



# SCALING METHODOLOGY AND SCALE REPORTING IN THE TREE2 PANEL SURVEY

Documentation of scales implemented in the baseline survey (2016) (Update 2023)

Stefan Sacchi

Dominique Krebs-Oesch



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University of Bern

Fabrikstr. 8

3012 Bern/Switzerland

www.tree.unibe.ch

tree.soz@unibe.ch

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#### **Abstract**

This documentation refers to the database of the 2<sup>nd</sup> TREE cohort's (TREE2) as published in the 2023 data release (TREE, 2023). It outlines the statistical models and estimation methods employed for scale construction and the calculation of student scores based on questionnaire items. Furthermore, we discuss the various metrics and indicators of relevant scale properties compiled in the technical appendix for all scales implemented in the TREE2 baseline survey.

The focus of the scale reporting is on the internal consistency of the scales and on the comparability of the measurements across survey languages, survey modes and survey settings involved. With very few exceptions, the results indicate at least sufficient or high internal consistency and measurement invariance of the scales used.

A complementary documentation covering the scales employed in later panel waves can be found in the 2023 TREE2 data release (Sacchi & Krebs-Oesch, 2023). With the exception of a few additional metrics of longitudinal measurement invariance over panels waves (ibid., sections 3.6, 4.2), it basically relies on the methods presented below.

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## Some practical guidelines for using the scales

For each scale administered in the TREE2 baseline survey, the technical appendix of this documentation provides a selection of relevant scale metrics and quality measures. Section 4 of the introductory text describes the type and calculation of the reported measures and gives some clues as to their interpretation. We thus intend to support data users in assessing measurement properties of the scales in question. Note that for some of the scales administered in the baseline survey, one or more repeated measurements from later panel waves are available, which are documented in Sacchi & Krebs-Oesch (2023).

The reported scale-specific measures focus primarily on reliability (in the sense of internal consistency) and measurement invariance across survey settings, modes and languages. What we do not address in this documentation is scale validity, as TREE mostly uses commonly accepted, well-established scales and validity is therefore not likely to be a major problem. In addition, the database offers researchers many opportunities to conduct external validations tailored to their specific analytical needs.

In some cases, several scales in the TREE2 scientific use file partly draw on one and the same items. The scales in question should therefore not be used simultaneously within the same multivariate model. This concerns some scales for which several versions exist (cf. section 2: scales surrounded by dotted lines in Table 3) as well as other scales composed of main and subdimensions (cf. section 2, Table 4).

Regarding the use of student scores in the context of multivariate models, we refer the reader to the remarks on this issue in section 3.2.2. Some scores represent item composites rather than scale scores (cf. Table 5), which may, however, be used similarly. The variable names (short names without wave-specific prefix) and labels of all items, student scores and composite variables in the technical appendix correspond with those in the TREE2 data release (TREE, 2023).

When estimating the confirmatory factor models and calculating the student scores, we imputed all missing item information, provided that at least one item of a given scale had a valid rating (see section 3.1.1b for details).

#### Introduction

This paper documents the questionnaire-based scales and item-based composites that have been collected on the occasion of the baseline survey administered to the second TREE cohort (TREE2) in 2016. First, the paper focuses on the methods and the estimation procedures that we have adopted for the calculation of the scale values published in the scientific use data files. Second, we describe the calculation of the scale-specific key figures and quality parameters (see appended tables) and provide some useful information for their interpretation.

The TREE2 baseline survey is composed of two surveys carried out at a short interval in spring/summer 2016. The first survey is a large-scale national assessment of mathematics skills administered to students who had reached the end of compulsory school (Assessment of the Attainment of Educational Standards, henceforth AES).¹ Beyond the assessment itself, the AES survey programme included a comprehensive student background questionnaire that collected a wide range of student background characteristics presumed to influence maths skills development and/or educational and labour-market pathways in the further (post-compulsory) life course. The second survey, which we refer to as extension survey, was conducted shortly after the first one. Its main purpose was to complete some student background characteristics that had not been collected among all respondents of the first survey. In doing so, TREE was able to substantially extend the size of the TREE2 starting cohort (see section 1 for details).

All parts of the AES student questionnaire include numerous item-based measures designed to capture latent (i.e., not directly observable) respondent, family or context characteristics. Instrument selection was largely restricted to instruments validated by previous research in the relevant research fields (see section 2 for details).

The documentation of scales pertaining to the AES survey was first published along with the AES data in 2017 (Sacchi & Oesch, 2017).<sup>2</sup> The present documentation covers the extended, more complex database of the TREE2 baseline survey, which also includes data from the extension survey described above. From a methodological point of view, this raises the issue of potential survey-mode and setting effects: The AES assessment was conducted in a uniform proctored classroom setting supervised by carefully instructed test administrators; the extension survey, by contrast, took place in an unproctored individual setting outside of school.

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The survey is part of an overarching assessment scheme implemented by the Swiss Conference of Cantonal Ministers of Education (EDK) to test basic skills in key subject areas at various stages of compulsory education. For details, see <a href="https://www.icer.unibe.ch">www.icer.unibe.ch</a> and <a href="http://wegk-schweiz.ch/">http://wegk-schweiz.ch/</a>).

<sup>&</sup>lt;sup>2</sup> See <u>forsbase.unil.ch/project/study-public-overview/16165/0/.</u>

Furthermore, the latter employed two sequentially applied survey modes (web survey and paperand-pencil questionnaire). With regard to scaling, this incongruence requires that we have to carefully check for measurement invariance across survey settings and modes. Consequently, this documentation includes a number of relevant invariance tests and parameters for all scales that are based on data from the extension survey.

Beyond psychometric scales stricto sensu, this documentation also includes a number of item sum scores based on two or more single items. However, we have not included scores of test results and other types of composite variables.<sup>3</sup>

For all scales and composites drawing exclusively on data of the AES assessment survey, we report the previously calculated parameters (Sacchi & Oesch 2017) in the technical appendix of this documentation. In doing so, we provide TREE2 data users with an overview of all scales and composite variables available in the TREE2 baseline survey in one single document (see particularly section 2). The introductory text describing the methods of calculation and estimation used and the parameters reported in the technical appendix largely corresponds to the 2017 AES documentation (ibid.).

For each of the scales, we report estimates (i.e., scores) of the individual scale values for all participating students. In addition, our documentation aims at enabling data users to assess the scales' quality and measurement invariance (cf. particularly the technical appendix). Last but not least, our documentation ought to allow scholars to replicate, if they wish to do so, the calculation of models, tests and scale parameters and compare them with alternative specifications.

In the following sections, we first specify some relevant aspects of the TREE2 baseline survey's design (1), the selection and adaptation of the scales (2) as well as the statistical modelling and calculation of the scale values (3). Finally, we specify how the scale-specific results, reliability and quality checks were calculated and give some information on how to interpret them (4).

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As for the scales, the extension survey considerably enlarges the database on which these scores rely.

## 1 Survey Design and Database

The data of the AES survey were collected by means of a computer-based classroom survey among a random sample of approximately 22,000 students who were in their last year of lower secondary education (i.e., the 11th year<sup>4</sup> of compulsory schooling).<sup>5</sup> The survey included a comprehensive test of basic mathematical skills, along with a computer-assisted self-interview (CASI) of approximately 45 minutes. Among other things, the student questionnaire covered a broad selection of psychometric and other item-based measures, which are the subject of this documentation.

AES implemented a modular design with two different versions of the questionnaire, each of which were administered to a randomised split-half of the total sample. The main building block of one version was the mathematics module, which mainly covered student, teacher and classroom characteristics relevant to the successful acquisition of mathematical skills during compulsory education and to related didactical and pedagogical research. The core of the second version was a student background module co-designed by TREE to collect information on a broad range of resources of the surveyed students, their families and the schools they were attending at the moment of the survey. This module was specifically developed for the TREE2 panel survey in order to measure, as comprehensively as possible, the starting conditions deemed to be relevant for the respondents' further education and labour-market careers and their life courses in general. Both questionnaire versions included a common core ('general questions') that was completed by all students participating in AES. The common core incorporated items that are of general interest for the research objectives of both modules.

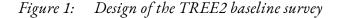
Due to the modular design of the AES questionnaire, a substantial part of the questionnaire pertaining to TREE-relevant starting conditions of post-compulsory pathways was administered to only half of the AES sample (see *Figure 1*). In order to complete the missing items for the respondents to the other half (termed 'maths sample split' in Figure 1), TREE carried out an out-of-school 'extension' survey immediately after the AES survey. With a few exceptions, the questionnaire used for this survey was equivalent to that of the background module in the AES

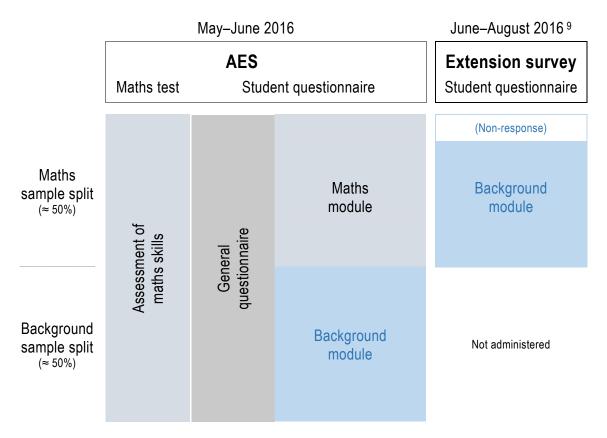
Including two years of kindergarten.

<sup>&</sup>lt;sup>5</sup> See Verner and Helbling (2019) for a detailed description of the sampling and the population.

The random assignment of the students to one questionnaire version was to guarantee that - within each school and each test session - both versions were evenly distributed over the 13 different test booklets used for the preceding mathematics assessment. Hence, from the students' perspective, booklet and questionnaire version were two independent, fully exogenous conditions.

survey, which was implemented in two 'standalone' versions, either in the form of a web or a paper-and-pencil questionnaire. The minor adaptations of the questionnaire under these changed setting and mode conditions included slightly modifying the order of instruments and adding a newly designed scale that had not been administered in the AES survey.<sup>7</sup> Apart from that, the web implementation was largely indistinguishable from the CASI instrument used by the AES.<sup>8</sup>





In every canton, the extension survey was carried out as soon as the AES survey had been concluded in all sampled schools. The web survey was implemented as the primary mode. Students who did not participate in the web survey received the questionnaire's paper-and-pencil version by mail as a secondary mode. As both survey modes are self-administered, they are well suited for the partly sensitive questionnaire items included in the extension survey. With this

Two additional elements were placed at the end of the questionnaire: a brief cognitive skills test (KFT 4–12 + R; Heller & Perleth, 2000) as well as an experimentally varied repeated measurement of parental education.

To maximise comparability with the AES CASI (and contrary to the web surveys in later TREE2 waves), the web mode was not adapted for smartphones (and respondents were asked to complete it on a computer).

The median lag between the AES and extension survey was 29 days. 98 % of respondents completed the questionnaire between June and August, with a few pencil-and-paper questionnaires being returned up to the end of October.

mixed-mode design, the extension survey achieved a total response rate of almost 75% (73.3% if we consider only complete questionnaires; see also Table 1). Taking the relevant methodological literature into consideration, we do not expect significant mode effects (de Leeuw & Hox, 2011; de Leeuw, 2018; for proctored surveys see also Colosante et al., 2019).

As *Table 1* illustrates, the extension survey enabled us to substantially enlarge the available initial TREE2 sample base with a comprehensive measurement of relevant starting conditions. Among other things, this also allows for a more precise estimation of the scaling models and parameters that are at the centre of this documentation.<sup>10</sup> In light of the sample structure displayed in Table 1, it is important to address the issue of measurement invariance across the various survey settings and modes. That is why this documentation also provides statistical tests and quality measures that are relevant to this end (see section 4 and the technical appendix). The estimation of *setting effects* thereby draws exclusively on the CASI and the web survey, which rely on virtually interchangeable survey modes (i.e., it excludes the paper and pencil questionnaires,  $n = 15\ 608$ ). And the estimation of *mode effects* draws exclusively on the extension survey (i.e., it excludes the classroom setting,  $n = 5\ 119$ ). In doing so, we avoid the risk that the estimations of mode and setting effects are mutually confounded.

Table 1: Sample size and structure of the TREE2 baseline survey

	AES	Exter	nsion survey 1)	Total
Survey Setting:	Proctored classroom survey	Unproctored i	ndividualised setting	
Survey Mode: CASI		Web survey	P&P questionnaire	
(Sub-)sample size 2)	11 124 <sup>3)</sup>	4 484	635	16 243

1) Including 89 incomplete questionnaires (with data for some scales only), which are treated as nonresponses when it comes to response statistics and the published sample weights (see also FN 10). 2) The number of cases for particular scales will generally be lower due to non-imputable missing values. 3) Background sample split (cf. Figure 1).

<sup>-</sup>

Regarding the scales partly relying on the extension survey, we draw on a customised sample weight tailored to the sample available for scaling purposes (cf. footer of Table 1). There are two types of non-negligible sample attrition, which exclusively affect the maths sample split (i.e., the unwillingness of AES respondents to provide their contact data for the TREE panel survey and non-participation in the extension survey). Given the high AES response rate of 93% (see Verner & Helbling, 2019: 39), the background split is therefore markedly less affected by attrition. The customised weight accounts for general and split-specific sources of attrition (see section 3.1.1a and FN 27 for further details).

These considerations do not affect the calculation of any of the scales administered in the general questionnaire and the AES maths module, as these scales do not rely on the extension survey. For calculations based on the general questionnaire, we can draw on data of the complete AES sample (approx. 22 000 students) and, for calculations based on the AES maths module, on the subsample to which the maths module was administered (approx. 11 000 students; cf. Figure 1). To ensure a statistically efficient estimate, the scaling models generally draw on the entire available sample base, including cases which, for various reasons, are not included in the scientific use files of the TREE2 dataset (Hupka-Brunner et al. 2023).<sup>11</sup>

In a survey administered in several languages, we also have to be careful regarding measurement invariance across survey languages (in our case German, French and Italian), which concerns all scales administered. <sup>12</sup> Basically, variance across languages can be the result of 'real' cultural or linguistic differences between language regions but also of inaccurate translations. That is why we report language-specific invariance tests and parameters (section 4 and appendix). As *Table 2* reveals, sample size substantially varies across survey languages.

Table 2: Breakdown of estimation samples by survey languages

Scales implemented in	General questionnaire	Background module	Math module
Available Estimation Sample 2)	Full AES sample	Baseline survey 2)	Math subsample
Survey Language:			
German	16 349	11 698	8 106
French	5 235	3 927	2 646
Italian	755	618	379

<sup>1)</sup> Number of cases for specific scales will in general be lower due to non-imputable missing values. 2) Cf. Table 1.

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Data users who wish to estimate or replicate scaling models drawing on the complete database may do so. As the data excluded from the published data files are highly confidential, however, this is possible only on the premises of the study's headquarter in Bern and using a specially protected computer workplace.

In the AES, the survey language is identical with the teaching language of the sampled schools. In the extension survey, respondents were able to choose the survey language. In a few cases, this led to the situation that the extension survey was not completed in the same (national) language as the AES survey.

## 2 Selection and Adaptation of Scales

The AES questionnaire incorporated a broad range of more than 90 item-based instruments from relevant research areas (for theoretical considerations regarding the selection of instruments, see Hupka-Brunner et al. [2015] and Hascher et al. [2019]). As a general rule, preference was given to well-established, cross-disciplinary validated instruments used in surveys both in Switzerland and abroad.

A first selection of instruments was thoroughly pretested in the year preceding the main survey (2015).<sup>13</sup> One important objective of the pretest was to assess measurement properties of the preliminary selection of questionnaire instruments and scales in the Swiss context. This included assessments of the dimensionality, reliability and the cross-language measurement invariance of the scales. Some of the scales had to be newly translated to make them available in all survey languages. In these cases, the pretest was used to check measurement invariance across language versions and to improve improper translations. Moreover, the pretest was used to clean up scales with dodgy items, to shorten others and, lastly, to narrow down and optimise the selection of instruments for the main survey. We shortened many scales to three or four items to ensure a comprehensive coverage of relevant concepts without unduly increasing response burden and interview duration.

Wherever possible, the original instruments were implemented without modification in order to preserve measurement properties of the selected scales and to maximise data comparability. However, given the multitude of aspects to be considered in questionnaire construction (Dillman, Smyth & Christian, 2014), slight adaptations of the original instruments often could not be avoided.<sup>14</sup>

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The main objective of the pretest was to improve the assessment of mathematical skills, the design of the student questionnaire and the fieldwork for the main survey. The pretest sample was split evenly across the three test languages, German, French and Italian, and included more than 2 000 students from 70 schools.

The manifold methodological, empirical and substantive reasons for such adaptations include the following: At the methodological level, there was the need to adapt instruments that were originally developed for a different survey mode (de Leeuw, Hox & Dillman, 2008: 311f.) and to standardise the format of each type of question in order to reduce the response burden and improve comprehensibility (Dillman, Smyth & Christian, 2014: 210f.). Empirically, the pretest in some instances uncovered insufficient cross-language measurement invariance, which suggested the need to check and, in some cases, improve the translations of the instruments. Finally, there was the requirement to closely replicate some of the instruments from the first TREE cohort (TREE1).

The modifications of the original instruments can pertain to both the question format and wording of stimuli as well as to the response scales and sometimes even to the items. In most cases, however, they are minor so that a substantial impact on the measurement properties and comparability of the resulting scales seems unlikely. It should also be noted that, for similar reasons, many popular scales are far less standardised in survey practice than generally perceived. Moreover, in the case of several circulating scale versions, the original version of the scale is not necessarily the most appropriate.

Table 3 conveys a topically ordered overview of all scales and item-based instruments that were implemented in the AES main field. The 'Positive Attitude towards Life' scale was administered in the extension survey only. In a few cases, several scales partly rely on the same items. Consequently, they should not be introduced in one and the same multivariate model. Apart from scales involving main and sub-dimensions, the scales in question are framed by a dotted line in Table 3. For the 'Global self-esteem scale' (and one of its subdimensions) a shortened version implemented in later waves of TREE2 is also available (see scale reporting in the appendix).

To enable comparative analyses between TREE1 and TREE2, the range of implemented instruments also includes some original scales used in the PISA 2000 survey, the baseline survey of the first TREE cohort (TREE1). For some of these scales (family wealth, social and cultural communication within the family), we implemented both the original version already used in PISA 2000 and an adapted version that was optimised for TREE2. The former is preferable for comparative analyses of both cohorts, the latter for analyses of the second cohort only.

Table 3: Item-based scales and composites (without scales for subdimensions)

Survey topic		AES question-			
Scale / composite	[Variable name] 1)	naire module 2)	Source <sup>3)</sup>		
Family background					
Family climate					
Emotional closeness to parents	[closep_comp]	Background	TREE1 - based on Szydlik, 2008		
Parental pressure to achieve	[press_fs]	Background	Böhm-Kasper et al., 2000		
Parents' achievement expectations	[expectp_fs]	Math	Hascher et al., 2019		
Mother's achievement expectations	[expectm_fs]	Math	Hascher et al., 2019		
Father's achievement expectations	[expectf_fs]	Math	Hascher et al., 2019		
Mother's social norms about mathematics	[socnormsm_fs]	Math	PISA 2012		
Father's social norms about mathematics	[socnormsf_fs]	Math	PISA 2012		
Family educational support (PISA2000) 4)	[famedsup_fs]	Background	PISA 2000		
Social communication (PISA2000) 4)	[soccom_fs]	Background	PISA 2000		
Social communication (adapted TREE2)	[soccom_m_fs]	Background	PISA 2000 (adapted TREE2)		
Social, cultural & economic resources					
Social capital (own)					
Perceived social network support	[closupp_fs]	Background	TREE2 (BHPS, ISSP 2003)		
Cultural capital (family of origin)					
Parents: reading interest	[joyreadp_comp]	Background	TREE2		
Cultural communication (PISA2000) 4)	[cultcom_fs]	Background	PISA 2000		
Cultural communication (adapted TREE2)	[cultcom_m_fs]	Background	PISA 2000 (adapted TREE2)		
Household possessions: classical culture (PISA2000) 4)	[cultposs_fs]	Background	PISA 2000		
Cultural capital (own)					
Embodied cultural capital	[inccap_fs]	Background	TREE2		
Cultural activities 5)	[cult_fs]	Background	PISA 2000 (partially adapted)		

<sup>1)</sup> Student score variable names from 2023 TREE2 data release. 2) Database by module: General  $\rightarrow$  full AES sample; background module  $\rightarrow$  TREE2 baseline sample; math module  $\rightarrow$  AES math sample split. 3) See technical appendix for a detailed list of sources. 4) Scales administered in the the first TREE cohort (TREE1). 5) A subscale of this scale has been adopted as is from PISA 2000 / TREE1 (cf. Table 4).

Table 3 (continued): Item-bases scales and comp Survey topic		AES question-	
Scale or composite	[Variable name] 1)	naire module 2)	Source 3)
Social, cultural & economic resources (continued)			
Economic capital (family of origin)			
Household possessions: family wealth (PISA2000) 4)	[wealth_fs]	Background	PISA 2000
Household possessions: family wealth (adapted TREE2)	1	Background	PISA 2000 (adapted TREE2)
Family affluence scale (FASIII)	[fasiii_comp]	Background	Hobza et al., 2017
Satisfaction and well-being	:		
Satisfaction			
Capabilities	[cap_fs]	Background	Sen, 1985; Anand & van Hees, 2006
School-related well-being			
Positive attitude towards school	[posatt_fs]	General	Hascher, 2004
Enjoyment in school	[enjoyschool_fs]	General	Hascher, 2004
Physical complaints in school	[physpain_fs]	General	Hascher, 2004
Worries about school	[trouschool_fs]	General	Hascher, 2004
Social problems in school	[socprob_fs]	General	Hascher, 2004
School reluctance	[schoolav_fs]	General	Hagenauer & Hascher, 2012 (modified)
Non-cognitive factors			
Motivational concepts			
Intrinsic achievement motivation	[achmoti_fs]	General	IGLU 2001
Extrinsic achievement motivation	[achmote_fs]	General	IGLU 2001
Instrumental learning motivation (PISA2000) 4)	[insmot_fs]	General	PISA 2000
Interest in reading (PISA2000) 4)	[intrea_fs]	General	PISA 2000
ICT interest	[ictintr_fs]	Math	ICILS 2013
Dispositional interest	[intsubj_fs]	Math	COACTIV 2008
Identified motivation (mathematics)	[instrumot_fs]	Math	PISA 2012
External motivation regulation	[extreg_fs]	Math	Ryan & Conell, 1989
Classroom participation	[engage_fs]	Math	Eder, 1995, 2007
Performance-approach goals (SELLMO)	[approxgoals_fs]	Math	SELLMO 2012
Learning goal orientation (SELLMO)	[learntarget_fs]	Math	SELLMO 2012
Work avoidance (SELLMO)	[avoidwork_fs]	Math	SELLMO 2012
Avoidance performance goals (SELLMO)	[avoidblame_fs]	Math	SELLMO 2012
Self-perception			
Global self-esteem 6)	[sel_fs]	Background	Rosenberg, 1979
General perceived self-efficacy scale (GSES)	[seef_fs]	Background	GSES (adapted TREE1)
Academic self-efficacy	[acaself_fs]	General	Hascher, 2004
Academic self-concept (PISA2000) 4)	[scacad_fs]	General	PISA 2000
Verbal self-concept (PISA2000) 4)	[scverb_fs]	General	PISA 2000
Maths self-concept	[matcon_fs]	General	PISA 2000 (adapted AES)
ICT self-concept	[ictabil_fs]	Math	ICILS 2013
Specific self-efficacy: numeracy	[selfeffa_fs]	(General) 7)	PISA 2012; Girnat, 2018
Specific self-efficacy: algebra	[selfeffb_fs]	(General) 7)	PISA 2012; Girnat, 2018
Specific self-efficacy: geometry	[selfeffc_fs]	(General) 7)	Girnat, 2018
Specific self-efficacy: probability	[selfeffd_fs]	(General) 7)	Girnat, 2018

<sup>1)</sup> Student score variable names from 2023 TREE2 data release. 2) Database by module: General  $\rightarrow$  full AES sample; background module  $\rightarrow$  TREE2 baseline sample; math module  $\rightarrow$  AES math sample split. 3) See technical appendix for a detailed list of sources. 4) Scales administered in the surveys of the first TREE cohort (TREE1). 6) Data and scale appendix also include a shortened 7-item-version of this scale. 7) Half of the items implemented in the math module.

Table 3 (continued): Item-bases scales and composites

Survey topic	1	AES question-	
Scale or composite	[Variable name] 1)	naire module 2)	Source 3)
Non-cognitive factors (continued)			
Emotions related to maths classes			
Mathematics anxiety	[anxmath_fs]	Math	PISA 2012
Mathematics boredom	[boredom_fs]	Math	AEQ-M (short-version)
Mathematics anger	[anger_fs]	Math	AEQ-M (short-version)
Mathematics enjoyment	[enjoymath_fs]	Math	AEQ-M (short-version)
Volitional strategies			
Perseverance	[persev_fs]	General	PISA 2012
Effort: learning (PISA2000) 4)	[effper_comp]	Background	PISA2000
Personality characteristics			
Big five: extraversion	[big5_e_comp]	Background	Rammstedt et al., 2014
Big five: agreeableness	[big5_a_comp]	Background	Rammstedt et al., 2014
Big five: conscientiousness	[big5_c_comp]	Background	Rammstedt et al., 2014
Big five: neuroticism	[big5_n_comp]	Background	Rammstedt et al., 2014
Big five: openness	[big5_o_comp]	Background	Rammstedt et al., 2014
Internal locus of control	[loci_comp]	Background	GESIS (short version)
External locus of control	[loce_comp]	Background	GESIS (short version)
Values & attitudes			
Work-related extrinsic value	[vawe_fs]	Background	TREE1 - based on Watermann, 2000
Work-related intrinsic value	[vawi_fs]	Background	TREE1 - based on Watermann, 2000
Family value	[vafa_comp]	Background	TREE1
Positive attitude towards life	[posl_fs]	Extension survey	TREE1; Grob et al., 1991
Attitudes related to mathematics classes			
Reality-based learning	[realref_fs]	Math	Girnat, 2015, 2017
Discovery / exploratory learning	[disclearn_fs]	Math	Girnat, 2015, 2017
Social learning	[soccomlearn_fs]	Math	Girnat, 2015, 2017
Instructivist learning	[instreplearn_fs]	Math	Girnat, 2015, 2017
System aspect	[sysformasp_fs]	Math	Girnat, 2015, 2017
Scheme aspect	[schemasp_fs]	Math	Girnat, 2015, 2017
Application aspect	[applyasp_fs]	Math	Girnat, 2015, 2017
Education and training			
Characteristics of maths lessons (end of lower secon	ndary education)		
Teacher: cognitive activation	[cogself_fs]	Math	COACTIV 2008
Teacher: classroom management	[classman_fs]	Math	COACTIV 2008
Teacher: individual learning support	[indsup_fs]	Math	COACTIV 2008
Teacher: instruction quality	[instqual_fs]	Math	PISA 2006
Situational interest	[intsit_fs]	Math	COACTIV 2008
Perceived autonomy support	[persuppauto_fs]	Math	Seidel, Prenzel & Kobarg, 2005
Perceived competence support	[persuppcomp_fs]	Math	Seidel, Prenzel & Kobarg, 2005
Perceived social relatedness	[persocincl_fs]	Math	Seidel, Prenzel & Kobarg, 2005
Classmates' appreciation of mathematics	[apprmath_fs]	Math	PISA 2012
Absenteeism / intention to change education			
Absenteeism / truancy	[truancy_fs]	General	PISA 2000, PISA 2012
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<sup>1)</sup> Student score variable names from 2023 TREE2 data release. 2) Database by module: General  $\rightarrow$  full AES sample; background module  $\rightarrow$  TREE2 baseline sample; math module  $\rightarrow$  AES math sample split. 3) See technical appendix for a detailed list of sources. 4) Scales administered in the first TREE cohort (TREE1).

In principle, all scales listed in Table 3 are one-dimensional, that is, they have been designed to measure *one* theoretical construct or latent dimension each.<sup>15</sup> However, some of the scales are composed of several sub-dimensions, each representing a facet of one overarching construct. As researchers may wish to distinguish between the sub-dimensions of these scales, the scientific use files of TREE2 also include student scores for each sub-dimension. The following table lists both the main and sub-dimensions of the scales in question.

Table 4 Scales with sub-dimensions

Scale – main dimension Variable name 1)		Subdimensions	Variable name 1)	
Background module scales				
Global self-esteem 2) 3)	[sel_fs]	Positive global self-esteem 4) Negative global self-esteem / depression 4) 5)	[sele_fs] [seld_fs]	
Embodied cultural capital	[inccap_fs]	Embodied cultural capital: manners Embodied cultural capital: verbal skills	[manners_fs] [verbskill_fs]	
Cultural activities	[cult_fs]	"Lowbrow" cultural activities "Highbrow" cultural activities (PISA2000) <sup>6)</sup>	[cultlow_fs] [culthigh_fs]	
Math module scales				
Parents' achievement expectations	[expectp_fs]	Mother's achievement expectations Father's achievement expectations	[expectm_fs] [expectf_fs]	
Instructivist learning	[instreplearn_fs]	Instructivist learning: teachers' instructions Instructivist learning: repetitive practice	[instrlearn_fs] [replearn_fs]	
Social learning	[soccomlearn_fs]	Social learning: social arrangement Social learning: communication	[soclearn_fs] [comlearn_fs]	
System aspect	[sysformasp_fs]	System aspect: logical thinking System aspect: formalism	[systasp_fs] [formasp_fs]	
Teacher: cognitive activation 7)	[cogself_fs]	Cogn. activation: finding solutions & arguing Cogn. activation: strategies and learning from mistakes	[cogselfa_fs] [cogselfb_fs]	

<sup>1)</sup> The short names of the student score variables in the TREE2 scientific use file are given in brackets. 2) In accordance with Huang et al. (2012) and Donnellan et al. (2016), this scale is clearly two-dimensional in the TREE2 baseline survey. 3) Data and appendix also include a shortened 7-Item-Version of this scale ( $sel\_m\_fs$ ). 4) Sub-dimension labels according to Huang et al. (2012). 5) Data and appendix also include a shortened 3-item-version of this subscale ( $seld\_m\_fs$ ). 6) Corresponds to 'Cultactv' scale in PISA 2000/TREE1. 7) As this scale is not one-dimensional in the AES survey, we distinguish two (inductively optimised) sub-dimensions.

Some of the instruments described in this documentation are based on two items only, making it impossible to fit any scaling model to the data. Henceforward, we call scores derived from

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One should note, however, that the one-dimensionality of the selected scales may be empirically controversial. For one scale, 'Global Self-Esteem' (according to Rosenberg, 1979; 2014), we are aware that this is the case (see von Collani & Herzberg, 2003; Huang & Dong, 2012; Donnellan, Ackerman & Brecheen, 2016). With respect to this scale, we decided to provide the student scores for both the one-dimensional model and for the two sub-dimensions described in the literature. Hence, we treat this scale the same way as other scales with sub-dimensions and leave it up to the data users to decide on the appropriate scaling solution.

mostly short, item-based instruments *item-based composites* (for an overview see *Table 5*).<sup>16</sup> In case of the 'Family affluence scale' in Table 5, the term «scale» is a misnomer as it represents de facto a sum score, i.e., an item-based composite (for details, see Hobca et al., 2017).<sup>17</sup>

Table 5: Item-based composites

Concept 1)			
Dimension	Variable name 2)	Number of items	
Big Five Inventory			
Extraversion	[big5_e_comp]	2	
Agreeableness	[big5_a_comp]	3 3)	
Conscientiousness	[big5_c_comp]	2	
Neuroticism	[big5_n_comp]	2	
Openness	[big5_o_comp]	2	
Locus of control			
Internal locus of control	[loci_comp]	2	
External locus of control	[loce_comp]	2	
Effort: learning (PISA2000) 4)	[effper_comp]	2	
Family values	[vafa_comp]	2	
Parents: reading interest	[joyreadp_comp]	2	
Emotional closeness to parents	[closep_comp]	2	
Family affluence scale (FASIII) FN17	[fasiii_comp]	6	

<sup>1)</sup> With the exception of 'Effort: learning' (general questionnaire, full sample), all composites belong to the background module. 2) The short variable names of the composite scores in the scientific use file are reported in brackets. 3) For the composite with one extra item, see Rammstedt and John (2007: 210). 4) This composite has been previously administered in the surveys of the first TREE cohort (TREE1).

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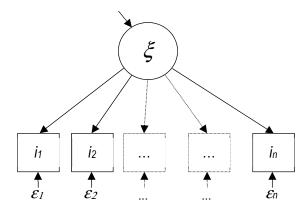
<sup>&</sup>lt;sup>16</sup> For item composites, student scores are calculated from imputed item ratings (cf. 3.1.1 b).

Note that this composite partly draws on the same items as the wealth scales in Table 3.

## 3 Statistical Modelling

As mentioned above, the scales in the AES questionnaire are item-based instruments intended to measure *one* theoretical construct each. Confirmatory factor analysis (CFA) is a common approach to the empirical estimation of latent (i.e., not directly observable) characteristics captured by such measurement instruments (see, e.g., Long, 1983; Schmitt, 2011). As our selection of scales is restricted to validated instruments that were designed to measure a common latent dimension, we limit ourselves to fitting a straightforward one-dimensional CFA model (see Aichholzer, 2017: 80–84) to each scale-specific item set. The CFA model illustrated in *Figure 2* relies on *n* items ( $i_1, i_2, ..., i_n$ ) with associated item-level measurement errors  $\mathcal{E}_n$ , which all measure the same latent dimension  $\xi$ . For scales with several subdimensions (see Table 4 above), a separate CFA model is fitted to each subdimension.<sup>18</sup>

Figure 2: One-dimensional confirmatory factor model



For every model estimated hereafter, selected model parameters, fit statistics and scale quality measures are reported in the technical appendix (p. 34ff.). This includes a test of one-dimensionality, various measures of internal scale consistency as well as tests and indices of measurement invariance across survey languages and, where appropriate, survey settings and modes. Throughout this documentation, our primary focus is the quality of the scales (and the corresponding student scores) rather than model fit. If the fit of the straightforward one-factor model turns out to be poor, we neither modify the model to improve fit nor do we test alternative (e.g., multi-dimensional) models. It is up to the data user to judge whether the one-dimensional CFA models are appropriate and whether the scales have the required properties.

An alternative approach would be to fit second-order CFA models to each dimension (Aichholzer, 2017: 89f.).

#### 3.1 Estimation of the confirmatory factor models

In its standard form, structural equation modelling - including CFA as a special case - relies on a number of quite restrictive assumptions that are hardly ever met in practice. Basically, the observations should be independent, and the indicators should be measured on a continuous scale (interval-level measurement) and follow a multi-normal distribution (see, e.g., Hoyle, 2000). As regards the database of the AES and the TREE2 baseline survey, none of these assumptions holds: The two-stage sampling procedure implies that observations are clustered within schools (see Verner & Helbling, 2019) and hence are not independent. Moreover, measurement of the indicators is at ordinal (or binary) level as it mostly relies on Likert-type rating scales. And last but not least, the skewed univariate distributions of many ratings are hardly consistent with the required multivariate normality.

The methodological literature offers a wide range of suggestions on how to relax some of the assumptions of the standard SEM model and how to deal with ordinal, binary or skewed indicators and clustered observations (cf., e.g., Bryant & Jöreskog, 2016).<sup>19</sup> In particular, the suggestions include two-stage estimation methods that exploit polychoric correlations and generalised structural equation models (GSEM) that are suited for short response scales and categorical indicators (Rhemtulla, Brosseau-Liard & Savalei, 2012; Bryant & Jöreskog, 2016). However, there is currently no well-established, generally accepted estimation approach tailored to both ordinal indicators that are not normally distributed and a complex sample with clustered observations.

We therefore follow the recommendations of Rhemtulla et al. (2012; similarly Harpe, 2015: 843) regarding the accurate estimation of CFA models on the basis of ordinal, Likert-type indicators. They suggest two different estimation strategies depending on the length of the rating scales. For item responses that rely on a rating scale with at least five points (i.e., ordered discrete response categories), they suggest a two-step estimation based on polychoric correlations. For item evaluations that rely on shorter rating scales with four or less points, a generalised structural equation model (GSEM) is in order. Below, we describe these estimation strategies in more detail.<sup>20</sup> As our primary goal is to estimate accurate student scores, we also implement some

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<sup>&</sup>lt;sup>9</sup> Clustered observations may not only affect variance estimation and model fit but also bias the estimation of model parameters (i.e., factor loadings; cf. Stochl et al., 2016; Muthén & Satorra, 1995; Wu & Kwok, 2012).

All calculations were performed using Stata version 15.0 (AES) and 16.1 (TREE2 baseline survey). For both strategies, model estimations in general converge without problems. In a few cases, mostly in multi-group models, it was necessary to constrain an error variance or to collapse smaller groups to achieve convergence (which is noted in the scale reporting of the scales concerned, see appendix).

sensitivity checks to assess the equivalence of student scores obtained via alternative modelestimation strategies (see section 3.2.1).

#### 3.1.1 Two-step estimation based on polychoric inter-item correlations

The two-step approach starts with the estimation of a matrix of polychoric correlations between all items of a given scale (tetrachoric correlations, respectively, in the case of dichotomous items).21 In the second step, maximum likelihood estimation is used to fit the one-dimensional CFA model from Figure 2 to the resulting correlation matrix.<sup>22</sup> The models are identified by setting the loading of the first item and the variance of the latent factor to one. The CFA models are also estimated separately for each of the three language subsamples. This allows for multigroup analysis designed to test and assess measurement invariance across the survey languages (see section 4 and, e.g., Steinmetz et al., 2008; Milfont & Fischer, 2015).

Below, we briefly describe how we deal with (a) the complex AES sample and (b) with missing item values in the context of the two-step estimation approach.

#### (a) Complex sample design and survey weighting

The AES survey relies on a random sample of students that was disproportionally stratified by cantons and type of cantonal curriculum (Verner & Helbling, 2019).<sup>23</sup> Furthermore, the samples analysed here are also affected by sample attrition. An unbiased estimation of any population characteristic therefore requires the application of an appropriate survey weight to account for the disproportional sampling design as well as for unit nonresponse. This also pertains to the estimation of polychoric correlations or the parameters of the CFA models to be estimated (e.g., factor loadings).24

A polychoric correlation is defined as the maximum likelihood estimate of the correlation between two hypothetical, normally distributed continuous latent variables derived from two corresponding ordinal indicators. Estimations were calculated using the Stata package "polychoric" by Stas Kolenikov (from http://staskolenikov.net/stata).

Maximum likelihood estimation has been found to be among the most appropriate estimation methods (together with ULS and DWLS; see Yang-Wallentin, Jöreskog & Luo, 2010) for analysing polychoric correlations derived from ordinal indicators.

Lower secondary schools in Switzerland are mostly "tracked", that is, students are enrolled in separate programmes with varying academic requirements.

<sup>&</sup>lt;sup>24</sup> Weighting would only be unnecessary in the case of a strict invariance of the postulated scaling model across subpopulations of any kind. If this strong assumption were met, the damage of unnecessarily applying survey weights would be limited to inflating the variances of the estimates to some degree (Bollen, Tueller & Oberski, 2013). Given the huge AES sample, this would not be too disturbing.

When estimating the polychoric correlations, we therefore use one out of three different survey weights, depending on whether a given scale is embedded in the background module, in the maths module or in the general questionnaire. For the scales from the latter two, we rely on the suitable AES weights.<sup>25</sup> With regard to AES, module-specific analyses require particular weights, as the sampling design of the randomised sample split for the distinct questionnaire modules (according to Figure 1) differs with respect to the shape of disproportional cantonal stratification.<sup>26</sup> On the basis of the module-specific AES weights, we have constructed an additional weight for the TREE2 baseline survey, which accounts not only for the AES sampling design and nonresponse but also for sample attrition in the extension survey.<sup>27</sup>

As regards the two-step estimation approach, it should be noted that variance estimation does not account for the clustering of observations within schools implied in the two-stage sampling (see Verner & Helbling, 2019).

#### (b) Handling of missing item values

Missing item values are not a major problem affecting the scales in the AES survey. As usual in surveys, however, there is a small share of missing item values, owing mainly to item non-response. With the exceptions mentioned below, the share of cases with missing information on at least one item of the scale does not exceed 5%. For two out of three scales, the percentage is below 1%.

A considerably higher share of missing values results for half of the items of each of the four scales that measure different facets of 'specific self-efficacy' in mathematics. This is a direct consequence of the questionnaire design (and therefore not a matter of methodological

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We use the respective non-response adjusted weights from the AES scientific use file ('smp\_w\_nrastubw' for the scales of the general questionnaire and 'smp\_w\_qmatb' for the scales of the maths module).

The reason is that the design of the two complementary sample splits has been optimised for two different purposes: The sample split drawn for the background module is designed to maximise statistical power at the national level, whereas the maths module split is optimised for separate analyses of cantons. In a nutshell, this was achieved by developing a disproportional subsampling scheme that further reinforces the general overrepresentation of small cantons among the sample split with the maths module and reduces it among the sample split with the background module. The weights for the sample splits then correspond to the general survey weight from the AES scientific use file ('smp\_w\_nrastubw') multiplied by the inverse of the within-canton subsampling fraction (see also Verner & Helbling, 2019).

