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# SCALING METHODOLOGY AND SCALE REPORTING IN THE TREE2 PANEL SURVEY

DOCUMENTATION OF SCALES IMPLEMENTED IN THE BASELINE SURVEY (2016)  
(Update 2023)

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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).<sup>1</sup> 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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<sup>1</sup> 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 [www.icer.unibe.ch](http://www.icer.unibe.ch) and <http://uegk-schweiz.ch/>.

<sup>2</sup> See [forsbase.unil.ch/project/study-public-overview/16165/0/](http://forsbase.unil.ch/project/study-public-overview/16165/0/).

Furthermore, the latter employed two sequentially applied survey modes (web survey and paper-and-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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<sup>3</sup> 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.<sup>6</sup> 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

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<sup>4</sup> Including two years of kindergarten.

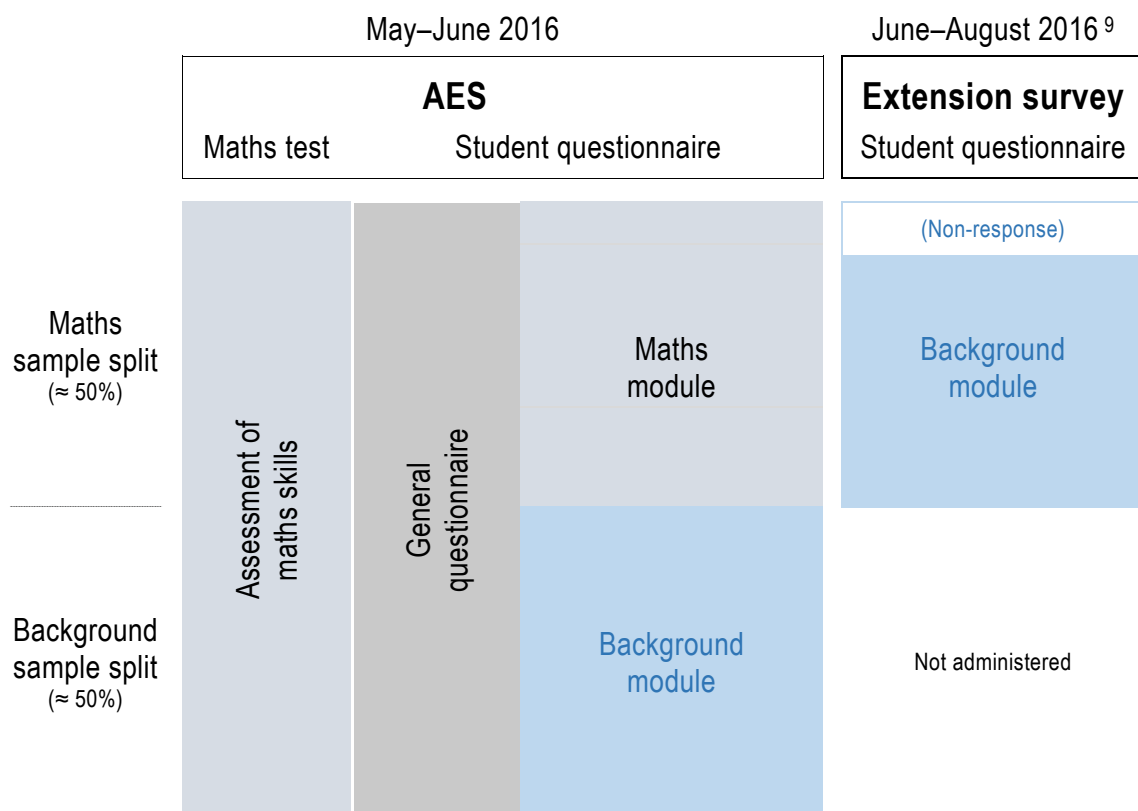
<sup>5</sup> See Verner and Helbling (2019) for a detailed description of the sampling and the population.

<sup>6</sup> 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>

Figure 1: Design of the TREE2 baseline survey



In every canton, the extension survey was carried out as soon as the AES survey had been concluded in all sampled schools.<sup>9</sup> 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

<sup>7</sup> 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.

<sup>8</sup> 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).

<sup>9</sup> 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	Extension survey <sup>1)</sup>		Total
Survey Setting:	Proctored classroom survey	Unproctored individualised setting		
Survey Mode:	CASI	Web survey	P&P questionnaire	
(Sub-)sample size <sup>2)</sup>	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>10</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 <sup>2)</sup>	Full AES sample	Baseline survey <sup>2)</sup>	Math subsample
Survey Language:			
German	16 349	11 698	8 106
French	5 235	3 927	2 646
Italian	755	618	379

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

<sup>11</sup> 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.

<sup>12</sup> 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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<sup>13</sup> 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.

<sup>14</sup> 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 Scale / composite	[Variable name] <sup>1)</sup>	AES question- naire module <sup>2)</sup>	Source <sup>3)</sup>
<b>Family background</b>			
<i>Family climate</i>			
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) <sup>4)</sup>	[famedsup_fs]	Background	PISA 2000
Social communication (PISA2000) <sup>4)</sup>	[socomm_fs]	Background	PISA 2000
Social communication (adapted TREE2)	[socomm_m_fs]	Background	PISA 2000 (adapted TREE2)
<b>Social, cultural &amp; economic resources</b>			
<i>Social capital (own)</i>			
Perceived social network support	[closupp_fs]	Background	TREE2 (BHPS, ISSP 2003)
<i>Cultural capital (family of origin)</i>			
Parents: reading interest	[joyreadp_comp]	Background	TREE2
Cultural communication (PISA2000) <sup>4)</sup>	[cultcom_fs]	Background	PISA 2000
Cultural communication (adapted TREE2)	[cultcom_m_fs]	Background	PISA 2000 (adapted TREE2)
Household possessions: classical culture (PISA2000) <sup>4)</sup>	[cultposs_fs]	Background	PISA 2000
<i>Cultural capital (own)</i>			
Embodied cultural capital	[inccap_fs]	Background	TREE2
Cultural activities <sup>5)</sup>	[cult_fs]	Background	PISA 2000 (partially adapted)

1) Student score variable names from 2023 TREE2 data release. 2) Database by module: General → full AES sample; background module → TREE2 baseline sample; math module → AES math sample split. 3) See technical appendix for a detailed list of sources. 4) Scales administered in 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 composites

Survey topic Scale or composite	[Variable name] <sup>1)</sup>	AES question- naire module <sup>2)</sup>	Source <sup>3)</sup>
<b>Social, cultural &amp; economic resources (continued)</b>			
<i>Economic capital (family of origin)</i>			
Household possessions: family wealth (PISA2000) <sup>4)</sup>	[wealth_fs]	Background	PISA 2000
Household possessions: family wealth (adapted TREE2)	[wealth_m_fs]	Background	PISA 2000 (adapted TREE2)
Family affluence scale (FASIII)	[fasiii_comp]	Background	Hobza et al., 2017
<b>Satisfaction and well-being</b>			
<i>Satisfaction</i>			
Capabilities	[cap_fs]	Background	Sen, 1985; Anand & van Hees, 2006
<i>School-related well-being</i>			
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)
<b>Non-cognitive factors</b>			
<i>Motivational concepts</i>			
Intrinsic achievement motivation	[achmoti_fs]	General	IGLU 2001
Extrinsic achievement motivation	[achmote_fs]	General	IGLU 2001
Instrumental learning motivation (PISA2000) <sup>4)</sup>	[insmot_fs]	General	PISA 2000
Interest in reading (PISA2000) <sup>4)</sup>	[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
<i>Self-perception</i>			
Global self-esteem <sup>6)</sup>	[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) <sup>4)</sup>	[scacad_fs]	General	PISA 2000
Verbal self-concept (PISA2000) <sup>4)</sup>	[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) <sup>7)</sup>	PISA 2012; Girnat, 2018
Specific self-efficacy: algebra	[selfeffb_fs]	(General) <sup>7)</sup>	PISA 2012; Girnat, 2018
Specific self-efficacy: geometry	[selfeffc_fs]	(General) <sup>7)</sup>	Girnat, 2018
Specific self-efficacy: probability	[selfeffd_fs]	(General) <sup>7)</sup>	Girnat, 2018