For the baseline survey, we use an entropy-balancing weight (cf. Hainmueller, 2012; Hainmueller & Xu, 2013) that compensates for the AES disproportionate sampling design (incl. non-response adjustments) and, as far as the math-sample split is concerned, for the non-response related to willingness to be (re-)contacted and to participate in the extension survey (for details, see the TREE2 documentation on weighting: Sacchi, forthcoming). For the purpose of scaling, the e-balancing weight for the TREE2 baseline survey was re-estimated by taking into account the somewhat looser definition of survey participation employed throughout the scaling process (see Table 1 and the explanatory text).

concern<sup>28</sup>), as half of the items of each of these scales were incorporated into the general questionnaire and the other half into the maths module. This implies that the share of missing item information is close to zero for the general questionnaire, whereas it rises to around 50% for the items implemented in the maths module.

A relatively high share of missing values is also observed for two measures in which students evaluate the items on a rating scale that includes an explicit 'don't know' option. This pertains to the scale measuring 'Perceived social network support' (closupp\_fs) and the two-item composite for 'Parental reading interest' (joyreadp\_comp). For both instruments, the share of missing information rises to 10.4 and 8.7%, respectively, when explicit don't-know answers are included.<sup>29</sup>

Finally, there are four instruments containing some items that could not be administered to a minor portion of the sample.<sup>30</sup> With one exception, the overall share of cases with at least one missing item does not exceed 5% in these instances.<sup>31</sup>

These special cases and exceptions notwithstanding, the fraction of missing items is low to very low for the bulk of the scales. Hence, the impact of missing item information is presumably limited.

We applied *multiple imputation* to cope with missing values when estimating the scaling models (Rubin, 1996; White, Royston & Wood, 2011). Basically, missing item information was imputed - scale-by-scale - on the basis of all valid items pertaining to the same scale. The imputed samples thus cover all cases with a valid response for at least one of the items of a given scale. Given the ordinal measurement level of the item ratings, we applied chained equations with an ordinal (or, in a few cases, binary) logit link to create samples with imputed values (Royston, 2011). Following the rules of thumb given in White et al. (2011: 388), we set the number of

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The randomised allocation of students to questionnaire modules ensures that the missing-at-random assumption (MAR), which is crucial for the imputation of missing values, is almost perfectly met here.

Missing item values owing to explicit don't-know answers and item non-response were imputed together.

Some items referring to specific relatives (e.g., the father) have not been administered when the students previously indicated that these relatives do not exist (this pertains to the instruments: Family Education Support, Parents Achievement Expectations, Parents Reading Interest and Emotional Closeness to Parents). The resulting missing values were treated the same way as other types of missing information. Although this is perhaps not an ideal solution in these cases, a substantial bias seems unlikely given the mostly very low number of cases to which this applies.

The exception is the 'Family Educational Support' scale (famedsup\_fs) for which the share of cases with at least one missing item amounts to 14.6%. This owes mainly to the item tapping sibling support, which was not administered among students who previously indicated that they have no siblings (see footnote 29).

imputations to five.<sup>32</sup> For each imputed dataset, we separately calculated a matrix of polychoric correlations and combined it to estimate the CFA models.<sup>33</sup>

For each scale-specific CFA model, we calculated statistics and indices describing factor structures, model-fit and scale properties (see section 4 and the technical appendix).

#### 3.1.2 Generalised structural equation model for short response scales

If scales rely on item evaluations with short response scales of four or less points (including binary items), they were analysed using a generalised structural equation model (GSEM), as recommended in the literature (Rhemtulla, Brosseau-Liard & Savalei, 2012; Bryant & Jöreskog, 2016). Model parameter estimates were derived in one step directly from the microdata through numeric integration.<sup>34</sup> Contrary to the two-step approach, this amounts to a full-information, true maximum likelihood method (Bryant & Jöreskog, 2016: 192). We henceforth adopted the GSEM version of a one-dimensional CFA model, mostly with an ordinal logit link to account for the ordinal measurement level of the item sets to be analysed.<sup>35</sup>

#### (a) Accounting for the complex survey design

GSEM, as implemented in Stata, is able to account for complex sample designs. In particular, we used survey weights (as described in 3.1.1a) to obtain unbiased population estimates of the model parameters and applied cluster-robust variance estimation, which controls for the clustering of students within schools. Still, we assume that there is no substantive variation in the measurement model across schools (cf. Wu & Kwok, 2012).

#### (b) Handling of missing item values

GSEM estimation proceeds on an equation-by-equation basis. In the context of a simple onedimensional CFA model, this amounts to an implicit treatment (i.e., imputation) of missing item values, as each item is represented by a separate equation.

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The relatively low number of imputations seems appropriate for two additional reasons: First, we are primarily interested in unbiased point estimates of population parameters (e.g., factor loadings) and to a lesser degree in between-imputation and sampling variances. Second, some exploratory reproducibility checks, as suggested by White et al. (2011: 387), indicate that the polychoric correlations and other point estimates are highly stable for an even smaller number of imputations.

After applying Fisher's z-transformation, we simply average the correlation matrices and transform them back (see also footnote 31).

Integration mostly relies on mean-variance Gauss-Hermite quadrature with seven integration points (StataCorp, 2017: 562)

<sup>35</sup> The ordinal logit link reduces to a simple logit link for the two scales that include binary items.

One drawback of the GSEM approach is that the calculation of most established statistics to describe model fit and scale properties is not straightforward. This is why we complemented the GSEM estimations for the item sets with short response scales by a separately estimated two-step model, as described in section 3.1. If the resulting factor structures and student scores do not substantially differ from those obtained via the GSEM approach, this may be taken as indirect evidence that the two-step approach works sufficiently well and its assumptions are met (in the appendix, we therefore also check for the equivalence of both types of student scores). Hence, the model and scale statistics taken from the two-step CFA model are likely to be valid approximations as well.

#### 3.2 Student scores

#### 3.2.1 Calculation and robustness of student scores

For instruments relying on item rating scales of 5 or more points, the student scores in the scientific use file (and the related descriptive statistics in the appendix) represent regression factor scores (see StataCorp, 2017: 582f. for details) from the two-step CFA models described in section 3.1.1. For scales based on item sets with short response scales (four or less categories), the student scores in the SUF are empirical Bayes means based on the GSEM models (ibid.: 566). The variable names assigned to the student scores in the scientific use file are composed of a prefix indicating the survey wave (e. g. 't2' in case of the 2nd follow-up survey), the root of the variable names of the involved items and the suffix '\_fs', which is used as a marker for student score variables. The corresponding suffix for the item composites from Table 5 is '\_comp'. The variable labels assigned to the student scores and item composites correspond to those contained in the scale-specific documentation in the appendix. For an unequivocal interpretation of the student scores in the TREE2 scientific use file, we recommend inspecting the factor loadings (see section 4). As a general rule, however, a high factor score will indicate that students score high on the latent dimension that is designated by the label of the student score variable.

For all scales, the model, scale and test statistics reported in the appendix rely on the two-step estimation approach described in section 3.1.1. This explicitly also applies to those instruments based on short response scales, where the student scores (and the related factor-score descriptives in the appendix) are derived from a GSEM model. We also check the calculation of student scores for robustness by reporting the shared variance of both types of student scores (from SEM and GSEM) as measured by the coefficient of determination (CD) (see appendix: Equivalence of Scores from Two-Step Approach). If their shared variance is close to 100% (i.e., CD approaches

1), one may safely conclude, first, that the different modelling strategies have a negligible impact on student scores and, second, that it also seems reasonable to take the various fit and scale statistics obtained from two-step estimation as good approximations. As documented scale by scale in the appendix, the coefficient of determination is indeed close to 1 for most scales (> .94 for 42 out of 48 involved scales). There are six exceptions, however, in which the shared variance is substantially lower (between 60 and 90%), thus indicating that some of the additional assumptions needed for the two-step model have probably been violated. This pertains to the scales measuring 'Absenteeism / truancy' (truancy\_fs), 'Family wealth' as indicated by home possessions (both scale versions: wealth\_fs, wealth\_m\_fs), 'Cultural activities' including one of its subscales (cult\_fs, culthigh\_fs) and students' 'Maths self-concept' (matcon\_fs). For these scales, the model and scale statistics reported in the appendix should be interpreted with great caution, if at all. Still, this does not indicate that the student scores estimated via the GSEM approach are biased in any way.

For an additional robustness check for the student scores, we re-estimated the confirmatory factor models in s single step directly from the student microdata by using the MLMV method (StataCorp, 2017: 574). This allows us to control for the complex survey design through weighting and cluster-robust estimation and, at the same time, to implement an alternative full-information maximum-likelihood approach to account for missing item values.

Let us again look at the shared variances between the student scores obtained via the MLMV method and those via the two-step approach described in section 3.1.1 (see appendix: Equivalence of Scores from Robust MLMV).<sup>36</sup> With the exception of the aforementioned wealth scale (both scale versions), the shared variances uniformly exceed 96% (i.e., CD > .96) for all of the 87 scales in this documentation. This can again be taken as indirect evidence that the additional assumptions of the two-step approach regarding multivariate normal distributions and the measurement level are mostly met and, hence, that the statistics and indices derived from it are valid. To sum up, the robustness checks imply that with the few exceptions mentioned above, student-score estimates are very robust across the three different estimation methods recommended for the type of data analysed here.<sup>37</sup>

#### 3.2.2 Inclusion of student scores in multivariate statistical models

Instead of using the scale-specific student scores, there are often good reasons to embed scale-specific CFA models into a more comprehensive structural equation model of substantive

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A disadvantage of this method is that many statistics to judge model fit and scale qualities are unavailable.

<sup>&</sup>lt;sup>37</sup> This may be due to the fact that we analyse short, one-dimensional scales based on a large sample.

interest and to fit them all together in one step (cf., e.g., Aichholzer, 2017). It should be noted, however, that simultaneous estimation of both the measurement and the substantive part of a structural equation model is not necessarily always the best choice (cf. Devlieger & Rosseel, 2017): When one analyses a subsample of limited size, for instance, robust estimation of more complex models may be impossible. Moreover, even when the sample is large, misspecification bias in one part of a complex model may spread to other parts when they are fitted in a single step. A two-step approach employing previously estimated factor scores to investigate the substantive part of the model may have methodological merits in this respect (ibid.). This approach also has methodological drawbacks, however, basically because it implicitly treats factor scores as error-free measures of the latent dimensions to be analysed.<sup>38</sup> Some of the resulting problems, possible biases and correction methods are discussed, for example, by Croon (2002), Lu and Thomas (2008), Jin et al. (2016), and Devlieger and Rossel (2017).

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A random extraction of plausible values from the posterior distributions of the CFA models could be a quite obvious solution to this. However, contrary to skills assessment, this is an uncommon approach in the scaling of questionnaire items, possibly because of the reduced convenience this entails for data analysis.

## 4 Scale-specific reporting: Content and interpretation

In this section, we outline the various statistics, indices and quality measures reported in the scale appendix. For each scale (or subscale; cf. Table 4), this report includes two pages with a variety of scale-specific statistics. Below, we take the scale that measures 'Parental pressure to achieve' as an example to illustrate the scope and interpretation of scale-specific results. Figure 3 and 4, respectively, display the two pages of results for this scale as they appear in the appendix. Each scale reporting is linked with the full list of scales available in the baseline survey, and vice versa (link in the lower right corner of Figure 3). Unless otherwise specified, all reported results refer to the two-step estimation of the CFA model according to Figure 2. However, the student-scores descriptives refer to the scores obtained from the GSEM model, as the 'press' items are rated on a four-point scale (see section 3.2.1). The header of each scale-specific results section includes the name of the scale that is also used to label the related student-score variable in the 2023 data release (TREE, 2023). Furthermore, the headers specify the sample basis on which the calculations for the respective scales draw (baseline survey sample<sup>39</sup>, full AES sample or maths sample split).

The *model and fit statistics* reported include two likelihood-ratio tests as well as various common goodness-of-fit statistics, as discussed in the SEM literature (cf. Schreiber et al., 2006). The likelihood-ratio tests compare the current against the saturated model and the baseline model (basically postulating uncorrelated items), respectively. Ideally, we would expect a nonsignificant likelihood-ratio test of the current against the saturated model, which, for the reasons given above, is an unlikely result, however (see also van der Eijk & Rose, 2015). Moreover, for a well-fitting model, we expect the *comparative fit index (CFI)* and the *Tucker–Lewis index (TLI)* to approach 1, whereas the root mean square error of approximation (RMSEA) and the standardised root mean squared residual (SRMR) should be close to 0. Conventional cut-off criteria indicating a good fit between the hypothesised model and the observed data are  $\geq$  .95 for CFI and TLI  $\leq$  .06 for RMSEA and  $\leq$  .08 for SRMR (see Hu & Bentler, 1999). Regarding Figure 3, one could tentatively conclude that the one-dimensional CFA model fits the 'Parental pressure to achieve' scale sufficiently well, with some reservations regarding RMSEA and TLI, however. Two fit measures designed to compare different models, Akaike's information criterion (AIC) and the Bayesian information criterion (BIC), are also reported. They may serve as a point of reference if data users wish to fit alternative scaling models to the data. Finally, the *coefficient* of determination (CD) may be considered as an alternative measure of composite reliability (in

That is, the combined sample composed of the background split-half sample of the AES and the AES extension survey.

the sense of internal consistency; cf. Bollen, 1989: 220f.), to be interpreted similarly to the reliability measures below.

Figure 3: Example of the reported scale-specific results (first results page)

0	ale: Parental pressure	to achie	eve			Ba	seline survey sample
M	odel and Fit Statistics				Reliability and	Dimensionality	′
1)	Likelihood-ratio tests	chi2	df	p > chi2	Ordinal Cronbac	h's Alpha	.811
	Model vs. saturated	462	2	.000	(Cronbach's alph	a = .751)	
	Baseline vs. saturated	20063	6	.000	McDonald's Om	ega	.811
e) Root mean squared error (RMSEA) .12			.122	Test of (one-)din	nensionality (par	rallel analysis)	
	90% Confidence interval: lower bound			.113	Criterion: Retain	factors with adj.	eigenvalue > o
	90% Confidence interval: ι	pper boun	d	.131	A	djusted eigenval	ue
	Probability RMSEA <= 0.05			.000	factor 1	1.95	
					factor 2	04	
3)	Akaike's Information Criter	ion (AIC)		142462	factor 3	09	
	Bayesian Information Crite	rion (BIC)		142554	factor 4	18	
4)	Baseline comparison						
	Comparative Fit Index (CFI)			.977			
	Tucker-Lewis Index (TLI)			.931			
5)	Size of residuals						
	Stand. root mean squared re	sidual (SRI	MR)	.026			
	Coefficient of determination	(CD)		.816			

Standardized fa	ctor loading	Item descriptives								
							Std.	Valid		
Indicators	Coef.	(SE)	[95% Conf.	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
press1	0.69	0.01	0.68	0.70	press1	2.2	1.0	1	4	15488
press2	0.69	0.01	0.68	0.71	press2	3.0	0.9	1	4	15491
press3	0.78	0.00	0.77	0.79	press3	3.0	0.8	1	4	15488
press4	0.71	0.01	0.70	0.72	press4	2.8	0.9	1	4	15490
Parameters of g	eneralized s	tructura	l equation	model (ordi	inal logit link)					
Indicators	Coef.	Cut <sub>1</sub>	Cut <sub>2</sub>	Cut <sub>3</sub>						
press1	1.66	-1.38	0.68	2.99						
press2	1.79	-3.56	-1.79	0.80						
press3	2.35	-5.01	-2.26	1.38						
press4	1.84	-3.48	-1.23	1.53				Li	st of scal	les (wave 0)

The output section to the right of the model-fit statistics presents the results on *scale reliability* and dimensionality. Among the various conceptualisations of measurement reliability discussed in the literature (e.g., Bollen, 1989), *internal scale consistency* is the most widely used in practical research. One important reason for this is certainly that internal consistency may be easily

assessed without additional re-test or parallel measurements of the indicators. It should also be noted, however, that consistency measures avoid several conceptual drawbacks of possible alternatives (see Bollen, 1989: 209ff.). We report three alternative measures of internal scale consistency: Cronbach's Alpha is still the most widespread, although much criticised, consistency measure (ibid.: 217; Sijtsma, 2009; Revelle & Zinbarg, 2009; Trizano-Hermosilla & Alvarado, 2016). In a nutshell, it is widely recognised that alpha underestimates internal consistency if the indicators are ordinal or congeneric (i.e., not tau-equivalent) as is typical of most practical research situations. We nevertheless do report the classical version of alpha as it is part of most survey documentations and — if interpreted as a lower-bound estimate of internal scale consistency — may still be useful for comparative purposes.<sup>40</sup> In addition, we also report Ordinal Cronbach's Alpha, which is calculated the same way as classical alpha but from the matrix of polychoric instead of Pearson correlations (see Gadermann, Guhn & Zumbo, 2012: 5). This avoids downward bias owing to ordinal measurement. Finally, we also report McDonald's Omega, which is one of the most recommended measures of internal consistency. Omega is calculated on the basis of the factor loadings of the one-dimensional CFA model (according to formula 1 in Trizano-Hermosilla & Alvarado, 2016), which implies that it is adjusted for ordinal measurement. As omega is appropriate for congeneric indicators, it is probably the most adequate measure overall of internal scale consistency in our context (see also Yang & Green, 2015). Basically, values close to 1 indicate high internal consistency for all three measures. Looking at Figure 3, many researchers would probably interpret the identical ordinal alpha and omega values of .811 each as an indication of a 'good', consistent scale. It should be noted, however, that the widely used rules of thumb to determine whether internal scale consistency can be considered 'acceptable' or 'good' (usually values above .7 and .8, respectively) are not without problems. First, there exist various such rules of thumb with different critical thresholds. Second, and more importantly, such rules should not be applied blindly, as the acceptable level of internal consistency depends strongly on the type of analysis to be performed (Lance, Butts & Michels, 2006).41

A crucial assumption of the estimated CFA models is that the analysed item set captures only one latent construct. Therefore, we have also included a *test of the assumed one-dimensionality*. However, assessing dimensionality of Likert-type items is quite 'risky business', as van der Eijk

<sup>&</sup>lt;sup>40</sup> The Stata package "Alphawgt", which allows for weights, was used to calculate alpha (Jann, 2004).

There are some rather dubious rules of thumb that distinguish different levels of internal scale consistency (i.e., Cronbach's alpha). A popular variant is:  $\alpha$  < .5: unacceptable; .5  $\leq$   $\alpha$  < .6: poor; .6  $\leq$   $\alpha$  < .7: questionable; .7  $\leq$   $\alpha$  < .8: acceptable; .8  $\leq$   $\alpha$  < .9: good; .9  $\leq$   $\alpha$ : excellent

<sup>(</sup>cf. https://en.wikipedia.org/wiki/Internal\_consistency, accessed on June 23, 2020).

and Rose (2015) put it. We used explorative factor analysis of polychoric correlations followed by Horn's parallel analysis to assess the dimensionality of the item sets, which proves to be a comparatively well-performing method (ibid.; Garrido, Abad & Ponsoda, 2013).<sup>42</sup> Basically, we applied an eigenvalue criterion that was corrected for random factors to account for sampling variance to determine the number of factors to be retained. In Figure 3, this approach gives us no reason to believe that the achievement-pressure scale is not one-dimensional, as only the eigenvalue of the first factor exceeds the critical value of zero. If we leave aside the scales composed of several sub-dimensions (cf. Table 4), the eigenvalues of the second factor are mostly below or only very slightly above zero for most of the scales in this documentation.<sup>43</sup> This being the case, we have no clear indication that the one-dimensionality assumption is violated.

The section below the model-fit statistics in Figure 3 documents the *standardised factor loadings* for each item, including standard errors and the confidence intervals. The item names correspond to those in the scientific use file (without the prefix-marker for the survey wave). High standardised loadings above, say, .6 or .7 indicate that neither measurement errors nor strong unique factors contribute excessively to the variance of the observed indicators. Almost all loadings reported in the appended scales reach this level. Occasionally, however, items show noticeably weaker loadings below .5 or even below .4, which some researchers may consider problematic. Eventually, the definition of an acceptable factor loading remains arbitrary and depends on the type of analysis, the number of scale items affected and the quality as well as the overall internal consistency of the scale (ibid.). As in other respects, we prefer to leave it to the data users to judge a particular scale's qualities.

To the right of the loadings, a number of *item descriptives* are reported, including the mean, the standard deviation, the range of the rating scale applied for item evaluation (min., max.) and the number of students with valid item data (see section 3.1.1b).

At the bottom of the first page of our scale-specific results, we report the *parameters of the categorical GSEM model* (cf. section 3.1.2) where it is estimated. Note that for this model, there are two types of item-specific parameters, namely, factor coefficients ('coef') that measure the effect of the latent variable on the indicator rating, and the estimated cut points ('cutx') on the logit distribution that separate the rating scale category 1 from category 2, category 2 from category 3 and so on. Hence, the number of estimated 'cut' parameters equals the number of ordered rating categories minus one. Remember that the GSEM model is used to generate

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<sup>&</sup>lt;sup>42</sup> The parallel analysis relies on the user-written "paran" package (Dinno, 2009).

Exception: the two wealth scales.

student scores (see section 3.1) where students' item evaluations rely on short rating scales with four or less points (as documented by the item descriptives).

A second page of scale-specific results (see Figure 4 below) is dedicated to tests and indices that assess measurement invariance across survey languages and, where appropriate, across survey settings and modes. This is an important facet of measurement quality, as student scores obviously should be comparable – i.e., measure the same concepts on a possibly invariant scale – across all kinds of measurement conditions and subsamples of the underlying student population. We focus on some of the most crucial tests suggested in the literature on the multigroup analysis of measurement invariance (e.g., Vandenberg & Lance, 2000; Milfont & Fischer, 2015) to assess cross-language measurement equivalence. On top of the second results page, we first report a chi-square test of the equality of the item-covariance matrices across survey languages

Figure 4: Example of the reported scale-specific results (second results page)

Scale: Parental pressure to achieve	(continu	Jed)				В	aseline	surve	y sample
Tests and Indices of Factorial Invariance ad	cross								
Equality of the									
variance-covariance matrices across	Surve	y lang	uages	Surv	ey sett	ings	Sur	vey m	odes
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
	1717	28	.000	105	14	.000	26	14	.027
Tests of measurement invariance across	Surve	y lang	uages	Surv	ey sett	ings	Sur	vey m	odes
	chi2		p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
Metric invariance (equal factor loadings)	31	6	.000	33	3	.000	11	3	.013
Strong invariance (plus equal intercepts)	923	6	.000	11	3	.010	4	3	.317
Strict invariance (plus equal error variances)	73	6	.000	12	3	.008	3	3	.413
Configural factor similarity across	Survey languages		Surv	Survey settings		Survey modes			
Tucker's congruence coefficient		, ,	TCC		′	TCC		,	TCC
Ge	rman vs. F	rench	.999	classroom vs.		0	W	eb vs.	
Fi	rench vs. It	talian	.997	unpro	ctored	.998		PAP	.996
Ita	lian vs. Ge	rman	.993						
Factor score equivalence: group									
specific vs. invariant models for	Survey languages			Surv	ey sett	ings	Sur	vey m	odes
Coefficient of determination			CD	CD		CD			CD
	Ge	rman	1.000	clas	sroom	1.000		web	1.000
	F	rench	1.000	unpro	tored	.999		PAP	.990
	I	talian	.980						
Factor score descriptives									
Std.									
Variable name Mean dev. Min. Max.	Obs.								
press_fs	15535								
Share of cases with imputed missing values:	0.6%								
(Equivalence of scores from robust MLMV: CD =									
(Equivalence of Scores from Two-Step-Approach	: CD = .984	1)							

(German, French, Italian; cf. Table 2) and, when a scale relies on the TREE2 baseline survey (including the AES extension survey), across survey settings (classroom vs. unproctored) and survey modes (web survey vs. paper-and-pencil questionnaire (PAP); cf. Table 1).<sup>44</sup> If the hypothesis of equal covariance matrices is not rejected, this would be a strong indication of measurement invariance, making any further tests obsolete (ibid.).

The chi-square tests assembled in the section below refer to the one-dimensional CFA model from section 3.1.1, which was re-estimated separately for each survey language and, where appropriate, for each survey setting and survey mode. Hence, the tests assume that a common latent dimension exists, and its invariance is investigated by means of multi-group analysis. The three tests are designed to distinguish different levels of measurement equivalence, as discussed in the literature (ibid.). The first test is for metric measurement invariance, that is, for equal factor loadings. A non-significant test indicates that there is no evidence against the postulated invariance of the factor loadings across the different survey conditions. The second test takes the model with invariant loadings as its baseline and tests it against an alternative model with invariant loadings and intercepts, which implies strong measurement invariance. Third and lastly, the latter model is tested against an alternative positing strict measurement invariance, which furthermore requires invariant error variances ( $\varepsilon_i$  in Figure 2). Given the nested structure of the compared models, strong invariance would require that the first two tests be not significant and strict invariance that all three tests be not significant. Although this is a rather standard approach to assess measurement equivalence, the reservations against chi-square-based fit statistics discussed above also extend to chi-square-based multi-group comparisons: Even if the cross-language variations in the model parameters are negligible, these tests will nearly always be significant given the mostly huge samples analysed here. That is to say, a level of measurement equivalence that would be adequate for nearly all practical research purposes would still not be enough to pass these tests. Against this background, it is rather surprising that, with regard to the 'Parental pressure to achieve' scale (see Figure 4), strong or even strict measurement invariance is not rejected (p < .01) with respect to survey modes (where, however, the test samples are smaller than for survey languages or settings; cf. Table 1).

Below the section with the chi-square-based invariance tests, we report two additional measures of factor equivalence, which will perhaps do better in meeting the practical needs of many data analysts. The first one, *Tucker's congruence coefficient (TCC)*, is a measure of *configural factor invariance* (calculated according to formula 1 in Lorenzo-Seva & ten Berge, 2006). Basically, it

Technically, this was achieved by specifying a multi-group model without a latent dimension and then testing a completely unconstrained model against a constrained one with equal variances and inter-item covariances.

is a pattern-similarity measure that approaches 1 when the loading patterns observed in two groups or conditions are identical. We report the coefficient separately for each pair of survey languages as well as for the pairs of survey settings and survey modes, where appropriate. According to Lorenzo-Seva and ten Bergen (ibid.: 61), two factors may be considered as approximately equal for practical purposes if TCC is .95 or higher. If we look at the scales documented in the appendix, this criterion is met for all pairwise comparisons across survey languages, survey settings and survey modes.

In addition, we also assess the degree of *micro-level factor equivalence at the level of student scores*. For this, we compare the student scores taken from an unconstrained model fitted separately for each language, setting or mode, respectively, with the student scores taken from a model for the entire sample on the assumption of strong measurement invariance (i.e., equal loadings and intercepts). If the differences between the former and latter are negligible across the analysed survey conditions, this is a strong indication that - from a practical point of view - the measurement can be regarded as sufficiently invariant. As a measure of micro-level agreement, we report – separately for each of the subsamples delineated by survey language, survey setting, and survey mode – the coefficient of determination (CD), which is calculated by regressing the student scores from the strong-invariance model on those from the unconstrained conditionspecific models. Where the CD indicates that both scores share, say, 98% of their variance (i.e.,  $CD \ge .98$ ), deviations from the postulated strong invariance model may be regarded as negligible. All scales in the appendix satisfy this criterion with respect to mode and setting effects. With regard to survey languages, there are some differences in a limited number of cases, which mostly concern the Italian language. It should be noted, however, that a perfect agreement cannot always be expected even if the 'true' measurement model was absolutely invariant as the estimated student scores also include some random error. This is particularly true for the scores gained through the separate analysis of small subsamples, as is the case for the Italian questionnaire (n = 379 - 755, cf. Table 2) and the paper-and-pencil mode (n = 635; cf. Table 1) of the extension survey (cf. Figure 1). Notably for these subsamples, the sampling errors in the factor loadings and hence also in the student scores are likely to be more substantial.<sup>45</sup> With this in mind, one could also accept a coefficient of determination of, say, .95 as an indication of a still fair level of measurement equivalence. Also with regard to language-specific invariance, almost

In combination with skewed item distributions, this is probably also the reason why a few of the models underlying the invariance tests did not converge so that the subsamples for the French and the Italian languages had to be collapsed for this purpose. We added an explanatory note at the end of the measurement-equivalence section in the appendix, which is shaded in grey in these cases (e.g., the 'School reluctance' scale).

all scales in the appendix satisfy this criterion.<sup>46</sup> In the case of the achievement-pressure scale in Figure 4, however, our results are unambiguous and do suggest a high degree of measurement equivalence across survey languages, settings and modes.

In the section following the measurement invariance tests and indices, we report the short variable names (*press\_fs* in Figure 3) of the student score variables in the scientific use file (from either ML-SEM or GSEM, depending on the length of the rating scales; see section 3.2.1).<sup>47</sup> The respective descriptive statistics refer to the sample base used for the calculation of the student scores (including cases not published in the scientific use files of the data release; cf. section 1).<sup>48</sup>

Either one or two measures of factor-score equivalence across different estimation methods are reported at the bottom of the second results page (see section 3.2.1), depending on the length of the rating scales applied for item evaluation. With regard to the achievement-pressure scale in Figure 4, they confirm a high degree of equivalence between the student scores from all three estimation procedures.

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Exceptions to the rule: the Italian versions of 'vawe', 'ictintr', 'cogselfb' and 'cultposs'. In the case of 'cultposs', this applies to the French version as well.

The full variable names include an additional prefix to distinguish TREE2 survey waves (e.g., "t2" for the second followup survey).

Relevant sample sizes are reported under "Factor score descriptives: Obs.". We also report the share of cases with imputed item values.

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### Extensions and minor corrections since version 2021\*

- The abstract has been revised and extended
- The shortened scales for 'Global self-esteem' (sel\_m\_fs) and for 'Negative global self-esteem' (seld\_m\_fs) used in later panel waves have been added to the scale appendix below and to the baseline survey data file (TREE2\_Data\_Wave\_0\_v2) in the data release (TREE, 2023).
- The reported Tucker coefficients for invariance across survey modes and survey settings include minor corrections.
- The student score variables *cogself1\_fs*, *cogself2\_fs*, *extregm\_fs* have been renamed according to TREE naming conventions (to *cogselfa\_fs*, *cogselfb\_fs*, and *extreg\_fs*).
- Some inconsistently used labels for scales and survey topics have been harmonised across the TREE2 data release (TREE, 2023).

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Sacchi, Stefan, Krebs-Oesch, Dominique (2021). Scaling methodology and scale reporting in the TREE2 panel survey. Documentation of scales implemented in the baseline survey (2016). University of Bern: TREE. http://dx.doi.org/10.48350/152055.

# **SCALE APPENDIX**

# Scales administered in the baseline survey

(<u>Scale names</u> linked with first page of scale-specific reporting)

cale-specific reportin	3/		
Variable Name	AES Module	Source	Page
[ closep_comp ]	Background	TREE1 - based on Szydlik, 2008	41
[ press_fs ]	Background	Böhm-Kasper et al., 2000	42
[ expectp_fs ]	Math	Hascher et al., 2019	44
[ expectm_fs ]	Math	Hascher et al., 2019	46
[ expectf_fs ]	Math	Hascher et al., 2019	48
[ socnormsm_fs ]	Math	PISA 2012	50
[ socnormsf_fs ]	Math	PISA 2012	52
[ famedsup_fs ]	Background	PISA 2000	54
[ soccom_fs ]	Background	PISA 2000	56
[ soccom_m_fs ]	Background	PISA 2000 (adapted)	58
[ closupp_fs ]	Background	TREE2 (BHPS, ISSP 2003)	60
[ joyreadp_comp ]	Background	TREE2	62
[ cultcom_fs ]	Background	PISA 2000	64
[ cultcom_m_fs ]	Background	PISA 2000 (adapted)	66
[ cultposs_fs ]	Background	PISA 2000	68
[ inccap_fs ]	Background	TREE2	70
[ manners_fs ]	Background	TREE2	72
[ verbskill_fs ]	Background	TREE2	74
[ cult_fs ]	Background	PISA 2000 (adapted)	76
[ cultlow_fs ]	Background	TREE2	78
[ culthigh_fs ]	Background	PISA 2000	80
	[ closep_comp] [ press_fs] [ expectp_fs] [ expectf_fs] [ expectf_fs] [ socnormsm_fs] [ socnormsf_fs] [ famedsup_fs] [ soccom_fs] [ soccom_m_fs] [ coltcom_m_fs] [ cultcom_m_fs] [ cultcom_m_fs] [ cultcom_m_fs] [ cultcom_mfs] [ cultcom_fs]	[ closep_comp] Background [ press_fs] Background [ expectp_fs] Math [ expectf_fs] Math [ socnormsm_fs] Math [ socnormsf_fs] Math [ famedsup_fs] Background [ soccom_fs] Background [ soccom_m_fs] Background [ coltcom_m_fs] Background [ cultcom_fs] Background [ cultcom_mfs] Background [ cultcom_mfs] Background [ cultcom_fs] Background [ manners_fs] Background [ verbskill_fs] Background [ cult_fs] Background [ cultlow_fs] Background	Variable Name       AES Module       Source         [ closep_comp]       Background       TREE1 - based on Szydlik, 2008         [ press_fs]       Background       Böhm-Kasper et al., 2000         [ expectp_fs]       Math       Hascher et al., 2019         [ expectf_fs]       Math       Hascher et al., 2019         [ expectf_fs]       Math       Hascher et al., 2019         [ socnormsm_fs]       Math       PISA 2012         [ socnormsf_fs]       Background       PISA 2012         [ famedsup_fs]       Background       PISA 2000         [ soccom_fs]       Background       PISA 2000 (adapted)         [ closupp_fs]       Background       TREE2         [ cultcom_fs]       Background       PISA 2000 (adapted)         [ cultcom_m_fs]       Background       PISA 2000 (adapted)         [ cultposs_fs]       Background       TREE2         [ manners_fs]       Background       TREE2         [ werbskill_fs]       Background       TREE2         [ cult_fs]       Background       PISA 2000 (adapted)         [ cultlow_fs]       Background       TREE2

Survey topics (continued)			Baseline survey	(2016)
Scale (or composite)	Variable Name	AES Module	Source	Page
5) Economic capital (family of origin)				
Household possessions: Family wealth (PISA2000)	[ wealth_fs ]	Background	PISA 2000	82
Household possessions: Family wealth (adapted TREE2)	[ wealth_m_fs ]	Background	PISA 2000 (adapted)	84
Family affluence scale (FASIII)	[ fasiii_comp ]	Background	Hobza et al., 2017	86
6) Satisfaction				
<u>Capabilities</u>	[ cap_fs ]	Background	Sen, 1985; Anand & van Hees, 2006	88
7) School-related well-being				
Positive attitude towards school	[ posatt_fs ]	General	Hascher, 2004	90
<u>Enjoyment in school</u>	[ enjoyschool_fs ]	General	Hascher, 2004	92
Physical complaints in school	[ physpain_fs ]	General	Hascher, 2004	94
Worries about school	[ trouschool_fs ]	General	Hascher, 2004	96
Social problems in school	[ socprob_fs ]	General	Hascher, 2004	98
School reluctance	[ schoolav_fs ]	General	Hagenauer & Hascher, 2012 (modified)	100
8) Motivational concepts				
Intrinsic achievement motivation	[ achmoti_fs ]	General	IGLU 2001	102
Extrinsic achievement motivation	[ achmote_fs ]	General	IGLU 2001	104
Instrumental learning motivation (PISA2000)	[ insmot_fs ]	General	PISA 2000	106
Interest in reading	[ intrea_fs ]	General	PISA 2000	108
ICT interest	[ ictintr_fs ]	Math	ICILS 2013	110
<u>Dispositional interest</u>	[ intsubj_fs ]	Math	COACTIV 2008	112
Identified motivation (mathematics)	[ instrumot_fs ]	Math	PISA 2012	114
External motivation regulation	[ extreg_fs ]	Math	Ryan & Conell, 1989	116
Classroom participation	[ engage_fs ]	Math	Eder, 1995, 2007	118
Performance-approach goals (SELLMO)	[ approxgoals_fs ]	Math	SELLMO 2012	120
Learning goal orientation (SELLMO)	[ learntarget_fs ]	Math	SELLMO 2012	122
Work avoidance (SELLMO)	[ avoidwork_fs ]	Math	SELLMO 2012	124
Avoidance performance goals (SELLMO)	[ avoidblame_fs ]	Math	SELLMO 2012	126

Baseline survey (2016)

Scale (or composite)	Variable Name	AES Module	Source	Page
9) Self-perception				
Global self-esteem	[ sel_fs ]	Background	Rosenberg, 1979 (translated)	128
Global self-esteem (shortened)	[ sel_m_fs ]	Background	Rosenberg, 1979 (translated)	130
Positive global self-esteem	[ sele_fs ]	Background	Rosenberg, 1979 (translated)	132
Negative global self-esteem	[ seld_fs ]	Background	Rosenberg, 1979 (translated)	134
Negative global self-esteem (shortened)	[ seld_m_fs ]	Background	Rosenberg, 1979 (translated)	136
General perceived self-efficacy scale (GSES)	[ seef_fs ]	Background	GSES (adapted TREE1)	138
Academic self-efficacy	[ acaself_fs ]	General	Hascher, 2004	140
Academic self-concept (PISA2000)	[ scacad_fs ]	General	PISA 2000	142
Verbal self-concept (PISA2000)	[ scverb_fs ]	General	PISA 2000	144
Maths self-concept [PISA 2000]	[ matcon_fs ]	General	PISA 2000	146
ICT self-concept	[ictabil_fs]	Math	ICILS 2013	148
Specific self-efficacy: numeracy	[ selfeffa_fs ]	General [Math]	PISA 2012; Girnat, 2018	150
Specific self-efficacy: algebra	[ selfeffb_fs ]	General [Math]	PISA 2012; Girnat, 2018	152
Specific self-efficacy: geometry	[ selfeffc_fs ]	General [Math]	Girnat, 2018	154
Specific self-efficacy: probability	[ selfeffd_fs ]	General [Math]	Girnat, 2018	156
10) Emotions related to maths classes				
Mathematics anxiety	[ anxmath_fs ]	Math	PISA 2012	158
<u>Mathematics boredom</u>	[ boredom_fs ]	Math	AEQ-M (short-version)	160
Mathematics anger	[ anger_fs ]	Math	AEQ-M (short-version)	162
Mathematics enjoyment	[ enjoymath_fs ]	Math	AEQ-M (short-version)	164
11) Volitional strategies				
<u>Perseverance</u>	[ persev_fs ]	General	PISA 2012	166
Effort: learning (PISA2000)	[ effper_comp ]	Background	PISA2000	168

, .				, , ,
Scale (or composite)	Variable Name	AES Module	Source	Page
12) Personality characteristics				
Big Five: extraversion	[ big5_e_comp ]	Background	Rammstedt et al., 2014	169
Big Five: agreeableness	[ big5_a_comp ]	Background	Rammstedt et al., 2014	169
Big Five: conscientiousness	[ big5_c_comp ]	Background	Rammstedt et al., 2014	169
Big Five: neuroticism	[ big5_n_comp ]	Background	Rammstedt et al., 2014	169
Big Five: openness	[ big5_o_comp ]	Background	Rammstedt et al., 2014	169
Internal locus of control	[ loci_comp ]	Background	GESIS (short-version)	170
External locus of control	[ loce_comp ]	Background	GESIS (short-version)	170
13) Values & attitudes				
Work-related extrinsic values	[ vawe_fs ]	Background	TREE1 - based on Watermann, 2000	172
Work-related intrinsic values	[ vawi_fs ]	Background	TREE1 - based on Watermann, 2000	174
<u>Family values</u>	[ vafa_comp ]	Background	TREE1	176
Positive attitude towards life	[ posl_fs ]	AES Extension Survey	TREE1; Grob et al., 1991	178
14) Attitudes related to mathematics classes				
Reality-based learning	[ realref_fs ]	Math	Girnat, 2015, 2017	180
Discovery / exploratory learning	[ disclearn_fs ]	Math	Girnat, 2015, 2018	182
Social learning	[ soccomlearn_fs ]	Math	Girnat, 2015, 2019	184
Social learning: social arrangement	[ soclearn_fs ]	Math	Girnat, 2015, 2020	186
Social learning: communication	[ comlearn_fs ]	Math	Girnat, 2015, 2021	188
Instructivist learning	[ instreplearn_fs ]	Math	Girnat, 2015, 2022	190
Instructivist learning: teachers instructions	[ instrlearn_fs ]	Math	Girnat, 2015, 2023	192
Instructivist learning: repetitive practice	[ replearn_fs ]	Math	Girnat, 2015, 2024	194
System aspect	[ sysformasp_fs ]	Math	Girnat, 2015, 2025	196
System aspect: logical thinking	[ systasp_fs ]	Math	Girnat, 2015, 2026	198
System aspect: formalism	[ formasp_fs ]	Math	Girnat, 2015, 2027	200
Scheme aspect	[ schemasp_fs ]	Math	Girnat, 2015, 2028	202
Application aspect	[ applyasp_fs ]	Math	Girnat, 2015, 2029	204

### **Survey topics**

Scale (or composite)	Variable Name	AES Module	Source	Page				
15) Characteristics of maths lessons (end of I	ower secondary educat	ion)						
Teacher: cognitive activation	[ cogself_fs ]	Math	COACTIV 2008	206				
Teacher cognitive activation: finding solutions & arguing	[ cogselfa_fs ]	Math	COACTIV 2008	208				
Teacher: cognitive activation: strategies & learning from mistakes	[ cogselfb_fs ]	Math	COACTIV 2008	210				
Teacher: classroom management	[ classman_fs ]	Math	COACTIV 2008	212				
Teacher: individual learning support	[ indsup_fs ]	Math	COACTIV 2008	214				
Teacher: instruction quality	[ instqual_fs ]	Math	PISA 2006	216				
<u>Situational interest</u>	[ intsit_fs ]	Math	COACTIV 2008	218				
Perceived autonomy support	[ persuppauto_fs ]	Math	Seidel, Prenzel & Kobarg, 2005	220				
Perceived competence support	[ persuppcomp_fs ]	Math	Seidel, Prenzel & Kobarg, 2005	222				
Perceived social relatedness	[ persocincl_fs ]	Math	Seidel, Prenzel & Kobarg, 2005	224				
<u>Classmates' appreciation of mathematics</u>	[ apprmath_fs ]	Math	PISA 2012	226				
15) Absenteeism/intention to change education								
Absenteeism / truancy	[ truancy_fs ]	General	PISA 2000, PISA 2012	228				
<u>List of Sources</u>				230				

Composite descriptives	Variable name	Mean	Std. dev.	Min.	Max.	Obs.		
Share of cases with impute	closep_comp	4.2 3.5%	0.8	1	5	15664		
								35
Item descriptives			Std.			Valid		

Item descriptives			Std.			Valid
	Indicators	Mean	dev.	Min.	Max.	obs.
	closef	4.1	1.1	1	5	15223
	closem	4.4	0.9	1	5	15558

# Scale: Parental pressure to achieve

Baseline survey sample

#### **Model and Fit Statistics**

# Reliability and Dimensionality

1)	Likelihood-ratio tests	chi2	df	p > chi2	Ordinal Cronbach's Alpha	.811
	Model vs. saturated	462	2	.000	(Cronbach's alpha = .751)	
	Baseline vs. saturated	20063	6	.000	McDonald's Omega	.811
2)	Root mean squared error (	RMSEA)		.122	Test of (one-)dimensionality	(parallel analysis)
	90% Confidence interval:	lower bound		.113	Criterion: Retain factors with a	ndj. eigenvalue > o
	90% Confidence interval:	upper bound		.131	Adjusted eiger	nvalue
	Probability RMSEA <= 0.05			.000	factor 1 1.95	
					factor 204	
3)	Akaike's Information Crite	rion (AIC)		142462	factor 309	
	Bayesian Information Crite	erion (BIC)		142554	factor 418	
4)	Baseline comparison					
	Comparative Fit Index (CFI)			.977		
	Tucker-Lewis Index (TLI)			.931		
5)	Size of residuals					
	Stand. root mean squared re	esidual (SRMR	2)	.026		
	Coefficient of determination	n (CD)		.816		

### Standardized factor loadings

### Item descriptives

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
press1	0.69	0.01	0.68	0.70	press1	2.2	1.0	1	4	15488
press2	0.69	0.01	0.68	0.71	press2	3.0	0.9	1	4	15491
press3	0.78	0.00	0.77	0.79	press3	3.0	0.8	1	4	15488
press4	0.71	0.01	0.70	0.72	press4	2.8	0.9	1	4	15490

## Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut1	Cut <sub>2</sub>	Cut <sub>3</sub>
press1	1.66	-1.38	0.68	2.99
press2	1.79	-3.56	-1.79	0.80
press3	2.35	-5.01	-2.26	1.38
nress.	184	-2 48	-1 22	1 52

Equal	ity	of t	he
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variance-covariance matrices across	Surve	Survey languages		Survey settings			Survey modes			
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2	
	1717	28	.000	105	14	.000	26	14	.027	
Tests of measurement invariance across	. Surve	y lan	guages	Surv	ey set	tings	Sur	vey m	odes	
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2	
Metric invariance (equal factor loadings)	31	6	.000	33	3	.000	11	3	.013	
Strong invariance (plus equal intercepts)	923	6	.000	11	3	.010	4	3	.317	
Strict invariance (plus equal error variances)	73	6	.000	12	3	.008	3	3	.413	
Configural factor similarity across	Survey languages		Surv	Survey settings			Survey modes			
Tucker's congruence coefficient			TCC			TCC			TCC	
	German vs. Fr	rench	.999	classro	om vs.	.998	W	eb vs.	.996	
	French vs. It	alian	.997	unprod	ctored	.990		PAP	.990	
	Italian vs. Ge	rman	.993							
Factor score equivalence: group										
specific vs. invariant models for	Surve	y lan	guages	Surv	ey set	tings	Sur	vey m	odes	
Coefficient of determination			CD			CD			CD	
	Ger	man	1.000	class	sroom	1.000		web	1.000	
	Fr	rench	1.000	unprod	ctored	.999		PAP	.990	
	It	talian	.980							

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. press\_fs 0.0 0.9 -2.4 1.7 15535 Share of cases with imputed missing values: 0.6% (Equivalence of scores from robust MLMV: CD = .997) (Equivalence of Scores from Two-Step-Approach: CD = .984)

# Scale: Parents' achievement expectations

Maths sample-split

#### **Model and Fit Statistics**

#### Reliability and Dimensionality

					•	•		
1)	Likelihood-ratio tests	chi2	df	p > chi2	Ordinal Cronbach's Alpha	.837		
	Model vs. saturated	8040	2	.000	(Cronbach's alpha = .774)			
	Baseline vs. saturated	24621	6	.000	McDonald's Omega	.834		
2)	Root mean squared error (RMSEA)			.606	Test of (one-)dimensionali	ty (parallel analysis)		
	90% Confidence interval:	lower bound		-595	Criterion: retain factors with	n adj. eigenvalue > o		
	90% Confidence interval:	upper bound		.617	Adjusted eigenvalue			
	Probability RMSEA <= 0.05			.000	Factor 1 2.	35		
					Factor 2 .	43		
3)	Akaike's Information Criter	ion (AIC)		77644	Factor 3	.11		
	Bayesian Information Crite	rion (BIC)		77731	Factor 4	19		
4)	Baseline comparison							
	Comparative Fit Index (CFI)			.673				
	Tucker–Lewis Index (TLI)			.020				
5)	Size of residuals							
٥,	Stand. root mean squared re	sidual (SRMI	R)	.108				
	Coefficient of determination		•	.854				
		•		٥,				

#### Standardized factor loadings

#### Item descriptives

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
expectf2	0.70	.007	0.69	0.72	expectf2	3.4	0.7	1	4	10568
expectf3	0.85	.005	0.84	0.86	expectf3	3.3	0.7	1	4	10566
expectm2	0.63	.009	0.62	0.65	expectm2	3.4	0.7	1	4	10862
expectm3	0.79	.005	0.78	0.80	expectm3	3.4	0.7	1	4	10864

#### Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut1	Cut <sub>2</sub>	Cut <sub>3</sub>
expectf2	2.12	-5.87	-4.04	-0.32
expectf3	2.31	-5.88	-3.69	0.30
expectm2	1.75	-5.42	-3.28	0.14
expectma	2.11	-6.40	-4.13	-0.12

#### Scale: Parents' achievement expectations (continued)

Maths sample-split

#### Tests and Indices of Factorial Invariance across Survey Languages

chi2	df	p > chi2
297	28	.000
chi2	df	p > chi2
15	6	.017
126	6	.000
12	6	.072
	297 chi2 15 126	297 28  chi2 df 15 6 126 6

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	1.000
French vs. Italian language version	.996
Italian vs. German language version	.995

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	.999
Language: Italian	.964

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. expectp\_fs 0.0 0.9 -3.1 1.1 10952 Share of cases with imputed missing values: 4.3% (Equivalence of scores from robust MLMV: CD = .991) (Equivalence of scores from two-step approach: CD = .941)

# Scale: Mother's achievement expectations

Maths sample-split

#### **Model and Fit Statistics**

Reliability and Dimensionality	and Dimensionality
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1)	<b>Likelihood-ratio tests</b> Model vs. saturated Baseline vs. saturated	chi2 0 4828	df o 3	p > chi2 .000	Ordinal Cronbach' (Cronbach's alpha : McDonald's Omeg	= .552)	.642 .663
2)	90% Confidence interval:	RMSEA) lower bound upper bound		.000 .000	-	ctors with adj. eig justed eigenvalue	envalue > o
	Probability RMSEA <= 0.05			1.000	Factor 1 Factor 2	1.01 07	
3)	Akaike's Information Criter Bayesian Information Crite	• •		67851 67917	Factor 3	22	
4)	Baseline comparison Comparative Fit Index (CFI) Tucker–Lewis Index (TLI)			1.000 1.000			
5)	Size of residuals Stand. root mean squared re Coefficient of determination		)	.000 .729			

#### Standardized factor loadings

Item descriptives	Item	descri	otives
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							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf.	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
expectm1	0.42	.010	0.40	0.44	expectm1	2.8	0.8	1	4	10859
expectm2	0.80	.013	0.77	0.82	expectm2	3.4	0.7	1	4	10862
expectma	0.65	.011	0.63	0.67	expectma	3.4	0.7	1	4	10864

#### Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut <sub>1</sub>	Cut <sub>2</sub>	Cut <sub>3</sub>
expectm1	0.83	-2.97	-0.79	1.48
expectm2	2.27	-6.07	-3.61	0.24
expectm3	1.68	-5.59	-3.50	-0.04

#### Scale: Mother's achievement expectations (continued)

Maths sample-split

#### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2 536	df 18	p > chi2 .000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	112	4	.000
Strong invariance (plus equal intercepts)	126	4	.000
Strict invariance (plus equal error variances)	66	4	.000

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.965
French vs. Italian language version	.982
Italian vs. German language version	.979

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	.964
Language: French	.961
Language: Italian	.970

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. expectm\_fs 0.0 0.8 -2.8 1.2 10864
Share of cases with imputed missing values: 0.1% (Equivalence of scores from robust MLMV: CD = .987) (Equivalence of scores from two-step approach: CD = .957)

# Scale: Father's achievement expectations

Maths sample-split

#### **Model and Fit Statistics**

#### **Reliability and Dimensionality**

1)	Likelihood-ratio tests	chi2	df	p > chi2	Ordinal Cronbach	n's Alpha	.738
	Model vs. saturated	0	0	•	(Cronbach's alpha	= .653)	-
	Baseline vs. saturated	75 <del>1</del> 7	3	.000	McDonald's Ome		.749
2)	Root mean squared error (I	RMSEA)		.000	Test of (one-)dim	ensionality (parallel	analysis)
	90% Confidence interval:	lower bound		.000	Criterion: retain fa	actors with adj. eigenv	alue > o
	90% Confidence interval:	upper bound		.000	Ac	djusted eigenvalue	
	Probability RMSEA <= 0.05			1.000	Factor 1	1.31	
					Factor 2	09	
3)	Akaike's Information Criter	ion (AIC)		65854	Factor 3	19	
	Bayesian Information Crite	rion (BIC)		65920			
4)	Baseline comparison						
	Comparative Fit Index (CFI)			1.000			
	Tucker–Lewis Index (TLI)			1.000			
5)	Size of residuals						
	Stand. root mean squared re	esidual (SRMF	?)	.000			
	Coefficient of determination	n (CD)		.791			

#### Standardized factor loadings

### Item descriptives

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
expectf1	0.55	.008	0.53	0.56	expectf1	2.9	0.9	1	4	10565
expectf2	0.83	.008	0.82	0.85	expectf2	3.4	0.7	1	4	10568
expectf3	0.72	.008	0.70	0.74	expectf3	3.3	0.7	1	4	10566

#### Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut1	Cut <sub>2</sub>	Cut <sub>3</sub>
expectf1	1.17	-3.07	-1.05	1.32
expectf2	3.04	-7.28	-4.84	-0.32
expectf3	1.92	-5.13	-3.06	0.33

#### Scale: Father's achievement expectations (continued)

Maths sample-split

#### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2
	429	18	.000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	100	4	.000
Strong invariance (plus equal intercepts)	57	4	.000
Strict invariance (plus equal error variances)	84	4	.000

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.986
French vs. Italian language version	.997
Italian vs. German language version	.990

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	.997
Language: French	.998
Language: Italian	.982

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. expectf\_fs 0.0 0.8 -2.7 1.2 10569
Share of cases with imputed missing values: 0.1% (Equivalence of scores from robust MLMV: CD = .988) (Equivalence of scores from two-step approach: CD = .957)

#### Scale: Mother's social norms about mathematics

Maths sample-split

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	0	0	
	Baseline vs. saturated	12780	3	.000

2) Root mean squared error (RMSEA) .000
90% Confidence interval: lower bound .000
90% Confidence interval: upper bound .000
Probability RMSEA <= 0.05 1.000

3) Akaike's Information Criterion (AIC) 66659Bayesian Information Criterion (BIC) 66724

4) Baseline comparison

Comparative Fit Index (CFI) 1.000
Tucker–Lewis Index (TLI) 1.000

5) Size of residuals

Stand. root mean squared residual (SRMR) .000
Coefficient of determination (CD) .881

#### **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.789
(Cronbach's alpha = .715)	
McDonald's Omega	.812

Test of (one-)dimensionality (parallel analysis)

Criterion: retain factors with adj. eigenvalue > o Adjusted eigenvalue

Factor 1 1.66
Factor 2 -.05
Factor 3 -.15

#### Standardized factor loadings

Indicators	Coef.	(SE)	[95% Conf	. interval]
socnormsm1	0.87	.006	o.86	0.88
socnormsm2	0.89	.006	0.88	0.91
socnormsm3	0.50	.008	0.49	0.52

Item descriptives

		Std.			Valid
Indicators	Mean	dev.	Min.	Max.	Obs.
socnormsm1	3.2	0.7	1	4	10833
socnormsm2	3.1	0.8	1	4	10834
socnormsm3	2.4	0.9	1	4	10795

Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut1	Cut <sub>2</sub>	Cut <sub>3</sub>
socnormsm1	3.95	-8.08	-4.66	1.62
socnormsm2	3.36	-5.95	-2.64	1.65
socnormsm3	0.99	-1.65	0.37	2.19

### Scale: Mother's social norms about mathematics (continued)

Maths sample-split

#### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2 195	df 18	p > chi2 .000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	11	4	.030
Strong invariance (plus equal intercepts)	44	4	.000
Strict invariance (plus equal error variances)	80	4	.000

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.999
French vs. Italian language version	.998
Italian vs. German language version	1.000

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	.999
Language: French	.990
Language: Italian	.999

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. socnormsm\_fs 0.1 0.9 -2.3 1.4 10847 Share of cases with imputed missing values: 0.6% (Equivalence of scores from robust MLMV: CD = .996) (Equivalence of scores from two-step approach: CD = .971)

#### Scale: Father's social norms about mathematics

Maths sample-split

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	0	0	
	Baseline vs. saturated	15486	3	.000
2)	Root mean squared error	(RMSEA)		.000
	90% Confidence interval:	lower bound		.000
	90% Confidence interval:	upper bound		.000
	Probability RMSEA <= 0.05			1.000
3)	Akaike's Information Crite	erion (AIC)		60431
	<b>Bayesian Information Crit</b>	erion (BIC)		60496
4)	Baseline comparison			
	Comparative Fit Index (CFI)	)		1.000
	Tucker–Lewis Index (TLI)			1.000
5)	Size of residuals			
	Stand, root mean squared in	residual (SRMI	۲)	.000

#### **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.837
(Cronbach's alpha = .771)	
McDonald's Omega	.851

#### Test of (one-)dimensionality (parallel analysis)

Criterion: retain factors with adj. eigenvalue > o
Adjusted eigenvalue

	.)	
Factor 1	1.85	
Factor 2	04	
Factor 3	14	

Stand. root mean squared residual (SRMR)	.000
Coefficient of determination (CD)	.922

#### Standardized factor loadings

#### Indicators Coef. (SE) [95% Conf. interval] socnormsf1 0.95 .004 0.96 0.94 0.86 socnormsf<sub>2</sub> 0.85 0.84 .005 socnormsf3 0.60 0.62 .007 0.59

#### Item descriptives

		Std.			Valid
Indicators	Mean	dev.	Min.	Max.	Obs.
socnormsf1	3.3	0.7	1	4	10576
socnormsf2	3.2	0.8	1	4	10572
socnormsf3	3.1	0.9	1	4	10567

#### Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut <sub>1</sub>	Cut <sub>2</sub>	Cut <sub>3</sub>
socnormsf1	4.84	-9.33	-5.83	1.21
socnormsf2	3.14	-5.97	-3.09	1.20
socnormsf3	1.25	-2.99	-1.28	0.85

### Scale: Father's social norms about mathematics (continued)

Maths sample-split

#### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2 198	df 18	p > chi2 .000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	15	4	.005
Strong invariance (plus equal intercepts)	85	4	.000
Strict invariance (plus equal error variances)	72	4	.000

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.999
French vs. Italian language version	.999
Italian vs. German language version	.996

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	.999
Language: French	.996
Language: Italian	.956

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. socnormsf\_fs 0.1 0.9 -2.4 1.2 10587 Share of cases with imputed missing values: 0.4% (Equivalence of scores from robust MLMV: CD = .992) (Equivalence of scores from two-step approach: CD = .960)

#### Scale: Family educational support (PISA2000)

Baseline survey sample

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	0	0	
	Baseline vs. saturated	16654	3	.000
2)	Root mean squared error (	(RMSEA)		.000
	90% Confidence interval:	lower bound		.000
	90% Confidence interval:	upper bound		.000
	Probability RMSEA <= 0.05			1.000
3)	Akaike's Information Crite	rion (AIC)		147278
	Bayesian Information Crit	erion (BIC)		147347
4)	Baseline comparison			
	Comparative Fit Index (CFI)			1.000
	Tucker-Lewis Index (TLI)			1.000

#### **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.785
(Cronbach's alpha = .746)	
McDonald's Omega	.803

#### Test of (one-)dimensionality (parallel analysis)

Criterion: Retain factors with adj. eigenvalue > o Adjusted eigenvalue

	,		
factor 1		1.60	
factor 2		07	
factor 3		16	

#### 5) Size of residuals

famedsup2

famedsup3

Stand. root mean squared residual (SRMR) .000 Coefficient of determination (CD) .861

#### Standardized factor loadings

#### Indicators \* Coef. (SE) [95% Conf. interval] famedsup1 0.88 0.87 0.89 0.01 0.86

0.01

0.01

0.84

0.51

0.54

0.85

0.53

#### Item descriptives

		Std.			Valid
Indicators *	Mean	dev.	Min.	Max.	Obs.
famedsup1	2.8	1.4	1	5	15462
famedsup2	2.6	1.4	1	5	15131
famedsup3	2.3	1.4	1	5	13709

<sup>\*</sup> Note: Replication of 'Famedsup'-Scale from TREE1 / PISA2000

Equal	lity	of	the
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variance-covariance matrices across	ce matrices across Survey languages		Survey settings			Survey modes			
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
	365	18	.000	101	9	.000	34	9	.000
Tests of measurement invariance across	. Survey	lang	guages	Surv	ey set	tings	Sur	vey m	odes
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
Metric invariance (equal factor loadings)	20	4	.001	9	2	.013	11	2	.005
Strong invariance (plus equal intercepts)	300	4	.000	32	2	.000	11	2	.003
Strict invariance (plus equal error variances)	12	4	.015	18	2	.000	2	2	.324
Configural factor similarity across	Survey	lang	guages	Surv	ey set	tings	Sur	vey m	odes
Tucker's congruence coefficient			TCC			TCC			TCC
	German vs. Fre	ench	.998	classro	om vs.	000	W	eb vs.	00/
	French vs. Ita	lian	.999	unprod	ctored	.999	PAP .95		.994
	Italian vs. Gerr	man	.999						
Factor score equivalence: group									
specific vs. invariant models for	Survey	lang	guages	Surv	ey set	tings	Sur	vey m	odes
Coefficient of determination			CD			CD			CD
	Gern	nan	1.000	class	sroom	1.000		web	1.000
	Fre	ench	.999	unprod	ctored	.998		PAP	.996
	lta	alian	.997						

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. famedsup\_fs 0.0 1.1 -1.6 2.2 15592 Share of cases with imputed missing values: 14.6% (Equivalence of scores from robust MLMV: CD = .998)

### Scale: Social communication (PISA2000)

Baseline survey sample

#### **Model and Fit Statistics**

1)	<b>Likelihood-ratio tests</b> Model vs. saturated Baseline vs. saturated	chi2 o 9734	df o 3	p > chi2
2)	Root mean squared error ( 90% Confidence interval: 90% Confidence interval: Probability RMSEA <= 0.05	lower bound		.000 .000 .000 1.000
3)	Akaike's Information Crite Bayesian Information Crite	, ,		124277 124346

#### **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.723
(Cronbach's alpha = .647)	
McDonald's Omega	.729

#### Test of (one-)dimensionality (parallel analysis)

Criterion: Retain factors with adj. eigenvalue > o
Adjusted eigenvalue

factor 1	1.24
factor 2	11
factor 3	20

#### 4) Baseline comparison

Comparative Fit Index (CFI)	1.000
Tucker-Lewis Index (TLI)	1.000

#### 5) Size of residuals

Stand. root mean squared residual (SRMR)	.000
Coefficient of determination (CD)	.750

#### Standardized factor loadings

Indicators *	Coef.	(SE)	[95% Conf	. interval]
soccom1	0.57	0.01	0.56	0.58
soccom2	0.71	0.01	0.69	0.72
soccom3	0.78	0.01	0.76	0.79
				_

<sup>\*</sup> Note: Replication of 'Soccom'-Scale from TREE1 / PISA2000

-			
Item	des	crinti	Ves

		Std.			Valid
Indicators *	Mean	dev.	Min.	Max.	Obs.
soccom1	3.9	1.1	1	5	15566
soccom2	4.6	0.9	1	5	15570
soccom3	4.0	1.1	1	5	15555

Equal	ity o	f the
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variance-covariance matrices across	Survey languages		Surv	Survey settings			Survey modes		
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
	626	18	.000	611	9	.000	20	9	.017
Tests of measurement invariance across	. Survey	/ lan	guages	Surv	ey set	tings	Sur	vey m	odes
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
Metric invariance (equal factor loadings)	31	4	.000	26	2	.000	9	2	.012
Strong invariance (plus equal intercepts)	228	4	.000	107	2	.000	3	2	.231
Strict invariance (plus equal error variances)	92	4	.000	201	2	.000	3	2	.258
Configural factor similarity across	Survey languages		Survey settings			Survey modes			
Tucker's congruence coefficient			TCC		classroom vs. unproctored ·995				TCC
	German vs. Fre	ench	1.000	classro			web vs.		206
	French vs. Ita	alian	.992	unpro				.986	
	Italian vs. Ger	= =							
Factor score equivalence: group									
specific vs. invariant models for	Survey	/ lang	guages	Survey settings		tings	Sur	vey m	odes
Coefficient of determination			CD			CD			CD
	Gerr	man	1.000	clas	sroom	.998		web	.999
	Fre	ench	1.000	unpro	ctored	.980		PAP	.925

Italian .973

#### Factor score descriptives

Std.

# Scale: Social communication (adapted TREE2)

Baseline survey sample

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	0	0	
	Baseline vs. saturated	26651	3	.000
2)	Root mean squared error	(RMSEA)		.000
	90% Confidence interval:	lower bound		.000
	90% Confidence interval:	upper bound		.000
	Probability RMSEA <= 0.05			1.000
3)	Akaike's Information Crite	erion (AIC)		119342
	<b>Bayesian Information Crit</b>		119411	
4)	Baseline comparison			
	Comparative Fit Index (CFI)	)		1.000
	Tucker-Lewis Index (TLI)			1.000
5)	Size of residuals			
	Stand. root mean squared i	esidual (SRMF	?)	.000

#### **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.889
(Cronbach's alpha = .851)	
McDonald's Omega	.889

#### Test of (one-)dimensionality (parallel analysis)

Criterion: Retain factors with adj. eigenvalue > o

Adjusted eigenvalue

	Aujusteu eigenvalue	
factor 1	2.06	
factor 2	11	
factor 3	11	

# Standardized factor loadings

Coefficient of determination (CD)

Indicators *	Coef.	(SE)	[95% Conf	. interval]
soccom3 **	0.84	0.00	0.84	0.85
soccom4	o.86	0.00	0.85	0.86
soccom5	o.86	0.00	0.86	0.87

\* Note: Scale from TREE1 / PISA2000 adapted for TREE2

#### Item descriptives

.890

		Std.			Valid
Indicators *	Mean	dev.	Min.	Max.	Obs.
soccom3 **	4.0	1.1	1	5	15555
soccom4	3.9	1.2	1	5	15560
soccom5	4.0	1.1	1	5	15563

<sup>\*\*</sup> Note: Original item from TREE1 / PISA2000

Equality	of the
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variance-covariance matrices across	Survey languages		Survey settings			Survey modes			
	chi2 c	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
	942 1	8۱	.000	159	9	.000	49	9	.000
Tests of measurement invariance across	Survey la	ang	juages	Surv	ey set	tings	Sur	vey m	odes
	chi2 c	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
Metric invariance (equal factor loadings)	50	4	.000	5	2	.094	2	2	.459
Strong invariance (plus equal intercepts)	129	4	.000	37	2	.000	2	2	.408
Strict invariance (plus equal error variances)	211	4	.000	19	2	.000	6	2	.041
Configural factor similarity across	Survey languages		Survey settings		tings	Survey modes			
Tucker's congruence coefficient			TCC			TCC			TCC
	German vs. Fren	ich	.999	classroom vs.		1.000	W	eb vs.	1.000
	French vs. Italia	an	.999	unprod	ctored	1.000	PAP		1.000
	Italian vs. Germ	an	-997						
Factor score equivalence: group									
specific vs. invariant models for	Survey la	ang	juages	Surv	ey set	tings	Sur	vey m	odes
Coefficient of determination			CD			CD			CD
	Germa	an	1.000	class	sroom	1.000		web	1.000
	Fren	ich	1.000	unprod	ctored	1.000		PAP	1.000
	Itali	an	.997						

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. soccom\_m\_fs 0.0 0.9 -2.6 0.9 15591 Share of cases with imputed missing values: 0.5% (Equivalence of scores from robust MLMV: CD = .997)

# Scale: Perceived social network support

Baseline survey sample

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	2147	5	.000
	Baseline vs. saturated	58182	10	.000
2)	Root mean squared error (	(RMSEA)		.169
	90% Confidence interval:	lower bound		.163
	90% Confidence interval:	upper bound		.175
	Probability RMSEA <= 0.05			.000
3)	Akaike's Information Crite	rion (AIC)		233311
	Bayesian Information Crite	erion (BIC)		233425
4)	Baseline comparison			
	Comparative Fit Index (CFI)			.963
	Tucker-Lewis Index (TLI)			.926
5)	Size of residuals			

Stand. root mean squared residual (SRMR)

Coefficient of determination (CD)

#### **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.920
(Cronbach's alpha = .896)	
McDonald's Omega	.920

#### Test of (one-)dimensionality (parallel analysis)

Criterion: Retain factors with adj. eigenvalue > o

	Adjusted eigenvalue	
factor 1	3.45	
factor 2	.09	
factor 3	.00	
factor 4	06	
factor 5	12	

#### Standardized factor loadings

Indicators	Coef.	(SE)	[95% Conf	. interval]
closupp1	0.81	0.00	0.80	0.81
closupp2	0.93	0.00	0.93	0.93
closupp3	0.88	0.00	0.88	o.88
closupp4	0.68	0.00	0.67	0.69
closupp5	0.86	0.00	0.86	0.87

#### Item descriptives

.035

.939

		Std.			Valid
Indicators	Mean	dev.	Min.	Max.	Obs.
closupp1	5.4	1.6	1	7	14695
closupp2	5.6	1.6	1	7	14756
closupp3	5.7	1.6	1	7	14760
closupp4	5.1	1.7	1	7	14086
closupp5	5.5	1.8	1	7	14430

Equality	of the
----------	--------

variance-covariance matrices across	Survey languages		Survey settings			Survey modes		
	chi2 df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
	635 40	.000	802	20	.000	105	20	.000
Tests of measurement invariance across	. Survey lar	guages	Surv	ey set	tings	Sur	vey m	odes
	chi2 df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
Metric invariance (equal factor loadings)	33 8	.000	87	4	.000	8	4	.075
Strong invariance (plus equal intercepts)	205 8	.000	219	4	.000	13	4	.014
Strict invariance (plus equal error variances)	291 8	.000	17	4	.002	26	4	.000
Configural factor similarity across	Survey languages		Survey settings		Survey modes			
Tucker's congruence coefficient		TCC			TCC			TCC
	German vs. French	1.000	classro	om vs.	.998	W	eb vs.	000
	French vs. Italian	.999	unprod	ctored	.990		PAP	.999
	Italian vs. Germar	n .999						
Factor score equivalence: group								
specific vs. invariant models for	Survey lar	guages	Surv	ey set	tings	Sur	vey m	odes
Coefficient of determination		CD			CD			CD
	German	1.000	class	sroom	1.000		web	1.000
	French	1.000	unprod	ctored	.999		PAP	1.000
	Italiar	1.000						

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. closupp\_fs 0.0 1.2 -3.9 1.2 15034 Share of cases with imputed missing values: 10.4% (Equivalence of scores from robust MLMV: CD = .999)

Composite descriptives	Variable name	Mean	Std. dev.	Min.	Max.	Obs.		
i	oyreadp_comp	3.1	0.8	1	4	15244		
Share of cases with imputed m (Including "don't know"-answe	nissing values:	8.7%	5.5	-	4	-)-44		
Item descriptives	Indicators	Mean	Std. dev.	Min.	Max.	Valid obs.		
	joyreadm joyreadf	3.4 2.9	0.9 1.1	1 1	4 4	15004 14164		

# Scale: Cultural communication (PISA2000)

Baseline survey sample

#### **Model and Fit Statistics**

1)	<b>Likelihood-ratio tests</b> Model vs. saturated  Baseline vs. saturated	chi2 0 8034	df o 3	p > chi2
2)	Root mean squared error ( 90% Confidence interval: 90% Confidence interval: Probability RMSEA <= 0.05	lower bound		.000 .000 .000 1.000
3)	Akaike's Information Crite Bayesian Information Crite	• •		146251 146320
4)	Baseline comparison Comparative Fit Index (CFI) Tucker-Lewis Index (TLI)			1.000 1.000

#### **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.677
(Cronbach's alpha = .606)	
McDonald's Omega	.690

#### Test of (one-)dimensionality (parallel analysis)

Criterion: Retain factors with adj. eigenvalue > o
Adjusted eigenvalue

	,		
factor 1		1.11	
factor 2		10	
factor 3		21	

# 5) **Size of residuals** Stand. root mean

Stand. root mean squared residual (SRMR)	.000
Coefficient of determination (CD)	.727

#### Standardized factor loadings

Indicators *	Coef.	(SE)	[95% Conf	. interval]			
cultcom1	0.72	0.01	0.70	0.73			
cultcom2	0.75	0.01	0.74	0.77			
cultcom3	0.47	0.01	0.45	0.49			
* Note: Replication of 'Cultcom'-Scale from TREE1 / PISA2000							

#### Item descriptives

		Std.			Valid
Indicators *	Mean	dev.	Min.	Max.	Obs.
cultcom1	3.0	1.3	1	5	15593
cultcom2	3.2	1.3	1	5	15578
cultcom3	1.7	1.2	1	5	15575

Equal	lity	of	the
-------	------	----	-----

variance-covariance matrices across	Survey languages		guages	Survey settings			Survey modes		
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
	369	18	.000	267	9	.000	42	9	.000
Tests of measurement invariance across	Surve	/ lan	guages	Surv	ey set	tings	Sur	vey m	odes
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
Metric invariance (equal factor loadings)	16	4	.003	8	2	.019	1	2	.673
Strong invariance (plus equal intercepts)	263	4	.000	141	2	.000	14	2	.001
Strict invariance (plus equal error variances)	30	4	.000	15	2	.001	13	2	.002
Configural factor similarity across	Survey languages		Survey settings			Survey modes			
Tucker's congruence coefficient			TCC			TCC			TCC
	German vs. Fr	-	55	classro		.999	W	eb vs.	.998
	French vs. Ita		.987	unpro	ctored			PAP	
	Italian vs. Ger	man	.996						
Factor score equivalence: group									
specific vs. invariant models for	Surve	/ lang	guages	Surv	ey set	tings	Sur	vey m	odes
Coefficient of determination			CD			CD			CD
	Ger	man	.999	clas	sroom	1.000		web	1.000
	Fr	ench	.996	unpro	ctored	.999		PAP	.996

Italian

.970

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. cultcom\_fs 0.0 0.8 -1.6 1.8 15601 Share of cases with imputed missing values: 0.3% (Equivalence of scores from robust MLMV: CD = .998)

### Scale: Cultural communication (adapted TREE2)

Baseline survey sample

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	0	0	
	Baseline vs. saturated	16199	3	.000
2)	Root mean squared error	(RMSEA)		.000
	90% Confidence interval:	lower bound		.000
	90% Confidence interval:	upper bound		.000
	Probability RMSEA <= 0.05			1.000
3)	Akaike's Information Crite	erion (AIC)		137695
	<b>Bayesian Information Crit</b>	erion (BIC)		137764
4)	Baseline comparison			
	Comparative Fit Index (CFI)	)		1.000
	Tucker-Lewis Index (TLI)			1.000
5)	Size of residuals			
	Stand. root mean squared i	esidual (SRM	R)	.000

#### **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.811
(Cronbach's alpha = .762)	
McDonald's Omega	.814

#### Test of (one-)dimensionality (parallel analysis)

Criterion: Retain factors with adj. eigenvalue > o

	Adjusted eigenvalue
factor 1	1.63
factor 2	11
factor 3	17

#### Standardized factor loadings

Coefficient of determination (CD)

Indicators *	Coef.	(SE)	[95% Conf.	. interval]
cultcom1 **	0.80	0.00	0.79	0.81
cultcom2 **	0.68	0.01	0.67	0.69
cultcom4	0.83	0.00	0.82	0.84

<sup>\*</sup> Note: Scale from TREE1 / PISA2000 adapted for TREE2

#### Item descriptives

.829

		Std.			Valid
Indicators *	Mean	dev.	Min.	Max.	Obs.
cultcom1 **	3.0	1.3	1	5	15593
cultcom2 **	3.2	1.3	1	5	15578
cultcom4	3.8	1.1	1	5	15571

<sup>\*\*</sup> Note: Original items from TREE1 / PISA2000

Equality of the	Equa	lity	of t	he
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variance-covariance matrices across	Survey languages		Survey settings			Survey modes			
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
	313	18	.000	333	9	.000	26	9	.002
Tests of measurement invariance across	Survey	/ lan	guages	Surv	ey set	tings	Sur	vey m	odes
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
Metric invariance (equal factor loadings)	14	4	.008	8	2	.015	5	2	.073
Strong invariance (plus equal intercepts)	206	4	.000	212	2	.000	1	2	.519
Strict invariance (plus equal error variances)	30	4	.000	24	2	.000	7	2	.032
Configural factor similarity across	Survey languages		Survey settings		tings	Survey modes			
Tucker's congruence coefficient			TCC	TCC		TCC			TCC
	German vs. French 1.000 French vs. Italian .997			classro		.999	W	eb vs.	.997
			unpro	ctored	333	PAP		337	
	Italian vs. Ger	man	.996						
Factor score equivalence: group									
specific vs. invariant models for	Survey	lan	guages	Surv	ey set	tings	Sur	vey m	odes
Coefficient of determination	_		CD			CD			CD
	Gerr	man	1.000	clas	sroom	1.000		web	1.000
	Fre	ench	1.000	unpro	ctored	.999		PAP	.998

Italian .996

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. cultcom\_m\_fs 0.0 0.9 -2.4 1.5 15610 Share of cases with imputed missing values: 0.4% (Equivalence of scores from robust MLMV: CD = .997)

# Scale: Household possessions: classical culture (PISA2000)

Baseline survey sample

#### **Model and Fit Statistics**

#### **Reliability and Dimensionality**

1)	<b>Likelihood-ratio tests</b> Model vs. saturated  Baseline vs. saturated	chi2 0 11545	df o 3	p > chi2	Ordinal Cronbach's A (Cronbach's alpha = .5. McDonald's Omega	•	.720 .742
2)	Root mean squared error (190% Confidence interval: 90% Confidence interval: Probability RMSEA <= 0.05	lower bound		.000 .000 .000 1.000	Test of (one-)dimensi Criterion: Retain facto Adjust factor 1 factor 2		
3)	Akaike's Information Crite Bayesian Information Crite	, ,		52733 52802	factor 3	20	
4)	Baseline comparison Comparative Fit Index (CFI) Tucker-Lewis Index (TLI)			1.000			
5)	Size of residuals Stand. root mean squared re Coefficient of determination	•	R)	.000 .817			

#### Standardized factor loadings

#### Item descriptives

							Std.			Valid
Indicators *	Coef.	(SE)	[95% Conf	. interval]	Indicators *	Mean	dev.	Min.	Max.	Obs.
cultposs1	0.74	0.01	0.73	0.76	cultposs1	0.4	0.5		1	15977
cultposs2	0.86	0.01	0.85	0.88	cultposs2	0.4	0.5		1	15990
cultposs3	0.46	0.01	0.45	0.48	cultposs3	0.7	0.4		1	16009

<sup>\*</sup> Note: Replication of 'Cultposs'-Scale from TREE1 / PISA2000

## Parameters of generalized structural equation model (ordinal logit link)

i arameters or	generanzea s	ci occoi ai	equation	mouci
Indicators	Coef.	Cut1	Cut <sub>2</sub>	Cut <sub>3</sub>
cultposs1	1.90	0.71		
cultposs2	3.51	0.55		
cultposs3	0.91	-1.23		

# Scale: Household possessions: classical culture (PISA2000) (cont.)

Baseline survey sample

Tests and Indices of Factorial Invariance across ...

Equality of the									
variance-covariance matrices across	Surve	y lang	juages	Surve	ey sett	ings	Sur	vey m	odes
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
	4574	18	.000	101	9	.000	79	9	.000
Tests of measurement invariance across	Surve	y lang	juages	Surve	ey sett	ings	Sur	vey m	odes
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
Metric invariance (equal factor loadings)	53	4	.000	1	2	.759	13	2	.002
Strong invariance (plus equal intercepts)	887	4	.000	52	2	.000	21	2	.000
Strict invariance (plus equal error variances)	366	4	.000	21	2	.000	19	2	.000
Configural factor similarity across	Surve	y lang	juages	Surve	ey sett	ings	Survey modes		
Tucker's congruence coefficient		,	TCC		•	TCC			TCC
-	German vs. F	rench	.996	classroo	om vs.		w	eb vs.	
	French vs. It	alian	1.000	unproc	tored	1.000		PAP	.996
	Italian vs. Ge	rman	.997						
Factor score equivalence: group									
specific vs. invariant models for	Surve	Survey languages		Surve	ey sett	inas	Sur	vey m	odes
Coefficient of determination	33.13	,	CD	55.11	.,	CD		, ,	CD
	Gei	man	.979	class	room	1.000		web	-
		rench	.890	unproc		1.000		PAP	.985
		talian	.819	op. o c		2.000			.505
Factor score descriptives			,				* Note:	The	
• Std.							calculation	on of i	model-
Variable name Mean dev. Min. N	Лах. Obs.						based in	varian	ce tests
cultposs_fs	1.1 16028						requires	that w	/e
Share of cases with imputed missing values	: 0.5%						constrair	the e	error
(Equivalence of scores from robust MLMV:	CD = .969)						variance	of cul	tposs2 to
(Equivalence of Scores from Two-Step-Appro	oach: CD = .96)						zero.		