1) Student score variable names from 2023 TREE2 data release. 2) Database by module: General → full AES sample; background module → TREE2 baseline sample; math module → 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 Scale or composite	[Variable name] <sup>1)</sup>	AES question- naire module <sup>2)</sup>	Source <sup>3)</sup>
<b>Non-cognitive factors (continued)</b>			
<i>Emotions related to maths classes</i>			
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)
<i>Volitional strategies</i>			
Perseverance	[persev_fs]	General	PISA 2012
Effort: learning (PISA2000) <sup>4)</sup>	[effper_comp]	Background	PISA2000
<i>Personality characteristics</i>			
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)
<i>Values &amp; attitudes</i>			
Work-related extrinsic value	[wawe_fs]	Background	TREE1 - based on Watermann, 2000
Work-related intrinsic value	[wawi_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
<i>Attitudes related to mathematics classes</i>			
Reality-based learning	[realref_fs]	Math	Girnat, 2015, 2017
Discovery / exploratory learning	[disclearn_fs]	Math	Girnat, 2015, 2017
Social learning	[socomlearn_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
<b>Education and training</b>			
<i>Characteristics of maths lessons (end of lower secondary education)</i>			
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	[persupcomp_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
<i>Absenteeism / intention to change education</i>			
Absenteeism / truancy	[truancy_fs]	General	PISA 2000, PISA 2012

1) Student score variable names from 2023 TREE2 data release. 2) Database by module: General → full AES sample; background module → TREE2 baseline sample; math module → 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 <sup>1)</sup>	Subdimensions	Variable name <sup>1)</sup>
<i>Background module scales</i>			
Global self-esteem <sup>2) 3)</sup>	[sel_fs]	Positive global self-esteem <sup>4)</sup> Negative global self-esteem / depression <sup>4) 5)</sup>	[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]
<i>Math module scales</i>			
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	[socomlearn_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 <sup>7)</sup>	[cogself_fs]	Cogn. activation: finding solutions & arguing Cogn. activation: strategies and learning from mistakes	[cogselfa_fs] [cogselfb_fs]

1) 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

<sup>15</sup> 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 <sup>1)</sup>	Dimension	Variable name <sup>2)</sup>	Number of items
Big Five Inventory			
	Extraversion	[big5_e_comp]	2
	Agreeableness	[big5_a_comp]	3 <sup>3)</sup>
	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) <sup>4)</sup>	[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) <sup>FN17</sup>	[fasiii_comp]	6

1) 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).

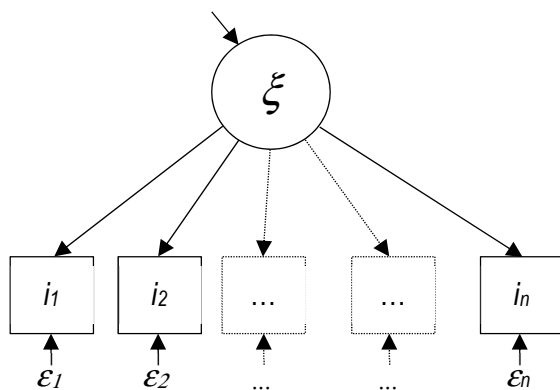
<sup>16</sup> For item composites, student scores are calculated from imputed item ratings (cf. 3.1.1 b).

<sup>17</sup> 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 ( $i_1, i_2, \dots, i_n$ ) with associated item-level measurement errors  $\varepsilon_n$ , 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.

<sup>18</sup> 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>19</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).

<sup>20</sup> 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 model-estimation 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).<sup>21</sup> 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 multi-group 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).<sup>24</sup>

<sup>21</sup> 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>).

<sup>22</sup> 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.

<sup>23</sup> Lower secondary schools in Switzerland are mostly “tracked”, that is, students are enrolled in separate programmes with varying academic requirements.

<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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<sup>25</sup> 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\_qmath'* for the scales of the maths module).

<sup>26</sup> 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).

<sup>27</sup> 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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<sup>28</sup> 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.

<sup>29</sup> Missing item values owing to explicit don’t-know answers and item non-response were imputed together.

<sup>30</sup> 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.

<sup>31</sup> 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 one-dimensional 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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<sup>32</sup> 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.

<sup>33</sup> After applying Fisher's z-transformation, we simply average the correlation matrices and transform them back (see also footnote 31).

<sup>34</sup> 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 2<sup>nd</sup> 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 a 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

<sup>36</sup> A disadvantage of this method is that many statistics to judge model fit and scale qualities are unavailable.

<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 Rosseel (2017).

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<sup>38</sup> 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 non-significant 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

<sup>39</sup> 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)

Scale: Parental pressure to achieve					Baseline survey sample					
<b>Model and Fit Statistics</b>					<b>Reliability and Dimensionality</b>					
1)	<b>Likelihood-ratio tests</b>	chiz	df	p > chiz	<b>Ordinal Cronbach's Alpha</b>	.811				
	Model vs. saturated	462	2	.000	(Cronbach's alpha = .751)					
	Baseline vs. saturated	20063	6	.000	<b>McDonald's Omega</b>	.811				
2)	<b>Root mean squared error (RMSEA)</b>				<b>Test of (one-)dimensionality (parallel analysis)</b>					
	90% Confidence interval: lower bound				Criterion: Retain factors with adj. eigenvalue > 0					
	90% Confidence interval: upper bound				Adjusted eigenvalue					
	Probability RMSEA <= 0.05				factor 1	1.95				
					factor 2	-.04				
3)	<b>Akaike's Information Criterion (AIC)</b>	142462			factor 3	-.09				
	<b>Bayesian Information Criterion (BIC)</b>	142554			factor 4	-.18				
4)	<b>Baseline comparison</b>									
	Comparative Fit Index (CFI)				.977					
	Tucker-Lewis Index (TLI)				.931					
5)	<b>Size of residuals</b>									
	Stand. root mean squared residual (SRMR)				.026					
	Coefficient of determination (CD)				.816					
<b>Standardized factor loadings</b>					<b>Item descriptives</b>					
Indicators	Coef.	(SE)	[95% Conf. interval]		Indicators	Mean	Std. dev.	Min.	Max.	Valid 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
<b>Parameters of generalized structural equation model (ordinal logit link)</b>										
Indicators	Coef.	Cut1	Cut2	Cut3						
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						

[List of scales \(wave 0\)](#)

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).<sup>41</sup>

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

<sup>41</sup> 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 (cf. [https://en.wikipedia.org/wiki/Internal\\_consistency](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>42</sup> The parallel analysis relies on the user-written "paran" package (Dinno, 2009).

<sup>43</sup> 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 multi-group 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 (continued)						Baseline survey sample					
<b>Tests and Indices of Factorial Invariance across ...</b>											
<b>Equality of the variance-covariance matrices across ...</b>											
	<b>Survey languages</b>			<b>Survey settings</b>			<b>Survey modes</b>				
	chi2	df	p > chi2	chi2	df	p > chi2	chi2	df	p > chi2		
	1717	28	.000	105	14	.000	26	14	.027		
<b>Tests of measurement invariance across ...</b>											
	<b>Survey languages</b>			<b>Survey settings</b>			<b>Survey modes</b>				
	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		
<b>Configural factor similarity across ...</b>											
<b>Tucker's congruence coefficient</b>			<b>Survey languages</b>			<b>Survey settings</b>			<b>Survey modes</b>		
			TCC			TCC			TCC		
German vs. French			.999			classroom vs. unproctored			.998		
French vs. Italian			.997						web vs. PAP		
Italian vs. German			.993						.996		
<b>Factor score equivalence: group specific vs. invariant models for ...</b>											
<b>Coefficient of determination</b>			<b>Survey languages</b>			<b>Survey settings</b>			<b>Survey modes</b>		
			CD			CD			CD		
German			1.000			classroom			1.000		
French			1.000			unproctored			.999		
Italian			.980						web		
									1.000		
									PAP		
									.990		
<b>Factor score descriptives</b>											
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)											

(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 ( $\epsilon_i$  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

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<sup>44</sup> 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 condition-specific models. Where the CD indicates that both scores share, say, 98% of their variance (i.e.,  $CD \geq .98$ ), 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

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<sup>45</sup> 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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<sup>46</sup> 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.

<sup>47</sup> The full variable names include an additional prefix to distinguish TREE2 survey waves (e.g., “t2” for the second follow-up survey).