#### Scale: Embodied cultural capital

Baseline survey sample

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	1455	9	.000
	Racolina vs. caturated	/2012	1.5	000

Baseline vs. saturated 42913 .000

2) Root mean squared error (RMSEA) .101 90% Confidence interval: lower bound .096 90% Confidence interval: upper bound .105 Probability RMSEA <= 0.05 .000

3) Akaike's Information Criterion (AIC) 166162 **Bayesian Information Criterion (BIC)** 166300

4) Baseline comparison Comparative Fit Index (CFI) .966 Tucker-Lewis Index (TLI) .944

5) Size of residuals

Stand. root mean squared residual (SRMR) .033 Coefficient of determination (CD) .883

#### **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.870
(Cronbach's alpha = .822)	
McDonald's Omega	.872

Test of (one-)dimensionality (parallel analysis)

Criterion: Retain factors with adj. eigenvalue > o

	Adjusted eigenvalue
factor 1	3.13
factor 2	.11
factor 3	04
factor 4	05
factor 5	12
factor 6	15
factor 3 factor 4 factor 5	.11 04 05 12

#### Standardized factor loadings

Indicators	Coef.	(SE)	[95% Conf.	. interval]
manners1	0.53	0.01	0.52	0.55
manners2	0.80	0.00	0.80	0.81
manners3	0.74	0.00	0.73	0.75
verbskill1	0.75	0.00	0.74	0.76
verbskill2	0.78	0.00	0.78	0.79
verbskill3	0.75	0.00	0.74	0.75

#### Item descriptives

		Std.			Valid
Indicators	Mean	dev.	Min.	Max.	Obs.
manners1	3.0	0.8	1	4	15819
manners2	3.1	0.7	1	4	15805
manners3	3.1	0.7	1	4	15807
verbskill1	3.0	0.7	1	4	15827
verbskill2	3.0	0.8	1	4	15817
verbskill3	2.9	0.7	1	4	15776

#### Parameters of generalized structural equation model (ordinal logit link)

. a.aetc.b o. ge		c. occo. a.	equation	
Indicators	Coef.	Cut1	Cut <sub>2</sub>	Cut <sub>3</sub>
manners1	1.21	-3.68	-1.95	1.19
manners2	2.57	-6.65	-2.90	1.98
manners3	2.10	-6.12	-2.90	1.50
verbskill1	2.13	-5.28	-2.04	1.80
verbskill2	2.39	-5.71	-2.08	1.73
verbskill3	2.13	-5.33	-1.79	2.15

Equal	lity	of	the
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variance-covariance matrices across	Survey languages		Survey settings			Survey modes			
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
	765	54	.000	221	27	.000	63	27	.000
Tests of measurement invariance across	. Surve	Survey languages		Survey settings		tings	Survey modes		odes
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
Metric invariance (equal factor loadings)	21	10	.018	36	5	.000	14	5	.018
Strong invariance (plus equal intercepts)	70	10	.000	24	5	.000	10	5	.085
Strict invariance (plus equal error variances)	197	10	.000	57	5	.000	15	5	.011
Configural factor similarity across	Survey languages		Survey settings			Survey modes			
Tucker's congruence coefficient			TCC			TCC			TCC
	German vs. Fr	ench	1.000	classro	om vs.	000	W	eb vs.	007
	French vs. It	alian	.999	unprod	nproctored <sup>.999</sup>		PAP .9		.997
	Italian vs. Ge	rman	.999						
Factor score equivalence: group									
specific vs. invariant models for	Surve	y lan	guages	Surv	ey set	tings	Sur	vey m	odes
Coefficient of determination			CD			CD			CD
	Ger	man	1.000	class	sroom	1.000		web	1.000
	Fr	ench	1.000	unprod	ctored	.999		PAP	.998
	lt	alian	.999						

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. inccap\_fs 0.0 0.9 1.8 15846 -3.2 Share of cases with imputed missing values: 0.9% (Equivalence of scores from robust MLMV: CD = .999) (Equivalence of Scores from Two-Step-Approach: CD = .989)

# Scale: Embodied cultural capital: manners

Baseline survey sample

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	0	0	
	Baseline vs. saturated	12618	3	.000
2)	Root mean squared error	(RMSEA)		.000
	90% Confidence interval:	lower bound		.000
	90% Confidence interval:	upper bound		.000
	Probability RMSEA <= 0.05			1.000
3)	Akaike's Information Crite	erion (AIC)		88215
	Bayesian Information Crit	erion (BIC)		88284
4)	Baseline comparison			
	Comparative Fit Index (CFI)	)		1.000
	Tucker-Lewis Index (TLI)			1.000

#### **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.763
(Cronbach's alpha = .684)	
McDonald's Omega	.769

#### Test of (one-)dimensionality (parallel analysis)

Criterion: Retain factors with adj. eigenvalue > o Adjusted eigenvalue

factor 1	1.41	
factor 2	10	
factor 3	20	

Comparative Fit Index (CFI)	1.000
Tucker-Lewis Index (TLI)	1.000

#### 5) Size of residuals

Stand. root mean squared residual (SRMR)	.000
Coefficient of determination (CD)	.798

### Standardized factor loadings

#### Item descriptives

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
manners1	0.60	0.01	0.58	0.61	manners1	3.0	0.8	1	4	15819
manners2	0.74	0.01	0.73	0.76	manners2	3.1	0.7	1	4	15805
manners3	0.83	0.01	0.81	0.84	manners3	3.1	0.7	1	4	15807

### Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut <sub>1</sub>	Cut <sub>2</sub>	Cut <sub>3</sub>
manners1	1.41	-3.87	-2.07	1.28
manners2	2.10	-5.87	-2.59	1.77
manners3	2.85	-7.40	-3.62	1.88

Equal	lity	of	the
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variance-covariance matrices across	Survey languages		Survey settings			Survey modes			
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
	470	18	.000	138	9	.000	15	9	.082
Tests of measurement invariance across	Surve	y lan	guages	Surv	ey sett	tings	Sur	vey m	odes
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
Metric invariance (equal factor loadings)	7	4	.160	1	2	.751	3	2	.231
Strong invariance (plus equal intercepts)	28	4	.000	16	2	.000	3	2	.280
Strict invariance (plus equal error variances)	40	4	.000	14	2	.001	4	2	.119
Configural factor similarity across	Surve	y lan	guages	Survey settings			Survey modes		
Tucker's congruence coefficient			TCC		•	TCC		•	TCC
	German vs. F	rench	.999	classro	om vs.	1 000	W	eb vs.	200
	French vs. It	alian	.999	unpro	ctored	1.000		PAP	.998
	Italian vs. Ge	rman	.999	•					
Factor score equivalence: group									
specific vs. invariant models for	Surve	y lan	guages	Surv	ey sett	tings	Sur	vey m	odes
Coefficient of determination			CD		-	CD		=	CD
	Gei	man	1.000	clas	sroom	1.000		web	1.000
	Fi	rench	.998	unpro	ctored	1.000		PAP	.998

Italian

.997

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. manners\_fs 0.0 0.8 -2.8 1.5 15843 Share of cases with imputed missing values: 0.5% (Equivalence of scores from robust MLMV: CD = .998) (Equivalence of Scores from Two-Step-Approach: CD = .988)

# Scale: Embodied cultural capital: verbal skills

Baseline survey sample

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	0	0	
	Baseline vs. saturated	16621	3	.000
2)	Root mean squared error (	(RMSEA)		.000
	90% Confidence interval:	lower bound		.000
	90% Confidence interval:	upper bound		.000
	Probability RMSEA <= 0.05			1.000
3)	Akaike's Information Crite	rion (AIC)		90127
	Bayesian Information Crite	erion (BIC)		90196
4)	Baseline comparison			
	Comparative Fit Index (CFI)			1.000
	Tucker-Lewis Index (TLI)			1.000
5)	Size of residuals			

#### **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.818
(Cronbach's alpha = .759)	
McDonald's Omega	.819

#### Test of (one-)dimensionality (parallel analysis)

Criterion: Retain factors with adj. eigenvalue > o
Adjusted eigenvalue

	,		
factor 1		1.64	
factor 2		14	
factor 3		15	

# Standardized factor loadings

Indicators	Coef.	(SE)	[95% Conf	. interval]
verbskill1	0.74	0.00	0.73	0.75
verbskill2	0.80	0.00	0.79	0.81
verbskilla	0.79	0.00	0.78	0.80

Stand. root mean squared residual (SRMR)

Coefficient of determination (CD)

#### Item descriptives

.000

.821

		Std.			Valid
Indicators	Mean	dev.	Min.	Max.	Obs.
verbskill1	3.0	0.7	1	4	15827
verbskill2	3.0	0.8	1	4	15817
verbskilla	2.9	0.7	1	4	15776

### Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut1	Cut <sub>2</sub>	Cut <sub>3</sub>
verbskill1	2.03	-5.16	-2.00	1.78
verbskill2	2.49	-5.91	-2.15	1.82
verbskill3	2.43	-5.80	-1.96	2.36

Equal	lity	of t	the
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variance-covariance matrices across	Survey languages		Surv	Survey settings			Survey modes		
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
	209	18	.000	24	9	.005	34	9	.000
Tests of measurement invariance across	Surve	y lan	guages	Surv	ey sett	tings	Sur	vey m	odes
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
Metric invariance (equal factor loadings)	6	4	.227	4	2	.137	12	2	.003
Strong invariance (plus equal intercepts)	36	4	.000	2	2	.425	4	2	.106
Strict invariance (plus equal error variances)	89	4	.000	13	2	.002	8	2	.023
Configural factor similarity across	Survey languages		Survey settings			Sur	vey m	odes	
Tucker's congruence coefficient			TCC		•	TCC		•	TCC
	German vs. Fr	ench	1.000	classro		1.000	W	eb vs.	.995
	French vs. Ita	alian	.998	unpro	ctored	2.000		PAP	.555
	Italian vs. Gei	rman	.999						
Factor score equivalence: group									
specific vs. invariant models for	Surve	y lan	guages	Surv	ey sett	tings	Sur	vey m	odes
Coefficient of determination			CD			CD			CD
	Ger	man	1.000	clas	sroom	1.000		web	1.000
	Fr	ench	1.000	unpro	ctored	1.000		PAP	.993

Italian

.998

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. verbskill\_fs 0.0 0.9 -2.7 1.6 15841
Share of cases with imputed missing values: 0.6% (Equivalence of scores from robust MLMV: CD = .999)
(Equivalence of Scores from Two-Step-Approach: CD = .992)

### Scale: Cultural activities

Baseline survey sample

#### **Model and Fit Statistics**

#### **Reliability and Dimensionality**

1)	Likelihood-ratio tests	chi2	df	p > chi2	Ordinal Cronbac	:h's Alpha	.743
	Model vs. saturated	7949	14	.000	(Cronbach's alph	a = .668)	
	Baseline vs. saturated	27943	21	.000	McDonald's Om	ega	.726
2)	Root mean squared error (	RMSEA)		.189	Test of (one-)dir	mensionality (para	llel analysis)
	90% Confidence interval:	lower bound		.186	Criterion: Retain	factors with adj. eig	genvalue > o
	90% Confidence interval:	upper bound		.193	A	Adjusted eigenvalue	<u> </u>
	Probability RMSEA <= 0.05			.000	factor 1	2.14	
					factor 2	.76	
3)	Akaike's Information Criter	rion (AIC)		260288	factor 3	.03	
	<b>Bayesian Information Crite</b>	rion (BIC)		260449	factor 4	02	
					factor 5	13	
4)	Baseline comparison				factor 6	20	
	Comparative Fit Index (CFI)			.716	factor 7	20	
	Tucker-Lewis Index (TLI)			.574			
5)	Size of residuals						
	Stand. root mean squared re	esidual (SRMI	?)	.118			
	Coefficient of determination	n (CD)		.809			

### Standardized factor loadings

#### Item descriptives

							Std.			Valid
Indicators *	Coef.	(SE)	[95% Conf	. interval]	Indicators *	Mean	dev.	Min.	Max.	Obs.
cult1 **	0.36	0.01	0.34	0.37	cult1 **	2.8	1.0	1	4	15787
cult2 **	0.70	0.01	0.69	0.71	cult2 **	1.8	0.9	1	4	15776
cult3 **	0.50	0.01	0.48	0.51	cult3 **	1.6	0.8	1	4	15769
cult4 **	0.77	0.00	0.76	0.78	cult4 **	1.3	0.6	1	4	15771
cult5 **	0.74	0.01	0.73	0.75	cult5 **	1.6	0.7	1	4	15761
cult7	0.29	0.01	0.27	0.31	cult7	2.6	1.0	1	4	15766
cult9	0.24	0.01	0.23	0.26	cult9	2.4	1.2	1	4	15761

<sup>\*</sup> Note: Scale from TREE1 / PISA2000 adapted for TREE2

#### Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut1	Cut <sub>2</sub>	Cut <sub>3</sub>
cult1	0.83	-2.69	-0.45	0.93
cult2	1.54	-0.32	2.13	3.59
cult3	1.17	0.43	2.48	3.64
cult4	1.93	2.19	4.18	5.39
cult5	1.76	0.12	3.13	4.74
cult7	0.70	-1.83	0.18	1.41
cult9	0.60	-0.93	0.30	1.17

<sup>\*\*</sup> Note: Original items from TREE1 / PISA2000

Equa	lity	of t	:he
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variance-covariance matrices across	Surve	Survey languages		Surv	Survey settings			Survey modes		
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2	
	1553	70	.000	737	35	.000	149	35	.000	
Tests of measurement invariance across	Surve	y lan	guages	Surv	ey set	tings	Sur	vey m	odes	
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2	
Metric invariance (equal factor loadings)	107	12	.000	30	6	.000	19	6	.005	
Strong invariance (plus equal intercepts)	1198	12	.000	231	6	.000	74	6	.000	
Strict invariance (plus equal error variances)	142	12	.000	269	6	.000	35	6	.000	
Configural factor similarity across	Surve	y lan	guages	Surv	ey set	tings	Sur	vey m	odes	
Tucker's congruence coefficient			TCC			TCC			TCC	
	German vs. F French vs. It Italian vs. Ge	alian	.996	classro unpro	om vs. ctored	.997	W	eb vs. PAP	.987	
Factor score equivalence: group specific vs. invariant models for Coefficient of determination	Surve	y lan	guages	Surv	ey set	_	Sur	vey m	odes CD	
Coefficient of determination	Ge	rman	CD 1.000	clas	sroom	CD 1.000		web		
	GC.	man	1.000	Clas	3100111	1.000		WED	1.000	

French

Italian

.995

.998

unproctored

.997

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. cult\_fs 0.0 0.8 -1.8 3.1 15797

Share of cases with imputed missing values: 0.6% (Equivalence of scores from robust MLMV: CD = .977) (Equivalence of Scores from Two-Step-Approach: CD = .886)

List of scales (wave 0)

PAP

.990

#### Scale: Lowbrow cultural activities

Baseline survey sample

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	0	0	
	Baseline vs. saturated	7348	3	.000

Baseline vs. saturated 7348 3 .000

2) Root mean squared error (RMSEA) .000

.000

.000

1.000

90% Confidence interval: lower bound 90% Confidence interval: upper bound Probability RMSEA <= 0.05

3) Akaike's Information Criterion (AIC) 124416 Bayesian Information Criterion (BIC) 124485

4) Baseline comparison

Comparative Fit Index (CFI) 1.000
Tucker-Lewis Index (TLI) 1.000

5) Size of residuals

Stand. root mean squared residual (SRMR) .000 Coefficient of determination (CD) .728

#### **Reliability and Dimensionality**

Ordinal Cronbach's Alpha .668 (Cronbach's alpha = .599) McDonald's Omega .679

Test of (one-)dimensionality (parallel analysis)

Criterion: Retain factors with adj. eigenvalue > o

Adjusted eigenvalue factor 1 1.05 factor 2 -.10 factor 3 -.22

#### Standardized factor loadings

Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicator
cult3 *	0.54	0.01	0.52	0.56	cult3 *
cult7	0.58	0.01	0.56	0.59	cult7
cult9	0.80	0.01	0.78	0.82	cult9

* Note: Original item from	TREE1 / PISA2000
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#### Item descriptives

		Std.			Valid
Indicators	Mean	dev.	Min.	Max.	Obs.
cult3 *	1.6	0.8	1	4	15769
cult7	2.6	1.0	1	4	15766
cult9	2.4	1.2	1	4	15761

# Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut1	Cut <sub>2</sub>	Cut <sub>3</sub>
cult3	1.11	0.43	2.46	3.56
cult7	1.27	-2.14	0.17	1.64
culta	2.25	-1.53	0.47	1.88

Equa	lity	of t	:he
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variance-covariance matrices across	Survey languages			Survey settings			Survey modes		
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
	993	18	.000	164	9	.000	50	9	.000
Tests of measurement invariance across	Surve	y lan	guages	Surv	ey set	tings	Sur	vey m	odes
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
Metric invariance (equal factor loadings)	65	4	.000	18	2	.000	13	2	.002
Strong invariance (plus equal intercepts)	674	4	.000	107	2	.000	24	2	.000
Strict invariance (plus equal error variances)	162	4	.000	13	2	.002	5	2	.071
Configural factor similarity across	Survey languages		Survey settings			Survey modes			
Tucker's congruence coefficient			TCC	TCC					TCC
	German vs. Fr French vs. It Italian vs. Ge	alian	.999	classro unpro		.996	W	eb vs. PAP	.949
Factor score equivalence: group specific vs. invariant models for			.gog quaqes	Surv	ey seti	tinas	Sur	vey m	odes
Coefficient of determination	23.72	,	CD		-,	CD		-,	CD
	Ger	man	.992	clas	sroom			web	.999
		rench		unpro		333		PAP	.852

Italian

.996

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. cultlow\_fs 0.0 0.8 -1.4 1.8 15788

Share of cases with imputed missing values: 0.3% (Equivalence of scores from robust MLMV: CD = .99)

(Equivalence of Scores from Two-Step-Approach: CD = .975)

### Scale: Highbrow cultural activities [PISA 2000]

Baseline survey sample

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	0	0	
	Baseline vs. saturated	14402	3	.000
2)	Root mean squared error	(RMSEA)		.000
	90% Confidence interval:	lower bound		.000
	90% Confidence interval:	upper bound		.000
	Probability RMSEA <= 0.05			1.000
3)	Akaike's Information Crite	erion (AIC)		90498
	Bayesian Information Crit		90567	
4)	Baseline comparison			
	Comparative Fit Index (CFI)	)		1.000
	Tucker-Lewis Index (TLI)			1.000

#### **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.793
(Cronbach's alpha = .690)	
McDonald's Omega	.795

#### Test of (one-)dimensionality (parallel analysis)

Criterion: Retain factors with adj. eigenvalue > o Adjusted eigenvalue

	_	-	
factor 1		1.53	
factor 2		13	
factor 3		17	

Comparative Fit Index (CFI)	1.000
Tucker-Lewis Index (TLI)	1.000

#### 5) Size of residuals

Stand. root mean squared residual (SRMR)	.000
Coefficient of determination (CD)	.805

#### Standardized factor loadings

# Item descriptives

							Std.			Valid
Indicators *	Coef.	(SE)	[95% Conf.	. interval]	Indicators *	Mean	dev.	Min.	Max.	Obs.
cult2	0.69	0.01	o.68	0.70	cult2	1.8	0.9	1	4	15776
cult4	0.82	0.01	0.81	0.83	cult4	1.3	0.6	1	4	15771
cult5	0.74	0.01	0.73	0.75	cult5	1.6	0.7	1	4	15761

<sup>\*</sup> Note: Replication of 'Cultactv'-Scale from TREE1 / PISA2000

#### Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut1	Cut <sub>2</sub>	Cut <sub>3</sub>
cult2	1.69	-0.33	2.26	3.75
cult4	2.53	2.64	4.95	6.28
cult5	2.01	0.15	3.41	5.05

Equal	lity	of	the
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variance-covariance matrices across	Survey languages		Survey settings			Survey modes			
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
	283	18	.000	436	9	.000	58	9	.000
Tests of measurement invariance across	Surve	y lan	guages	Surv	ey set	tings	Sur	vey m	odes
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
Metric invariance (equal factor loadings)	40	4	.000	5	2	.085	1	2	.518
Strong invariance (plus equal intercepts)	125	4	.000	48	2	.000	10	2	.008
Strict invariance (plus equal error variances)	48	4	.000	176	2	.000	13	2	.001
Configural factor similarity across	Survey languages		Survey settings			Survey modes			
Tucker's congruence coefficient			TCC			TCC			TCC
	German vs. Fr French vs. It	-	.997 .999	classro unpro	om vs. ctored	.999	W	eb vs. PAP	.999
	Italian vs. Ge	rman							
Factor score equivalence: group									
specific vs. invariant models for	Surve	y lan	guages	Surv	ey set	tings	Sur	vey m	odes
Coefficient of determination			CD			CD			CD
	Ger	man	1.000	clas	sroom	1.000		web	1.000
	Fr	rench	.993	unpro	ctored	.999		PAP	.996

Italian

.999

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. culthigh\_fs 0.0 0.8 -0.9 2.6 15788

Share of cases with imputed missing values: 0.3% (Equivalence of scores from robust MLMV: CD = .98) (Equivalence of Scores from Two-Step-Approach: CD = .886)

# Scale: Household Possessions: Family Wealth (PISA2000)

Baseline survey sample

#### **Model and Fit Statistics**

#### **Reliability and Dimensionality**

1)	<b>Likelihood-ratio tests</b> Model vs. saturated Baseline vs. saturated	chi2 12119 41971	df 27 36	p > chi2 .000 .000	Ordinal Cronbach (Cronbach's alpha McDonald's Ome	ı = .565)	.782 .789
2)	Root mean squared error (R			.167		nensionality (paralle	•
	90% Confidence interval: logo% Confidence interval: u	ower bound opper bound		.000		actors with adj. eiger djusted eigenvalue	ivalue > 0
	Probability RMSEA <= 0.05			.000	factor 1 factor 2	2.83	
3)	Akaike's Information Criteri	ion (AIC)		138697	factor 3	.49 .40	
	Bayesian Information Criter	rion (BIC)		138904	factor 4 factor 5	.08	
4)	Baseline comparison				factor 6	.07 .02	
	Comparative Fit Index (CFI)			.712	factor 7	10	
	Tucker-Lewis Index (TLI)			.616	factor 8	15	
					factor 9	25	
5)	Size of residuals						
	Stand. root mean squared re	sidual (SRMR	)	.079			
	Coefficient of determination	(CD)		.839			

#### Item descriptives

							Std.			Valid
Indicators *	Coef.	(SE)	[95% Conf.	. interval]	Indicators *	Mean	dev.	Min.	Max.	Obs.
wealth1	0.71	0.00	0.70	0.72	wealth1	0.9	0.3		1	16040
wealth2	0.57	0.01	0.56	0.58	wealth2	0.9	0.3		1	16039
wealth3	0.31	0.01	0.29	0.32	wealth3	0.6	0.5		1	15942
wealth4	0.81	0.00	0.80	0.82	wealth4	1.0	0.1		1	16043
wealthn1	0.59	0.01	0.58	0.61	wealthn1	3.9	0.4	1	4	16037
wealthn2	0.35	0.01	0.33	0.36	wealthn2	2.8	0.8	1	4	16037
wealthn3	0.50	0.01	0.49	0.51	wealthn3	3.3	0.8	1	4	16032
wealthn4	0.42	0.01	0.41	0.44	wealthn4	2.7	0.8	1	4	16030
wealthn5	0.55	0.01	0.54	0.56	wealthn5	2.9	0.7	1	4	16037

<sup>\*</sup> Note: Replication of 'Wealth'-Scale from TREE1 / PISA2000

# Parameters of generalized structural equation model (ordinal logit link)

Parameters of ge	neralized s	tructurai	equation	moaei (
Indicators	Coef.	Cut1	Cut <sub>2</sub>	Cut <sub>3</sub>
wealth1	1.64	-3.46		
wealth2	1.08	-2.75		
wealth3	0.29	-0.36		
wealth4	1.76	-5.87		
wealthn1	1.46	-6.37	-4.51	-3.29
wealthn2	0.79	-3.58	-0.51	1.35
wealthn3	1.01	-4.94	-1.65	-0.01
wealthn4	1.18	-3.18	-0.25	2.19
wealthn5	1.48	-6.23	-1.26	2.00

### Scale: Household Possessions: Family Wealth (PISA2000) (cont.)

Baseline survey sample

Tests and Indices of Factorial Invariance across ...

Equal	lity	of t	the
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variance-covariance matrices across	Surve	y lang	guages	Surv	ey set	tings	Sur	vey m	odes
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
	4879	108	.000	1025	54	.000	1065	54	.000
Tests of measurement invariance across	Surve	y lan	guages	Surv	ey set	tings	Sur	vey m	odes
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
Metric invariance (equal factor loadings)	139	16	.000	92	8	.000	103	8	.000
Strong invariance (plus equal intercepts)	499	16	.000	74	8	.000	44	8	.000
Strict invariance (plus equal error variances)	1367	16	.000	270	8	.000	147	8	.000
Configural factor similarity across	Surve	y lan	guages	Surv	ey set	tings	Sur	vey m	odes
Tucker's congruence coefficient			TCC			TCC			TCC
	German vs. F		.989	classro		.997	W	eb vs.	.965
	French vs. It	talian	.992	unpro	ctored	-557		PAP	-5-5
	Italian vs. Ge	erman	.991						
Factor score equivalence: group									
specific vs. invariant models for	Surve	y lan	guages	Surv	ey set	tings	Sur	vey m	odes
Coefficient of determination			CD			CD			CD
	Ge	rman	1.000	clas	sroom	1.000		web	.997
	F	rench	.999	unpro	ctored	.995		PAP	.964

Italian

.959

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. wealth\_fs 0.0 0.8 -4.0 1.8 16057

Share of cases with imputed missing values: 1.0% (Equivalence of scores from robust MLMV: CD = .641) (Equivalence of Scores from Two-Step-Approach: CD = .508)

# Scale: Household Possessions: Family Wealth (adapted TREE2)

Baseline survey sample

#### **Model and Fit Statistics**

#### **Reliability and Dimensionality**

1)	<b>Likelihood-ratio tests</b> Model vs. saturated  Baseline vs. saturated	chi2 df 8521 14 38309 21	p > chi2 .000 .000	Ordinal Cronbach's A (Cronbach's alpha = .5. McDonald's Omega	,
	Root mean squared error (RI 90% Confidence interval: lo 90% Confidence interval: up Probability RMSEA <= 0.05	wer bound	.195 .191 .198 .000	Criterion: Retain facto	onality (parallel analysis) rs with adj. eigenvalue > o red eigenvalue 2.76 .46
4)	Akaike's Information Criterion Bayesian Information Criterion Baseline comparison Comparative Fit Index (CFI)		59604 59765 .778	factor 3 factor 4 factor 5 factor 6 factor 7	.20 .02 07 12
5)	Tucker-Lewis Index (CFI)  Size of residuals  Stand. root mean squared res  Coefficient of determination (		.7/° .667 .079 .837	ractor /	24

#### Standardized factor loadings

#### Item descriptives

	_				•		Std.			Valid
Indicators *	Coef.	(SE)	[95% Conf	. interval]	Indicators *	Mean	dev.	Min.	Max.	Obs.
wealth1 **	0.77	0.00	0.76	0.77	wealth1 **	0.9	0.3		1	16040
wealth2 **	0.62	0.01	0.61	0.63	wealth2 **	0.9	0.3		1	16039
wealth4 **	0.75	0.00	0.74	0.76	wealth4 **	1.0	0.1		1	16043
wealth5	0.61	0.01	0.60	0.62	wealth5	0.7	0.5		1	16021
wealthn1 **	0.51	0.01	0.50	0.52	wealthn1 **	3.9	0.4	1	4	16037
wealthn3 **	0.47	0.01	0.46	0.49	wealthn3 **	3.3	8.0	1	4	16032
wealthn5 **	0.60	0.01	0.59	0.61	wealthn5 **	2.9	0.7	1	4	16037

<sup>\*</sup> Note: Scale from TREE1 / PISA2000 adapted for TREE2

#### Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut1	Cut <sub>2</sub>	Cut <sub>3</sub>
wealth1	2.07	-3.91		
wealth2	1.43	-3.03		
wealth4	2.04	-6.28		
wealth5	1.44	-0.76		
wealthn1	1.07	-5.80	-4.09	-2.96
wealthn3	0.87	-4.81	-1.60	-0.01
wealthn5	1.79	-6.65	-1.40	2.20

<sup>\*\*</sup> Note: Original items from TREE1/PISA2000

Equality of the	Equa	lity	of t	he
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variance-covariance matrices across	Survey	lan	guages	Surv	ey set	tings	Sur	vey m	odes
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
	2014	70	.000	777	35	.000	890	35	.000
Tests of measurement invariance across	Survey	lan	guages	Surv	ey set	tings	Sur	vey m	odes
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
Metric invariance (equal factor loadings)	168	12	.000	144	6	.000	74	6	.000
Strong invariance (plus equal intercepts)	329	12	.000	65	6	.000	25	6	.000
Strict invariance (plus equal error variances)	983	12	.000	175	6	.000	140	6	.000
Configural factor similarity across	Survey	lan	guages	Surv	ey set	tings	Sur	vey m	odes
Tucker's congruence coefficient			TCC			TCC			TCC
	German vs. Fre		.996 .975	classro unpro	om vs. ctored	.992	W	eb vs. PAP	.965
	Italian vs. Ger			·					
Factor score equivalence: group									
specific vs. invariant models for	Survey	lan	guages	Surv	ey set	tings	Sur	vey m	odes
Coefficient of determination			CD			CD			CD
	Gerr	nan	.999	clas	sroom	.999		web	.999
	Fre	ench	.978	unpro	ctored	.991		PAP	.947

Italian

.902

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. wealth\_m\_fs 0.0 0.8 -3.6 1.3 16056 Share of cases with imputed missing values: 0.4% (Equivalence of scores from robust MLMV: CD = .83) (Equivalence of Scores from Two-Step-Approach: CD = .692)

Composite descriptives	Variable name	Mean	Std. dev.	Min.	Max.	Obs.	
Share of cases with imputed	fasiii_comp missing values:	9·5 o.5%	2.1	0	13	16059	
Item descriptives			Std.			Valid	
·	Indicators	Mean	dev.	Min.	Max.	obs.	
						_	
	wealthn4	1.5	0.6	0	2	- 5 -	*
	wealth2	0.9	0.3	0	1	16039	
	wealthn3	2.3	0.8	0	3	16032	*
	wealthn5	1.9	0.7	0	3	16037	*
	wealth1	0.9	0.3	0	1	16040	
	Wealth	0.9	0.5	•	-	10040	

<sup>\*</sup> Items recoded for composite calculation (see Hobza et al. 2017)

Scale: Capabilities	Baseline survey	samp
Scale: Capabilities	Baseline survey	Sa

p > chi2

.000

.000

df

5

10

#### **Model and Fit Statistics**

1) Likelihood-ratio tests

Model vs. saturated

Baseline vs. saturated

Ordinal Cronbach's
(Cronbach's alpha =

#### 's Alpha .871 .845) McDonald's Omega .871

**Reliability and Dimensionality** 

#### 2) Root mean squared error (RMSEA) .145 90% Confidence interval: lower bound .139 90% Confidence interval: upper bound .151 Probability RMSEA <= 0.05 .000

chi2

1666

37134

#### Test of (one-)dimensionality (parallel analysis) Criterion: Retain factors with adj. eigenvalue > o Adjusted eigenvalue

3)	Akaike's Information Criterion (AIC)	221347
	Bayesian Information Criterion (BIC)	221462

factor 1	2.79
factor 2	.10
factor 3	07
factor 4	13
factor 5	13

#### 4) Baseline comparison

Comparative Fit Index (CFI) .955 Tucker-Lewis Index (TLI) .911

#### 5) Size of residuals

Stand. root mean squared residual (SRMR) .038 Coefficient of determination (CD) .874

#### Standardized factor loadings

#### Item descriptives

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
cap1	0.76	0.00	0.75	0.77	cap1	5.9	1.3	1	7	15756
cap2	0.78	0.00	0.77	0.79	cap2	5.7	1.2	1	7	15733
cap3	0.79	0.00	0.78	0.80	cap3	5.9	1.2	1	7	15732
cap4	0.69	0.00	0.68	0.70	cap4	5.3	1.3	1	7	15714
cap5	0.76	0.00	0.75	0.77	cap5	5.7	1.2	1	7	15738

Equal	ity o	f the
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variance-covariance matrices across	Survey languages		Survey settings			Survey modes			
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
	1233	40	.000	412	20	.000	32	20	.042
Tests of measurement invariance across	Surve	y lang	guages	Surv	ey set	tings	Sur	vey m	odes
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
Metric invariance (equal factor loadings)	106	8	.000	21	4	.000	7	4	.145
Strong invariance (plus equal intercepts)	601	8	.000	75	4	.000	11	4	.025
Strict invariance (plus equal error variances)	216	8	.000	15	4	.005	4	4	.456
Configural factor similarity across	Survey languages		Survey settings		tings	Survey modes		odes	
Tucker's congruence coefficient			TCC			TCC			TCC
	German vs. Fr	ench	.996	classro	om vs.	000	W	eb vs.	000
	French vs. Ita	alian	.997	unpro	ctored	.999		PAP	.998
	Italian vs. Ger	man	.997						
Factor score equivalence: group									
specific vs. invariant models for	Surve	y lan	guages	Surv	ey set	tings	Sur	vey m	odes
Coefficient of determination			CD			CD			CD
	Ger	man	1.000	clas	sroom	1.000		web	1.000
	Fr	ench	.998	unpro	ctored	1.000		PAP	.998

Italian

.999

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. cap\_fs 0.0 0.9 -4.3 1.2 15783 Share of cases with imputed missing values: 0.7% (Equivalence of scores from robust MLMV: CD = .997)

#### Scale: Positive attitude towards school

Full AES sample

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	0	0	
	Baseline vs. saturated	22788	3	.000
2)	Root mean squared error (	RMSEA)		.000
	90% Confidence interval:	lower bound		.000
	90% Confidence interval:	upper bound		.000
	Probability RMSEA <= 0.05			1.000
3)	Akaike's Information Crite	rion (AIC)		205667
	Bayesian Information Crite	erion (BIC)		205739
4)	Baseline comparison			
	Comparative Fit Index (CFI)			1.000
	Tucker–Lewis Index (TLI)			1.000

#### **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.809
(Cronbach's alpha = .784)	
McDonald's Omega	.813

#### Test of (one-)dimensionality (parallel analysis)

Criterion: retain factors with adj. eigenvalue > o Adjusted eigenvalue

	,	_
Factor 1		1.61
Factor 2		10
Factor 3		17

5) Size of residuals

Indicators

posatt1

posatt2

posatt3

Stand. root mean squared residual (SRMR) .000 Coefficient of determination (CD) .835

(SE)

.004

.004

.004

[95% Conf. interval]

0.73

0.85

0.69

0.75

0.87

0.71

Coef.

0.74

0.86

0.70

#### Standardized factor loadings

# Item descriptives

		Std.			Valid
Indicators	Mean	dev.	Min.	Max.	Obs.
posatt1	3.8	1.3	1	6	22295
posatt2	4.1	1.3	1	6	22288
posatt3	4.6	1.3	1	6	22287

#### Scale: Positive attitude towards school (continued)

Full AES sample

#### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2 998	df 18	p > chi2 .000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	17	4	.002
Strong invariance (plus equal intercepts)	172	4	.000
Strict invariance (plus equal error variances)	217	4	.000

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.999
French vs. Italian language version	1.000
Italian vs. German language version	.999

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	.999
Language: Italian	1.000

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. posatt\_fs 0.0 0.9 -2.5 1.4 22299 Share of cases with imputed missing values: 0.1% (Equivalence of scores from robust MLMV: CD = .999)

# Scale: Enjoyment in school

Full AES sample

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	0	0	
	Baseline vs. saturated	24844	3	.000
2)	Root mean squared error (	(RMSEA)		.000
	90% Confidence interval:	lower bound		.000
	90% Confidence interval:	upper bound		.000
	Probability RMSEA <= 0.05			1.000
3)	Akaike's Information Crite	rion (AIC)		216963
	Bayesian Information Crit	erion (BIC)		217035
4)	Baseline comparison			
	Comparative Fit Index (CFI)			1.000

#### **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.821
(Cronbach's alpha = .796)	
McDonald's Omega	.825

#### Test of (one-)dimensionality (parallel analysis)

Criterion: retain factors with adj. eigenvalue > o Adjusted eigenvalue

	-	•
Factor 1		1.67
Factor 2		08
Factor 3		16

Tucker-Lewis Index (TLI) 1.000

#### 5) Size of residuals

Stand. root mean squared residual (SRMR) .000 Coefficient of determination (CD) .856

### Standardized factor loadings

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
enjoyschool1	0.76	.004	0.75	0.77	enjoyschool1	3.2	1.5	1	6	22254
enjoyschool2	0.89	.004	0.88	0.89	enjoyschool2	3.5	1.4	1	6	22252
enjoyschool3	0.69	.004	0.68	0.70	enjoyschool3	3.9	1.4	1	6	22257

Item descriptives

# Scale: Enjoyment in school (continued)

Full AES sample

#### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2
	506	18	.000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	33	4	.000
Strong invariance (plus equal intercepts)	258	4	.000
Strict invariance (plus equal error variances)	34	4	.000

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.999
French vs. Italian language version	.992
Italian vs. German language version	.996

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	.998
Language: Italian	.994

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. enjoyschool\_fs 0.0 1.1 -2.1 2.1 22267
Share of cases with imputed missing values: 0.1% (Equivalence of scores from robust MLMV: CD = .999)

#### Scale: Physical complaints in school

Full AES sample

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2	Ordinal
	Model vs. saturated	29	2	.000	(Cronba
	Baseline vs. saturated	36796	6	.000	McDona

Ordinal Cronbach's Alpha	.847
(Cronbach's alpha = .772)	
McDonald's Omega	.849

**Reliability and Dimensionality** 

#### 2) Root mean squared error (RMSEA) .025 90% Confidence interval: lower bound .017 90% Confidence interval: upper bound .033

#### Test of (one-)dimensionality (parallel analysis) Criterion: retain factors with adj. eigenvalue > o

Adjusted eigenvalue

-.12

Factor 1 2.22 Factor 2 -.09 Factor 3 -.10

Factor 4

Probability RMSEA <= 0.05	1.000
Akaike's Information Criterion (AIC) Bayesian Information Criterion (BIC)	272002 272098
	Akaike's Information Criterion (AIC)

# 4) Baseline comparison

Comparative Fit Index (CFI)	.999
Tucker–Lewis Index (TLI)	.998

# 5) Size of residuals

Indicators

physpain1

physpain2

physpain3

physpain4

Stand. root mean squared residual (SRMR)	.005
Coefficient of determination (CD)	.857

Standardized factor loadings

(SE)

.003

.003

.003

.004

[95% Conf. interval]

0.79

0.79

0.82

0.68

0.77

0.78

0.81

0.66

Coef.

0.78

0.79

0.82

0.67

# Item descriptives

		Std.			Valid
Indicators	Mean	dev.	Min.	Max.	Obs.
physpain1	1.7	1.3	1	6	22260
physpain2	1.7	1.4	1	6	22249
physpain3	1.7	1.3	1	6	22222
physpain4	2.3	1.6	1	6	22245

# Scale: Physical complaints in school (continued)

Full AES sample

#### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2 1179	df 28	p > chi2 .000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	76	6	.000
Strong invariance (plus equal intercepts)	188	6	.000
Strict invariance (plus equal error variances)	542	6	.000

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.999
French vs. Italian language version	.997
Italian vs. German language version	.996

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	.999
Language: Italian	.988

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. physpain\_fs 0.0 0.8 -.6 3.5 22271 Share of cases with imputed missing values: 0.3% (Equivalence of scores from robust MLMV: CD = .995)

			,				_			
Sca	Δ.	\/\	orr	IAC	al	20	ı ıtı	CC	hα	10

Full AES sample

#### **Model and Fit Statistics**

1)

Likelihood-ratio tests	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	21848	3	.000

2) Root mean squared error (RMSEA) .ooo
90% Confidence interval: lower bound .ooo
90% Confidence interval: upper bound .ooo
Probability RMSEA <= 0.05 1.000

3) Akaike's Information Criterion (AIC) 240309 Bayesian Information Criterion (BIC) 240381

4) Baseline comparison

Comparative Fit Index (CFI) 1.000
Tucker–Lewis Index (TLI) 1.000

5) Size of residuals

Stand. root mean squared residual (SRMR) .000
Coefficient of determination (CD) .836

#### **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.795
(Cronbach's alpha = .753)	
McDonald's Omega	.802

Test of (one-)dimensionality (parallel analysis)

Criterion: retain factors with adj. eigenvalue > o

	Adjusted eigenvalue
Factor 1	1.57
Factor 2	09
Factor 3	18

Standardized factor loadings

ltam	descri	ptives
iteiii	uescii	puves

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf.	interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
trouschool1	0.78	.004	0.78	0.79	trouschool1	2.9	1.6	1	6	22260
trouschool2	0.86	.004	0.85	0.87	trouschool2	3.2	1.7	1	6	22263
trouschoola	0.62	.005	0.61	0.63	trouschoola	3.4	1.9	1	6	22263

#### Scale: Worries about school (continued)

Full AES sample

#### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2 1522	df 18	p > chi2 .000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	51	4	.000
Strong invariance (plus equal intercepts)	889	4	.000
Strict invariance (plus equal error variances)	295	4	.000

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.998
French vs. Italian language version	.999
Italian vs. German language version	.999

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	.999
Language: Italian	.996

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. trouschool\_fs 0.0 1.2 -1.9 2.5 22270 Share of cases with imputed missing values: 0.1% (Equivalence of scores from robust MLMV: CD = .997)

# Scale: Social problems in school

Full AES sample

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	0	0	
	Baseline vs. saturated	39687	3	.000
2)	Root mean squared error (	(RMSEA)		.000
	90% Confidence interval:	lower bound		.000
	90% Confidence interval:	upper bound		.000
	Probability RMSEA <= 0.05			1.000
3)	Akaike's Information Crite	• •		164458
	Bayesian Information Crit	erion (BIC)		164530
4)	Baseline comparison			
	Comparative Fit Index (CFI)			1.000
	Tucker–Lewis Index (TLI)			1.000

#### **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.886
(Cronbach's alpha = .817)	
McDonald's Omega	.889

#### Test of (one-)dimensionality (parallel analysis)

Criterion: retain factors with adj. eigenvalue > o
Adjusted eigenvalue

Factor 1	2.07
Factor 2	05
Factor 3	12

### Standardized factor loadings

5) Size of residuals

Indicators	Coef.	(SE)	[95% Conf	. interval]
socprob1	0.95	.002	0.95	0.95
socprob2	0.84	.003	0.84	0.85
socprob3	0.76	.003	0.75	0.77

Stand. root mean squared residual (SRMR)

Coefficient of determination (CD)

_	_	
ltem	descr	riptives

.000

.929

		Std.			Valid
Indicators	Mean	dev.	Min.	Max.	Obs.
socprob1	1.5	1.0	1	6	22244
socprob2	1.7	1.2	1	6	22259
socprob3	1.5	1.1	1	6	22239

# Scale: Social problems in school (continued)

Full AES sample

#### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2 466	df 18	p > chi2 .000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	16	4	.003
Strong invariance (plus equal intercepts)	129	4	.000
Strict invariance (plus equal error variances)	157	4	.000

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	1.000
French vs. Italian language version	.999
Italian vs. German language version	.999

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	.999
Language: Italian	1.000

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. socprob\_fs 0.0 0.9 -0.5 4.3 22265 Share of cases with imputed missing values: 0.2% (Equivalence of scores from robust MLMV: CD = .991)

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Sca	Δ.	Sch	$\sim$	r۵	luctan	0
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Full AES sample

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MADA	l and	Fit St	atistics
WOUL	aliu	ロルコン	ausucs

1)	<b>Likelihood-ratio tests</b> Model vs. saturated Baseline vs. saturated	chi2 0 14239	df o 3	p > chi2
2)	Root mean squared error 90% Confidence interval: 90% Confidence interval: Probability RMSEA <= 0.05	lower bound upper bound		.000 .000 .000 1.000
3)	Akaike's Information Crite Bayesian Information Crit	, ,		245338 245410

# Reliability and Dimensionality

Ordinal Cronbach's Alpha	.702
(Cronbach's alpha = .661)	
McDonald's Omega	.727

#### Test of (one-)dimensionality (parallel analysis)

Criterion: retain factors with adj. eigenvalue > o

Adjusted eigenvalue

Factor 1	1.23
Factor 2	05
Factor 3	22

#### 4) Baseline comparison

Comparative Fit Index (CFI)	1.000
Tucker–Lewis Index (TLI)	1.000

### 5) Size of residuals

Stand. root mean squared residual (SRMR)	.000
Coefficient of determination (CD)	.835

### Standardized factor loadings

#### Item descriptives

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
schoolav1	0.89	.007	0.88	0.91	schoolavı	3.1	1.8	1	6	22245
schoolav2	0.67	.007	0.66	0.69	schoolav2	3.7	1.9	1	6	22248
schoolav3	0.46	.006	0.45	0.47	schoolav3	2.2	1.5	1	6	22235

#### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2 1451	df 9	p > chi2 .000		
Tests of measurement invariance	chi2	df	p > chi2		
Metric invariance (equal factor loadings)	99	2	.000		
Strong invariance (plus equal intercepts)	981	2	.000		
Strict invariance (plus equal error variances)	49	2	.000		
Configural factor similarity					
Tucker's Congruence Coefficient	TCC				
German vs. French language version	.999				
French vs. Italian language version					
Italian vs. German language version					
Factor score equivalence: group specific vs. inva	riant mode	ls			
Coefficient of determination	CD				
Language: German	.994				
Language: French/ Italian	.981				
* <b>Note:</b> Due to sparse tables for the italian version converge and were reestimated with colla		•		d to	

# Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. schoolav\_fs 0.0 1.4 -2.0 2.6 22266 Share of cases with imputed missing values: 0.2% (Equivalence of scores from robust MLMV: CD = .999)

#### Scale: Intrinsic achievement motivation

Full AES sample

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	0	0	
	Baseline vs. saturated	12995	3	.000
۵١	Do at many agreed array (	DMCEAN		

#### 2) Root mean squared error (RMSEA) .000 90% Confidence interval: lower bound .000 90% Confidence interval: upper bound .000 Probability RMSEA <= 0.05 1.000

#### 3) Akaike's Information Criterion (AIC) 152039 **Bayesian Information Criterion (BIC)** 152111

### 4) Baseline comparison

Comparative Fit Index (CFI) 1.000 Tucker-Lewis Index (TLI) 1.000

#### 5) Size of residuals

Stand. root mean squared residual (SRMR)	.000
Coefficient of determination (CD)	.795

#### **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.703
(Cronbach's alpha = .652)	
McDonald's Omega	.718

#### Test of (one-)dimensionality (parallel analysis)

Criterion: retain factors with adj. eigenvalue > o Adjusted eigenvalue

	Aujusted eigenvalue	
Factor 1	1.19	
Factor 2	08	
Factor 3	22	

#### Standardized factor loadings

Standardized factor loadings					Item descriptives					
	_						Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
achmot2	0.54	.006	0.52	0.55	achmot2	3.0	0.8	1	4	22249
achmot4	0.62	.006	0.60	0.63	achmot4	2.8	0.8	1	4	22242
achmot6	o 86	007	o 85	0.87	achmot6	2.6	0.0	1	/.	22220

#### Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut <sub>1</sub>	Cut <sub>2</sub>	Cut <sub>3</sub>
achmot2	1.16	-3.58	-1.45	1.12
achmot4	1.47	-3.30	-0.89	2.11
achmot6	2.88	-4.12	-0.77	3.70

#### Scale: Intrinsic achievement motivation (continued)

Full AES sample

#### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2
	1286	18	.000
<u></u>	1.1	16	1.1
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	14	4	.007
Strong invariance (plus equal intercepts)	956	4	.000
Strict invariance (plus equal error variances)	141	4	.000

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.999
French vs. Italian language version	.993
Italian vs. German language version	.996

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	.999
Language: French	.999
Language: Italian	.990

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. achmoti\_fs 0.0 0.9 -2.2 1.8 22262

Share of cases with imputed missing values: 0.2% (Equivalence of scores from robust MLMV: CD = .994) (Equivalence of scores from two-step approach: CD = .982)

#### Scale: Extrinsic achievement motivation

Full AES sample

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	0	0	
	Baseline vs. saturated	12774	3	.000

2) Root mean squared error (RMSEA) .000
90% Confidence interval: lower bound .000
90% Confidence interval: upper bound .000
Probability RMSEA <= 0.05 1.000

3) Akaike's Information Criterion (AIC) 148710 Bayesian Information Criterion (BIC) 148782

4) Baseline comparison

Comparative Fit Index (CFI) 1.000
Tucker–Lewis Index (TLI) 1.000

5) Size of residuals

Stand. root mean squared residual (SRMR) .000
Coefficient of determination (CD) .792

#### **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.648
(Cronbach's alpha = .589)	
McDonald's Omega	.690

Test of (one-)dimensionality (parallel analysis)

Criterion: retain factors with adj. eigenvalue > o
Adjusted eigenvalue

	r lajostea elgelli.
Factor 1	1.14
Factor 2	04
Factor 3	22

#### Standardized factor loadings

Indicators	Coef.	(SE)	[95% Conf	. interval]
achmot1	0.33	.007	0.32	0.34
achmot3	0.73	.009	0.72	0.75
achmot5	0.85	.009	0.83	o.86

# Item descriptives

		Std.			Valid
Indicators	Mean	dev.	Min.	Max.	Obs.
achmot1	3.2	0.7	1	4	22263
achmot3	1.8	0.8	1	4	22239
achmot5	1.9	0.9	1	4	22235

#### Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut <sub>1</sub>	Cut <sub>2</sub>	Cut <sub>3</sub>
achmot1	0.58	-3.66	-2.13	0.51
achmot3	2.18	-0.50	2.38	5.22
achmot5	2.49	-0.62	2.16	5.11

# Scale: Extrinsic achievement motivation (continued)

Full AES sample

#### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2 1767	df 18	p > chi2 .000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	36	4	.000
Strong invariance (plus equal intercepts)	954	4	.000
Strict invariance (plus equal error variances)	211	4	.000

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.982
French vs. Italian language version	.995
Italian vs. German language version	.996

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	.979
Language: French	.961
Language: Italian	.993

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. achmote\_fs 0.0 0.8 -1.3 2.3 22266

Share of cases with imputed missing values: 0.2% (Equivalence of scores from robust MLMV: CD = .990) (Equivalence of scores from two-step approach: CD = .981)

# Scale: Instrumental learning motivation (PISA2000)

Full AES sample

#### **Model and Fit Statistics**

Reliability and	l Dimensionali	ity
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1)	Likelihood-ratio tests	chi2	df	p > chi2	Ordinal Cronbach's Alp	oha .848
	Model vs. saturated	0	0	•	(Cronbach's alpha = .79)	6)
	Baseline vs. saturated	28969	3	.000	McDonald's Omega	.850
	Buseline vs. satoratea	20909	3	.000	mezonala s omega	.050
2)	Root mean squared error (	RMSEA)		.000	Test of (one-)dimension	nality (parallel analysis)
	90% Confidence interval:	lower bound		.000	Criterion: retain factors	with adj. eigenvalue > o
	90% Confidence interval:	upper bound		.000	Adjuste	d eigenvalue
	Probability RMSEA <= 0.05			1.000	Factor 1	1.81
					Factor 2	10
3)	Akaike's Information Crite	rion (AIC)		144091	Factor 3	14
	Bayesian Information Crite	erion (BIC)		144163		
4)	Baseline comparison					
	Comparative Fit Index (CFI)			1.000		
	Tucker–Lewis Index (TLI)			1.000		
5)	Size of residuals					
٥.	Stand. root mean squared re	esidual (SRMF	?)	.000		
	Coefficient of determination			.865		

# Standardized factor loadings

Item descriptives	5
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							Std.			Valid
Indicators *	Coef.	(SE)	[95% Conf. i	interval]	Indicators *	Mean	dev.	Min.	Max.	Obs.
insmot1	0.75	0.00	0.74	0.76	insmot1	2.8	0.9	1	4	22246
insmot2	0.79	0.00	0.78	0.80	insmot2	2.9	0.9	1	4	22220
insmot3	0.88	0.00	0.88	0.89	insmot3	3.1	0.9	1	4	22220

<sup>\*</sup> Note: Replication of 'Insmot'-Scale from TREE1 / PISA2000

# Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut1	Cut <sub>2</sub>	Cut <sub>3</sub>
insmot1	2.05	-3.82	-0.83	2.13
insmot2	2.35	-3.90	-1.28	1.70
insmoto	2 48	-6.22	-2.28	0.80

# Scale: Instrumental learning motivation (PISA2000) (continued)

Full AES sample

#### Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b> chi2 df p >	
347 18 .	000
Tests of measurement invariance chi2 df p >	> chi2
Metric invariance (equal factor loadings) 29 4 .	000
Strong invariance (plus equal intercepts) 136 4 .	000
Strict invariance (plus equal error variances) 55 4.	000

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	1.000
French vs. Italian language version	.997
Italian vs. German language version	.994

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	1.000
Language: Italian	.982

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. insmot\_fs 0.0 0.9 -2.2 1.4 22265

Share of cases with imputed missing values: 0.4% (Equivalence of scores from robust MLMV: CD = .996) (Equivalence of scores from two-step approach: CD = .978)

# Scale: Interest in reading

Full AES sample

#### **Model and Fit Statistics**

Reliability and	l Dimens	ionality
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1)	Likelihood-ratio tests	chi2	df	p > chi2	Ordinal Cronbach's Alpha	.906
	Model vs. saturated	0	0		(Cronbach's alpha = .864)	
	Baseline vs. saturated	44643	3	.000	McDonald's Omega	.907
2)	Root mean squared error (l	RMSEA)		.000	Test of (one-)dimensional	ity (parallel analysis)
	90% Confidence interval:	lower bound		.000	Criterion: retain factors wit	h adj. eigenvalue > o
	90% Confidence interval:	upper bound		.000	Adjusted e	igenvalue
	Probability RMSEA <= 0.05			1.000	Factor 1	2.19
					Factor 2	07
3)	Akaike's Information Criter	rion (AIC)		153979	Factor 3	11
	Bayesian Information Crite	rion (BIC)		154051		
4)	Baseline comparison					
	Comparative Fit Index (CFI)			1.000		
	Tucker–Lewis Index (TLI)			1.000		
5)	Size of residuals					
	Stand. root mean squared re	esidual (SRMI	₹)	.000		
	Coefficient of determination	n (CD)		.924		

Standardized factor	Item descriptives									
							Std.			Valid
Indicators *	Coef.	(SE)	[95% Conf. i	nterval]	Indicators *	Mean	dev.	Min.	Max.	Obs.
intrea1	o.86	.002	0.85	o.86	intrea1	2.2	1.0	1	4	22180
intrea2	0.94	.002	0.93	0.94	intrea2	2.1	1.1	1	4	22178
intrea3	0.83	.003	0.82	0.83	intrea3	2.3	1.1	1	4	22165

<sup>\*</sup> Note: Replication of 'Intrea'-Scale from TREE1 / PISA2000

# Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut <sub>1</sub>	Cut <sub>2</sub>	Cut <sub>3</sub>
intrea1	3.03	-1.81	0.96	3.55
intrea2	5.35	-1.65	2.08	5.65
intrea3	2.63	-1.67	0.17	2.61

#### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2 732	df 18	p > chi2 .000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	94	4	.000
Strong invariance (plus equal intercepts)	560	4	.000
Strict invariance (plus equal error variances)	7	4	.155

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.999
French vs. Italian language version	1.000
Italian vs. German language version	.999

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	.999
Language: French	.998
Language: Italian	.998

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. intrea\_fs 0.0 0.9 -1.3 1.7 22200

Share of cases with imputed missing values: 0.3% (Equivalence of scores from robust MLMV: CD = .997) (Equivalence of scores from two-step approach: CD = .973)

Scale: ICT interest

Maths sample-split

#### **Model and Fit Statistics**

# **Reliability and Dimensionality**

1)	Likelihood-ratio tests	chi2	df	p > chi2	Ordinal Cronbach'	•	.855
	Model vs. saturated	0	0		(Cronbach's alpha =	= .797)	
	Baseline vs. saturated	15929	3	.000	McDonald's Omeg	ja –	.860
2)	Root mean squared error (	RMSEA)		.000	Test of (one-)dime	ensionality (paralle	analysis)
	90% Confidence interval:	lower bound		.000	Criterion: retain fac	ctors with adj. eigen	value > o
	90% Confidence interval:	upper bound		.000	Adj	justed eigenvalue	
	Probability RMSEA <= 0.05			1.000	Factor 1	1.88	
	,				Factor 2	09	
3)	Akaike's Information Crite	rion (AIC)		69317	Factor 3	13	
	Bayesian Information Crite	erion (BIC)		69383			
4)	Baseline comparison						
	Comparative Fit Index (CFI)			1.000			
	Tucker–Lewis Index (TLI)			1.000			
5)	Size of residuals						
٥,	Stand. root mean squared r	esidual (SRM	R)	.000			

.884

# Standardized factor loadings

Coefficient of determination (CD)

# Item descriptives

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf.	interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
ictmot2	0.69	.006	0.68	0.71	ictmot2	3.2	0.7	1	4	11068
ictmot3	0.88	.004	0.87	0.89	ictmot3	2.4	1.0	1	4	11065
ictmot4	0.87	.004	0.86	o.88	ictmot4	2.8	0.9	1	4	11060

# Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut1	Cut <sub>2</sub>	Cut <sub>3</sub>
ictmot2	1.77	-4.71	-2.62	0.94
ictmot3	3.41	-3.34	0.41	3.52
ictmot4	3.42	-4.79	-1.57	2.83

#### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2 408	df 18	p > chi2 .000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	69	4	.000
Strong invariance (plus equal intercepts)	95	4	.000
Strict invariance (plus equal error variances)	34	4	.000

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.995
French vs. Italian language version	.997
Italian vs. German language version	.995

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	.994
Language: Italian	.892

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. ictintr\_fs 0.0 0.9 -2.1 1.6 11071 Share of cases with imputed missing values: 0.2% (Equivalence of scores from robust MLMV: CD = .999) (Equivalence of scores from two-step approach: CD = .992)

# Scale: Dispositional interest

Maths sample-split

#### **Model and Fit Statistics**

# Reliability and Dimensionality

1)	Likelihood-ratio tests	chi2	df	p > chi2	Ordinal Cronbach	's Alpha	.875
	Model vs. saturated	1805	9	.000	(Cronbach's alpha	= .836)	
	Baseline vs. saturated	31076	15	.000	McDonald's Ome	ga	.876
2)	Root mean squared error (	RMSEA)		.135	Test of (one-)dime	ensionality (parall	el analysis)
	90% Confidence interval:	lower bound		.130	Criterion: retain fa	ctors with adj. eige	nvalue > o
	90% Confidence interval:	upper bound		.140	Ad	ljusted eigenvalue	
	Probability RMSEA <= 0.05			.000	Factor 1	3.19	
					Factor 2	.14	
3)	Akaike's Information Criter	rion (AIC)		137195	Factor 3	01	
	Bayesian Information Crite	rion (BIC)		137326	Factor 4	05	
					Factor 5	13	
4)	Baseline comparison				Factor 6	14	
	Comparative Fit Index (CFI)			.942			
	Tucker–Lewis Index (TLI)			.904			
5)	Size of residuals						
	Stand. root mean squared re	esidual (SRMI	۲)	.041			
	Coefficient of determination	n (CD)		.888			

# Standardized factor loadings

# Item descriptives

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
intsubj1	0.84	.004	0.83	0.85	intsubj1	2.5	0.9	1	4	10889
intsubj2	0.65	.006	0.64	0.66	intsubj2	3.2	0.7	1	4	10922
intsubj3	0.75	.005	0.74	0.76	intsubj3	2.9	0.8	1	4	10845
intsubj4	0.66	.006	0.65	0.67	intsubj4	2.6	0.9	1	4	10842
intsubj5	0.69	.006	0.68	0.71	intsubj5	2.8	0.8	1	4	10905
intsubj6	0.80	.004	0.80	0.81	intsubj6	2.4	1.0	1	4	10853

# Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut1	Cut <sub>2</sub>	Cut <sub>3</sub>
intsubj1	2.92	-3.37	-0.29	3.76
intsubj2	1.58	-4.54	-2.81	0.59
intsubj3	2.12	-4.06	-1.70	1.90
intsubj4	1.63	-2.34	-0.39	2.29
intsubj5	1.80	-3.88	-0.89	2.43
intsubj6	2.53	-2.10	0.31	3.26

# Scale: Dispositional interest (continued)

Maths sample-split

## Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2 885	df 54	p > chi2 .000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	91	10	.000
Strong invariance (plus equal intercepts)	332	10	.000
Strict invariance (plus equal error variances)	77	10	.000

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.996
French vs. Italian language version	.995
Italian vs. German language version	.998

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	.999
Language: Italian	.999

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. intsubj\_fs 0.0 0.9 -2.6 2.1 10949

Share of cases with imputed missing values: 1.6% (Equivalence of scores from robust MLMV: CD = .999) (Equivalence of scores from two-step approach: CD = .988)

# Scale: Identified motivation (mathematics)

Maths sample-split

#### **Model and Fit Statistics**

1)

)	Likelihood-ratio tests	chi2	df	p > chi2	Ordinal Cronbach's Alpha	.946
	Model vs. saturated	45	2	.000	(Cronbach's alpha = .918)	
	Baseline vs. saturated	43936	6	.000	McDonald's Omega	.947

2)	Root mean squared error (RMSEA)	.044	Tes
	90% Confidence interval: lower bound	.034	Crit
	90% Confidence interval: upper bound	.056	
	Probability RMSEA <= 0.05	.777	Fac
			Fac
3)	Akaike's Information Criterion (AIC)	72033	Fac
	Bayesian Information Criterion (BIC)	72121	Fac

# est of (one-)dimensionality (parallel analysis) riterion: retain factors with adj. eigenvalue > o Adjusted eigenvalue

Factor 1	3.20
Factor 2	04
Factor 3	05
Factor 4	04

**Reliability and Dimensionality** 

# 4) Baseline comparison

Comparative Fit Index (CFI)	.999
Tucker–Lewis Index (TLI)	.997

#### 5) Size of residuals

Stand. root mean squared residual (SRMR)	.004
Coefficient of determination (CD)	.955

# Standardized factor loadings

#### Item descriptives

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
instrumot1	0.95	.001	0.94	0.95	instrumot1	2.9	0.9	1	4	11018
instrumot2	0.93	.002	0.93	0.94	instrumot2	2.9	0.9	1	4	11020
instrumot3	0.89	.002	0.88	0.89	instrumot3	2.8	0.9	1	4	11030
instrumot4	0.85	.003	0.84	0.85	instrumot4	2.9	0.9	1	4	11013

#### Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut1	Cut <sub>2</sub>	Cut <sub>3</sub>
instrumot1	4.16	-7.00	-2.77	2.59
instrumot2	3.66	-5.86	-2.07	1.94
instrumot3	2.86	-5.38	-1.92	2.16
instrumot4	2.49	-5.04	-2.19	1.86

# Scale: Identified motivation (mathematics) (continued)

Maths sample-split

#### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2		
	387	14	.000		
Tests of measurement invariance	chi2	df	p > chi2		
Metric invariance (equal factor loadings)	111	3	.000		
Strong invariance (plus equal intercepts)	75	3	.000		
Strict invariance (plus equal error variances)	135	3	.000		
Configural factor similarity					
Tucker's Congruence Coefficient	TCC				
German vs. French language version	1.000				
French vs. Italian language version					
Italian vs. German language version					
Factor score equivalence: group specific vs. inv	ariant mode	als			
Coefficient of determination	CD	5			
Language: German	1.000				
Language: French/ Italian	1.000				
* Note: Due to sparse tables for the italian version		•		d to	
converge and were reestimated with col	iapsed italia	n and frei	nch versions.		

## Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. instrumot\_fs -0.1 1.0 -2.4 1.5 11033

Share of cases with imputed missing values: 0.3% (Equivalence of scores from robust MLMV: CD = .999) (Equivalence of scores from two-step approach: CD = .985)

# Scale: External motivation regulation

Maths sample-split

# **Model and Fit Statistics**

#### **Reliability and Dimensionality**

_ikelihood-ratio tests	chi2	df	p > chi2	Ordinal Cronbach's A	Alpha .820
Model vs. saturated	687	2	.000	(Cronbach's alpha = .7	764)
Baseline vs. saturated	16452	6	.000	McDonald's Omega	.826
Root mean squared error (F	RMSEA)		.177	Test of (one-)dimens	sionality (parallel analysis)
90% Confidence interval:	ower bound		.166	Criterion: retain facto	ors with adj. eigenvalue > o
90% Confidence interval: ι	upper bound		.188	Adjus	sted eigenvalue
Probability RMSEA <= 0.05			.000	Factor 1	2.06
				Factor 2	.06
Akaike's Information Criter	ion (AIC)		100910	Factor 3	15
Bayesian Information Crite	rion (BIC)		100998	Factor 4	15
Baseline comparison					
Comparative Fit Index (CFI)			.958		
Tucker–Lewis Index (TLI)			.875		
Size of residuals					
	esidual (SRM	R)	.038		
·		,	_		
	Baseline vs. saturated  Root mean squared error (I 20% Confidence interval: 1 20% Confidence interval:	Baseline vs. saturated 16452  Root mean squared error (RMSEA)  30% Confidence interval: lower bound  30% Confidence interval: upper bound  30% Con	Baseline vs. saturated 16452 6  Root mean squared error (RMSEA)  90% Confidence interval: lower bound  90% Confidence interval: upper bound  Probability RMSEA <= 0.05  Akaike's Information Criterion (AIC)  Bayesian Information Criterion (BIC)  Comparative Fit Index (CFI)  Fucker-Lewis Index (TLI)  Size of residuals  Stand. root mean squared residual (SRMR)	Model vs. saturated 687 2 .000 Baseline vs. saturated 16452 6 .000  Root mean squared error (RMSEA) .177 90% Confidence interval: lower bound .166 90% Confidence interval: upper bound .188 Probability RMSEA <= 0.05 .000  Akaike's Information Criterion (AIC) 100910 Bayesian Information Criterion (BIC) 100998  Baseline comparison Comparative Fit Index (CFI) .958 Fucker-Lewis Index (TLI) .875  Size of residuals Stand. root mean squared residual (SRMR) .038	Model vs. saturated 687 2 .000 (Cronbach's alpha = .7) Baseline vs. saturated 16452 6 .000 McDonald's Omega  Root mean squared error (RMSEA) .177 Test of (one-)dimense of Confidence interval: lower bound .166 Criterion: retain factor of Confidence interval: upper bound .188 Adjust Probability RMSEA <= 0.05 .000 Factor 1 Factor 2  Akaike's Information Criterion (AIC) 100910 Factor 3  Bayesian Information Criterion (BIC) 100998 Factor 4  Baseline comparison Comparative Fit Index (CFI) .958  Fucker—Lewis Index (TLI) .875  Size of residuals  Stand. root mean squared residual (SRMR) .038

# Standardized factor loadings

# Item descriptives

	_						Std.			Valid
Indicators *	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
extreg2	0.76	.005	0.75	0.77	extreg2	1.9	0.9	1	4	10901
extreg3	0.81	.005	0.80	0.82	extreg3	2.0	0.9	1	4	10830
extreg4	0.58	.008	0.56	0.59	extreg4	2.4	0.9	1	4	10841
extreg5	0.78	.005	0.77	0.79	extreg5	1.8	0.9	1	4	10827

<sup>\*</sup> Note: Items Extreg1 and Extreg6 Excluded to Improve Scale Quality

# Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut <sub>1</sub>	Cut <sub>2</sub>	Cut <sub>3</sub>
extreg2	2.11	-0.76	1.62	4.25
extreg3	2.55	-1.03	1.52	4.56
extreg4	1.28	-1.75	0.01	2.39
extreg5	2.34	-0.17	2.28	4.99

# Scale: External motivation regulation (continued)

Maths sample-split

#### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2
	222	28	.000
		16	
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	46	6	.000
Strong invariance (plus equal intercepts)	113	6	.000
Strict invariance (plus equal error variances)	35	6	.000

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.998
French vs. Italian language version	.990
Italian vs. German language version	.996

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	.998
Language: Italian	.997

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. extreg\_fs 0.0 0.9 -1.4 2.5 10930 Share of cases with imputed missing values: 1.5% (Equivalence of scores from robust MLMV: CD = .999) (Equivalence of scores from two-step approach: CD = .977)

# Scale: Classroom participation

Maths sample-split

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	584	5	.000
	Baseline vs. saturated	28718	10	.000
2)	Root mean squared error (F	RMSEA)		.103

2)	Root mean squared error	(RMSEA)	.103
	90% Confidence interval:	lower bound	.096
	90% Confidence interval:	upper bound	.110
	Probability RMSEA <= 0.05		.000

3)	Akaike's Information Criterion (AIC)	97128
	Bayesian Information Criterion (BIC)	97238

## 4) Baseline comparison Comparative Fit Index (CFI)

Comparative Fit Index (CFI) .980 Tucker–Lewis Index (TLI) .960

#### 5) Size of residuals

Stand. root mean squared residual (SRMR) .024
Coefficient of determination (CD) .890

# **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.888
(Cronbach's alpha = .848)	
McDonald's Omega	.888

#### Test of (one-)dimensionality (parallel analysis)

Criterion: retain factors with adj. eigenvalue > o

	Adjusted eigenvalue	
Factor 1	2.95	
Factor 2	.02	
Factor 3	05	
Factor 4	11	
Factor 5	11	

#### Standardized factor loadings

Indicators	Coef.	(SE)	[95% Conf	. interval]
engage1	0.76	.005	0.75	0.77
engage2	0.83	.004	0.82	0.84
engage3	0.75	.005	0.74	0.76
engage4	0.80	.004	0.79	0.81
engage5	0.77	.005	0.76	0.78

### Item descriptives

		Std.			Valid
Indicators	Mean	dev.	Min.	Max.	Obs.
engage1	2.9	0.8	1	4	10897
engage2	2.9	0.7	1	4	10852
engage3	3.0	0.7	1	4	10907
engage4	3.0	0.8	1	4	10898
engage5	2.8	0.8	1	4	10829

#### Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut1	Cut <sub>2</sub>	Cut <sub>3</sub>
engage1	2.22	-4.53	-1.82	2.06
engage2	2.82	-5.44	-2.01	3.03
engage3	2.14	-4.97	-2.11	1.89
engage4	2.51	-5.30	-2.40	2.21
engage5	2.28	-4.28	-1.30	3.10

# Scale: Classroom participation (continued)

Maths sample-split

#### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2
	938	40	.000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	51	8	.000
Strong invariance (plus equal intercepts)	31	8	.000
Strict invariance (plus equal error variances)	149	8	.000

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.997
French vs. Italian language version	.997
Italian vs. German language version	.999

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	.999
Language: Italian	.999

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. engage\_fs 0.0 0.9 -2.7 1.9 10936
Share of cases with imputed missing values: 1.5% (Equivalence of scores from robust MLMV: CD = .996) (Equivalence of scores from two-step approach: CD = .984)

# Scale: Performance-approach goals (SELLMO)

Maths sample-split

.834

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2	Ordinal Cronbac	ch's Alpha
	Model vs. saturated	620	2	.000	(Cronbach's alph	na = .804)
	Baseline vs. saturated	17637	6	.000	McDonald's Om	iega
2)	Root mean squared error (RM	ISEA)		.171	Test of (one-)di	mensionalit
	90% Confidence interval: lov	wer bound		.159	Criterion: retain	factors with
	90% Confidence interval: up	per bound	i	.182	,	Adjusted eig
	Probability RMSEA <= 0.05			.000	Factor 1	2.
	,				Factor 2	
3)	Akaike's Information Criterio	n (AIC)		117025	Factor 3	
-	Bayesian Information Criterio	on (BIC)		117112	Factor 4	
4)	Baseline comparison					
17	Comparative Fit Index (CFI)			.965		
	Tucker–Lewis Index (TLI)			.895		
5)	Size of residuals					
١,	Stand root mean squared resi	dual (SRM	ID)	040		

# alpha = .804)

s Omega .837

#### e-)dimensionality (parallel analysis)

etain factors with adj. eigenvalue > o Adjusted eigenvalue

Factor 1	2.16
Factor 2	.05
Factor 3	15
Factor 4	13

**Reliability and Dimensionality** 

Stand. root mean squared residual (SRMR) .040 Coefficient of determination (CD) .865

# Standardized factor loadings

# Item descriptives

		-								
							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
approxgoals1	0.74	.006	0.73	0.75	approxgoals1	2.8	1.2	1	5	10608
approxgoals2	0.84	.004	0.83	0.84	approxgoals2	2.5	1.2	1	5	10478
approxgoals3	0.57	.008	0.55	0.58	approxgoals3	3.3	1.1	1	5	10596
approxgoals4	0.84	.004	0.83	0.85	approxgoals4	2.7	1.2	1	5	10474

# Scale: Performance-approach goals (SELLMO) (continued)

Maths sample-split

#### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2 370	df 28	p > chi2 .000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	51	6	.000
Strong invariance (plus equal intercepts)	89	6	.000
Strict invariance (plus equal error variances)	76	6	.000

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.999
French vs. Italian language version	.988
Italian vs. German language version	.985

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	.999
Language: Italian	.991

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. approxgoals\_fs 0.0 0.8 -1.4 1.9 10628

Share of cases with imputed missing values: 1.8% (Equivalence of scores from robust MLMV: CD = .999)

# Scale: Learning goal orientation (SELLMO)

Maths sample-split

.839

.839

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	396	2	.000
	Baseline vs. saturated	16559	6	.000
2)	Root mean squared error (	(RMSEA)		.136
	90% Confidence interval:	lower bound		.125
	90% Confidence interval:	upper bound		.147
	Probability RMSEA <= 0.05			.000
3)	Akaike's Information Crite Bayesian Information Crite	• •		113590 113677
4)	Baseline comparison			
	Comparative Fit Index (CFI)			.976
	Tucker–Lewis Index (TLI)			.929
5)	Size of residuals			

#### **Reliability and Dimensionality**

Ordinal Cronbach's Alpha

(Cronbach's alpha = .808) **McDonald's Omega** 

Test of (one-)dimensionality (parallel analysis) Criterion: retain factors with adj. eigenvalue > 0					
Adjusted eigenvalue					
Factor 1	2.15				
Factor 2	01				
Factor 3	15				

-.13

# Standardized factor loadings

Indicators	Coef.	(SE)	[95% Conf	. interval]
learntarget1	0.74	.006	0.72	0.75
learntarget2	0.76	.006	0.75	0.77
learntarget3	0.73	.006	0.72	0.74
learntarget4	0.78	.005	0.77	0.79

Stand. root mean squared residual (SRMR)

Coefficient of determination (CD)

# Item descriptives

Factor 4

.028

.841

		Std.			Valid
Indicators	Mean	dev.	Min.	Max.	Obs.
learntarget1	3.3	1.1	1	5	10637
learntarget2	3.4	1.1	1	5	10481
learntarget3	3.3	1.1	1	5	10606
learntarget4	3.1	1.1	1	5	10485

# Scale: Learning goal orientation (SELLMO) (continued)

Maths sample-split

### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2 887	df 28	p > chi2 .000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	12	6	.072
Strong invariance (plus equal intercepts)	421	6	.000
Strict invariance (plus equal error variances)	254	6	.000

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	1.000
French vs. Italian language version	.999
Italian vs. German language version	.998

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	1.000
Language: Italian	.997

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. learntarget\_fs 0.0 0.7 -2.0 1.5 10649
Share of cases with imputed missing values: 1.8% (Equivalence of scores from robust MLMV: CD = .998)

# Scale: Work avoidance (SELLMO)

Maths sample-split

#### **Model and Fit Statistics**

Reliability and Dimensionality	
Ordinal Cronbach's Alpha	.747

1)	<b>Likelihood-ratio tests</b> Model vs. saturated	chi2 370	df 2	p > chi2 .000	Ordinal Cronback (Cronbach's alpha	•	.747
	Baseline vs. saturated	9625	6	.000	McDonald's Ome	•	.750
2)	Root mean squared error (I 90% Confidence interval: 90% Confidence interval:	lower bound		.131 .120 .143	Criterion: retain f	nensionality (para actors with adj. eig djusted eigenvalue	envalue > o
	Probability RMSEA <= 0.05	opper booma		.000	Factor 1 Factor 2	1.59 02	•
3)	Akaike's Information Criter Bayesian Information Crite	• •		122140 122227	Factor 3 Factor 4	09 22	
4)	Baseline comparison Comparative Fit Index (CFI) Tucker–Lewis Index (TLI)			.962 .885			
5)	Size of residuals Stand. root mean squared re Coefficient of determination		R)	.033 .761			

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							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
avoidwork1	0.53	.009	0.51	0.54	avoidwork1	2.9	1.1	1	5	10615
avoidwork2	0.70	.007	0.68	0.71	avoidwork2	3.1	1.1	1	5	10483
avoidwork3	0.67	.008	0.66	0.69	avoidwork3	3.2	1.2	1	5	10599
avoidwork4	0.71	.007	0.70	0.72	avoidwork4	3.1	1.1	1	5	10480

# Scale: Work avoidance (SELLMO) (continued)

Maths sample-split

### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2	
	611	28	.000	
Tests of measurement invariance	chi2	df	p > chi2	
Metric invariance (equal factor loadings)	11	6	.087	
Strong invariance (plus equal intercepts)	282	6	.000	
Strict invariance (plus equal error variances)	170	6	.000	

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.999
French vs. Italian language version	.989
Italian vs. German language version	.994

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	.998
Language: Italian	.991

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. avoidwork\_fs 0.0 0.5 -1.2 1.2 10637
Share of cases with imputed missing values: 1.8% (Equivalence of scores from robust MLMV: CD = .996)

# Scale: Avoidance performance goals (SELLMO)

Maths sample-split

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	550	2	.000
	Baseline vs. saturated	20651	6	.000
2)	Root mean squared error (	(RMSEA)		.160
	90% Confidence interval:	lower bound		.149
	90% Confidence interval:	upper bound		.172
	Probability RMSEA <= 0.05			.000
3)	Akaike's Information Crite	rion (AIC)		117023
	Bayesian Information Crit	erion (BIC)		117111
4)	Baseline comparison			
	Comparative Fit Index (CFI)			.973
	Tucker–Lewis Index (TLI)			.920

#### **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.866
(Cronbach's alpha = .830)	
McDonald's Omega	.867

#### Test of (one-)dimensionality (parallel analysis)

Criterion: retain factors with adj. eigenvalue > o Adjusted eigenvalue

2.37
.01
09
14

#### 5) Size of residuals

Stand. root mean squared residual (SRMR)	.027
Coefficient of determination (CD)	.877

#### Standardized factor loadings Item descriptives

		-								
							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
avoidblame1	0.73	.005	0.72	0.74	avoidblame1	2.6	1.2	1	5	10594
avoidblame2	0.75	.005	0.74	0.76	avoidblame2	2.6	1.3	1	5	10496
avoidblame3	0.86	.004	0.85	0.87	avoidblame3	2.5	1.2	1	5	10604
avoidblame4	0.81	.005	0.80	0.81	avoidblame4	2.3	1.1	1	5	10509

# Scale: Avoidance performance goals (SELLMO) (continued)

Maths sample-split

#### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2 378	df 28	p > chi2 .000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	19	6	.004
Strong invariance (plus equal intercepts)	120	6	.000
Strict invariance (plus equal error variances)	161	6	.000

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.999
French vs. Italian language version	.997
Italian vs. German language version	1.000

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	.999
Language: Italian	1.000

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. avoidblame\_fs 0.0 0.8 -1.2 2.1 10642
Share of cases with imputed missing values: 1.9% (Equivalence of scores from robust MLMV: CD = .998)

# Scale: Global self-esteem

Baseline survey sample

#### **Model and Fit Statistics**

Reliability and Dimensionality	

1)	Likelihood-ratio tests	chi2	df	p > chi2	Ordinal Cronbach's	Alpha	.859
	Model vs. saturated	20015	20	.000	(Cronbach's alpha =	.820)	
	Baseline vs. saturated	64288	28	.000	McDonald's Omega	1	.852
2)	Root mean squared error (I	RMSEA)		.250	Test of (one-)dimer	nsionality (paralle	el analysis)
	90% Confidence interval:	ower bound		.000	Criterion: Retain fac	tors with adj. eige	nvalue > o
	90% Confidence interval:	upper bound			Adju	ısted eigenvalue	
	Probability RMSEA <= 0.05			.000	factor 1	3.56	
					factor 2	1.12	
3)	Akaike's Information Criter	ion (AIC)		329588	factor 3	.07	
	<b>Bayesian Information Crite</b>	rion (BIC)		329772	factor 4	05	
					factor 5	09	
4)	Baseline comparison				factor 6	10	
	Comparative Fit Index (CFI)			.689	factor 7	12	
	Tucker-Lewis Index (TLI)			.564	factor 8	13	
5)	Size of residuals						
	Stand. root mean squared re	esidual (SRMI	₹)	.147			
	Coefficient of determination	n (CD)		.887			

# Standardized factor loadings

# Item descriptives

	_						Std.			Valid
Indicators *	Coef.	(SE)	[95% Conf	. interval]	Indicators *	Mean	dev.	Min.	Max.	Obs.
seleı	0.63	0.01	0.62	0.64	seleı	4.0	0.9	1	5	15991
sele2	0.51	0.01	0.49	0.52	sele2	4.1	0.8	1	5	15961
sele3	0.44	0.01	0.43	0.46	sele3	3.9	0.8	1	5	15957
sele4	0.49	0.01	0.48	0.51	sele4	3.8	1.0	1	5	15946
seld1	0.85	0.00	0.84	0.85	seldı	3.8	1.2	1	5	15972
seld3	0.75	0.00	0.74	0.75	seld3	3.2	1.2	1	5	15953
seld4	0.65	0.01	0.64	0.66	seld4	3.2	1.3	1	5	15902
seld5	0.80	0.00	0.79	0.81	seld5	4.0	1.2	1	5	15943

<sup>\*</sup> Note: Reversed categories for all seld-items

## Tests and Indices of Factorial Invariance across ...

Equal	lity c	of the
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variance-covariance matrices across	Surve	y lan	guages	Surv	ey set	tings	Sur	vey m	odes
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
	5550	88	.000	693	44	.000	136	44	.000
Tests of measurement invariance across	. Surve	y lan	guages	Surv	ey set	tings	Sur	vey m	odes
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
Metric invariance (equal factor loadings)	85	14	.000	27	7	.000	38	7	.000
Strong invariance (plus equal intercepts)	3216	14	.000	618	7	.000	42	7	.000
Strict invariance (plus equal error variances)	415	14	.000	205	7	.000	25	7	.001
Configural factor similarity across	Surve	y lan	guages	Surv	ey set	tings	Sur	vey m	odes
Tucker's congruence coefficient			TCC			TCC			TCC
	German vs. F French vs. It		.999 .998	classro unpro	om vs. ctored	.999	W	eb vs. PAP	.991
	Italian vs. Ge	rman	.996						
Factor score equivalence: group									
specific vs. invariant models for	Surve	y lan	guages	Surv	ey set	tings	Sur	vey m	odes
Coefficient of determination			CD			CD			CD
	Ge	rman	1.000	clas	sroom	1.000		web	1.000
	F	rench	.994	unpro	ctored	.998		PAP	.985

Italian

.989

# Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. sel\_fs 0.0 0.5 -1.8 0.8 16003 Share of cases with imputed missing values: 1.2% (Equivalence of scores from Robust MLMV: CD = .997)

# Scale: Global self-esteem (shortened)

Baseline survey sample

#### **Model and Fit Statistics**

# Reliability and Dimensionality

1)	Likelihood-ratio tests	chi2	df	p > chi2	Ordinal Cronbach's Alpha	.852
	Model vs. saturated	17789	14	.000	(Cronbach's alpha = .809)	
	Baseline vs. saturated	55337	21	.000	McDonald's Omega	.852
2)	Root mean squared error (I	RMSEA)		.282	Test of (one-)dimensionality	(parallel analysis)
	90% Confidence interval:	ower bound		.000	Criterion: Retain factors with a	adj. eigenvalue > o
	90% Confidence interval:	upper bound			Adjusted eige	nvalue
	Probability RMSEA <= 0.05			.000	factor 1 3.24	+
					factor 2 .97	7
3)	Akaike's Information Criter	ion (AIC)		283054	factor 3o1	L
	<b>Bayesian Information Crite</b>	rion (BIC)		283215	factor 4o6	j
					factor 511	L
4)	Baseline comparison				factor 612	2
	Comparative Fit Index (CFI)			.679	factor 714	ŧ
	Tucker-Lewis Index (TLI)			.518		
5)	Size of residuals					
	Stand. root mean squared re	sidual (SRM	R)	.133		
	Coefficient of determination	(CD)		.860		

# Standardized factor loadings

# Item descriptives

		-								
							Std.			Valid
Indicators *	Coef.	(SE)	[95% Conf	interval]	Indicators *	Mean	dev.	Min.	Max.	Obs.
sele1	0.77	0.00	0.76	0.78	sele1	4.0	0.9	1	5	15991
sele2	0.72	0.01	0.71	0.73	sele2	4.1	0.8	1	5	15961
sele3	0.67	0.01	0.66	0.68	sele3	3.9	0.8	1	5	15957
sele4	0.68	0.01	0.67	0.69	sele4	3.8	1.0	1	5	15946
seld1	0.66	0.01	0.64	0.67	seld1	3.8	1.2	1	5	15972
seld3	0.56	0.01	0.55	0.58	seld3	3.2	1.2	1	5	15953
seld5	0.63	0.01	0.62	0.64	seld5	4.0	1.2	1	5	15943

<sup>\*</sup> Note: Reversed categories for all seld-items

# Parameters of generalized structural equation model (ordinal logit link)

Indicators Coef. Cut1 Cut2 Cut3

Tests and Indices of Factorial Invariance across ...