<sup>48</sup> 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 *cogsself1\_fs*, *cogsself2\_fs*, *extregm\_fs* have been renamed according to TREE naming conventions (to *cogsselfa\_fs*, *cogsselfb\_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

([Scale names](#) linked with first page of scale-specific reporting)

### Survey topics

<i>Scale (or composite)</i>	<i>Variable Name</i>	<i>AES Module</i>	<i>Source</i>	<i>Page</i>
<b>1) Family climate</b>				
<a href="#">Emotional closeness to parents</a>	[ closep_comp ]	Background	TREE1 - based on Szydlik, 2008	41
<a href="#">Parental pressure to achieve</a>	[ press_fs ]	Background	Böhm-Kasper et al., 2000	42
<a href="#">Parents' achievement expectations</a>	[ expectp_fs ]	Math	Hascher et al., 2019	44
<a href="#">Mother's achievement expectations</a>	[ expectm_fs ]	Math	Hascher et al., 2019	46
<a href="#">Father's achievement expectations</a>	[ expectf_fs ]	Math	Hascher et al., 2019	48
<a href="#">Mother's social norms about mathematics</a>	[ socnormsm_fs ]	Math	PISA 2012	50
<a href="#">Father's social norms about mathematics</a>	[ socnormsf_fs ]	Math	PISA 2012	52
<a href="#">Family educational support (PISA2000)</a>	[ famedsup_fs ]	Background	PISA 2000	54
<a href="#">Social communication (PISA2000)</a>	[ soccom_fs ]	Background	PISA 2000	56
<a href="#">Social communication (adapted TREE2)</a>	[ soccom_m_fs ]	Background	PISA 2000 (adapted)	58
<b>2) Social capital (own)</b>				
<a href="#">Perceived social network support</a>	[ closupp_fs ]	Background	TREE2 (BHPS, ISSP 2003)	60
<b>3) Cultural capital (family of origin)</b>				
<a href="#">Parental reading interest</a>	[ joyreadp_comp ]	Background	TREE2	62
<a href="#">Cultural communication (PISA2000)</a>	[ cultcom_fs ]	Background	PISA 2000	64
<a href="#">Cultural communication (adapted TREE2)</a>	[ cultcom_m_fs ]	Background	PISA 2000 (adapted)	66
<a href="#">Household possessions: classical culture (PISA2000)</a>	[ cultposs_fs ]	Background	PISA 2000	68
<b>4) Cultural capital (own)</b>				
<a href="#">Embodied cultural capital</a>	[ inccap_fs ]	Background	TREE2	70
<a href="#">Embodied cultural capital: manners</a>	[ manners_fs ]	Background	TREE2	72
<a href="#">Embodied cultural capital: verbal skills</a>	[ verbskill_fs ]	Background	TREE2	74
<a href="#">Cultural activities</a>	[ cult_fs ]	Background	PISA 2000 (adapted)	76
<a href="#">Lowbrow cultural activities</a>	[ cultlow_fs ]	Background	TREE2	78
<a href="#">Highbrow cultural activities [PISA 2000]</a>	[ culthigh_fs ]	Background	PISA 2000	80



**Survey topics (continued)**

Baseline survey (2016)

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

**Survey topics (continued)**

Baseline survey (2016)

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

## Survey topics

Baseline survey (2016)

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

## Survey topics

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

<b>Composite descriptives</b>		Mean	Std. dev.	Min.	Max.	Obs.
Variable name						
closep_comp		4.2	0.8	1	5	15664
Share of cases with imputed missing values:		3.5%				

35

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

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	462	2	.000
Baseline vs. saturated	20063	6	.000
2) <b>Root mean squared error (RMSEA)</b>			.122
90% Confidence interval: lower bound			.113
90% Confidence interval: upper bound			.131
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			142462
<b>Bayesian Information Criterion (BIC)</b>			142554
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.977
Tucker-Lewis Index (TLI)			.931
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.026
Coefficient of determination (CD)			.816

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.811
(Cronbach's alpha = .751)	
<b>McDonald's Omega</b>	.811
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: Retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
factor 1	1.95
factor 2	-.04
factor 3	-.09
factor 4	-.18

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
press1	0.69	0.01	0.68	0.70
press2	0.69	0.01	0.68	0.71
press3	0.78	0.00	0.77	0.79
press4	0.71	0.01	0.70	0.72

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
press1	2.2	1.0	1	4	15488
press2	3.0	0.9	1	4	15491
press3	3.0	0.8	1	4	15488
press4	2.8	0.9	1	4	15490

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

Indicators	Coef.	Cut1	Cut2	Cut3
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

[List of scales \(wave 0\)](#)

Tests and Indices of Factorial Invariance across ...

Equality of the

variance-covariance matrices across ...

Survey languages

chi2	df	p > chi2
1717	28	.000

Survey settings

chi2	df	p > chi2
105	14	.000

Survey modes

chi2	df	p > chi2
26	14	.027

Tests of measurement invariance across ...

Survey languages

chi2	df	p > chi2
31	6	.000
923	6	.000
73	6	.000

Survey settings

chi2	df	p > chi2
33	3	.000
11	3	.010
12	3	.008

Survey modes

chi2	df	p > chi2
11	3	.013
4	3	.317
3	3	.413

Metric invariance (equal factor loadings)

Strong invariance (plus equal intercepts)

Strict invariance (plus equal error variances)

Configural factor similarity across ...

Survey languages

	TCC
German vs. French	.999
French vs. Italian	.997
Italian vs. German	.993

Survey settings

	TCC
classroom vs. unproctored	.998

Survey modes

	TCC
web vs. PAP	.996

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

Coefficient of determination

Survey languages

	CD
German	1.000
French	1.000
Italian	.980

Survey settings

	CD
classroom	1.000
unproctored	.999

Survey modes

	CD
web	1.000
PAP	.990

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)

[List of scales \(wave 0\)](#)

## Model and Fit Statistics

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	8040	2	.000
Baseline vs. saturated	24621	6	.000
2) <b>Root mean squared error (RMSEA)</b>			.606
90% Confidence interval: lower bound			.595
90% Confidence interval: upper bound			.617
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			77644
<b>Bayesian Information Criterion (BIC)</b>			77731
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.673
Tucker–Lewis Index (TLI)			.020
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.108
Coefficient of determination (CD)			.854

## Reliability and Dimensionality

<b>Ordinal Cronbach's Alpha</b>	.837
(Cronbach's alpha = .774)	
<b>McDonald's Omega</b>	.834
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	2.35
Factor 2	.43
Factor 3	.11
Factor 4	-.19

## Standardized factor loadings

Indicators	Coef.	(SE)	[95% Conf. interval]	
expectf2	0.70	.007	0.69	0.72
expectf3	0.85	.005	0.84	0.86
expectm2	0.63	.009	0.62	0.65
expectm3	0.79	.005	0.78	0.80

## Item descriptives

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
expectf2	3.4	0.7	1	4	10568
expectf3	3.3	0.7	1	4	10566
expectm2	3.4	0.7	1	4	10862
expectm3	3.4	0.7	1	4	10864

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

Indicators	Coef.	Cut1	Cut2	Cut3
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
expectm3	2.11	-6.40	-4.13	-0.12

[List of scales \(wave 0\)](#)



## Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	297	28	.000

<b>Tests of measurement invariance</b>	chi2	df	p > chi2
Metric invariance (equal factor loadings)	15	6	.017
Strong invariance (plus equal intercepts)	126	6	.000
Strict invariance (plus equal error variances)	12	6	.072

**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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chiz	df	p > chiz
Model vs. saturated	0	0	
Baseline vs. saturated	4828	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			67851
<b>Bayesian Information Criterion (BIC)</b>			67917
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker–Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.729

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.642
(Cronbach's alpha = .552)	
<b>McDonald's Omega</b>	.663
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	1.01
Factor 2	-.07
Factor 3	-.22

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
expectm1	0.42	.010	0.40	0.44
expectm2	0.80	.013	0.77	0.82
expectm3	0.65	.011	0.63	0.67

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
expectm1	2.8	0.8	1	4	10859
expectm2	3.4	0.7	1	4	10862
expectm3	3.4	0.7	1	4	10864

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

Indicators	Coef.	Cut1	Cut2	Cut3
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

[List of scales \(wave 0\)](#)

## Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	536	18	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chiz	df	p > chiz
Model vs. saturated	0	0	
Baseline vs. saturated	7517	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			65854
<b>Bayesian Information Criterion (BIC)</b>			65920
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker–Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.791

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.738
(Cronbach's alpha = .653)	
<b>McDonald's Omega</b>	.749
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	1.31
Factor 2	-.09
Factor 3	-.19

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
expectf1	0.55	.008	0.53	0.56
expectf2	0.83	.008	0.82	0.85
expectf3	0.72	.008	0.70	0.74

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
expectf1	2.9	0.9	1	4	10565
expectf2	3.4	0.7	1	4	10568
expectf3	3.3	0.7	1	4	10566

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

Indicators	Coef.	Cut1	Cut2	Cut3
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

[List of scales \(wave 0\)](#)

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

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	429	18	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chiz	df	p > chiz
Model vs. saturated	0	0	
Baseline vs. saturated	12780	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			66659
<b>Bayesian Information Criterion (BIC)</b>			66724
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker–Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.881

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.789
(Cronbach's alpha = .715)	
<b>McDonald's Omega</b>	.812
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
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	0.86	0.88
socnormsm2	0.89	.006	0.88	0.91
socnormsm3	0.50	.008	0.49	0.52

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid 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	Cut2	Cut3
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

[List of scales \(wave 0\)](#)

Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	195	18	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	15486	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			60431
<b>Bayesian Information Criterion (BIC)</b>			60496
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker–Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.922

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.837
(Cronbach's alpha = .771)	
<b>McDonald's Omega</b>	.851
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	1.85
Factor 2	-.04
Factor 3	-.14

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
socnormsf1	0.95	.004	0.94	0.96
socnormsf2	0.85	.005	0.84	0.86
socnormsf3	0.60	.007	0.59	0.62

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid 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.	Cut1	Cut2	Cut3
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

[List of scales \(wave 0\)](#)



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

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	198	18	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	16654	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			147278
<b>Bayesian Information Criterion (BIC)</b>			147347
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker-Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.861

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.785
(Cronbach's alpha = .746)	
<b>McDonald's Omega</b>	.803
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: Retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
factor 1	1.60
factor 2	-.07
factor 3	-.16

**Standardized factor loadings**

Indicators *	Coef.	(SE)	[95% Conf. interval]	
famedsup1	0.88	0.01	0.87	0.89
famedsup2	0.85	0.01	0.84	0.86
famedsup3	0.53	0.01	0.51	0.54

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

**Item descriptives**

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

[List of scales \(wave 0\)](#)

Tests and Indices of Factorial Invariance across ...

Equality of the

variance-covariance matrices across ...

Survey languages

chi2	df	p > chi2
365	18	.000

Survey settings

chi2	df	p > chi2
101	9	.000

Survey modes

chi2	df	p > chi2
34	9	.000

Tests of measurement invariance across ...

Survey languages

chi2	df	p > chi2
20	4	.001
300	4	.000
12	4	.015

Survey settings

chi2	df	p > chi2
9	2	.013
32	2	.000
18	2	.000

Survey modes

chi2	df	p > chi2
11	2	.005
11	2	.003
2	2	.324

Metric invariance (equal factor loadings)

Strong invariance (plus equal intercepts)

Strict invariance (plus equal error variances)

Configural factor similarity across ...

Survey languages

	TCC
German vs. French	.998
French vs. Italian	.999
Italian vs. German	.999

Survey settings

	TCC
classroom vs. unproctored	.999

Survey modes

	TCC
web vs. PAP	.994

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

Survey languages

	CD
German	1.000
French	.999
Italian	.997

Survey settings

	CD
classroom	1.000
unproctored	.998

Survey modes

	CD
web	1.000
PAP	.996

Factor score descriptives

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	9734	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			124277
<b>Bayesian Information Criterion (BIC)</b>			124346
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker-Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.750

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.723
(Cronbach's alpha = .647)	
<b>McDonald's Omega</b>	.729
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: Retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
factor 1	1.24
factor 2	-.11
factor 3	-.20

**Standardized factor loadings**

Indicators *	Coef.	(SE)	[95% Conf. interval]	
socom1	0.57	0.01	0.56	0.58
socom2	0.71	0.01	0.69	0.72
socom3	0.78	0.01	0.76	0.79

\* **Note:** Replication of 'Socom'-Scale from TREE1 / PISA2000

**Item descriptives**

Indicators *	Mean	Std. dev.	Min.	Max.	Valid Obs.
socom1	3.9	1.1	1	5	15566
socom2	4.6	0.9	1	5	15570
socom3	4.0	1.1	1	5	15555

[List of scales \(wave 0\)](#)

Tests and Indices of Factorial Invariance across ...

Equality of the

variance-covariance matrices across ...

Survey languages		
chi2	df	p > chi2
626	18	.000

Survey settings		
chi2	df	p > chi2
611	9	.000

Survey modes		
chi2	df	p > chi2
20	9	.017

Tests of measurement invariance across ...

Survey languages		
chi2	df	p > chi2
31	4	.000
228	4	.000
92	4	.000

Survey settings		
chi2	df	p > chi2
26	2	.000
107	2	.000
201	2	.000

Survey modes		
chi2	df	p > chi2
9	2	.012
3	2	.231
3	2	.258

Configural factor similarity across ...

Tucker's congruence coefficient

Survey languages		TCC
German vs. French		1.000
French vs. Italian		.992
Italian vs. German		.988

Survey settings		TCC
classroom vs. unproctored		.995

Survey modes		TCC
web vs. PAP		.986

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

Coefficient of determination

Survey languages		CD
German		1.000
French		1.000
Italian		.973

Survey settings		CD
classroom		.998
unproctored		.980

Survey modes		CD
web		.999
PAP		.925

Factor score descriptives

Variable name	Mean	Std. dev.	Min.	Max.	Obs.
socom_fs	0.0	0.5	-2.1	0.5	15588
Share of cases with imputed missing values:					0.4%
(Equivalence of scores from robust MLMV: CD = .986)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	26651	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			119342
<b>Bayesian Information Criterion (BIC)</b>			119411
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker-Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.890

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.889
(Cronbach's alpha = .851)	
<b>McDonald's Omega</b>	.889
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: Retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
factor 1	2.06
factor 2	-.11
factor 3	-.11

**Standardized factor loadings**

Indicators *	Coef.	(SE)	[95% Conf. interval]	
socom3 **	0.84	0.00	0.84	0.85
socom4	0.86	0.00	0.85	0.86
socom5	0.86	0.00	0.86	0.87

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

\*\* **Note:** Original item from TREE1 / PISA2000

**Item descriptives**

Indicators *	Mean	Std. dev.	Min.	Max.	Valid Obs.
socom3 **	4.0	1.1	1	5	15555
socom4	3.9	1.2	1	5	15560
socom5	4.0	1.1	1	5	15563

[List of scales \(wave 0\)](#)

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

## Equality of the

## variance-covariance matrices across ...

## Survey languages

chi2	df	p > chi2
942	18	.000

## Survey settings

chi2	df	p > chi2
159	9	.000

## Survey modes

chi2	df	p > chi2
49	9	.000

## Tests of measurement invariance across ...

## Survey languages

chi2	df	p > chi2
------	----	----------

Metric invariance (equal factor loadings)

50	4	.000
----	---	------

Strong invariance (plus equal intercepts)

129	4	.000
-----	---	------

Strict invariance (plus equal error variances)

211	4	.000
-----	---	------

## Survey settings

chi2	df	p > chi2
------	----	----------

5	2	.094
---	---	------

37	2	.000
----	---	------

19	2	.000
----	---	------

## Survey modes

chi2	df	p > chi2
------	----	----------

2	2	.459
---	---	------

2	2	.408
---	---	------

6	2	.041
---	---	------

## Configural factor similarity across ...

## Survey languages

TCC

German vs. French .999

French vs. Italian .999

Italian vs. German .997

## Survey settings

TCC

classroom vs.

unproctored 1.000

## Survey modes

TCC

web vs.

PAP 1.000

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

## Survey languages

CD

German 1.000

French 1.000

Italian .997

## Survey settings

CD

classroom 1.000

unproctored 1.000

## Survey modes

CD

web 1.000

PAP 1.000

## Factor score descriptives

Std.

Variable name	Mean	dev.	Min.	Max.	Obs.
socomm_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)

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	2147	5	.000
Baseline vs. saturated	58182	10	.000
2) <b>Root mean squared error (RMSEA)</b>			.169
90% Confidence interval: lower bound			.163
90% Confidence interval: upper bound			.175
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			233311
<b>Bayesian Information Criterion (BIC)</b>			233425
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.963
Tucker-Lewis Index (TLI)			.926
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.035
Coefficient of determination (CD)			.939

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.920
(Cronbach's alpha = .896)	
<b>McDonald's Omega</b>	.920
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: Retain factors with adj. eigenvalue > 0	
	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	0.88
closupp4	0.68	0.00	0.67	0.69
closupp5	0.86	0.00	0.86	0.87

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid 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

[List of scales \(wave 0\)](#)



Tests and Indices of Factorial Invariance across ...