Equality of the	Equa	lity	of '	the
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variance-covariance matrices across	Survey	lang	guages	Surv	ey set	tings	Sur	vey m	odes
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
	4643	70	.000	628	35	.000	125	35	.000
Tests of measurement invariance across	Survey	land	quages	Surv	ey seti	tings	Sur	vey m	odes
	chi2	df .		chi2	df	p > chi2	chi2	df	p > chi2
Metric invariance (equal factor loadings)	130	12	.000	40	6	.000	12	6	.069
Strong invariance (plus equal intercepts)	1838	12	.000	589	6	.000	52	6	.000
Strict invariance (plus equal error variances)	320	12	.000	142	6	.000	15	6	.017
Configural factor similarity across	Survey languages		Survey settings			Survey modes			
Tucker's congruence coefficient	_		TCC		-	TCC		-	TCC
	German vs. Fre		33	classro		.996	W	eb vs. PAP	.996
	French vs. Ital		.983	unpro	ctored			PAP	
	Italian vs. Gerr	nan	.966						
Factor score equivalence: group									
specific vs. invariant models for	Survey	lan	guages	Surv	ey set	tings	Sur	vey m	odes
Coefficient of determination			CD			CD			CD
	Gern	nan	.999	clas	sroom	.999		web	1.000
	Fre	nch	.997	unpro	ctored	.991		PAP	.997

Italian .826

# Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. sel\_m\_fs 0.0 0.6  $^{-2.7}$  1.0 16003 Share of cases with imputed missing values: 1.0% (Equivalence of Scores from Robust MLMV: CD = .997)

# Scale: Positive global self-esteem

Baseline survey sample

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	329	2	.000
	Baseline vs. saturated	26567	6	.000
2)	Root mean squared error (	(RMSEA)		.101
	90% Confidence interval:	lower bound		.092
	90% Confidence interval:	upper bound		.110
	Probability RMSEA <= 0.05			.000
3)	Akaike's Information Crite	rion (AIC)		140371
	Bayesian Information Crit	erion (BIC)		140463
4)	Baseline comparison			
•	Comparative Fit Index (CFI)			.988
	Tucker-Lewis Index (TLI)			.963
	c'			

## **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.848
(Cronbach's alpha = .801)	
McDonald's Omega	.849

#### Test of (one-)dimensionality (parallel analysis)

Criterion: Retain factors with adj. eigenvalue > o Adjusted eigenvalue

factor 1	2.21
factor 2	06
factor 3	07
factor 4	15

#### 5) Size of residuals

Indicators

seleı

sele2

sele3

sele4

Stand. root mean squared residual (SRMR) .018 Coefficient of determination (CD) .856

(SE)

0.00

0.00

0.00

0.00

# Standardized factor loadings

Coef.

0.72

0.83

0.78

0.72

# Item descriptives

				Std.			Valid
[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
0.71	0.73	sele1	4.0	0.9	1	5	15991
0.82	0.83	sele2	4.1	8.0	1	5	15961
0.78	0.79	sele3	3.9	0.8	1	5	15957
0.71	0.73	sele4	3.8	1.0	1	5	15946

Tests and Indices of Factorial Invariance across ...

Equal	ity o	f the
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variance-covariance matrices across	Survey languages		guages	Survey settings			Survey modes		
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
	1803	28	.000	346	14	.000	35	14	.002
Tests of measurement invariance across	Survey	Survey languages		Survey settings			Survey modes		odes
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
Metric invariance (equal factor loadings)	21	6	.002	11	3	.013	1	3	.769
Strong invariance (plus equal intercepts)	1214	6	.000	140	3	.000	8	3	.052
Strict invariance (plus equal error variances)	216	6	.000	123	3	.000	10	3	.017
Configural factor similarity across	Survey	/ lan	guages	Survey settings			Survey modes		
Tucker's congruence coefficient			TCC	TCC					TCC
	German vs. Fre	-		classroom vs.		1.000	W	eb vs.	1.000
	French vs. Ita		.998	unpro	ctored			PAP	
	Italian vs. Ger	man	.997						
Factor score equivalence: group									
specific vs. invariant models for	Survey languages		Survey settings		tings	Sur	vey m	odes	
Coefficient of determination			CD			CD			CD
	Gerr	man	1.000	clas	sroom	1.000		web	1.000
	Fre	ench	.998	unpro	ctored	1.000		PAP	1.000

Italian

.992

# Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. sele\_fs 0.0 0.6  $^{-2.5}$  0.9 15997 Share of cases with imputed missing values: 0.6% (Equivalence of scores from robust MLMV: CD = .996)

# Scale: Negative global self-esteem

Baseline survey sample

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	712	2	.000
	Baseline vs. saturated	31810	6	.000
2)	Root mean squared error of 90% Confidence interval: 90% Confidence interval: Probability RMSEA <= 0.05	lower bound		.149 .140 .158 .000
3)	Akaike's Information Crite Bayesian Information Crit			175983 176075
4)	Baseline comparison Comparative Fit Index (CFI) Tucker-Lewis Index (TLI)			.978 .933
5)	Size of residuals	escidual (CDM)	2)	229

## **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.866
(Cronbach's alpha = .824)	
McDonald's Omega	.868

#### Test of (one-)dimensionality (parallel analysis)

Criterion: Retain factors with adj. eigenvalue > o Adjusted eigenvalue

factor 1	2.39
factor 2	.02
factor 3	13
factor 4	12

Stand. root mean squared residual (SRMR)	.028
Coefficient of determination (CD)	.887

# Standardized factor loadings Item descriptives

	_						Std.			Valid
Indicators *	Coef.	(SE)	[95% Conf	. interval]	Indicators *	Mean	dev.	Min.	Max.	Obs.
seld1	0.88	0.00	0.88	0.89	seld1	3.8	1.2	1	5	15972
seld3	0.79	0.00	0.78	0.80	seld3	3.2	1.2	1	5	15953
seld4	0.67	0.01	0.66	0.68	seld4	3.2	1.3	1	5	15902
seld5	0.80	0.00	0.80	0.81	seld5	4.0	1.2	1	5	15943

<sup>\*</sup> Note: Reversed Item Categories

Tests and Indices of Factorial Invariance across ...

Equal	lity	of t	:he
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variance-covariance matrices across	Survey languages		Survey settings			Survey modes			
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
	4554	28	.000	140	14	.000	59	14	.000
Tests of measurement invariance across	. Surve	Survey languages		Survey settings			Survey modes		odes
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
Metric invariance (equal factor loadings)	107	6	.000	4	3	.235	7	3	.064
Strong invariance (plus equal intercepts)	2496	6	.000	86	3	.000	27	3	.000
Strict invariance (plus equal error variances)	355	6	.000	1	3	.707	7	3	.089
Configural factor similarity across	Surve	y lan	guages	Survey settings			Survey modes		
Tucker's congruence coefficient			TCC	TCC		TCC			TCC
	German vs. French .997 French vs. Italian 1.000		classroom vs. unproctored		1.000	W	eb vs. PAP	.998	
	Italian vs. Gei			onproctored				. ,	
Factor score equivalence: group									
specific vs. invariant models for	Surve	y lan	guages	Surv	ey set	tings	Sur	vey m	odes
Coefficient of determination			CD			CD			CD
	Ger	man	1.000	clas	sroom	1.000		web	1.000
	Fr	ench	.990	unpro	ctored	1.000		PAP	.999

Italian

.980

# Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. seld\_fs 0.0 1.0 -2.6 1.3 15995 Share of cases with imputed missing values: 0.9% (Equivalence of scores from robust MLMV: CD = .993)

# Scale: Negative global self-esteem (shortened)

Baseline survey sample

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	0	0	
	Baseline vs. saturated	23184	3	.000
2)	Root mean squared error (I		.000	

2) Root mean squared error (RMSEA) .000
90% Confidence interval: lower bound .000
90% Confidence interval: upper bound .000
Probability RMSEA <= 0.05 1.000

3) Akaike's Information Criterion (AIC) 130616 Bayesian Information Criterion (BIC) 130685

4) Baseline comparison

Comparative Fit Index (CFI) 1.000
Tucker-Lewis Index (TLI) 1.000

5) Size of residuals

Stand. root mean squared residual (SRMR) .000
Coefficient of determination (CD) .885

# **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.863
(Cronbach's alpha = .816)	
McDonald's Omega	.865

Test of (one-)dimensionality (parallel analysis)

Criterion: Retain factors with adj. eigenvalue > 0

Adjusted eigenvalue

	Aujusteu eigenvalue	
factor 1	1.90	
factor 2	08	
factor 3	14	

# Standardized factor loadings

#### Item descriptives

							Std.			Valid
Indicators *	Coef.	(SE)	[95% Conf	interval]	Indicators *	Mean	dev.	Min.	Max.	Obs.
seld1	0.91	0.00	0.90	0.91	seld1	3.8	1.2	1	5	15972
seld3	0.80	0.00	0.79	0.80	seld3	3.2	1.2	1	5	15953
seld5	0.77	0.00	0.76	0.78	seld5	4.0	1.2	1	5	15943

<sup>\*</sup> Note: Reversed Item Categories

#### Parameters of generalized structural equation model (ordinal logit link)

Indicators Coef. Cut1 Cut2 Cut3

Tests and Indices of Factorial Invariance across ...

Equal	lity	of t	:he
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variance-covariance matrices across	Surve	guages	Survey settings			Survey modes			
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
	2872	18	.000	104	9	.000	53	9	.000
Tests of measurement invariance across	Surve	y lan	guages	Surv	ey seti	tings	Sur	vey m	odes
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
Metric invariance (equal factor loadings)	61	4	.000	1	2	.749	6	2	.061
Strong invariance (plus equal intercepts)	1218	4	.000	62	2	.000	26	2	.000
Strict invariance (plus equal error variances)	27	4	.000	1	2	.511	5	2	.087
Configural factor similarity across	Surve	y lan	guages	Survey settings			Survey modes		
Tucker's congruence coefficient			TCC	TCC					TCC
	German vs. Fr		.998 1.000	classroom vs. unproctored		1.000	web vs. PAP <sup>.998</sup>		.998
	Italian vs. Ge	rman	.998	•					
Factor score equivalence: group									
specific vs. invariant models for	Surve	y lan	guages	Surv	ey set	tings	Sur	vey m	odes
Coefficient of determination			CD			CD			CD
	Ger	man	.999	class	sroom	1.000		web	1.000
	Fi	rench	.989	unprod	tored	1.000		PAP	.997

Italian

.980

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. seld\_m\_fs 0.0 1.0 -2.6 1.2 15994 Share of cases with imputed missing values: 0.5% (Equivalence of Scores from Robust MLMV: CD = .999)

# Scale: General perceived self-efficacy scale (GSES)

Baseline survey sample

# **Model and Fit Statistics**

Reliability and	Dimensionality

1)	Likelihood-ratio tests	chi2	df	p > chi2	Ordinal Cronbach's Alpha	a .835
	Model vs. saturated	63	2	.000	(Cronbach's alpha = .772)	
	Baseline vs. saturated	23581	6	.000	McDonald's Omega	.835
2)	Root mean squared error (	RMSEA)		.044	Test of (one-)dimensiona	lity (parallel analysis)
	90% Confidence interval:	lower bound		.035	Criterion: Retain factors w	vith adj. eigenvalue > o
	90% Confidence interval:	upper bound		.053	Adjusted 6	eigenvalue
	Probability RMSEA <= 0.05			.847	factor 1	2.10
					factor 2	08
3)	Akaike's Information Crite	rion (AIC)		104477	factor 3	12
	Bayesian Information Crite	erion (BIC)		104569	factor 4	13
4)	Baseline comparison					
	Comparative Fit Index (CFI)			.997		
	Tucker-Lewis Index (TLI)			.992		
5)	Size of residuals					
	Stand. root mean squared re	esidual (SRM	₹)	.009		
	Coefficient of determination	n (CD)		.836		

# Standardized factor loadings

# Item descriptives

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
seef1	0.73	0.00	0.72	0.74	seef1	3.1	0.6	1	4	15941
seef2	0.77	0.00	0.76	0.78	seef2	3.1	0.7	1	4	15928
seef3	0.76	0.00	0.75	0.77	seef3	2.8	0.7	1	4	15916
seef4	0.73	0.00	0.72	0.74	seef4	3.0	0.7	1	4	15923

# Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut1	Cut <sub>2</sub>	Cut <sub>3</sub>
seef1	2.04	-6.05	-3.17	2.22
seef2	2.28	-6.20	-2.91	1.82
seef3	2.14	-5.09	-1.43	2.66
seef4	2.03	-5.56	-2.00	2.27

Tests and Indices of Factorial Invariance across ...

Equali	tv of	the
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variance-covariance matrices across	Survey languages			Surv	tings	Survey modes			
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
	1049	28	.000	104	14	.000	24	14	.044
Tests of measurement invariance across	. Survey languages		guages	Survey settings		tings	Survey mod		odes
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
Metric invariance (equal factor loadings)	47	6	.000	1	3	.763	4	3	.252
Strong invariance (plus equal intercepts)	448	6	.000	10	3	.018	2	3	.652
Strict invariance (plus equal error variances)	230	6	.000	12	3	.008	4	3	.303
Configural factor similarity across	Surve	y lan	guages	Survey settings			Survey modes		
Tucker's congruence coefficient			TCC	TCC		TCC	TC		TCC
	German vs. Fr	rench	.998	classro	om vs.	1.000	W	eb vs.	000
	French vs. It	alian	.995	unpro	unproctored		PAP		.999
	Italian vs. Ge	rman	.996						
Factor score equivalence: group									
specific vs. invariant models for	Surve	y lan	guages	Surv	ey set	tings	Sur	vey m	odes
Coefficient of determination			CD			CD			CD
	Ger	man	1.000	clas	sroom	1.000		web	1.000
	Fr	rench	.997	unpro	ctored	1.000		PAP	.999

Italian

.993

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. seef\_fs 0.0 0.9 -3.0 1.8 15951 Share of cases with imputed missing values: 0.4% (Equivalence of scores from robust MLMV: CD = .996) (Equivalence of Scores from Two-Step-Approach: CD = .989)

# Scale: Academic self-efficacy

Full AES sample

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	0	0	
	Baseline vs. saturated	32752	3	.000
2)	Root mean squared error (	(RMSEA)		.000
	90% Confidence interval:	lower bound		.000
	90% Confidence interval:	upper bound		.000
	Probability RMSEA <= 0.05			1.000
3)	Akaike's Information Crite	erion (AIC)		179405
	Bayesian Information Crit	erion (BIC)		179477
4)	Baseline comparison			
	Comparative Fit Index (CFI)			1.000

## **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.868
(Cronbach's alpha = .836)	
McDonald's Omega	.869

#### Test of (one-)dimensionality (parallel analysis)

Criterion: retain factors with adj. eigenvalue > o
Adjusted eigenvalue

	-	_	
Factor 1		1.92	
Factor 2		11	
Factor 3		13	

Comparative Fit Index (CFI) 1.000
Tucker–Lewis Index (TLI) 1.000

# 5) Size of residuals

Stand. root mean squared residual (SRMR) .000 Coefficient of determination (CD) .874

# Standardized factor loadings Item descriptives

					Std.			Valid		
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
acaself1	0.81	.003	0.80	0.81	acaself1	4.7	1.1	1	6	22256
acaself2	0.87	.003	0.87	0.88	acaself2	4.1	1.2	1	6	22248
acaself3	0.81	.003	0.80	0.81	acaself3	4.3	1.2	1	6	22252

# Scale: Academic self-efficacy (continued)

Full AES sample

#### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2 774	df 18	p > chi2 .000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	77	4	.000
Strong invariance (plus equal intercepts)	250	4	.000
Strict invariance (plus equal error variances)	318	4	.000

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.998
French vs. Italian language version	.998
Italian vs. German language version	.996

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	.999
Language: Italian	.989

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. acaself\_fs 0.0 0.8  $^{-2.7}$  1.4 22264 Share of cases with imputed missing values: 0.1% (Equivalence of scores from robust MLMV: CD = .999)

# Scale: Academic self-concept (PISA2000)

Full AES sample

#### **Model and Fit Statistics**

#### **Reliability and Dimensionality**

					,	
1)	<b>Likelihood-ratio tests</b> Model vs. saturated	chi2 0	df o	p > chi2	Ordinal Cronbach's Alpha (Cronbach's alpha = .795)	.856
	Baseline vs. saturated	31794	3	.000	McDonald's Omega	.860
2)	Root mean squared error 90% Confidence interval: 90% Confidence interval: Probability RMSEA <= 0.05	ower bound upper bound		.000 .000 .000 1.000		h adj. eigenvalue > o
3)	Akaike's Information Crit Bayesian Information Crit	, ,		111791 111863	Factor 3	14
4)	Baseline comparison Comparative Fit Index (CFI Tucker–Lewis Index (TLI)	)		1.000 1.000		
5)	Size of residuals Stand. root mean squared Coefficient of determination		MR)	.000 .884		

# Standardized factor loadings

# Item descriptives

							Std.			Valid
Indicators *	Coef.	(SE)	[95% Conf	. interval]	Indicators *	Mean	dev.	Min.	Max.	Obs.
scacad1	0.70	.004	0.70	0.71	scacad1	2.9	0.7	1	4	22202
scacad2	0.89	.003	0.89	0.90	scacad2	2.9	0.7	1	4	22175
scacad3	0.85	.003	0.84	o.86	scacad3	2.9	0.7	1	4	22168
* Note: Replication of 'Scacad'-Scale from TREE1 / PISA2000										

Parameters of generalized structural equation model (ordinal logit link)					
Indicators	Coef.	Cut1	Cut <sub>2</sub>	Cut <sub>3</sub>	
scacad1	1.87	-4.54	-1.94	2.37	
scacad2	3.96	-7.57	-2.86	3.92	
scacada	3.05	-6.36	-2.61	2.71	

# Scale: Academic self-concept (PISA2000) (continued)

Full AES sample

#### Tests and Indices of Factorial Invariance across Survey Languages

chi2	df	p > chi2
1571	18	.000
chi2	df	p > chi2
76	4	.000
768	4	.000
427	4	.000
	1571 chi2 76 768	1571 18  chi2 df 76 4 768 4

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.998
French vs. Italian language version	.999
Italian vs. German language version	1.000

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	.999
Language: French	.987
Language: Italian	.996

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. scacad\_fs 0.0 0.9 -2.5 1.7 22210 Share of cases with imputed missing values: 0.3% (Equivalence of scores from robust MLMV: CD = .997) (Equivalence of scores from two-step approach: CD = .986)

# Scale: Verbal self-concept (PISA2000)

Full AES sample

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests chi2  Model vs. saturated o  Baseline vs. saturated 32226	df o	p > chi2	Ordinal Cronbach's (Cronbach's alpha =
	Baseline vs. saturated 32226	3	.000	McDonald's Omega
2)	Root mean squared error (RMSEA)		.000	Test of (one-)dimen
	90% Confidence interval: lower bound		.000	Criterion: retain fact
	90% Confidence interval: upper bound		.000	Adju
	Probability RMSEA <= 0.05		1.000	Factor 1
				Factor 2
3)	Akaike's Information Criterion (AIC)		128063	Factor 3
	Bayesian Information Criterion (BIC)		128135	
4)	Baseline comparison			
	Comparative Fit Index (CFI)		1.000	
	Tucker–Lewis Index (TLI)		1.000	
5)	Size of residuals			
	Stand. root mean squared residual (SRM	R)	.000	
	Coefficient of determination (CD)		.888	

# **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.856
(Cronbach's alpha = .795)	
McDonald's Omega	.861

## Test of (one-)dimensionality (parallel analysis)

Criterion: retain factors with adj. eigenvalue > o

Adjusted eigenvalue

	Aujusteu eigenvalue	
Factor 1	1.90	
Factor 2	08	
Factor 3	14	

# Standardized factor loadings

	J						Std.			Valid
Indicators *	Coef.	(SE)	[95% Conf	. interval]	Indicators *	Mean	dev.	Min.	Max.	Obs.
scverb1 **	0.70	0.00	0.69	0.70	scverb1 **	3.2	0.8	1	4	22196
scverb2	0.90	0.00	0.89	0.90	scverb2	2.8	0.8	1	4	22173
scverb3	0.86	0.00	0.85	0.86	scverb3	2.9	0.8	1	4	22171

Item descriptives

# Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut <sub>1</sub>	Cut <sub>2</sub>	Cut <sub>3</sub>
scverb1	1.84	-4.49	-2.24	0.34
scverb2	3.52	-6.01	-1.79	3.39
scverb3	2.89	-5.94	-2.37	2.79

<sup>\*</sup> Note: Replication of 'Scverb'-Scale from TREE1/PISA2000

<sup>\*\*</sup> Note: Reversed Categories for Item Scverb1

# Scale: Verbal self-concept (PISA2000) (continued)

Full AES sample

#### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2
	621	18	.000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	30	4	.000
Strong invariance (plus equal intercepts)	58	4	.000
Strict invariance (plus equal error variances)	215	4	.000

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	1.000
French vs. Italian language version	.989
Italian vs. German language version	.986

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	.998
Language: Italian	.998

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. scverb\_fs 0.0 0.9 -2.4 1.6 22205

Share of cases with imputed missing values: 0.3% (Equivalence of scores from robust MLMV: CD = .999) (Equivalence of scores from two-step approach: CD = .988)

# Scale: Maths self-concept [PISA 2000]

Full AES sample

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	0	0	
	Baseline vs. saturated	57824	3	.000

2) Root mean squared error (RMSEA) .000 90% Confidence interval: lower bound .000 90% Confidence interval: upper bound .000 Probability RMSEA <= 0.05 1.000

Akaike's Information Criterion (AIC) 134733 **Bayesian Information Criterion (BIC)** 134805

4) Baseline comparison

Comparative Fit Index (CFI) 1.000 Tucker-Lewis Index (TLI) 1.000

Size of residuals

Stand. root mean squared residual (SRMR) .000 Coefficient of determination (CD) .980

## **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.927
(Cronbach's alpha = .888)	
McDonald's Omega	.930

Test of (one-)dimensionality (parallel analysis)

Criterion: retain factors with adj. eigenvalue > o Adjusted eigenvalue

	Adjusted eigenval
Factor 1	2.38
Factor 2	01
Factor 3	08

# Standardized factor loadings

Indicators *	Coef.	(SE)	[95% Conf	. interval]
matcon1	0.90	.002	0.90	0.90
matcon2	0.99	.001	0.99	0.99
matcon3	0.82	.002	0.81	0.82

Hattonii	0.90	.002	0.90	0.90
natcon2	0.99	.001	0.99	0.99
natcon3	0.82	.002	0.81	0.82
Note: Replication	of 'Matcon'	-Scale fror	n TREE1 / P	ISA2000

# Item descriptives

		Std.			Valid
Indicators *	Mean	dev.	Min.	Max.	Obs.
matcon1	2.7	0.9	1	4	22183
matcon2	2.4	1.1	1	4	22187
matcon3	2.4	1.0	1	4	22180

#### Parameters of generalized structural equation model (ordinal logit link)

				•
Indicators	Coef.	Cut <sub>1</sub>	Cut <sub>2</sub>	Cut <sub>3</sub>
matcon1	3.38	-4.50	-1.06	2.95
matcon2	4.96	-3.20	0.21	4.25
matcon3	2.40	-2.30	0.21	2.53

#### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2	
	937	18	.000	
Tests of measurement invariance	chi2	df	p > chi2	
Metric invariance (equal factor loadings)	335	4	.000	
Strong invariance (plus equal intercepts)	47	4	.000	
Strict invariance (plus equal error variances)	241	2	.000	
Configural factor similarity				
Tucker's Congruence Coefficient	TCC			
German vs. French language version	.998			
French vs. Italian language version	.997			
Italian vs. German language version	.999			
Factor score equivalence: group specific vs. inv	ariant mode	els		
Coefficient of determination	CD			
Language: German	1.000			
Language: French	1.000			
Language: Italian	1.000			
* Note: Language-specific models do not conve	erge and the r	elated in	variance tests	and indices may not be

# Factor score descriptives

Std.

calculated unless the error variance of item matcon2 is constrained to zero.

Variable name Mean dev. Min. Max. Obs. matcon\_fs 0.0 1.0 -1.7 1.6 22193

Share of cases with imputed missing values: 0.1% (Equivalence of scores from robust MLMV: CD = .967) (Equivalence of scores from two-step approach: CD = .899)

# Scale: ICT self-concept

Maths sample-split

#### **Model and Fit Statistics**

Re	liabi	lity ar	nd Dime	ensional	lity
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1)	Likelihood-ratio tests chi2	df	p > chi2	Ordinal Cronbach's A	Alpha .896
	Model vs. saturated o	0	•	(Cronbach's alpha = .8	349)
	Baseline vs. saturated 20861	3	.000	McDonald's Omega	.898
	20001	5			
2)	Root mean squared error (RMSEA)		.000	Test of (one-)dimens	ionality (parallel analysis)
	90% Confidence interval: lower bound		.000	Criterion: retain facto	rs with adj. eigenvalue > o
	90% Confidence interval: upper bound		.000		ted eigenvalue
	Probability RMSEA <= 0.05		1.000	Factor 1	2.12
	•			Factor 2	08
3)	Akaike's Information Criterion (AIC)		68148	Factor 3	10
	Bayesian Information Criterion (BIC)		68214		
4)	Baseline comparison				
	Comparative Fit Index (CFI)		1.000		
	Tucker–Lewis Index (TLI)		1.000		
5)	Size of residuals				
	Stand. root mean squared residual (SRM	R)	.000		
	Coefficient of determination (CD)		.912		

# Standardized factor loadings

# Item descriptives

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
ictmot6	0.78	.004	0.77	0.79	ictmot6	2.9	0.9	1	4	11064
ictmot7	0.90	.003	0.89	0.90	ictmot7	2.2	0.9	1	4	11057
ictmot8	0.91	.003	0.90	0.91	ictmot8	2.4	0.9	1	4	11058

# Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut1	Cut <sub>2</sub>	Cut <sub>3</sub>
ictmot6	2.31	-4.15	-1.43	1.80
ictmot7	3.82	-2.56	1.99	5.06
ictmot8	4.06	-2 7/.	0.04	7. 72

#### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2 628	df 18	p > chi2 .000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	82	4	.000
Strong invariance (plus equal intercepts)	47	4	.000
Strict invariance (plus equal error variances)	170	4	.000

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.996
French vs. Italian language version	.987
Italian vs. German language version	.997

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	.999
Language: Italian	.996

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. ictabil\_fs 0.0 0.9 -1.8 1.8 11067

Share of cases with imputed missing values: 0.2% (Equivalence of scores from robust MLMV: CD = .997) (Equivalence of scores from two-step approach: CD = .989)

# Scale: Specific self-efficacy: numeracy

Full AES sample

#### **Model and Fit Statistics**

#### **Reliability and Dimensionality**

IVI	Juei and Tit Statistics				Reliability and Dill	ilensionancy	
1)	Likelihood-ratio tests	chi2	df	p > chi2	Ordinal Cronbach's	Alpha	.851
	Model vs. saturated	536	2	.000	(Cronbach's alpha = .	831)	
	Baseline vs. saturated	36814	6	.000	McDonald's Omega		.852
2)	Root mean squared error (I	RMSEA)		.110	Test of (one-)dimen	el analysis)	
	90% Confidence interval:	lower bound		.103	Criterion: retain facto	ors with adj. eige	nvalue > o
	90% Confidence interval:	upper bound	l	.118	Adjus	sted eigenvalue	
	Probability RMSEA <= 0.05			.000	Factor 1	2.23	
					Factor 2	05	
3)	Akaike's Information Criter	rion (AIC)		196455	Factor 3	08	
	Bayesian Information Crite	rion (BIC)		196551	Factor 4	16	
4)	Baseline comparison						
	Comparative Fit Index (CFI)			.985			
	Tucker–Lewis Index (TLI)			.956			
5)	Size of residuals						
	Stand. root mean squared re	esidual (SRM	IR)	.020			
	Coefficient of determination	n (CD)		.854			

# Standardized factor loadings

# Item descriptives

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
selfeffoı	0.77	.004	0.76	0.77	selfeffoı	3.3	0.9	1	4	21801
selfeffo2	0.77	.004	0.76	0.78	selfeffo2	3.0	0.9	1	4	21827
selfeffo3	0.80	.003	0.79	0.81	selfeffo3	2.8	0.9	1	4	10734
selfeffo4	0.73	.004	0.72	0.74	selfeffo4	2.7	0.9	1	4	10755

# Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut1	Cut <sub>2</sub>	Cut <sub>3</sub>
selfeffo1	2.35	-4.76	-2.62	-0.16
selfeffo2	2.38	-4.13	-1.77	1.07
selfeffo3	3.03	-5.40	-1.83	2.94
selfeffo4	2.27	-4.13	-1.09	2.49

# Scale: Specific self-efficacy: numeracy (continued)

Full AES sample

#### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2
	651	28	.000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	72	6	.000
Strong invariance (plus equal intercepts)	85	6	.000
Strict invariance (plus equal error variances)	33	6	.000

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.998
French vs. Italian language version	.998
Italian vs. German language version	1.000

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	.998
Language: Italian	1.000

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. selfeffa\_fs 0.0 0.9 -2.4 1.6 21881

Share of cases with imputed missing values: 51.2% (Equivalence of scores from robust MLMV: CD = .995) (Equivalence of scores from two-step approach: CD = .976)

# Scale: Specific self-efficacy: algebra

Full AES sample

#### **Model and Fit Statistics**

#### **Reliability and Dimensionality**

Model and Tit Statistics					Reliability and E			
1)	Likelihood-ratio tests	chi2	df	p > chi2	Ordinal Cronbach	•	.947	
	Model vs. saturated	3889	2	.000	(Cronbach's alpha	= .926)		
	Baseline vs. saturated	92426	6	.000	McDonald's Ome	ga	.948	
2)	Root mean squared error (RMSEA) .298				Test of (one-)dimensionality (parallel analysis)			
	90% Confidence interval:	lower bound		.290	Criterion: retain fa	ctors with adj. ei	igenvalue > o	
	90% Confidence interval:	upper bound	i	.306	Ad	ljusted eigenvalu	je	
	Probability RMSEA <= 0.05			.000	Factor 1	3.24		
	•				Factor 2	.07		
3)	Akaike's Information Crite	rion (AIC)		147967	Factor 3	06		
	Bayesian Information Crite	erion (BIC)		148063	Factor 4	06		
4)	Baseline comparison							
-	Comparative Fit Index (CFI)			.958				
	Tucker–Lewis Index (TLI)			.874				
5)	Size of residuals							
-	Stand. root mean squared re	esidual (SRN	IR)	.026				
	Coefficient of determination			.957				

# Standardized factor loadings

# Item descriptives

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
selfeffo5	0.86	.002	0.86	0.87	selfeffo5	3.3	0.9	1	4	21809
selfeffo6	0.95	.001	0.95	0.96	selfeffo6	3.0	1.0	1	4	21794
selfeffo7	0.88	.002	0.88	0.89	selfeffo7	2.8	1.0	1	4	10747
selfeffo8	0.92	.001	0.92	0.93	selfeffo8	3.2	0.9	1	4	10730

# Parameters of generalized structural equation model (ordinal logit link)

				•
Indicators	Coef.	Cut <sub>1</sub>	Cut <sub>2</sub>	Cut <sub>3</sub>
selfeffo5	3.39	-5.99	-3.58	-0.95
selfeffo6	8.35	-11.55	-5.35	1.58
selfeffo7	4.65	-6.43	-2.51	1.99
selfeffo8	5.99	-9.89	-5.56	-0.57

# Scale: Specific self-efficacy: algebra (continued)

Full AES sample

#### Tests and Indices of Factorial Invariance across Survey Languages

506	df 28	p > chi2 .000
chi2	df	p > chi2
17	6	.010
116	6	.000
238	6	.000
	chi2 17 116	506 28  chi2 df 17 6 116 6

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	1.000
French vs. Italian language version	1.000
Italian vs. German language version	1.000

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	1.000
Language: Italian	.998

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. selfeffb\_fs -0.1 0.9 -2.2 1.1 21872

Share of cases with imputed missing values: 51.2% (Equivalence of scores from robust MLMV: CD = .998) (Equivalence of scores from two-step approach: CD = .957)

# Scale: Specific self-efficacy: geometry

Full AES sample

#### **Model and Fit Statistics**

## **Reliability and Dimensionality**

Model and Fit Statistics					Reliability and Diffiensionality				
1)	<b>Likelihood-ratio tests</b> Model vs. saturated	chi2 229	df 2	p > chi2 .000	Ordinal Cronback (Cronbach's alpha	=	.823		
	Baseline vs. saturated	30977	6	.000	McDonald's Ome	=	.825		
2)	Root mean squared error (I	RMSEA)		.072	Test of (one-)dimensionality (parallel analysis)				
	90% Confidence interval:	lower boun	d	.064	Criterion: retain f	actors with adj. ei	genvalue > o		
	90% Confidence interval:	upper boun	d	.080	Α	djusted eigenvalu	e		
	Probability RMSEA <= 0.05			.000	Factor 1	2.05			
					Factor 2	07			
3)	Akaike's Information Criter	ion (AIC)		203347	Factor 3	09			
	Bayesian Information Crite	rion (BIC)		203443	Factor 4	16			
4)	Baseline comparison								
	Comparative Fit Index (CFI)			.993					
	Tucker–Lewis Index (TLI)			.978					
5)	Size of residuals								
	Stand. root mean squared re	esidual (SRI	ЛR)	.015					
	Coefficient of determination	n (CD)		.836					

# Standardized factor loadings

# Item descriptives

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
selfeffo9	0.81	.004	0.80	0.81	selfeffo9	3.3	0.9	1	4	10752
selfeff10	0.76	.004	0.75	0.76	selfeff10	3.2	0.9	1	4	21783
selfeff11	0.75	.004	0.74	0.75	selfeff11	3.0	1.0	1	4	21802
selfeff12	0.63	.005	0.62	0.64	selfeff12	2.6	0.9	1	4	10751

# Parameters of generalized structural equation model (ordinal logit link)

				•
Indicators	Coef.	Cut1	Cut <sub>2</sub>	Cut <sub>3</sub>
selfeffo9	3.22	-6.78	-3.69	-0.03
selfeff10	2.24	-4.55	-2.29	0.17
selfeff11	2.15	-3.88	-1.49	0.85
selfeff12	1.75	-3.32	-0.62	2.77

# Scale: Specific self-efficacy: geometry (continued)

Full AES sample

#### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2
	3499	28	.000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	59	6	.000
Strong invariance (plus equal intercepts)	2400	6	.000
Strict invariance (plus equal error variances)	320	6	.000

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.999
French vs. Italian language version	.997
Italian vs. German language version	.993

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	.993
Language: Italian	.988

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. selfeffc\_fs 0.0 0.9 -2.5 1.5 21875 Share of cases with imputed missing values: 51.3% (Equivalence of scores from robust MLMV: CD = .995) (Equivalence of scores from two-step approach: CD = .965)

# Scale: Specific self-efficacy: probability

Full AES sample

## Model and Fit Statistics

# Reliability and Dimensionality

Mo	odel and Fit Statistics				Reliability and	Dimensionality	
1)	<b>Likelihood-ratio tests</b> Model vs. saturated	chi2	df	p > chi2	Ordinal Cronbac	•	.917
	Baseline vs. saturated	1326 63299	2 6	.000	(Cronbach's alpha McDonald's Ome	3 7.	.917
2)	2) Root mean squared error (RMSEA) .174 90% Confidence interval: lower bound .166				Test of (one-)dimensionality (parallel analysis) Criterion: retain factors with adj. eigenvalue > o		
	90% Confidence interval: Probability RMSEA <= 0.05	opper boom	u	.182 .000	Factor 1 Factor 2	djusted eigenvalu 2.86 .01	=
3)	Akaike's Information Crite	rion (AIC)		178726	Factor 3	09	
	Bayesian Information Crite	erion (BIC)		178821	Factor 4	10	
4)	Baseline comparison						
	Comparative Fit Index (CFI)			.979			
	Tucker–Lewis Index (TLI)			.937			
5)	Size of residuals						
	Stand. root mean squared re	esidual (SRN	⁄IR)	.022			
	Coefficient of determination	n (CD)		.919			

# Standardized factor loadings

# Item descriptives

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
selfeff13	0.87	.002	0.86	0.87	selfeff13	2.7	1.0	1	4	21778
selfeff14	0.84	.002	0.83	0.84	selfeff14	2.6	1.0	1	4	10754
selfeff15	0.89	.002	0.88	0.89	selfeff15	2.8	0.9	1	4	21776
selfeff16	0.83	.003	0.83	0.84	selfeff16	2.5	0.9	1	4	10751

# Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut <sub>1</sub>	Cut <sub>2</sub>	Cut <sub>3</sub>
selfeff13	3.46	-4.44	-0.85	2.41
selfeff14	3.65	-4.88	-0.67	3.58
selfeff15	3.96	-5.27	-1.24	2.74
selfeff16	3.51	-4.69	-0.45	3.96

# Scale: Specific self-efficacy: probability (continued)

Full AES sample

#### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2
	118	28	.000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	11	6	.102
Strong invariance (plus equal intercepts)	42	6	.000
Strict invariance (plus equal error variances)	21	6	.002

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	1.000
French vs. Italian language version	1.000
Italian vs. German language version	1.000

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	1.000
Language: Italian	1.000

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. selfeffd\_fs 0.0 0.9 -2.0 1.7 21858

Share of cases with imputed missing values: 51.2% (Equivalence of scores from robust MLMV: CD = .997) (Equivalence of scores from two-step approach: CD = .986)

# Scale: Mathematics anxiety

Maths sample-split

#### **Model and Fit Statistics**

Reliability and Dimensionality
--------------------------------

					,	,			
1)	Likelihood-ratio tests	chi2	df	p > chi2	Ordinal Cronba	ch's Alpha	.914		
	Model vs. saturated	1904	5	.000	(Cronbach's alph	na = .877)			
	Baseline vs. saturated	37885	10	.000	McDonald's Om	nega	.914		
2)	Root mean squared error (I	RMSEA)		.186	.186 Test of (one-)dimensionality (paral				
	90% Confidence interval:		.179	Criterion: retain	factors with adj. ei	genvalue > o			
	90% Confidence interval:	upper bound		.193	,	Adjusted eigenvalu	e		
	Probability RMSEA <= 0.05			.000	Factor 1	3.35			
					Factor 2	.10			
3)	Akaike's Information Criter	ion (AIC)		114426	Factor 3	03			
	<b>Bayesian Information Crite</b>	rion (BIC)		114535	Factor 4	10			
					Factor 5	12			
4)	Baseline comparison								
	Comparative Fit Index (CFI)			.950					
	Tucker–Lewis Index (TLI)			.900					
5)	Size of residuals								
٥,	Stand. root mean squared re	sidual (SRMF	۲)	.035					
	Coefficient of determination		•	.916					
		• •							