Equality of the

variance-covariance matrices across ...

Survey languages

chi2	df	p > chi2
635	40	.000

Survey settings

chi2	df	p > chi2
802	20	.000

Survey modes

chi2	df	p > chi2
105	20	.000

Tests of measurement invariance across ...

Survey languages

chi2	df	p > chi2
33	8	.000
205	8	.000
291	8	.000

Survey settings

chi2	df	p > chi2
87	4	.000
219	4	.000
17	4	.002

Survey modes

chi2	df	p > chi2
8	4	.075
13	4	.014
26	4	.000

Metric invariance (equal factor loadings)

Strong invariance (plus equal intercepts)

Strict invariance (plus equal error variances)

Configural factor similarity across ...

Survey languages

	TCC
German vs. French	1.000
French vs. Italian	.999
Italian vs. German	.999

Survey settings

	TCC
classroom vs. unproctored	.998

Survey modes

	TCC
web vs. PAP	.999

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

Survey languages

	CD
German	1.000
French	1.000
Italian	1.000

Survey settings

	CD
classroom	1.000
unproctored	.999

Survey modes

	CD
web	1.000
PAP	1.000

Coefficient of determination

Factor score descriptives

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Composite descriptives**

Variable name	Mean	Std. dev.	Min.	Max.	Obs.
joyreadp_comp	3.1	0.8	1	4	15244
Share of cases with imputed missing values: (Including "don't know"-answers for one parent)		8.7%			

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid obs.
joyreadm	3.4	0.9	1	4	15004
joyreadf	2.9	1.1	1	4	14164

[List of scales \(wave 0\)](#)



**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	8034	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			146251
<b>Bayesian Information Criterion (BIC)</b>			146320
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker-Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.727

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.677
(Cronbach's alpha = .606)	
<b>McDonald's Omega</b>	.690
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: Retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
factor 1	1.11
factor 2	-.10
factor 3	-.21

**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**

Indicators *	Mean	Std. dev.	Min.	Max.	Valid 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

[List of scales \(wave 0\)](#)

Tests and Indices of Factorial Invariance across ...

Equality of the

variance-covariance matrices across ...

Survey languages

chi2	df	p > chi2
369	18	.000

Survey settings

chi2	df	p > chi2
267	9	.000

Survey modes

chi2	df	p > chi2
42	9	.000

Tests of measurement invariance across ...

Survey languages

chi2	df	p > chi2
16	4	.003
263	4	.000
30	4	.000

Survey settings

chi2	df	p > chi2
8	2	.019
141	2	.000
15	2	.001

Survey modes

chi2	df	p > chi2
1	2	.673
14	2	.001
13	2	.002

Metric invariance (equal factor loadings)

Strong invariance (plus equal intercepts)

Strict invariance (plus equal error variances)

Configural factor similarity across ...

Survey languages

	TCC
German vs. French	.998
French vs. Italian	.987
Italian vs. German	.996

Survey settings

	TCC
classroom vs. unproctored	.999

Survey modes

	TCC
web vs. PAP	.998

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

Survey languages

	CD
German	.999
French	.996
Italian	.970

Survey settings

	CD
classroom	1.000
unproctored	.999

Survey modes

	CD
web	1.000
PAP	.996

Factor score descriptives

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	16199	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			137695
<b>Bayesian Information Criterion (BIC)</b>			137764
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker-Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.829

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.811
(Cronbach's alpha = .762)	
<b>McDonald's Omega</b>	.814
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: Retain factors with adj. eigenvalue > 0	
	Adjusted eigenvalue
factor 1	1.63
factor 2	-.11
factor 3	-.17

**Standardized factor loadings**

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

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

\*\* **Note:** Original items from TREE1 / PISA2000

**Item descriptives**

Indicators *	Mean	Std. dev.	Min.	Max.	Valid 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

[List of scales \(wave 0\)](#)

Tests and Indices of Factorial Invariance across ...

Equality of the

variance-covariance matrices across ...

Survey languages

chi2	df	p > chi2
313	18	.000

Survey settings

chi2	df	p > chi2
333	9	.000

Survey modes

chi2	df	p > chi2
26	9	.002

Tests of measurement invariance across ...

Survey languages

chi2	df	p > chi2
14	4	.008
206	4	.000
30	4	.000

Survey settings

chi2	df	p > chi2
8	2	.015
212	2	.000
24	2	.000

Survey modes

chi2	df	p > chi2
5	2	.073
1	2	.519
7	2	.032

Metric invariance (equal factor loadings)

Strong invariance (plus equal intercepts)

Strict invariance (plus equal error variances)

Configural factor similarity across ...

Survey languages

	TCC
German vs. French	1.000
French vs. Italian	.997
Italian vs. German	.996

Survey settings

	TCC
classroom vs. unproctored	.999

Survey modes

	TCC
web vs. PAP	.997

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

Survey languages

	CD
German	1.000
French	1.000
Italian	.996

Survey settings

	CD
classroom	1.000
unproctored	.999

Survey modes

	CD
web	1.000
PAP	.998

Factor score descriptives

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	11545	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			52733
<b>Bayesian Information Criterion (BIC)</b>			52802
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker-Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.817

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.720
(Cronbach's alpha = .556)	
<b>McDonald's Omega</b>	.742
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: Retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
factor 1	1.30
factor 2	-.06
factor 3	-.20

**Standardized factor loadings**

Indicators *	Coef.	(SE)	[95% Conf. interval]	
cultposs1	0.74	0.01	0.73	0.76
cultposs2	0.86	0.01	0.85	0.88
cultposs3	0.46	0.01	0.45	0.48

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

**Item descriptives**

Indicators *	Mean	Std. dev.	Min.	Max.	Valid Obs.
cultposs1	0.4	0.5		1	15977
cultposs2	0.4	0.5		1	15990
cultposs3	0.7	0.4		1	16009

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

Indicators	Coef.	Cut1	Cut2	Cut3
cultposs1	1.90	0.71		
cultposs2	3.51	0.55		
cultposs3	0.91	-1.23		

[List of scales \(wave 0\)](#)



Tests and Indices of Factorial Invariance across ...

Equality of the

variance-covariance matrices across ...

Survey languages

chi2	df	p > chi2
4574	18	.000

Survey settings

chi2	df	p > chi2
101	9	.000

Survey modes

chi2	df	p > chi2
79	9	.000

Tests of measurement invariance across ...

Survey languages

chi2	df	p > chi2
53	4	.000
887	4	.000
366	4	.000

Survey settings

chi2	df	p > chi2
1	2	.759
52	2	.000
21	2	.000

Survey modes

chi2	df	p > chi2
13	2	.002
21	2	.000
19	2	.000

Metric invariance (equal factor loadings)

Strong invariance (plus equal intercepts)

Strict invariance (plus equal error variances)

Configural factor similarity across ...

Survey languages

	TCC
German vs. French	.996
French vs. Italian	1.000
Italian vs. German	.997

Survey settings

	TCC
classroom vs. unproctored	1.000

Survey modes

	TCC
web vs. PAP	.996

Tucker's congruence coefficient

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

Survey languages

	CD
German	.979
French	.890
Italian	.819

Survey settings

	CD
classroom	1.000
unproctored	1.000

Survey modes

	CD
web	.999
PAP	.985

Coefficient of determination

Factor score descriptives

Variable name	Mean	Std. dev.	Min.	Max.	Obs.
cultposs_fs	0.0	0.8	-1.0	1.1	16028
Share of cases with imputed missing values:					0.5%
(Equivalence of scores from robust MLMV: CD = .969)					
(Equivalence of Scores from Two-Step-Approach: CD = .96)					

**\* Note:** The calculation of model-based invariance tests requires that we constrain the error variance of *cultposs2* to zero.