# Standardized factor loadings

# Item descriptives

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
anxmath1	0.83	.004	0.82	0.84	anxmath1	2.4	1.0	1	4	10999
anxmath2	0.79	.004	0.79	0.80	anxmath2	1.9	0.9	1	4	10996
anxmath3	0.84	.004	0.83	0.85	anxmath3	1.8	0.9	1	4	10992
anxmath4	0.80	.004	0.79	0.81	anxmath4	2.5	1.0	1	4	10995
anxmath5	o.86	.003	0.85	o.86	anxmath5	2.1	1.0	1	4	10994

# Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut <sub>1</sub>	Cut <sub>2</sub>	Cut <sub>3</sub>
anxmath1	2.83	-2.61	0.30	3.40
anxmath2	2.48	-0.85	1.87	4.52
anxmath3	2.94	-0.26	2.70	5.39
anxmath4	2.50	-2.49	-0.24	2.32
anxmath5	3.11	-1.59	1.60	4.41

# Scale: Mathematics anxiety (continued)

Maths sample-split

#### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2
	1137	40	.000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	141	8	.000
Strong invariance (plus equal intercepts)	502	8	.000
Strict invariance (plus equal error variances)	151	8	.000

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.998
French vs. Italian language version	.995
Italian vs. German language version	.988

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	.999
Language: Italian	.980

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. anxmath\_fs 0.0 0.9 -1.6 2.3 11005

Share of cases with imputed missing values: 0.2% (Equivalence of scores from robust MLMV: CD = .999) (Equivalence of scores from two-step approach: CD = .976)

Scale.	<b>Mathematics</b>	horedom
Julie.	Mathematics	DOICUOIII

Maths sample-split

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	689	2	.000
	Baseline vs. saturated	20215	6	.000
2)	Root mean squared error (	RMSEA)		.178
	90% Confidence interval:	lower bound		.167
	90% Confidence interval:	upper bound		.189
	Probability RMSEA <= 0.05			.000
3)	Akaike's Information Crite	rion (AIC)		125128
	Bayesian Information Crite	erion (BIC)		125216
4)	Baseline comparison			
	Comparative Fit Index (CFI)			.966

# **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.863
(Cronbach's alpha = .831)	
McDonald's Omega	.863

#### Test of (one-)dimensionality (parallel analysis)

Criterion: retain factors with adj. eigenvalue > o Adjusted eigenvalue

Factor 1	2.34
Factor 2	.02
Factor 3	11
Factor 4	15

Tucker-Lewis Index (TLI) .898

#### 5) Size of residuals

Stand. root mean squared residual (SRMR) .032 Coefficient of determination (CD) .863

# Standardized factor loadings

#### Item descriptives

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
boredom1	0.78	.005	0.77	0.79	boredom1	2.9	1.3	1	5	10877
boredom2	0.78	.005	0.77	0.79	boredom2	2.6	1.2	1	5	10834
boredom3	0.80	.005	0.79	0.81	boredom3	2.5	1.3	1	5	10813
boredom4	0.77	.005	0.76	0.78	boredom4	3.0	1.3	1	5	10877

# Scale: Mathematics boredom (continued)

Maths sample-split

#### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2
	815	28	.000
Tests of measurement invariance Metric invariance (equal factor loadings) Strong invariance (plus equal intercepts)	chi2	df	p > chi2
	15	6	.022
	599	6	.000
Strict invariance (plus equal error variances)	166	6	.000

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.999
French vs. Italian language version	.997
Italian vs. German language version	.999

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	.999
Language: Italian	.995

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. boredom\_fs 0.0 0.9  $^{-1.5}$  1.9 10902 Share of cases with imputed missing values: 1.1% (Equivalence of scores from robust MLMV: CD = .998)

# Scale: Mathematics anger

Model vs. saturated

Maths sample-split

.895

#### **Model and Fit Statistics**

1) Likelihood-ratio tests

Ordinal Cronbach's Alpha	
(Cronbach's alpha = .865)	

**Reliability and Dimensionality** 

	Baseline vs. saturated	27251	6	.000
2)	Root mean squared error 90% Confidence interval: 90% Confidence interval: Probability RMSEA <= 0.05	lower bound upper bound		.059 .049 .071 .073

chi2

79

df

2

p > chi2

.000

Factor 3

Factor 4

McDonald's Omega	.897
Test of (one-)dimensionality (par	allel analysis)

3) Akaike's Information Criterion (AIC) 120644 **Bayesian Information Criterion (BIC)** 120732

Criterion: retai	n factors with adj. eigen	value > o
	Adjusted eigenvalue	
Factor 1	2.66	
Factor 2	- 05	

-.08

-.09

# 4) Baseline comparison

Comparative Fit Index (CFI) Tucker-Lewis Index (TLI) .992

.997

#### 5) Size of residuals

Stand. root mean squared residual (SRMR) .010 Coefficient of determination (CD) .915

Standardized factor loadings	Item descriptive
Standardized factor loadings	Item descriptive

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
anger1	0.70	.005	0.69	0.71	anger1	2.6	1.2	1	5	10891
anger2	0.89	.003	0.89	0.90	anger2	2.4	1.3	1	5	10815
anger3	0.89	.003	o.88	0.89	anger3	2.5	1.3	1	5	10810
anger4	0.82	.004	0.82	0.83	anger4	2.5	1.4	1	5	10869

# Scale: Mathematics anger (continued)

Maths sample-split

#### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2
	1045	28	.000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	52	6	.000
Strong invariance (plus equal intercepts)	264	6	.000
Strict invariance (plus equal error variances)	48	6	.000

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.998
French vs. Italian language version	.997
Italian vs. German language version	.998

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	.999
Language: Italian	.996

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. anger\_fs 0.0 0.9  $^{-1.4}$  2.1 10902 Share of cases with imputed missing values: 1.1% (Equivalence of scores from robust MLMV: CD = .999)

# Scale: Mathematics enjoyment

Maths sample-split

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	191	2	.000
	Baseline vs. saturated	23069	6	.000
2)	Root mean squared error (F	RMSEA)		.093

2)	Root mean squared error (RMSEA)		
	90% Confidence interval:	lower bound	.082
	90% Confidence interval:	upper bound	.104
	Probability RMSEA <= 0.05	<u></u>	.000

3)	Akaike's Information Criterion (AIC)	114281
	Bayesian Information Criterion (BIC)	114369

# 4) Baseline comparison

Comparative Fit Index (CFI)	.992
Tucker–Lewis Index (TLI)	.975

#### 5) Size of residuals

Stand. root mean squared residual (SRMR)	.014
Coefficient of determination (CD)	.892

# **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.877
(Cronbach's alpha = .845)	
McDonald's Omega	.879

#### Test of (one-)dimensionality (parallel analysis)

Criterion: retain factors with adj. eigenvalue > o Adjusted eigenvalue

, lajostea eigenitaise	
2.47	
04	
09	
11	
	2.47 04 09

Item descriptives

# Standardized factor loadings

							Std.		
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.
enjoymath1	o.86	.004	o.86	0.87	enjoymath1	2.5	1.2	1	5
enjoymath2	0.86	.004	o.86	0.87	enjoymath2	2.5	1.2	1	5
enjoymath3	0.73	.005	0.72	0.74	enjoymath3	2.3	1.2	1	5
enjoymath4	0.75	.005	0.74	0.76	enjoymath4	2.3	1.1	1	5

List of scales (wave 0)

Valid

Obs.

10880

10830

10882

10823

# Scale: Mathematics enjoyment (continued)

Maths sample-split

#### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2 333	df 28	p > chi2 .000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	44	6	.000
Strong invariance (plus equal intercepts)	152	6	.000
Strict invariance (plus equal error variances)	40	6	.000

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.998
French vs. Italian language version	1.000
Italian vs. German language version	.998

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	.999
Language: Italian	.998

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. enjoymath\_fs 0.0 0.9 -1.4 2.5 10907 Share of cases with imputed missing values: 1.0% (Equivalence of scores from robust MLMV: CD = .999)

Scale: Perseverance Full AES sample

#### **Model and Fit Statistics**

# **Reliability and Dimensionality**

1)	<b>Likelihood-ratio tests</b> Model vs. saturated  Baseline vs. saturated	chi2 0 18182	df o 3	p > chi2 .000	Ordinal Cronbac (Cronbach's alpha McDonald's Ome	a = .731)
2)	Root mean squared error (I			.000	Test of (one-)din	
	90% Confidence interval:			.000		djusted eig
	Probability RMSEA <= 0.05	о <b>рре</b> г 200а		1.000	Factor 1	1.4
3)	Akaike's Information Criter	ion (AIC)		168695	Factor 2 Factor 3	C 2
	Bayesian Information Crite	rion (BIC)		168767		
4)	Baseline comparison					
"	Comparative Fit Index (CFI)			1.000		
	Tucker–Lewis Index (TLI)			1.000		
5)	Size of residuals					

Test of (one-)d	limensionality (parallel analysis)
Criterion: retail	n factors with adj. eigenvalue > o
	Adjusted eigenvalue
Eactora	1 /2

.767

.775

Factor 1	1.43
Factor 2	09
Factor 3	20

# Standardized factor loadings

Stand. root mean squared residual (SRMR)

Coefficient of determination (CD)

# Item descriptives

	_						Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
persev1	0.67	.005	0.66	0.68	persev1	3.5	0.9	1	5	22268
persev2	0.87	.005	0.86	0.88	persev2	3.4	1.0	1	5	22269
perseva	0.64	.005	0.63	0.65	perseva	2.9	1.0	1	5	22265

.000

.825

#### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2
	2678	18	.000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	79	4	.000
Strong invariance (plus equal intercepts)	1498	4	.000
Strict invariance (plus equal error variances)	207	4	.000

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.997
French vs. Italian language version	.999
Italian vs. German language version	.994

#### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	.998
Language: French	.990
Language: Italian	.989

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. persev\_fs 0.0 0.5 -1.5 1.1 22280 Share of cases with imputed missing values: 0.1% (Equivalence of scores from robust MLMV: CD = .997)

Composite descriptives	Variable name	Mean	Std. dev.	Min.	Max.	Obs.		
	effper_comp	2.8	0.8	1	4	22265		
Share of cases with imputed r	nissing values:	0.2%						
Item descriptives	Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.		
	effper1 *	2.7	0.8	1	4	22243		
	effper4 *	2.9	0.9	1	4	22249		

<sup>\*</sup> Note: Original items from TREE1 / PISA2000

Composite descriptives			Std.			
	Variable name	Mean	dev.	Min.	Max.	Obs.
Big five: extraversion						
p. c	big5_e_comp	3.3	0.9	1	5	15915
Big five: agreeableness	big5_a_comp	2.5	0.7	1	-	15015
Big five: conscientiousness		3.5	0.7	1	5	15915
<b>g</b>	big5_c_comp	3.2	0.8	1	5	15915
Big five: neuroticism						
	big5_n_comp	2.9	0.9	1	5	15915
Big five: openness	1.					
	big5_o_comp	3.3	0.9	1	5	15915
Share of cases with imputed	missing values:	1.4%				

Std.   Valid   Valid   Indicators   Mean   dev.   Min.   Max.   obs.
big five: extraversion       bigfive1       3.1       1.1       1       5       15890       *         big five6       3.6       1.0       1       5       15851         Big five: agreeableness
bigfive1 3.1 1.1 1 5 15890 * bigfive6 3.6 1.0 1 5 15851  Big five: agreeableness
bigfive6 3.6 1.0 1 5 15851  Big five: agreeableness
bigfive6 3.6 1.0 1 5 15851  Big five: agreeableness
Big five: agreeableness
higfiyoo 22 11 1 5 15870
bigfive2 3.2 1.1 1 5 15879
bigfive7 3.3 1.0 1 5 15854 *
bigfive11 3.8 1.0 1 5 15838
Big five: conscientiousness
bigfive3 2.8 1.1 1 5 15863 *
bigfive8 3.6 o.9 1 5 15854
Big five: neuroticism
bigfive4 2.8 1.1 1 5 15875 *
bigfive9 3.0 1.1 1 5 15869
Big five: openness
bigfive5 3.0 1.4 1 5 15875 *
bigfive10 3.7 1.1 1 5 15864

<sup>\*</sup> Item category order reversed for composit calculation (see Rammstedt et al., 2007)

Locus of control Baseline survey sample

Composite descriptives	Variable name	Mann	Std.	Min	Max	Oha	
	variable flaffle	Mean	dev.	Min.	Max.	Obs.	
Internal locus of control							
	loci_comp	4.0	0.7	1	5	15833	
External locus of control							
External locos of control	loce_comp	2.5	0.9	1	5	15833	
	loce_comp	2.5	0.9	-	5	±3°33	
Share of cases with imputed	missing values	0.6%					
Share of cases with imputed	illissing values.	0.070					
Item descriptives			Std.			Valid	
icem descriptives	Indicators	Mean	dev.	Min.	Max.	Obs.	
Internal locus of control	indicators	ivieali	uev.	IVIIII.	ividX.	Obs.	
internal locus of control							
	loci1	3.9	0.9	1	5	15811	
	loci2	4.2	0.8	1	5	15812	
<b>External locus of control</b>							

5

15793

15777

loce1

loce2

2.3

2.6

1.1

# Scale: Work-related extrinsic values

Baseline survey sample

**Reliability and Dimensionality** 

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2	Ordinal Cronba	ch's Alpha	.655	
	Model vs. saturated	0	0		(Cronbach's alpl	na = .56o)		
	Baseline vs. saturated	6673	3	.000	McDonald's On	nega	.658	
2)	Root mean squared error (I	RMSEA)		.000	Test of (one-)di	mensionality (paralle	el analysis)	
	90% Confidence interval:	ower bound		.000	Criterion: Retair	n factors with adj. eige	nvalue > o	
	90% Confidence interval:	upper bound		.000		Adjusted eigenvalue *	<del>,</del>	
	Probability RMSEA <= 0.05			1.000	factor 1	.96		
					factor 2	14		
3)	Akaike's Information Criter	ion (AIC)		96617	factor 3	20		
	<b>Bayesian Information Crite</b>	rion (BIC)		96686	* No component with an			
						adjusted eigenvalu	e ≥ 1	
4)	Baseline comparison							
	Comparative Fit Index (CFI)			1.000				
	Tucker-Lewis Index (TLI)			1.000				
_\	Cina of worldwala							
5)	Size of residuals		<b>-</b> .					
	Stand. root mean squared re	esidual (SRMI	R)	.000				

Coefficient of determination (CD)

	_						Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
vawe1	0.70	0.01	0.68	0.71	vawe1	3.2	0.7	1	4	16066
vawe2	0.62	0.01	0.60	0.63	vawe2	3.7	0.6	1	4	16064
vawe4	0.56	0.01	0.54	0.58	vawe4	2.9	0.9	1	4	16065

Item descriptives

.668

# Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut1	Cut <sub>2</sub>	Cut <sub>3</sub>
vawe1	1.80	-5.36	-2.46	1.06
vawe2	1.42	-5.41	-3.92	-1.02
vawe4	1.19	-3.30	-0.98	1.39

Tests and Indices of Factorial Invariance across ...

Equalit	y of	the
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variance-covariance matrices across	Survey languages		Survey settings			Survey modes			
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
	273	18	.000	237	9	.000	19	9	.026
Tests of measurement invariance across	Surve	ey lang	guages	Survey settings			Survey modes		
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
Metric invariance (equal factor loadings)	12	4	.016	7	2	.033	1	2	.629
Strong invariance (plus equal intercepts)	86	4	.000	21	2	.000	О	2	.815
Strict invariance (plus equal error variances)	90	4	.000	6	2	.050	6	2	.043
Configural factor similarity across	Survey languages		Survey settings			Survey modes			
Tucker's congruence coefficient			TCC			TCC			TCC
	man vs. F		.997	classro		.999		eb vs.	.997
	ench vs. I		.988	unprod	ctored	333		PAP	33,
lta	lian vs. Ge	erman	.997						
Factor score equivalence: group									
specific vs. invariant models for	Surve	ey lang	guages	Surv	ey set	tings	Sur	vey m	odes
Coefficient of determination			CD			CD			CD
	Ge	rman	1.000	class	sroom	1.000		web	1.000
	F	rench	.994	unprod	ctored	.995		PAP	.988
	ļ	Italian	.977						

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. vawe\_fs 0.0 0.7 -2.8 1.2 16084 Share of cases with imputed missing values: 0.3% (Equivalence of scores from robust MLMV: CD = .996) (Equivalence of Scores from Two-Step-Approach: CD = .975)

# Scale: Work-related intrinsic values

Baseline survey sample

#### **Model and Fit Statistics**

1)	<b>Likelihood-ratio tests</b> Model vs. saturated	chi2 0	df o	p > chi2
	Baseline vs. saturated	14560	3	.000
2)	Root mean squared error ( 90% Confidence interval: 90% Confidence interval: Probability RMSEA <= 0.05	lower bound		.000 .000 .000 1.000
3)	Akaike's Information Crite Bayesian Information Crit			80533 80602

#### **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.789
(Cronbach's alpha = .705)	
McDonald's Omega	.793

#### Test of (one-)dimensionality (parallel analysis)

Criterion: Retain factors with adj. eigenvalue > o Adjusted eigenvalue

factor 1	1.52
factor 2	11
factor a	18

#### 4) Baseline comparison

Comparative Fit Index (CFI)	1.000
Tucker-Lewis Index (TLI)	1.000

#### 5) Size of residuals

Stand. root mean squared residual (SRMR)	.000
Coefficient of determination (CD)	.818

# Standardized factor loadings

# Item descriptives

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
vawi1	0.72	0.01	0.71	0.73	vawi1	3.2	0.7	1	4	16078
vawi2	0.85	0.01	0.84	0.86	vawi2	3.5	0.6	1	4	16071
vawi5	0.67	0.01	0.66	0.68	vawi5	3.5	0.6	1	4	16065

# Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut1	Cut <sub>2</sub>	Cut <sub>3</sub>
vawi1	1.83	-5.30	-2.78	0.95
vawi2	3.18	-8.88	-6.16	-0.70
vawis	1.64	-5.46	-3.70	-0.35

Tests and Indices of Factorial Invariance across ...

Equal	ity o	of the
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variance-covariance matrices across	Surv	Survey languages		Survey settings			Survey modes			
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2	
	376	18	.000	413	9	.000	32	9	.000	
Tests of measurement invariance across	Surv	Survey languages		Surve	Survey settings			Survey modes		
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2	
Metric invariance (equal factor loadings)	2	4	.727	5	2	.075	24	2	.000	
Strong invariance (plus equal intercepts)	179	4	.000	109	2	.000	1	2	.760	
Strict invariance (plus equal error variance	s) 81	4	.000	3	2	.236	5	2	.070	
Configural factor similarity across	Surv	ey lang	guages	Surve	ey set	tings	Sur	vey m	odes	
Tucker's congruence coefficient			TCC			TCC			TCC	
	German vs.	French	1.000	classro	om vs.	000	W	eb vs.	.985	
	French vs.	Italian	1.000	unprod	ctored	.999		PAP	.905	
	Italian vs. G	erman	1.000							
Factor score equivalence: group										
specific vs. invariant models for	Surv	ey lang	guages	Surve	ey set	tings	Sur	vey m	odes	
Coefficient of determination			CD			CD			CD	
	G	erman	1.000	class	sroom	1.000		web	.999	
		French	1.000	unprod	ctored	.999		PAP	.962	
		Italian	1.000							

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. vawi\_fs 0.0 0.8 -3.0 1.1 16086

Share of cases with imputed missing values: 0.2% (Equivalence of scores from robust MLMV: CD = .993) (Equivalence of Scores from Two-Step-Approach: CD = .964)

Family values Baseline survey sample

Composite descriptives	Variable name	Mean	Std. dev.	Min.	Max.	Obs.	
Share of cases with imputed	vafa_comp missing values:	3.1 0.2%	0.8	1	4	16075	
Item descriptives	Indicators	Mean	Std. dev.	Min.	Max.	Valid obs.	
	vafa1 vafa2	3·3 3.0	o.8 o.9	1	4 4	16064 16051	

### Scale: Positive attitude towards life

AES Extension Survey

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	1110	5	.000
	Baseline vs. saturated	13955	10	.000
2)	Root mean squared error	(RMSEA)		.208
	90% Confidence interval:	lower bound		.198
	90% Confidence interval:	upper bound		.218
	Probability RMSEA <= 0.05			.000
3)	Akaike's Information Crite	erion (AIC)		57850
	<b>Bayesian Information Crit</b>	erion (BIC)		57948

### **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.880
(Cronbach's alpha = .844)	
McDonald's Omega	.881

### Test of (one-)dimensionality (parallel analysis)

Criterion: Retain factors with adj. eigenvalue > o
Adjusted eigenvalue

	Aujusteu eigenvalue
factor 1	2.91
factor 2	.18
factor 3	03
factor 4	13
factor 5	11

## 5) Size of residuals

4) Baseline comparison

Comparative Fit Index (CFI)

Tucker-Lewis Index (TLI)

5120 01 105100015	
Stand. root mean squared residual (SRMR)	.050
Coefficient of determination (CD)	.893

### Standardized factor loadings

### Item descriptives

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
posl1	0.72	0.01	0.70	0.74	posl1	5.0	0.9	1	6	5106
posl <sub>2</sub>	0.84	0.01	0.83	0.85	posl2	5.4	0.9	1	6	5107
posl3	0.78	0.01	0.76	0.79	posl3	4.8	1.0	1	6	5106
posl5	0.67	0.01	0.65	0.69	posl5	4.6	1.1	1	6	5108
posl6	0.85	0.01	0.84	o.86	posl6	5.0	1.1	1	6	5103

.921

.841

Tests and Indices of Factorial Invariance across ...

Equal	lity	of	the
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variance-covariance matrices across	Survey languages			Survey settings			Survey modes		
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
	933	40	.000	1		1	146	20	.000
Tests of measurement invariance across	. Survey	lang	guages	Surve	ey set	tings	Sur	vey m	odes
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2
Metric invariance (equal factor loadings)	9	8	.385	1		1	17	4	.002
Strong invariance (plus equal intercepts)	311	8	.000	1		1	7	4	.113
Strict invariance (plus equal error variances)	282	8	.000	1		1	20	4	.001
Configural factor similarity across	Survey languages		Survey settings			Survey modes		odes	
Tucker's congruence coefficient			TCC			TCC			TCC
	German vs. Fre		.999 .998	classroo unprod		1	W	eb vs. PAP	.998
	Italian vs. Ger			'					
Factor score equivalence: group									
specific vs. invariant models for	Survey	lang	guages	Surve	ey set	tings	Sur	vey m	odes
Coefficient of determination			CD			CD			CD
	Gerr	man	1.000	class	room	1		web	1.000
	Fre	ench	1.000	unprod	tored	ı		PAP	.999
	lta	alian	.999						

### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. posl\_fs 0.0 0.6 -3.0 0.7 5114 Share of cases with imputed missing values: 0.5% (Equivalence of scores from robust MLMV: CD = .997)

### Scale: Reality-based learning

Maths sample-split

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2	Ordinal
	Model vs. saturated	129	2	.000	(Cronba
	Baseline vs. saturated	14527	6	.000	McDona
2)	Root mean squared error (F	RMSEA)		.076	Test of (

2)	Root mean squared error	(RMSEA)	.076
	90% Confidence interval:	lower bound	.065
	90% Confidence interval:	upper bound	.087
	Probability RMSEA <= 0.05	5	.000

3)	Akaike's Information Criterion (AIC)	145766
	Bayesian Information Criterion (BIC)	145853

### 4) Baseline comparison

Comparative Fit Index (CFI)	.991
Tucker–Lewis Index (TLI)	.974

#### 5) Size of residuals

Stand. root mean squared residual (SRMR)	.016
Coefficient of determination (CD)	.832

### **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.807
(Cronbach's alpha = .779)	
McDonald's Omega	.811

### Test of (one-)dimensionality (parallel analysis)

Criterion: retain factors with adj. eigenvalue > o
Adjusted eigenvalue

Factor 1	1.94
Factor 2	04
Factor 3	11
Factor 4	15

# Standardized factor loadings

	_	_	_	
Item	des	crip	tive	S

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
realref1	0.61	.007	0.60	0.63	realref1	3.8	1.5	1	6	11042
realref2	0.65	.007	0.64	0.66	realref2	3.9	1.4	1	6	10995
realref3	0.80	.005	0.79	0.81	realref3	3.7	1.5	1	6	10984
realref4	0.80	.005	0.79	0.81	realref4	4.1	1.5	1	6	11035

### Scale: Reality-based learning (continued)

Maths sample-split

### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2
	388	28	.000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	210	6	.000
Strong invariance (plus equal intercepts)	116	6	.000
Strict invariance (plus equal error variances)	78	6	.000

### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.983
French vs. Italian language version	.993
Italian vs. German language version	.998

### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	.999
Language: French	.989
Language: Italian	1.000

### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. realref\_fs 0.0 0.8 -2.1 1.6 11063

Share of cases with imputed missing values: 1.1% (Equivalence of scores from robust MLMV: CD = .998)

### Scale: Discovery / exploratory learning

Maths sample-split

### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	132	2	.000
	Baseline vs. saturated	19790	6	.000
2)	Root mean squared error	(RMSEA)		.076
	90% Confidence interval:	lower bound		.066
	90% Confidence interval:	upper bound		.088
	Probability RMSEA <= 0.05			.000
3)	Akaike's Information Crite		143687	
	Bayesian Information Crit		143775	
4)	Baseline comparison			
	Comparative Fit Index (CFI)	)		.993
	Tucker–Lewis Index (TLI)			.980
۲)	Size of residuals			

### **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.858
(Cronbach's alpha = .836)	
McDonald's Omega	.859

### Test of (one-)dimensionality (parallel analysis)

Criterion: retain factors with adj. eigenvalue > o
Adjusted eigenvalue

Factor 1	2.30
Factor 2	06
Factor 3	09
Factor 4	13

#### 5) Size of residuals

Stand. root mean squared residual (SRMR)	.013
Coefficient of determination (CD)	.867

### Standardized factor loadings Item descriptives

	3				•	•	Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
disclearn1	0.73	.005	0.72	0.74	disclearn1	3.5	1.6	1	6	11049
disclearn2	0.84	.004	0.83	0.85	disclearn2	3.5	1.5	1	6	10986
disclearn3	0.81	.004	0.80	0.82	disclearn3	3.6	1.5	1	6	11002
disclearn4	0.72	.005	0.71	0.74	disclearn4	3.7	1.5	1	6	11006

### Scale: Discovery / exploratory learning (continued)

Maths sample-split

### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2
	712	28	.000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	83	6	.000
Strong invariance (plus equal intercepts)	126	6	.000
Strict invariance (plus equal error variances)	190	6	.000

### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.985
French vs. Italian language version	.992
Italian vs. German language version	.998

### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	.993
Language: Italian	1.000

### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. disclearn\_fs 0.0 1.1 -2.3 2.1 11067
Share of cases with imputed missing values: 1.1% (Equivalence of scores from robust MLMV: CD = .998)

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Scal	e: '	$\mathbf{S}$	เลโ โ	earn	ın	П

Maths sample-split

#### **Model and Fit Statistics**

1)	<b>Likelihood-ratio tests</b> Model vs. saturated Baseline vs. saturated	chi2 5090 36459	df 9 15	p > chi2 .000 .000	Ordinal ( (Cronbac McDonal
2)	Root mean squared error (F	RMSEA)		.226	Test of (

2)	Root mean squared error (RMSEA)		.226
	90% Confidence interval:	lower bound	.221
	90% Confidence interval:	upper bound	.231
	Probability RMSEA <= 0.0	5	.000

3)	Akaike's Information Criterion (AIC)	211536
	Bayesian Information Criterion (BIC)	211668

4)	Baseline comparison	
	Comparative Fit Index (CFI)	.861
	Tucker–Lewis Index (TLI)	.768

### 5) Size of residuals

5.20 01 105.000.5	
Stand. root mean squared residual (SRMR)	.096
Coefficient of determination (CD)	012

### **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.869
(Cronbach's alpha = .849)	
McDonald's Omega	.865

### Test of (one-)dimensionality (parallel analysis)

Criterion: retain factors with adj. eigenvalue > o

	Adjusted eigenvalue	
Factor 1	3.20	
Factor 2	.48	
Factor 3	06	
Factor 4	08	
Factor 5	09	
Factor 6	13	

### Standardized factor loadings

Indicators	Coef.	(SE)	[95% Conf. i	nterval]	Indicators
comlearn1	0.54	.007	0.52	0.55	comlearn1
comlearn2	0.51	.008	0.50	0.53	comlearn2
comlearn3	0.62	.006	0.61	0.64	comlearn3
soclearn1	0.83	.004	0.83	0.84	soclearn1
soclearn2	0.88	.003	0.87	0.89	soclearn2
soclearn3	0.87	.003	0.87	0.88	soclearn3

Item	des	crip	tives
		P	

		Std.			Valid
Indicators	Mean	dev.	Min.	Max.	Obs.
comlearn1	3.8	1.5	1	6	11035
comlearn2	3.5	1.5	1	6	11009
comlearn3	3.7	1.5	1	6	10993
soclearn1	4.0	1.6	1	6	11039
soclearn2	4.3	1.5	1	6	11004
soclearn3	4.2	1.5	1	6	10990

### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2 580	df 54	p > chi2 .000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	55	10	.000
Strong invariance (plus equal intercepts)	202	10	.000
Strict invariance (plus equal error variances)	155	10	.000

### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.998
French vs. Italian language version	.997
Italian vs. German language version	.997

### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	1.000
Language: Italian	.998

### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. soccomlearn\_fs 0.0 0.8 -1.9 1.2 11065
Share of cases with imputed missing values: 1.2% (Equivalence of scores from robust MLMV: CD = .999)

### Scale: Social learning: social arrangement

Maths sample-split

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	0	0	
	Baseline vs. saturated	21585	3	.000
2)	Root mean squared error (	(RMSEA)		.000
	90% Confidence interval:	lower bound		.000
	90% Confidence interval:	upper bound		.000
	Probability RMSEA <= 0.05			1.000
3)	Akaike's Information Crite	erion (AIC)		100479
	Bayesian Information Crit	erion (BIC)		100545
4)	Baseline comparison			
	Comparative Fit Index (CFI)			1.000
	Tucker–Lewis Index (TLI)			1.000

### **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.904
(Cronbach's alpha = .882)	
McDonald's Omega	.905

### Test of (one-)dimensionality (parallel analysis)

Criterion: retain factors with adj. eigenvalue > o Adjusted eigenvalue

Factor 1	2.16
Factor 2	07
Factor 3	11

### Standardized factor loadings

5) Size of residuals

Indicators	Coef.	(SE)	[95% Conf	. interval]
soclearn1	0.85	.003	0.84	o.86
soclearn2	0.92	.003	0.92	0.93
soclearn3	0.84	.004	0.84	0.85

Stand. root mean squared residual (SRMR)

Coefficient of determination (CD)

ltem	descr	iptives

.000

.914

		Std.			Valid
Indicators	Mean	dev.	Min.	Max.	Obs.
soclearn1	4.0	1.6	1	6	11039
soclearn2	4.3	1.5	1	6	11004
soclearn3	4.2	1.5	1	6	10990

### Scale: Social learning: social arrangement (continued)

Maths sample-split

### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2
	142	18	.000
		16	
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	25	4	.000
Strong invariance (plus equal intercepts)	54	4	.000
Strict invariance (plus equal error variances)	21	4	.000

### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.999
French vs. Italian language version	.999
Italian vs. German language version	.998

### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	1.000
Language: Italian	.999

### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. soclearn\_fs 0.0 1.2  $^{-2.9}$  1.7 11060 Share of cases with imputed missing values: 1.0% (Equivalence of scores from robust MLMV: CD = .999)

### Scale: Social learning: communication

Maths sample-split

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	0	0	
	Baseline vs. saturated	9617	3	.000
2)	Root mean squared error	.000		
	90% Confidence interval:	lower bound		.000

90% Confidence interval: lower bound .000
90% Confidence interval: upper bound .000
Probability RMSEA <= 0.05 1.000

3) Akaike's Information Criterion (AIC) 111136

3) Akaike's Information Criterion (AIC) 111136
Bayesian Information Criterion (BIC) 111202

4) Baseline comparison

Comparative Fit Index (CFI) 1.000
Tucker–Lewis Index (TLI) 1.000

5) Size of residuals

Stand. root mean squared residual (SRMR) .000
Coefficient of determination (CD) .816

### **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.782
(Cronbach's alpha = .751)	
McDonald's Omega	.786

Test of (one-)dimensionality (parallel analysis)

Criterion: retain factors with adj. eigenvalue > o
Adjusted eigenvalue

Factor 1	1.47
Factor 2	10
Factor 3	18

### Standardized factor loadings

Indicators	Coef.	(SE)	[95% Conf	. interval]
comlearn1	0.70	.007	0.69	0.72
comlearn2	0.66	.007	0.65	0.68
comlearn3	0.85	.007	0.84	0.87

### Item descriptives

		Std.			Valid
Indicators	Mean	dev.	Min.	Max.	Obs.
comlearn1	3.8	1.5	1	6	11035
comlearn2	3.5	1.5	1	6	11009
comlearn3	3.7	1.5	1	6	10993

### Scale: Social learning: communication (continued)

Maths sample-split

### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2
	261	18	.000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	9	4	.070
Strong invariance (plus equal intercepts)	53	4	.000
Strict invariance (plus equal error variances)	17	4	.002

### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.999
French vs. Italian language version	.999
Italian vs. German language version	1.000

### Factor score equivalence: group specific vs. invariant models

CD
1.000
.999
1.000

### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. comlearn\_fs 0.0 0.9  $^{-2.1}$  1.8 11062 Share of cases with imputed missing values: 1.0% (Equivalence of scores from robust MLMV: CD = .999)

### Scale: Instructivist learning

Maths sample-split

### **Model and Fit Statistics**

### **Reliability and Dimensionality**

1) Likelihood-ratio test	chi2	df	p > chi2	Ordinal Cronbach's	Alpha .841
Model vs. satu	ated 4517	20	.000	(Cronbach's alpha =	.818)
Baseline vs. satu	ated 29913	28	.000	McDonald's Omega	.842
2) Root mean squared 6	error (RMSEA)		.143	Test of (one-)dimen	nsionality (parallel analysis)
90% Confidence inter	val: lower bound	d	.139	Criterion: retain fact	ors with adj. eigenvalue > o
90% Confidence inter	val: upper boun	d	.146	Adju	sted eigenvalue
Probability RMSEA <=	0.05		.000	Factor 1	3.18
				Factor 2	.36
3) Akaike's Information	Criterion (AIC)		286311	Factor 3	.21
Bayesian Information	Criterion (BIC)		286487	Factor 4	.05
				Factor 5	10
4) Baseline comparison				Factor 6	14
Comparative Fit Index	(CFI)		.850	Factor 7	14
Tucker–Lewis Index (7	LI)		.789	Factor 8	20
5) Size of residuals					
Stand. root mean squ	ared residual (SRN	/IR)	.066		
Coefficient of determine		,,,,	.848		

### Standardized factor loadings

### Item descriptives

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf.	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
instrlearn1	0.65	.006	0.64	0.67	instrlearn1	4.6	1.4	1	6	11031
instrlearn2	0.65	.007	0.63	0.66	instrlearn2	3.8	1.4	1	6	11001
instrlearn3	0.48	.008	0.47	0.50	instrlearn3	3.3	1.5	1	6	10993
instrlearn4	0.70	.006	0.69	0.71	instrlearn4	4.6	1.4	1	6	11052
replearn1	0.67	.006	0.66	0.68	replearn1	4.4	1.4	1	6	11041
replearn2	0.59	.007	0.58	0.61	replearn2	4.3	1.3	1	6	10990
replearn3	0.60	.007	0.59	0.62	replearn3	3.6	1.4	1	6	10991
replearn4	0.70	.006	0.69	0.71	replearn4	4.3	1.4	1	6	11010

### Scale: Instructivist learning (continued)

Maths sample-split

### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2
	4066	88	.000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	117	14	.000
Strong invariance (plus equal intercepts)	1511	14	.000
Strict invariance (plus equal error variances)	337	14	.000

### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.994
French vs. Italian language version	.996
Italian vs. German language version	.990

### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	.999
Language: French	.998
Language: Italian	.993

### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. instreplearn\_fs 0.0 0.8 -2.7 1.5 11069
Share of cases with imputed missing values: 1.3% (Equivalence of scores from robust MLMV: CD = .997)

### Scale: Instructivist learning: teachers instructions

Maths sample-split

#### **Model and Fit Statistics**

1)	<b>Likelihood-ratio tests</b> Model vs. saturated  Baseline vs. saturated	chi2 605 9077	df 2 6	p > chi2 .000 .000
2)	Root mean squared error ( 90% Confidence interval: 90% Confidence interval: Probability RMSEA <= 0.05	lower bound		.165 .154 .176 .000
3)	Akaike's Information Crite Bayesian Information Crite	` '		147556 147643
4)	Baseline comparison Comparative Fit Index (CFI) Tucker–Lewis Index (TLI)			·934 .801

### **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.723
(Cronbach's alpha = .683)	
McDonald's Omega	.727

### Test of (one-)dimensionality (parallel analysis)

Criterion: retain factors with adj. eigenvalue > o Adjusted eigenvalue

1.48	
.05	
12	
22	
	.05

### 5) Size of residuals

Indicators

Stand. root mean squared residual (SRMR)	.045
Coefficient of determination (CD)	.741

(SE)

### Standardized factor loadings

Coef.

		Std.			Valid
Indicators	Mean	dev.	Min.	Max.	Obs.
instrlearn1	4.6	1.4	1	6	11031

Item descriptives

	200	(0-)	[]]/ 0 00							•
instrlearn1	0.66	.008	0.65	0.68	instrlearn1	4.6	1.4	1	6	11031
instrlearn2	0.68	.008	0.67	0.70	instrlearn2	3.8	1.4	1	6	11001
instrlearn3	0.49	.009	0.47	0.51	instrlearn3	3.3	1.5	1	6	10993
instrlearn4	0.69	.008	0.67	0.70	instrlearn4	4.6	1.4	1	6	11052

[95% Conf. interval]

### Scale: Instructivist learning: teachers instructions (continued)

Maths sample-split

### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2
	2118	28	.000
Tooks of management invasions	alaia	<b>ا</b> د	منطم د م
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	49	6	.000
Strong invariance (plus equal intercepts)	466	6	.000
Strict invariance (plus equal error variances)	146	6	.000

### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.994
French vs. Italian language version	.975
Italian vs. German language version	.978

### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	.998
Language: French	.998
Language: Italian	.958

### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. instrlearn\_fs 0.0 0.8 -2.6 1.4 11064
Share of cases with imputed missing values: 1.1% (Equivalence of scores from robust MLMV: CD = .989)

Scale: Instructivist learning: repetitive practice							Maths sample-split
М	Model and Fit Statistics				Reliability and	Dimensionali	ty
1)	Likelihood-ratio tests	chi2	df	p > chi2	Ordinal Cronbac	•	·745
	Model vs. saturated Baseline vs. saturated	24 9920	2 6	.000	(Cronbach's alph <b>McDonald's O</b> m	· -	.751
2)	Root mean squared error (R			.032	Test of (one-)di		•
90% Confidence interval: lower bound 90% Confidence interval: upper bound		.021 .043	Criterion: retain factors with adj. eigenvalue > o Adjusted eigenvalue				
	Probability RMSEA <= 0.05			.996	Factor 1 Factor 2	1.58 08	
3)	Akaike's Information Criter			145662	Factor 3	10	
	Bayesian Information Crite	rion (BIC)		145750	Factor 4	16	
4)	Baseline comparison						
	Comparative Fit Index (CFI)			.998			
	Tucker–Lewis Index (TLI)			.993			
5)	Size of residuals						
	Stand. root mean squared re		R)	.008			
	Coefficient of determination	(CD)		.774			

Standardized factor loadings					Item descriptives					
	_						Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
replearn1	0.76	.007	0.75	0.78	replearn1	4.4	1.4	1	6	11041
replearn2	0.71	.007	0.70	0.72	replearn2	4.3	1.3	1	6	10990
replearn3	0.49	.009	0.48	0.51	replearn3	3.6	1.4	1	6	10991
replearn4	0.64	.007	0.63	0.66	replearn4	4.3	1.4	1	6	11010

### Scale: Instructivist learning: repetitive practice (continued)

Maths sample-split

### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2 1353	df 28	p > chi2 .000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	36	6	.000
Strong invariance (plus equal intercepts)	965	6	.000
Strict invariance (plus equal error variances)	209	6	.000

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.996
French vs. Italian language version	.999
Italian vs. German language version	.996

### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	.999
Language: French	.996
Language: Italian	.997

### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. replearn\_fs 0.0 0.9  $^{-2.8}$  1.5 11067 Share of cases with imputed missing values: 1.1% (Equivalence of scores from robust MLMV: CD = .997)

_				
C I	_	<b>C</b> -		aspect
~~		• 1	CTAM	SCHACE
<b>3</b> Cal	<b>.</b>		vateiii	aspect

Maths sample-split

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	2443	9	.000
	Baseline vs. saturated	31459	15	.000

2)	Root mean squared error	(RMSEA)	.157
	90% Confidence interval:	lower bound	.152
	90% Confidence interval:	upper bound	.162
	Probability RMSEA <= 0.09	.000	

3)	Akaike's Information Criterion (AIC)	185422
	Bayesian Information Criterion (BIC)	185553

# 4) Baseline comparison Comparative Fit Index (CFI) .923 Tucker–Lewis Index (TLI) .871

### 5) Size of residuals

Stand. root mean squared residual (SRMR) .050
Coefficient of determination (CD) .879

### **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.878
(Cronbach's alpha = .854)	
McDonald's Omega	.878

### Test of (one-)dimensionality (parallel analysis)

Criterion: retain factors with adj. eigenvalue > o

	Adjusted eigenvalue	
Factor 1	3.21	
Factor 2	.22	
Factor 3	03	
Factor 4	06	
Factor 5	13	
Factor 6	15	

### Standardized factor loadings

Indicators	Coef.	(SE)	[95% Conf	. interval]	Ir
formasp1	0.71	.006	0.70	0.73	fo
formasp2	0.72	.005	0.71	0.73	f
formasp3	0.75	.005	0.74	0.76	f
systasp1	0.74	.005	0.73	0.75	S
systasp2	0.76	.005	0.75	0.77	S
systasp3	0.75	.005	0.74	0.76	S

Item	descri	ptives
	acse	P C. 1 C.