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	1455	9	.000
Baseline vs. saturated	42913	15	.000
2) <b>Root mean squared error (RMSEA)</b>			.101
90% Confidence interval: lower bound			.096
90% Confidence interval: upper bound			.105
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			166162
<b>Bayesian Information Criterion (BIC)</b>			166300
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.966
Tucker-Lewis Index (TLI)			.944
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.033
Coefficient of determination (CD)			.883

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.870
(Cronbach's alpha = .822)	
<b>McDonald's Omega</b>	.872
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: Retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
factor 1	3.13
factor 2	.11
factor 3	-.04
factor 4	-.05
factor 5	-.12
factor 6	-.15

**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**

Indicators	Mean	Std. dev.	Min.	Max.	Valid 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)**

Indicators	Coef.	Cut1	Cut2	Cut3
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

[List of scales \(wave 0\)](#)

Tests and Indices of Factorial Invariance across ...

Equality of the

variance-covariance matrices across ...

Survey languages

chi2	df	p > chi2
765	54	.000

Survey settings

chi2	df	p > chi2
221	27	.000

Survey modes

chi2	df	p > chi2
63	27	.000

Tests of measurement invariance across ...

Survey languages

chi2	df	p > chi2
21	10	.018
70	10	.000
197	10	.000

Survey settings

chi2	df	p > chi2
36	5	.000
24	5	.000
57	5	.000

Survey modes

chi2	df	p > chi2
14	5	.018
10	5	.085
15	5	.011

Metric invariance (equal factor loadings)

Strong invariance (plus equal intercepts)

Strict invariance (plus equal error variances)

Configural factor similarity across ...

Survey languages

	TCC
German vs. French	1.000
French vs. Italian	.999
Italian vs. German	.999

Survey settings

	TCC
classroom vs. unproctored	.999

Survey modes

	TCC
web vs. PAP	.997

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

Survey languages

	CD
German	1.000
French	1.000
Italian	.999

Survey settings

	CD
classroom	1.000
unproctored	.999

Survey modes

	CD
web	1.000
PAP	.998

Coefficient of determination

Factor score descriptives

Variable name	Mean	Std. dev.	Min.	Max.	Obs.
inccap_fs	0.0	0.9	-3.2	1.8	15846
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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	12618	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			88215
<b>Bayesian Information Criterion (BIC)</b>			88284
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker-Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.798

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.763
(Cronbach's alpha = .684)	
<b>McDonald's Omega</b>	.769
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: Retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
factor 1	1.41
factor 2	-.10
factor 3	-.20

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
manners1	0.60	0.01	0.58	0.61
manners2	0.74	0.01	0.73	0.76
manners3	0.83	0.01	0.81	0.84

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid 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

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

Indicators	Coef.	Cut1	Cut2	Cut3
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

[List of scales \(wave 0\)](#)

Tests and Indices of Factorial Invariance across ...

Equality of the

variance-covariance matrices across ...

Survey languages

chi2	df	p > chi2
470	18	.000

Survey settings

chi2	df	p > chi2
138	9	.000

Survey modes

chi2	df	p > chi2
15	9	.082

Tests of measurement invariance across ...

Survey languages

chi2	df	p > chi2
7	4	.160
28	4	.000
40	4	.000

Survey settings

chi2	df	p > chi2
1	2	.751
16	2	.000
14	2	.001

Survey modes

chi2	df	p > chi2
3	2	.231
3	2	.280
4	2	.119

Metric invariance (equal factor loadings)

Strong invariance (plus equal intercepts)

Strict invariance (plus equal error variances)

Configural factor similarity across ...

Survey languages

	TCC
German vs. French	.999
French vs. Italian	.999
Italian vs. German	.999

Survey settings

	TCC
classroom vs. unproctored	1.000

Survey modes

	TCC
web vs. PAP	.998

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

Survey languages

	CD
German	1.000
French	.998
Italian	.997

Survey settings

	CD
classroom	1.000
unproctored	1.000

Survey modes

	CD
web	1.000
PAP	.998

Factor score descriptives

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	16621	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			90127
<b>Bayesian Information Criterion (BIC)</b>			90196
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker-Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.821

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.818
(Cronbach's alpha = .759)	
<b>McDonald's Omega</b>	.819
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: Retain factors with adj. eigenvalue > 0	
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
verbskill3	0.79	0.00	0.78	0.80

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
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)**

Indicators	Coef.	Cut1	Cut2	Cut3
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

[List of scales \(wave 0\)](#)

Tests and Indices of Factorial Invariance across ...

Equality of the

variance-covariance matrices across ...

Survey languages

chi2	df	p > chi2
209	18	.000

Survey settings

chi2	df	p > chi2
24	9	.005

Survey modes

chi2	df	p > chi2
34	9	.000

Tests of measurement invariance across ...

Survey languages

chi2	df	p > chi2
6	4	.227
36	4	.000
89	4	.000

Survey settings

chi2	df	p > chi2
4	2	.137
2	2	.425
13	2	.002

Survey modes

chi2	df	p > chi2
12	2	.003
4	2	.106
8	2	.023

Metric invariance (equal factor loadings)

Strong invariance (plus equal intercepts)

Strict invariance (plus equal error variances)

Configural factor similarity across ...

Survey languages

	TCC
German vs. French	1.000
French vs. Italian	.998
Italian vs. German	.999

Survey settings

	TCC
classroom vs. unproctored	1.000

Survey modes

	TCC
web vs. PAP	.995

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

Coefficient of determination

Survey languages

	CD
German	1.000
French	1.000
Italian	.998

Survey settings

	CD
classroom	1.000
unproctored	1.000

Survey modes

	CD
web	1.000
PAP	.993

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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	7949	14	.000
Baseline vs. saturated	27943	21	.000
2) <b>Root mean squared error (RMSEA)</b>			.189
90% Confidence interval: lower bound			.186
90% Confidence interval: upper bound			.193
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			260288
<b>Bayesian Information Criterion (BIC)</b>			260449
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.716
Tucker-Lewis Index (TLI)			.574
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.118
Coefficient of determination (CD)			.809

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.743
(Cronbach's alpha = .668)	
<b>McDonald's Omega</b>	.726
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: Retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
factor 1	2.14
factor 2	.76
factor 3	.03
factor 4	-.02
factor 5	-.13
factor 6	-.20
factor 7	-.20

**Standardized factor loadings**

Indicators *	Coef.	(SE)	[95% Conf. interval]	
cult1 **	0.36	0.01	0.34	0.37
cult2 **	0.70	0.01	0.69	0.71
cult3 **	0.50	0.01	0.48	0.51
cult4 **	0.77	0.00	0.76	0.78
cult5 **	0.74	0.01	0.73	0.75
cult7	0.29	0.01	0.27	0.31
cult9	0.24	0.01	0.23	0.26

**Item descriptives**

Indicators *	Mean	Std. dev.	Min.	Max.	Valid Obs.
cult1 **	2.8	1.0	1	4	15787
cult2 **	1.8	0.9	1	4	15776
cult3 **	1.6	0.8	1	4	15769
cult4 **	1.3	0.6	1	4	15771
cult5 **	1.6	0.7	1	4	15761
cult7	2.6	1.0	1	4	15766
cult9	2.4	1.2	1	4	15761

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

\*\* **Note:** Original items from TREE1 / PISA2000

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

Indicators	Coef.	Cut1	Cut2	Cut3
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

[List of scales \(wave 0\)](#)



Tests and Indices of Factorial Invariance across ...

Equality of the

variance-covariance matrices across ...

Survey languages

chi2	df	p > chi2
1553	70	.000

Survey settings

chi2	df	p > chi2
737	35	.000

Survey modes

chi2	df	p > chi2
149	35	.000

Tests of measurement invariance across ...

Survey languages

chi2	df	p > chi2
107	12	.000
1198	12	.000
142	12	.000

Survey settings

chi2	df	p > chi2
30	6	.000
231	6	.000
269	6	.000

Survey modes

chi2	df	p > chi2
19	6	.005
74	6	.000
35	6	.000

Metric invariance (equal factor loadings)

Strong invariance (plus equal intercepts)

Strict invariance (plus equal error variances)

Configural factor similarity across ...

Survey languages

	TCC
German vs. French	.992
French vs. Italian	.996
Italian vs. German	.992

Survey settings

	TCC
classroom vs. unproctored	.997

Survey modes

	TCC
web vs. PAP	.987

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

Survey languages

	CD
German	1.000
French	.995
Italian	.998

Survey settings

	CD
classroom	1.000
unproctored	.997

Survey modes

	CD
web	1.000
PAP	.990

Factor score descriptives

Variable name	Mean	Std. 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 \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	7348	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			124416
<b>Bayesian Information Criterion (BIC)</b>			124485
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker-Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.728

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.668
(Cronbach's alpha = .599)	
<b>McDonald's Omega</b>	.679
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: Retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
factor 1	1.05
factor 2	-.10
factor 3	-.22

**Standardized factor loadings**

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

\* **Note:** Original item from TREE1 / PISA2000

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid 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	Cut2	Cut3
cult3	1.11	0.43	2.46	3.56
cult7	1.27	-2.14	0.17	1.64
cult9	2.25	-1.53	0.47	1.88

[List of scales \(wave 0\)](#)

Tests and Indices of Factorial Invariance across ...

Equality of the

variance-covariance matrices across ...

Survey languages

chi2	df	p > chi2
993	18	.000

Survey settings

chi2	df	p > chi2
164	9	.000

Survey modes

chi2	df	p > chi2
50	9	.000

Tests of measurement invariance across ...

Survey languages

chi2	df	p > chi2
65	4	.000
674	4	.000
162	4	.000

Survey settings

chi2	df	p > chi2
18	2	.000
107	2	.000
13	2	.002

Survey modes

chi2	df	p > chi2
13	2	.002
24	2	.000
5	2	.071

Metric invariance (equal factor loadings)

Strong invariance (plus equal intercepts)

Strict invariance (plus equal error variances)

Configural factor similarity across ...

Survey languages

	TCC
German vs. French	.985
French vs. Italian	.999
Italian vs. German	.989

Survey settings

	TCC
classroom vs. unproctored	.996

Survey modes

	TCC
web vs. PAP	.949

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

Survey languages

	CD
German	.992
French	.975
Italian	.996

Survey settings

	CD
classroom	.999
unproctored	.990

Survey modes

	CD
web	.999
PAP	.852

Coefficient of determination

Factor score descriptives

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

## Model and Fit Statistics

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	14402	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			90498
<b>Bayesian Information Criterion (BIC)</b>			90567
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker-Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.805

## Reliability and Dimensionality

<b>Ordinal Cronbach's Alpha</b>	.793
(Cronbach's alpha = .690)	
<b>McDonald's Omega</b>	.795
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: Retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
factor 1	1.53
factor 2	-.13
factor 3	-.17

## Standardized factor loadings

Indicators *	Coef.	(SE)	[95% Conf. interval]	
cult2	0.69	0.01	0.68	0.70
cult4	0.82	0.01	0.81	0.83
cult5	0.74	0.01	0.73	0.75

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

## Item descriptives

Indicators *	Mean	Std. dev.	Min.	Max.	Valid Obs.
cult2	1.8	0.9	1	4	15776
cult4	1.3	0.6	1	4	15771
cult5	1.6	0.7	1	4	15761

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

Indicators	Coef.	Cut1	Cut2	Cut3
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

[List of scales \(wave 0\)](#)

Tests and Indices of Factorial Invariance across ...

Equality of the

variance-covariance matrices across ...

Survey languages

chi2	df	p > chi2
283	18	.000

Survey settings

chi2	df	p > chi2
436	9	.000

Survey modes

chi2	df	p > chi2
58	9	.000

Tests of measurement invariance across ...

Survey languages

chi2	df	p > chi2
40	4	.000
125	4	.000
48	4	.000

Survey settings

chi2	df	p > chi2
5	2	.085
48	2	.000
176	2	.000

Survey modes

chi2	df	p > chi2
1	2	.518
10	2	.008
13	2	.001

Metric invariance (equal factor loadings)

Strong invariance (plus equal intercepts)

Strict invariance (plus equal error variances)

Configural factor similarity across ...

Survey languages

	TCC
German vs. French	.997
French vs. Italian	.999
Italian vs. German	.999

Survey settings

	TCC
classroom vs. unproctored	.999

Survey modes

	TCC
web vs. PAP	.999

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

Survey languages

	CD
German	1.000
French	.993
Italian	.999

Survey settings

	CD
classroom	1.000
unproctored	.999

Survey modes

	CD
web	1.000
PAP	.996

Factor score descriptives

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	12119	27	.000
Baseline vs. saturated	41971	36	.000
2) <b>Root mean squared error (RMSEA)</b>			.167
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			138697
<b>Bayesian Information Criterion (BIC)</b>			138904
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.712
Tucker-Lewis Index (TLI)			.616
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.079
Coefficient of determination (CD)			.839

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.782
(Cronbach's alpha = .565)	
<b>McDonald's Omega</b>	.789
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: Retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
factor 1	2.83
factor 2	.49
factor 3	.40
factor 4	.08
factor 5	.07
factor 6	.02
factor 7	-.10
factor 8	-.15
factor 9	-.25

**Standardized factor loadings**

Indicators *	Coef.	(SE)	[95% Conf. interval]	
wealth1	0.71	0.00	0.70	0.72
wealth2	0.57	0.01	0.56	0.58
wealth3	0.31	0.01	0.29	0.32
wealth4	0.81	0.00	0.80	0.82
wealthn1	0.59	0.01	0.58	0.61
wealthn2	0.35	0.01	0.33	0.36
wealthn3	0.50	0.01	0.49	0.51
wealthn4	0.42	0.01	0.41	0.44
wealthn5	0.55	0.01	0.54	0.56

**Item descriptives**

Indicators *	Mean	Std. dev.	Min.	Max.	Valid Obs.
wealth1	0.9	0.3		1	16040
wealth2	0.9	0.3		1	16039
wealth3	0.6	0.5		1	15942
wealth4	1.0	0.1		1	16043
wealthn1	3.9	0.4	1	4	16037
wealthn2	2.8	0.8	1	4	16037
wealthn3	3.3	0.8	1	4	16032
wealthn4	2.7	0.8	1	4	16030
wealthn5	2.9	0.7	1	4	16037

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

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

Indicators	Coef.	Cut1	Cut2	Cut3
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

[List of scales \(wave 0\)](#)

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

## Equality of the

## variance-covariance matrices across ...

## Survey languages

chi2	df	p > chi2
4879	108	.000

## Survey settings

chi2	df	p > chi2
1025	54	.000

## Survey modes

chi2	df	p > chi2
1065	54	.000

## Tests of measurement invariance across ...

## Survey languages

chi2	df	p > chi2
139	16	.000
499	16	.000
1367	16	.000

## Survey settings

chi2	df	p > chi2
92	8	.000
74	8	.000
270	8	.000

## Survey modes

chi2	df	p > chi2
103	8	.000
44	8	.000
147	8	.000

Metric invariance (equal factor loadings)

Strong invariance (plus equal intercepts)

Strict invariance (plus equal error variances)

## Configural factor similarity across ...

## Survey languages

	TCC
German vs. French	.989
French vs. Italian	.992
Italian vs. German	.991

## Survey settings

	TCC
classroom vs. unproctored	.997

## Survey modes

	TCC
web vs. PAP	.965

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

## Coefficient of determination

## Survey languages

	CD
German	1.000
French	.999
Italian	.959

## Survey settings

	CD
classroom	1.000
unproctored	.995

## Survey modes

	CD
web	.997
PAP	.964

## 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	8521	14	.000
Baseline vs. saturated	38309	21	.000
2) <b>Root mean squared error (RMSEA)</b>			.195
90% Confidence interval: lower bound			.191
90% Confidence interval: upper bound			.198
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			59604
<b>Bayesian Information Criterion (BIC)</b>			59765
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.778
Tucker-Lewis Index (TLI)			.667
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.079
Coefficient of determination (CD)			.837

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.813
(Cronbach's alpha = .548)	
<b>McDonald's Omega</b>	.815
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: Retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
factor 1	2.76
factor 2	.46
factor 3	.20
factor 4	.02
factor 5	-.07
factor 6	-.12
factor 7	-.24

**Standardized factor loadings**

Indicators *	Coef.	(SE)	[95% Conf. interval]	
wealth1 **	0.77	0.00	0.76	0.77
wealth2 **	0.62	0.01	0.61	0.63
wealth4 **	0.75	0.00	0.74	0.76
wealth5	0.61	0.01	0.60	0.62
wealthn1 **	0.51	0.01	0.50	0.52
wealthn3 **	0.47	0.01	0.46	0.49
wealthn5 **	0.60	0.01	0.59	0.61

**Item descriptives**

Indicators *	Mean	Std. dev.	Min.	Max.	Valid Obs.
wealth1 **	0.9	0.3		1	16040
wealth2 **	0.9	0.3		1	16039
wealth4 **	1.0	0.1		1	16043
wealth5	0.7	0.5		1	16021
wealthn1 **	3.9	0.4	1	4	16037
wealthn3 **	3.3	0.8	1	4	16032
wealthn5 **	2.9	0.7	1	4	16037

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

\*\* **Note:** Original items from TREE1 / PISA2000

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

Indicators	Coef.	Cut1	Cut2	Cut3
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

[List of scales \(wave 0\)](#)



Tests and Indices of Factorial Invariance across ...