		Std.			Valid
Indicators	Mean	dev.	Min.	Max.	Obs.
formasp1	4.3	1.3	1	6	10946
formasp2	4.1	1.3	1	6	10932
formasp3	4.4	1.2	1	6	10965
systasp1	5.0	1.2	1	6	10967
systasp2	4.7	1.2	1	6	10925
systasp3	4.7	1.2	1	6	10975

### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2 478	df 54	p > chi2 .000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	64	10	.000
Strong invariance (plus equal intercepts)	171	10	.000
Strict invariance (plus equal error variances)	45	10	.000

### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.998
French vs. Italian language version	.998
Italian vs. German language version	.996

### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	.999
Language: Italian	.998

### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. sysformasp\_fs 0.0 0.8 -3.2 1.3 11006

Share of cases with imputed missing values: 1.3% (Equivalence of scores from robust MLMV: CD = .999)

### Scale: System aspect: logical thinking

Maths sample-split

### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	0	0	
	Baseline vs. saturated	12550	3	.000
2)	Root mean squared error	(RMSEA)		.000
	90% Confidence interval:	lower bound		.000
	90% Confidence interval:	upper bound		.000
	Probability RMSEA <= 0.05			1.000
3)	Akaike's Information Crite	erion (AIC)		92905
	Bayesian Information Crit	erion (BIC)		92970
4)	Baseline comparison			
	Comparative Fit Index (CFI)	1		1.000
	Tucker–Lewis Index (TLI)			1.000
5)	Size of residuals			

### **Reliability and Dimensionality**

Ordinal Cronbach's Alpha		
(Cronbach's alpha = .792)		
McDonald's Omega	.832	

### Test of (one-)dimensionality (parallel analysis)

Criterion: retain factors with adj. eigenvalue > o Adjusted eigenvalue

	,	_
Factor 1		1.70
Factor 2		13
Factor 3		14

Standardized factor loadings

Stand. root mean squared residual (SRMR)	.000
Coefficient of determination (CD)	.833

### Item descriptives

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
systasp1	0.76	.006	0.75	0.78	systasp1	5.0	1.2	1	6	10967
systasp2	0.81	.005	0.79	0.82	systasp2	4.7	1.2	1	6	10925
systasp3	0.80	.005	0.79	0.81	systasp3	4.7	1.2	1	6	10975

### Scale: System aspect: logical thinking (continued)

Maths sample-split

### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2
	210	18	.000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	35	4	.000
Strong invariance (plus equal intercepts)	84	4	.000
Strict invariance (plus equal error variances)	13	4	.012

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.997
French vs. Italian language version	1.000
Italian vs. German language version	.996

### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	.997
Language: Italian	.995

### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. systasp\_fs 0.0 0.8 -3.1 1.0 11004
Share of cases with imputed missing values: 1.0% (Equivalence of scores from robust MLMV: CD = .999)

### Scale: System aspect: formalism

Maths sample-split

### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	0	0	
	Baseline vs. saturated	11712	3	.000
2)	Root mean squared error (	(RMSEA)		.000
	90% Confidence interval:	lower bound		.000
	90% Confidence interval:	upper bound		.000
	Probability RMSEA <= 0.05			1.000
3)	Akaike's Information Crite	rion (AIC)		97123
	Bayesian Information Crit	erion (BIC)		97189
4)	Baseline comparison			
••	Comparative Fit Index (CFI)			1.000
	Tucker Lowis Index (TLI)			1 000

### **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.821
(Cronbach's alpha = .791)	
McDonald's Omega	.821

### Test of (one-)dimensionality (parallel analysis)

Criterion: retain factors with adj. eigenvalue > o Adjusted eigenvalue

	rajostea eigenvanoe	
Factor 1	1.65	
Factor 2	14	
Factor 3	14	

Tucker–Lewis Index (TLI) 1.000

### 5) Size of residuals

Stand. root mean squared residual (SRMR) .000 Coefficient of determination (CD) .822

### Standardized factor loadings

### Item descriptives

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
formasp1	0.78	.006	0.77	0.79	formasp1	4.3	1.3	1	6	10946
formasp2	0.79	.006	0.78	0.80	formasp2	4.1	1.3	1	6	10932
formasp3	0.77	.006	0.76	0.78	formasp3	4.4	1.2	1	6	10965

### Scale: System aspect: formalism (continued)

Maths sample-split

### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2 193	df 18	p > chi2 .000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	11	4	.025
Strong invariance (plus equal intercepts)	83	4	.000
Strict invariance (plus equal error variances)	14	4	.008

### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	1.000
French vs. Italian language version	.994
Italian vs. German language version	.993

### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	1.000
Language: Italian	.985

### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. formasp\_fs 0.0 0.9  $^{-2.7}$  1.5 10992 Share of cases with imputed missing values: 0.9% (Equivalence of scores from robust MLMV: CD = 1.00)

Ccal	^.	Cak	2000	0 20	nact
Scal	e.	ЭU	ш	e as	pect

Maths sample-split

#### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	0	0	
	Baseline vs. saturated	12713	3	.000
2)	Root mean squared error	(RMSEA)		.000
	90% Confidence interval:	lower bound		.000
	90% Confidence interval:	upper bound		.000
	Probability RMSEA <= 0.05			1.000
3)	Akaike's Information Crite	erion (AIC)		100471
	Bayesian Information Crit	erion (BIC)		100537
4)	Baseline comparison			
	Comparative Fit Index (CFI)	)		1.000
	Tucker–Lewis Index (TLI)			1.000

### **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.832
(Cronbach's alpha = .806)	
McDonald's Omega	.833

### Test of (one-)dimensionality (parallel analysis)

Criterion: retain factors with adj. eigenvalue > o
Adjusted eigenvalue

	,	5	
Factor 1		1.72	
Factor 2		11	
Factor 3		16	

### 5) Size of residuals

Stand. root mean squared residual (SRMR) .000 Coefficient of determination (CD) .843

### Standardized factor loadings

### Item descriptives

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
schemasp1	0.76	.006	0.75	0.77	schemasp1	3.9	1.4	1	6	10967
schemasp2	0.76	.006	0.75	0.77	schemasp2	4.0	1.3	1	6	10926
schemasp3	0.85	.005	0.84	0.86	schemasp3	3.7	1.4	1	6	10927

### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2
	313	18	.000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	8	4	.092
Strong invariance (plus equal intercepts)	98	4	.000
Strict invariance (plus equal error variances)	25	4	.000

### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	1.000
French vs. Italian language version	1.000
Italian vs. German language version	.999

### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	.999
Language: Italian	.998

### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. schemasp\_fs 0.0 0.9 -2.4 1.8 10990 Share of cases with imputed missing values: 0.9% (Equivalence of scores from robust MLMV: CD = .999)

### Scale: Application aspect

Maths sample-split

#### **Model and Fit Statistics**

1)	<b>Likelihood-ratio tests</b> Model vs. saturated	chi2 316	df 2	p > chi2 .000
	Baseline vs. saturated	20302	6	.000
2)	Root mean squared error of 90% Confidence interval: 90% Confidence interval: Probability RMSEA <= 0.05	lower bound upper bound		.119 .109 .131 .000
<b>5</b> )	Akaike's Information Crite	erion (AIC)		120771

	,	
3)	Akaike's Information Criterion (AIC) Bayesian Information Criterion (BIC)	129471 129559

# Tucker–Lewis Index (TLI)

4) Baseline comparison

Comparative Fit Index (CFI)

5)	Size of residuals	
	Stand. root mean squared residual (SRMR)	.021
	Coefficient of determination (CD)	.866

### **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.863
(Cronbach's alpha = .839)	
McDonald's Omega	.864

### Test of (one-)dimensionality (parallel analysis)

Criterion: retain factors with adj. eigenvalue > o Adjusted eigenvalue

Factor 1	2.33
Factor 2	03
Factor 3	11
Factor 4	13

Item descriptives

### Standardized factor loadings

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
applyasp1	0.80	.005	0.79	0.81	applyasp1	4.2	1.3	1	6	10982
applyasp2	0.79	.005	0.78	0.80	applyasp2	4.6	1.3	1	6	10933
applyasp3	0.73	.005	0.72	0.74	applyasp3	3.9	1.4	1	6	10958
applyasp4	0.81	.005	0.80	0.82	applyasp4	4.3	1.3	1	6	10924

.985

.954

### Scale: Application aspect (continued)

Maths sample-split

### Tests and Indices of Factorial Invariance across Survey Languages

chi2 498	df 28	p > chi2 .000
chi2	df	p > chi2
70	6	.000
151	6	.000
53	6	.000
	chi2 70 151	498 28  chi2 df 70 6 151 6

### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.997
French vs. Italian language version	.992
Italian vs. German language version	.998

### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	.999
Language: Italian	.999

### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. applyasp\_fs 0.0 1.0 -3.0 1.6 11007 Share of cases with imputed missing values: 1.1% (Equivalence of scores from robust MLMV: CD = .999)

### Scale: Teacher: cognitive activation

Maths sample-split

### **Model and Fit Statistics**

### **Reliability and Dimensionality**

IVI	oder and Fit Statistics			Reliability and Dimensionality			
1)	Likelihood-ratio tests	chi2	df	p > chi2	Ordinal Cronbach's	<b>s Alpha</b> .873	
	Model vs. saturated	5636	20	.000	(Cronbach's alpha =	.844)	
	Baseline vs. saturated	38613	28	.000	McDonald's Omega	a .872	
2)	Root mean squared error (I	RMSEA)		.164	Test of (one-)dime	nsionality (parallel analysis)	
	90% Confidence interval:	lower bound	t t	.160	Criterion: retain fact	tors with adj. eigenvalue > o	
	90% Confidence interval:	upper boun	d	.167	Adju	usted eigenvalue	
	Probability RMSEA <= 0.05			.000	Factor 1	3.74	
					Factor 2	.52	
3)	Akaike's Information Criter	rion (AIC)		176245	Factor 3	.15	
	<b>Bayesian Information Crite</b>	rion (BIC)		176419	Factor 4	03	
					Factor 5	07	
4)	Baseline comparison				Factor 6	13	
	Comparative Fit Index (CFI)			.854	Factor 7	14	
	Tucker–Lewis Index (TLI)			.796	Factor 8	14	
5)	Size of residuals						
	Stand. root mean squared re	esidual (SRN	⁄IR)	.078			
	Coefficient of determination	n (CD)		.894			

### Standardized factor loadings

### Item descriptives

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
cogself1	0.83	.004	0.82	0.83	cogself1	2.8	0.9	1	4	10443
cogself2	0.50	.008	0.48	0.51	cogself2	2.6	0.8	1	4	10290
cogself3	0.56	.007	0.54	0.57	cogself3	2.7	0.9	1	4	10324
cogself4	0.75	.005	0.74	0.76	cogself4	2.9	0.8	1	4	10423
cogself5	0.82	.004	0.81	0.83	cogself5	2.8	0.9	1	4	10428
cogself6	0.66	.006	0.64	0.67	cogself6	2.9	0.8	1	4	10432
cogself7	0.62	.007	0.61	0.63	cogself7	2.7	0.8	1	4	10271
cogself8	0.67	.006	0.66	o.68	cogself8	2.7	0.8	1	4	10278

### Parameters of Generalized Structural Equation Model (Ordinal Logit Link)

Parameters of Gene	eranzed Str	uctural Eq	uation Mo	aei (Orai
Indicators	Coef.	Cut <sub>1</sub>	Cut <sub>2</sub>	Cut <sub>3</sub>
cogself1	2.85	-4.53	-1.62	2.57
cogself2	1.13	-2.48	-0.26	2.42
cogself3	1.29	-2.66	-0.59	2.07
cogself4	2.17	-3.98	-1.53	1.87
cogself5	2.75	-4.35	-1.27	2.61
cogself6	1.67	-3.68	-1.26	1.58
cogself7	1.56	-3.22	-0.69	2.66
cogself8	1.77	-3.44	-0.88	2.53

### Scale: Teacher: cognitive activation (continued)

Maths sample-split

### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2 943	df 88	p > chi2 .000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	46	14	.000
Strong invariance (plus equal intercepts)	495	14	.000
Strict invariance (plus equal error variances)	321	14	.000

#### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.998
French vs. Italian language version	.998
Italian vs. German language version	.996

### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	1.000
Language: Italian	.999

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. cogself\_fs 0.0 0.9 -2.8 2.3 10496 Share of cases with imputed missing values: 3.2% (Equivalence of scores from robust MLMV: CD = .998) (Equivalence of scores from two-step approach: CD = .983)

### Scale: Cogn. activation: finding solutions & arguing

Maths sample-split

### **Model and Fit Statistics**

### **Reliability and Dimensionality**

1)	Likelihood-ratio testschi2Model vs. saturated332Baseline vs. saturated19997	df 2 6	p > chi2 .000 .000	Ordinal Cronbach's A (Cronbach's alpha = .8 McDonald's Omega	•
2)	Root mean squared error (RMSEA) 90% Confidence interval: lower bo 90% Confidence interval: upper bo Probability RMSEA <= 0.05	und	.125 .114 .137 .000	Criterion: retain facto	ionality (parallel analysis) rs with adj. eigenvalue > 0 ted eigenvalue 2.3702
3)	Akaike's Information Criterion (AIC Bayesian Information Criterion (BIC	•	85451 85538	Factor 3 Factor 4	08 15
4)	Baseline comparison Comparative Fit Index (CFI) Tucker–Lewis Index (TLI)		.984 .951		
5)	Size of residuals Stand. root mean squared residual (S Coefficient of determination (CD)	SRMR)	.023 .878		

### Standardized factor loadings

### Item descriptives

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
cogself1	0.83	.004	0.83	0.84	cogself1	2.8	0.9	1	4	10443
cogself4	0.75	.005	0.74	0.76	cogself4	2.9	0.8	1	4	10423
cogself5	0.86	.004	0.85	0.86	cogself5	2.8	0.9	1	4	10428
cogself6	0.69	.006	0.68	0.71	cogself6	2.9	0.8	1	4	10432

### Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut <sub>1</sub>	Cut <sub>2</sub>	Cut <sub>3</sub>
cogself1	2.72	-4.62	-1.74	2.31
cogself4	2.19	-4.18	-1.71	1.71
cogself5	3.03	-4.91	-1.50	2.61
cogself6	1.91	-4.07	-1.49	1.53

### Scale: Cogn. activation: finding solutions & arguing (continued)

Maths sample-split

### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2
	351	28	.000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	24	6	.000
Strong invariance (plus equal intercepts)	110	6	.000
Strict invariance (plus equal error variances)	105	6	.000

### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	1.000
French vs. Italian language version	.997
Italian vs. German language version	.997

### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	1.000
Language: Italian	.995

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. cogselfa\_fs -0.1 0.9 -2.4 1.7 10467 Share of cases with imputed missing values: 0.9% (Equivalence of scores from robust MLMV: CD = .999) (Equivalence of scores from two-step approach: CD = .985)

### Scale: Cogn. activation: strategies & learning from mistakes

Maths sample-split

### **Model and Fit Statistics**

### **Reliability and Dimensionality**

1)	Likelihood-ratio tests	chi2	df	p > chi2	Ordinal Cronbach	's Alpha	.788
	Model vs. saturated	1037	2	.000	(Cronbach's alpha	= .743)	
	Baseline vs. saturated	12679	6	.000	McDonald's Ome	ga	.787
2)	Root mean squared error (I	RMSEA)		.224	Test of (one-)dim	ensionality (parall	el analysis)
	90% Confidence interval:	lower bound		.212	Criterion: retain fa	ctors with adj. eige	envalue > o
	90% Confidence interval:	upper bound		.235	Ad	ljusted eigenvalue	
	Probability RMSEA <= 0.05			.000	Factor 1	1.84	
					Factor 2	.12	
3)	Akaike's Information Criter	rion (AIC)		90475	Factor 3	18	
	Bayesian Information Crite	erion (BIC)		90562	Factor 4	17	
4)	Baseline comparison						
	Comparative Fit Index (CFI)			.918			
	Tucker–Lewis Index (TLI)			.755			
5)	Size of residuals						
	Stand. root mean squared re	esidual (SRMI	R)	.061			
	Coefficient of determination	n (CD)		.816			

Standard	lized factor	loadings

### Item descriptives

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
cogself2	0.60	.008	0.59	0.62	cogself2	2.6	0.8	1	4	10290
cogself3	0.58	.008	0.56	0.59	cogself3	2.7	0.9	1	4	10324
cogself7	0.76	.006	0.75	0.78	cogself7	2.7	0.8	1	4	10271
cogself8	0.81	.006	0.80	0.82	cogself8	2.7	0.8	1	4	10278

### Parameters of generalized structural equation model (ordinal logit link)

				•
Indicators	Coef.	Cut1	Cut <sub>2</sub>	Cut <sub>3</sub>
cogself2	1.45	-2.80	-0.35	2.62
cogself3	1.36	-2.79	-0.64	2.10
cogself7	2.13	-3.95	-0.89	3.12
coaself8	2.37	-4.28	-1.14	3.01

### Scale: Cogn. activation: strategies & learning from mistakes (continued)

Maths sample-split

### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2
	402	28	.000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	70	6	.000
Strong invariance (plus equal intercepts)	151	6	.000
Strict invariance (plus equal error variances)	124	6	.000

### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.998
French vs. Italian language version	.962
Italian vs. German language version	.975

### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	.999
Language: Italian	.936

#### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. cogselfb\_fs 0.0 0.9 -2.4 2.1 10334 Share of cases with imputed missing values: 0.9% (Equivalence of scores from robust MLMV: CD = .996) (Equivalence of scores from two-step approach: CD = .985)

### Scale: Teacher: classroom management

Maths sample-split

.882

#### **Model and Fit Statistics**

1)	<b>Likelihood-ratio tests</b> Model vs. saturated  Baseline vs. saturated	chi2 0 16993	df o 3	p > chi2	Ordinal Cronbac (Cronbach's alph McDonald's Om	a = .842)
2)	Root mean squared error (	RMSEA)		.000	Test of (one-)dir	mensionalit
	90% Confidence interval:	lower bound	I	.000	Criterion: retain	factors with
	90% Confidence interval: upper bound			.000	A	Adjusted eig
	Probability RMSEA <= 0.05			1.000	Factor 1	2.
					Factor 2	0
3)	Akaike's Information Criter	rion (AIC)		63509	Factor 3	
	Bayesian Information Crite	erion (BIC)		63574		
4)	Baseline comparison					
	Comparative Fit Index (CFI)			1.000		
	Tucker–Lewis Index (TLI)			1.000		

### **Reliability and Dimensionality**

(Cronbach's alpha = .042)	
McDonald's Omega	.883
Test of (one-)dimensionality (n	arallel analysi

#### ne-)dimensionality (parallel analysis)

retain factors with adj. eigenvalue > o Adjusted eigenvalue

Factor 1	2.02
Factor 2	09
Factor 3	12

# Standardized factor loadings

Stand. root mean squared residual (SRMR)

Coefficient of determination (CD)

5) Size of residuals

Item o	descriptives
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							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf.	interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
classman1	0.79	.005	0.78	0.80	classman1	2.4	0.9	1	4	10313
classman2	0.85	.004	0.84	0.85	classman2	2.4	0.9	1	4	10295
classman3	0.90	.004	0.89	0.90	classman3	2.3	0.9	1	4	10272

.000

.892

Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut <sub>1</sub>	Cut <sub>2</sub>	Cut <sub>3</sub>
classman1	2.48	-3.02	0.54	3.83
classman2	3.05	-3.19	0.28	4.06
classman3	3.96	-3.53	0.98	5.59

# Scale: Teacher: classroom management (continued)

Maths sample-split

### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2
	267	18	.000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	6	4	.169
Strong invariance (plus equal intercepts)	58	4	.000
Strict invariance (plus equal error variances)	13	4	.010

### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	1.000
French vs. Italian language version	.999
Italian vs. German language version	.999

### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	1.000
Language: Italian	.999

### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. classman\_fs 0.0 0.9 -1.7 2.0 10343

Share of cases with imputed missing values: 0.9% (Equivalence of scores from robust MLMV: CD = .999) (Equivalence of scores from two-step approach: CD = .992)

# Scale: Teacher: individual learning support

Maths sample-split

### **Model and Fit Statistics**

Reliability and Dimensional	ity
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	oder and the Statistics			Renability and Dillie	sionancy
1)	Likelihood-ratio tests chi2  Model vs. saturated 121	df 5	p > chi2 .000	Ordinal Cronbach's Alp (Cronbach's alpha = .90)	333
	Baseline vs. saturated 42736	10	.000	McDonald's Omega	.935
2)	Root mean squared error (RMSEA) 90% Confidence interval: lower bound 90% Confidence interval: upper bound Probability RMSEA <= 0.05		.047 .040 .055 .730	Criterion: retain factors	nality (parallel analysis) with adj. eigenvalue > 0 d eigenvalue 3.6304
3)	Akaike's Information Criterion (AIC) Bayesian Information Criterion (BIC)		94824 94932	Factor 3 Factor 4 Factor 5	06 05 06
4)	Baseline comparison Comparative Fit Index (CFI) Tucker–Lewis Index (TLI)		.997 .995		
5)	Size of residuals Stand. root mean squared residual (SRMR) Coefficient of determination (CD)	ı	.007 .936		

# Standardized factor loadings

# Item descriptives

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
indsup1	o.86	.003	0.85	0.86	indsup1	2.7	0.9	1	4	10434
indsup2	0.89	.003	0.88	0.89	indsup2	3.0	0.9	1	4	10436
indsup3	0.87	.003	0.87	0.88	indsup3	2.8	0.9	1	4	10464
indsup4	0.87	.003	0.86	0.87	indsup4	2.8	0.9	1	4	10439
indsup5	0.82	.004	0.81	0.83	indsup5	2.9	0.9	1	4	10423

# Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut <sub>1</sub>	Cut <sub>2</sub>	Cut <sub>3</sub>
indsup1	3.14	-4.44	-1.26	2.84
indsup2	3.72	-5.69	-2.62	1.91
indsup3	3.43	-4.89	-1.86	2.64
indsup4	3.29	-4.42	-1.53	2.12
indsup5	2.74	-4.43	-1.76	2.14

# Scale: Teacher: individual learning support (continued)

Maths sample-split

### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2
	515	40	.000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	35	8	.000
Strong invariance (plus equal intercepts)	196	8	.000
Strict invariance (plus equal error variances)	57	8	.000

### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.999
French vs. Italian language version	.999
Italian vs. German language version	1.000

### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	1.000
Language: Italian	1.000

### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. indsup\_fs 0.0 0.9 -2.2 1.6 10486 Share of cases with imputed missing values: 1.0% (Equivalence of scores from robust MLMV: CD = 1.00) (Equivalence of scores from two-step approach: CD = .981)

# Scale: Teacher: instruction quality

Maths sample-split

### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	0	0	
	Baseline vs. saturated	9348	3	.000

2) Root mean squared error (RMSEA) .000
90% Confidence interval: lower bound .000
90% Confidence interval: upper bound .000
Probability RMSEA <= 0.05 1.000

3) Akaike's Information Criterion (AIC) 71991
 Bayesian Information Criterion (BIC) 72056

4) Baseline comparison

Comparative Fit Index (CFI) 1.000
Tucker–Lewis Index (TLI) 1.000

5) Size of residuals

Stand. root mean squared residual (SRMR) .000
Coefficient of determination (CD) .829

# **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.765
(Cronbach's alpha = .712)	
McDonald's Omega	.780

Test of (one-)dimensionality (parallel analysis)

Criterion: retain factors with adj. eigenvalue > o Adjusted eigenvalue

Factor 1 1.47
Factor 2 -.08
Factor 3 -.18

Standardized factor loadings

Item	descri	ptives
icciii	acscii	Puves

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
instqual1	0.80	.007	0.79	0.82	instqual1	2.8	0.9	1	4	10426
instqual2	0.85	.007	0.84	0.87	instqual2	2.8	0.8	1	4	10285
instqual3	0.53	.008	0.51	0.54	instqual3	2.6	0.9	1	4	10266

Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut <sub>1</sub>	Cut <sub>2</sub>	Cut <sub>3</sub>
instqual1	2.52	-3.80	-1.28	1.99
instqual2	3.09	-4.94	-1.54	3.53
instqual3	1.15	-2.11	-0.25	2.18

# Scale: Teacher: instruction quality (continued)

Maths sample-split

### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2 432	df 18	p > chi2 .000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	31	4	.000
Strong invariance (plus equal intercepts)	310	4	.000
Strict invariance (plus equal error variances)	21	4	.000

### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.996
French vs. Italian language version	.999
Italian vs. German language version	.999

### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	.998
Language: Italian	.996

### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. instqual\_fs 0.0 0.9 -2.0 1.7 10473

Share of cases with imputed missing values: 2.6% (Equivalence of scores from robust MLMV: CD = .999) (Equivalence of scores from two-step approach: CD = .988)

### Scale: Situational interest

Maths sample-split

### **Model and Fit Statistics**

Ordinal Cronbach's Alpha	

**Reliability and Dimensionality** 

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	0	0	
	Baseline vs. saturated	11000	3	.000

Ordinal Cronbach's Alpha	.806
(Cronbach's alpha = .757)	
McDonald's Omega	.810

2) Root mean squared error (RMSEA) .000
90% Confidence interval: lower bound .000
90% Confidence interval: upper bound .000
Probability RMSEA <= 0.05 1.000

**Test of (one-)dimensionality (parallel analysis)** Criterion: retain factors with adj. eigenvalue > 0

Adjusted eigenvalue

Probability RMSEA <= 0.05 1.000

3) Akaike's Information Criterion (AIC) 76347
Bayesian Information Criterion (BIC) 76413

Factor 1 1.60 Factor 2 -.10 Factor 3 -.17

4) Baseline comparison

Comparative Fit Index (CFI)
Tucker–Lewis Index (TLI)

1.000 1.000

5) Size of residuals

Stand. root mean squared residual (SRMR) .000 Coefficient of determination (CD) .834

Standardized factor loadings

Item	descri	ptives

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
intsit1	0.75	.006	0.73	0.76	intsit1	2.6	0.9	1	4	10891
intsit2	0.68	.007	0.67	0.70	intsit2	2.3	0.9	1	4	10836
intsit3	0.86	.006	0.85	0.87	intsit3	2.4	0.9	1	4	10897

Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut1	Cut <sub>2</sub>	Cut <sub>3</sub>
intsit1	2.09	-3.06	-0.39	3.19
intsit2	1.82	-1.86	0.46	3.11
intsit3	3.24	-2.76	0.54	4.35

### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2
	801	18	.000
	1.1	16	1.1
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	282	4	.000
Strong invariance (plus equal intercepts)	61	4	.000
Strict invariance (plus equal error variances)	251	4	.000

### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.974
French vs. Italian language version	.999
Italian vs. German language version	.983

### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	.998
Language: French	.971
Language: Italian	.995

### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. intsit\_fs 0.0 0.9 -1.7 2.0 10926
Share of cases with imputed missing values: 1.2% (Equivalence of scores from robust MLMV: CD = .996) (Equivalence of scores from two-step approach: CD = .988)

# Scale: Perceived autonomy support

Maths sample-split

### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2	Ord
	Model vs. saturated	0	0		(Cro
	Baseline vs. saturated	10030	3	.000	McI
2)	Root mean squared error (F	•		.000	Tes

2) Root mean squared error (RMSEA) .000
90% Confidence interval: lower bound .000
90% Confidence interval: upper bound .000
Probability RMSEA <= 0.05 1.000

Akaike's Information Criterion (AIC) 72281 Bayesian Information Criterion (BIC) 72346

4) Baseline comparison

Comparative Fit Index (CFI) 1.000
Tucker–Lewis Index (TLI) 1.000

5) Size of residuals

Stand. root mean squared residual (SRMR) .000
Coefficient of determination (CD) .809

# **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.799
(Cronbach's alpha = .748)	
McDonald's Omega	.800

Test of (one-)dimensionality (parallel analysis)

Criterion: retain factors with adj. eigenvalue > o Adjusted eigenvalue

Factor 1 1.55
Factor 2 -.13
Factor 3 -.17

# Standardized factor loadings

ltem	deca	rrint	TIVAC
100111	ucs	יקיי	

							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
persuppauto1	0.74	.006	0.73	0.76	persuppauto1	2.7	0.9	1	4	10665
persuppauto2	0.82	.006	0.81	0.83	persuppauto2	2.9	0.9	1	4	10627
persuppauto3	0.70	.007	0.69	0.72	persuppauto3	3.0	0.8	1	4	10655

# Parameters of generalized structural equation model (ordinal logit link)

Coef.	Cut1	Cut <sub>2</sub>	Cut <sub>3</sub>
2.02	-3.46	-0.78	2.16
2.67	-4.43	-1.76	2.02
1.88	-4.13	-1.81	1.12
	2.02 2.67	2.02 -3.46 2.67 -4.43	2.02 -3.46 -0.78 2.67 -4.43 -1.76

# Scale: Perceived autonomy support (continued)

Maths sample-split

### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2
	229	18	.000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	34	4	.000
Strong invariance (plus equal intercepts)	142	4	.000
Strict invariance (plus equal error variances)	28	4	.000

### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.996
French vs. Italian language version	.994
Italian vs. German language version	.998

### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	.993
Language: Italian	.993

### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. persuppauto\_fs 0.0 0.9 -2.2 1.5 10674
Share of cases with imputed missing values: 0.5%
(Equivalence of scores from robust MLMV: CD = .999)
(Equivalence of scores from two-step approach: CD = .987)

# Scale: Perceived competence support

Maths sample-split

### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	0	0	
	Baseline vs. saturated	19504	3	.000

2) Root mean squared error (RMSEA) .000
90% Confidence interval: lower bound .000
90% Confidence interval: upper bound .000
Probability RMSEA <= 0.05 1.000

3) Akaike's Information Criterion (AIC) 61112Bayesian Information Criterion (BIC) 61178

4) Baseline comparison

Comparative Fit Index (CFI) 1.000
Tucker–Lewis Index (TLI) 1.000

5) Size of residuals

Stand. root mean squared residual (SRMR) .000
Coefficient of determination (CD) .951

# **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.888
(Cronbach's alpha = .842)	
McDonald's Omega	.892

Test of (one-)dimensionality (parallel analysis)

Criterion: retain factors with adj. eigenvalue > o

Adjusted eigenvalue

	Aujusteu eigenvalue	
Factor 1	2.09	
Factor 2	03	
Factor 3	13	

Standardized factor loadings

Indicators	Coef.	(SE)	[95% Conf	. interval]
persuppcomp1	0.97	.003	0.96	0.98
persuppcomp2	0.77	.005	0.77	0.78
persuppcomp3	0.82	.004	0.81	0.83

Item descriptives

		Std.			Valid
Indicators	Mean	dev.	Min.	Max.	Obs.
persuppcomp1	2.9	8.0	1	4	10639
persuppcomp2	2.7	0.9	1	4	10639
persuppcompa	3.0	0.8	1	4	10645

Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut <sub>1</sub>	Cut <sub>2</sub>	Cut <sub>3</sub>
persuppcomp1	4.74	-7.76	-3.07	3.05
persuppcomp2	2.29	-3.63	-0.99	2.34
persuppcomp3	2.73	-5.44	-2.51	1.35

# Scale: Perceived competence support (continued)

Maths sample-split

### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2 281	df 18	p > chi2 .000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	61	4	.000
Strong invariance (plus equal intercepts)	124	4	.000
Strict invariance (plus equal error variances)	43	4	.000

### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.998
French vs. Italian language version	.998
Italian vs. German language version	.997

### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	.998
Language: Italian	.982

### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. persuppcomp\_fs 0.0 0.9 -2.2 1.5 10665 Share of cases with imputed missing values: 0.5% (Equivalence of scores from robust MLMV: CD = .994) (Equivalence of scores from two-step approach: CD = .953)

# Scale: Perceived social relatedness

Maths sample-split

### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2	Ordinal Cronbach	's Alpha	.858
	Model vs. saturated	О	0	•	(Cronbach's alpha	= .814)	
	Baseline vs. saturated	15653	3	.000	McDonald's Omeg	ja –	.862
2)	Root mean squared error (	RMSEA)		.000	Test of (one-)dime	ensionality (paralle	el analysis)
	90% Confidence interval:	lower bound		.000	Criterion: retain fac	ctors with adj. eige	nvalue > o
	90% Confidence interval:	upper bound		.000	Ad	justed eigenvalue	
	Probability RMSEA <= 0.05			1.000	Factor 1	1.90	
					Factor 2	08	
3)	Akaike's Information Criter	rion (AIC)		69393	Factor 3	13	
	Bayesian Information Crite	erion (BIC)		69459			
4)	Baseline comparison						
	Comparative Fit Index (CFI)			1.000			
	Tucker–Lewis Index (TLI)			1.000			
5)	Size of residuals						

.000

.886

0.89

0.71

0.88

persocincl<sub>3</sub>

[95% Conf. interval]

0.88

0.69

0.86

# Standardized factor loadings

Indicators

persocincl<sub>1</sub>

persocincl<sub>2</sub>

persocincl3

Stand. root mean squared residual (SRMR)

Coef.

0.89

0.70

0.87

Coefficient of determination (CD)

Item descriptives							
		Std.			Valid		
Indicators	Mean	dev.	Min.	Max.	Obs.		
persocincl1	2.7	0.9	1	4	10635		
persocincl2	2.7	0.9	1	4	10640		

0.9

2.4

**Reliability and Dimensionality** 

# Parameters of generalized structural equation model (ordinal logit link)

(SE)

.004

.006

.004

				(
Indicators	Coef.	Cut <sub>1</sub>	Cut <sub>2</sub>	Cut <sub>3</sub>
persocincl1	3.65	-4.81	-1.00	3.37
persocincl2	1.82	-3.18	-0.78	2.11
persocincla	3.34	-2.89	0.28	4.36

List of scales (wave 0)

10632

# Scale: Perceived social relatedness (continued)

Maths sample-split

### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2
	1205	18	.000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	74	4	.000
Strong invariance (plus equal intercepts)	745	4	.000
Strict invariance (plus equal error variances)	216	4	.000

### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.993
French vs. Italian language version	.993
Italian vs. German language version	1.000

### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	1.000
Language: French	.992
Language: Italian	1.000

### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. persocincl\_fs 0.0 0.9 -1.9 1.8 10684
Share of cases with imputed missing values: 0.9% (Equivalence of scores from robust MLMV: CD = .996) (Equivalence of scores from two-step approach: CD = .987)

# Scale: Classmates' appreciation of mathematics

Maths sample-split

Model and Fit Statistics				Reliability and Di	mensionality		
1)	<b>Likelihood-ratio tests</b> Model vs. saturated  Baseline vs. saturated	chi2 0 19804	df o 3	p > chi2	Ordinal Cronbach's (Cronbach's alpha = McDonald's Omeg	: .776)	
2)	Root mean squared error (Rigo% Confidence interval: low go% Confidence interval: up Probability RMSEA <= 0.05	wer bound		.000 .000 .000 1.000	Test of (one-)dimensionality (parallel analysis) Criterion: retain factors with adj. eigenvalue > o Adjusted eigenvalue Factor 1 1.94		
3)	Akaike's Information Criterion Bayesian Information Criterion	• •		53455 53521	Factor 2 Factor 3	02 08	
4)	Baseline comparison Comparative Fit Index (CFI)			1.000			

# Standardized factor loadings

Tucker-Lewis Index (TLI)

Stand. root mean squared residual (SRMR)

Coefficient of determination (CD)

5) Size of residuals

Standardized factor loadings				Item descriptives						
							Std.			Valid
Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean	dev.	Min.	Max.	Obs.
apprmath1	0.92	.004	0.92	0.93	apprmath1	2.0	0.7	1	4	10778
apprmath2	0.96	.004	0.95	0.97	apprmath2	2.0	0.7	1	4	10775
apprmath3	0.53	.007	0.51	0.54	apprmath3	2.7	0.8	1	4	10776

1.000

.000

.946

### Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut <sub>1</sub>	Cut <sub>2</sub>	Cut <sub>3</sub>
apprmath1	4.34	-2.78	3.80	8.49
apprmath2	4.83	-2.94	4.63	9.65
apprmath3	1.14	-2.82	-0.55	2.41

# Scale: Classmates' appreciation of mathematics (continued)

Maths sample-split

### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2			
	320	9	.000			
Tests of measurement invariance	chi2	df	p > chi2			
Metric invariance (equal factor loadings)	13	2	.001			
Strong invariance (plus equal intercepts)	67	2	.000			
Strict invariance (plus equal error variances)	5	2	.082			
Configural factor similarity						
Tucker's Congruence Coefficient	TCC					
German vs. French language version	1.000					
French vs. Italian language version						
Italian vs. German language version						
Factor score equivalence: group specific vs. inva	ariant mode	els				
Coefficient of determination	CD					
Language: German	.999					
Language: French/ Italian	.991					
* Note: Due to sparse tables for the italian version of the scale, equivalence tests failed to						

converge and were reestimated with collapsed italian and french versions.

### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. apprmath\_fs 0.0 0.9 -1.6 2.4 10784 Share of cases with imputed missing values: 0.1% (Equivalence of scores from robust MLMV: CD = .997) (Equivalence of scores from two-step approach: CD = .980)

# Scale: Absenteeism / truancy

Full AES sample

### **Model and Fit Statistics**

1)	Likelihood-ratio tests	chi2	df	p > chi2
	Model vs. saturated	0	0	
	Baseline vs. saturated	30122	3	.000

2)	Root mean squared error	(RMSEA)	.000
	90% Confidence interval:	lower bound	.000
	90% Confidence interval:	upper bound	.000
	Probability RMSEA <= 0.0	5	1.000

3)	Akaike's Information Criterion (AIC)	84033
	Bayesian Information Criterion (BIC)	84105

## 4) Baseline comparison

Comparative Fit Index (CFI)	1.000
Tucker–Lewis Index (TLI)	1.000

### 5) Size of residuals

Stand. root mean squared residual (SRMR)	.000
Coefficient of determination (CD)	.923

# **Reliability and Dimensionality**

Ordinal Cronbach's Alpha	.819
(Cronbach's alpha = .648)	
McDonald's Omega	.837

### Test of (one-)dimensionality (parallel analysis)

Criterion: retain factors with adj. eigenvalue > o

Adjusted eigenvalue

Std.

0.4

0.5

0.8

dev. Min. Max.

1

4

4

	rajostea eigenvaloe	
Factor 1	1.77	
Factor 2	03	
Factor 3	14	

Item descriptives

# Standardized factor loadings

\* Note: Original item from TREE1 / PISA2000

Indicators	Coef.	(SE)	[95% Conf	. interval]	Indicators	Mean
truancy1	0.84	.004	0.83	0.85	truancy1	1.1
truancy2 *	0.95	.004	0.94	0.96	truancy2 *	1.2
truancy3 *	0.56	.005	0.55	0.57	truancy3 *	1.5

Parameters of generalized structural equation model (ordinal logit link)

Indicators	Coef.	Cut1	Cut <sub>2</sub>	Cut <sub>3</sub>
truancy1	3.27	4.85	7.51	8.62
truancy2	4.63	5.31	8.79	10.99
truancy3	1.16	0.54	2.44	3.49

List of scales (wave 0)

Valid

Obs.

22242

22245

22251

### Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2 2001	df 18	p > chi2 .000
Tests of measurement invariance	chi2	df	p > chi2
Metric invariance (equal factor loadings)	38	4	.000
Strong invariance (plus equal intercepts)	734	4	.000
Strict invariance (plus equal error variances)	680	4	.000

### Configural factor similarity

Tucker's Congruence Coefficient	TCC
German vs. French language version	.999
French vs. Italian language version	.998
Italian vs. German language version	1.000

### Factor score equivalence: group specific vs. invariant models

Coefficient of determination	CD
Language: German	.997
Language: French	.988
Language: Italian	.954

### Factor score descriptives

Std.

Variable name Mean dev. Min. Max. Obs. truancy\_fs 0.0 0.7 -0.5 2.8 22254

Share of cases with imputed missing values: 0.1% (Equivalence of scores from robust MLMV: CD = .995) (Equivalence of scores from two-step approach: CD = .780)

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List of scales (wave 0)

Appendix: List of sources [232]