Equality of the

variance-covariance matrices across ...

Survey languages

chi2	df	p > chi2
2014	70	.000

Survey settings

chi2	df	p > chi2
777	35	.000

Survey modes

chi2	df	p > chi2
890	35	.000

Tests of measurement invariance across ...

Survey languages

chi2	df	p > chi2
168	12	.000
329	12	.000
983	12	.000

Survey settings

chi2	df	p > chi2
144	6	.000
65	6	.000
175	6	.000

Survey modes

chi2	df	p > chi2
74	6	.000
25	6	.000
140	6	.000

Metric invariance (equal factor loadings)

Strong invariance (plus equal intercepts)

Strict invariance (plus equal error variances)

Configural factor similarity across ...

Survey languages

	TCC
German vs. French	.996
French vs. Italian	.975
Italian vs. German	.989

Survey settings

	TCC
classroom vs. unproctored	.992

Survey modes

	TCC
web vs. PAP	.965

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

Survey languages

	CD
German	.999
French	.978
Italian	.902

Survey settings

	CD
classroom	.999
unproctored	.991

Survey modes

	CD
web	.999
PAP	.947

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)					

[List of scales \(wave 0\)](#)

Composite descriptives		Mean	Std. dev.	Min.	Max.	Obs.
Variable name						
	faiiii_comp	9.5	2.1	0	13	16059
Share of cases with imputed missing values:		0.5%				

Item descriptives		Mean	Std. dev.	Min.	Max.	Valid obs.
Indicators						
	wealthn4	1.5	0.6	0	2	16030 *
	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
	holyn	1.9	1.0	0	3	16028 *

\* Items recoded for composite calculation (see Hobza et al. 2017)

[List of scales \(wave 0\)](#)



**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	1666	5	.000
Baseline vs. saturated	37134	10	.000
2) <b>Root mean squared error (RMSEA)</b>			.145
90% Confidence interval: lower bound			.139
90% Confidence interval: upper bound			.151
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			221347
<b>Bayesian Information Criterion (BIC)</b>			221462
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.955
Tucker-Lewis Index (TLI)			.911
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.038
Coefficient of determination (CD)			.874

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.871
(Cronbach's alpha = .845)	
<b>McDonald's Omega</b>	.871
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: Retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
factor 1	2.79
factor 2	.10
factor 3	-.07
factor 4	-.13
factor 5	-.13

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
cap1	0.76	0.00	0.75	0.77
cap2	0.78	0.00	0.77	0.79
cap3	0.79	0.00	0.78	0.80
cap4	0.69	0.00	0.68	0.70
cap5	0.76	0.00	0.75	0.77

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
cap1	5.9	1.3	1	7	15756
cap2	5.7	1.2	1	7	15733
cap3	5.9	1.2	1	7	15732
cap4	5.3	1.3	1	7	15714
cap5	5.7	1.2	1	7	15738

[List of scales \(wave 0\)](#)

Tests and Indices of Factorial Invariance across ...

Equality of the

variance-covariance matrices across ...

Survey languages

chi2	df	p > chi2
1233	40	.000

Survey settings

chi2	df	p > chi2
412	20	.000

Survey modes

chi2	df	p > chi2
32	20	.042

Tests of measurement invariance across ...

Survey languages

chi2	df	p > chi2
106	8	.000
601	8	.000
216	8	.000

Survey settings

chi2	df	p > chi2
21	4	.000
75	4	.000
15	4	.005

Survey modes

chi2	df	p > chi2
7	4	.145
11	4	.025
4	4	.456

Metric invariance (equal factor loadings)

Strong invariance (plus equal intercepts)

Strict invariance (plus equal error variances)

Configural factor similarity across ...

Survey languages

	TCC
German vs. French	.996
French vs. Italian	.997
Italian vs. German	.997

Survey settings

	TCC
classroom vs. unproctored	.999

Survey modes

	TCC
web vs. PAP	.998

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

Survey languages

	CD
German	1.000
French	.998
Italian	.999

Survey settings

	CD
classroom	1.000
unproctored	1.000

Survey modes

	CD
web	1.000
PAP	.998

Coefficient of determination

Factor score descriptives

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	22788	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			205667
<b>Bayesian Information Criterion (BIC)</b>			205739
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker–Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.835

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.809
(Cronbach's alpha = .784)	
<b>McDonald's Omega</b>	.813
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	1.61
Factor 2	-.10
Factor 3	-.17

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
posatt1	0.74	.004	0.73	0.75
posatt2	0.86	.004	0.85	0.87
posatt3	0.70	.004	0.69	0.71

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid 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

[List of scales \(wave 0\)](#)

Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	998	18	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	24844	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			216963
<b>Bayesian Information Criterion (BIC)</b>			217035
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker–Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.856

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.821
(Cronbach's alpha = .796)	
<b>McDonald's Omega</b>	.825
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	1.67
Factor 2	-.08
Factor 3	-.16

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
enjoyschool1	0.76	.004	0.75	0.77
enjoyschool2	0.89	.004	0.88	0.89
enjoyschool3	0.69	.004	0.68	0.70

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
enjoyschool1	3.2	1.5	1	6	22254
enjoyschool2	3.5	1.4	1	6	22252
enjoyschool3	3.9	1.4	1	6	22257

[List of scales \(wave 0\)](#)



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

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	506	18	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	29	2	.000
Baseline vs. saturated	36796	6	.000
2) <b>Root mean squared error (RMSEA)</b>			.025
90% Confidence interval: lower bound			.017
90% Confidence interval: upper bound			.033
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			272002
<b>Bayesian Information Criterion (BIC)</b>			272098
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.999
Tucker–Lewis Index (TLI)			.998
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.005
Coefficient of determination (CD)			.857

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.847
(Cronbach's alpha = .772)	
<b>McDonald's Omega</b>	.849
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	2.22
Factor 2	-.09
Factor 3	-.10
Factor 4	-.12

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
physpain1	0.78	.003	0.77	0.79
physpain2	0.79	.003	0.78	0.79
physpain3	0.82	.003	0.81	0.82
physpain4	0.67	.004	0.66	0.68

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid 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

[List of scales \(wave 0\)](#)

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

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	1179	28	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	21848	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			240309
<b>Bayesian Information Criterion (BIC)</b>			240381
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker–Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.836

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.795
(Cronbach's alpha = .753)	
<b>McDonald's Omega</b>	.802
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	1.57
Factor 2	-.09
Factor 3	-.18

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
trouschool1	0.78	.004	0.78	0.79
trouschool2	0.86	.004	0.85	0.87
trouschool3	0.62	.005	0.61	0.63

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
trouschool1	2.9	1.6	1	6	22260
trouschool2	3.2	1.7	1	6	22263
trouschool3	3.4	1.9	1	6	22263

[List of scales \(wave 0\)](#)

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

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	1522	18	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	39687	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			164458
<b>Bayesian Information Criterion (BIC)</b>			164530
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker–Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.929

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.886
(Cronbach's alpha = .817)	
<b>McDonald's Omega</b>	.889
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
	Adjusted eigenvalue
Factor 1	2.07
Factor 2	-.05
Factor 3	-.12

**Standardized factor loadings**

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

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid 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

[List of scales \(wave 0\)](#)

Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	466	18	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	14239	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			245338
<b>Bayesian Information Criterion (BIC)</b>			245410
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker–Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.835

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.702
(Cronbach's alpha = .661)	
<b>McDonald's Omega</b>	.727
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	1.23
Factor 2	-.05
Factor 3	-.22

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
schoolav1	0.89	.007	0.88	0.91
schoolav2	0.67	.007	0.66	0.69
schoolav3	0.46	.006	0.45	0.47

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
schoolav1	3.1	1.8	1	6	22245
schoolav2	3.7	1.9	1	6	22248
schoolav3	2.2	1.5	1	6	22235

[List of scales \(wave 0\)](#)



## Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2
	1451	9	.000

<b>Tests of measurement invariance</b>	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. invariant models**

Coefficient of determination	CD
Language: German	.994
Language: French/ Italian	.981

\* **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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	12995	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			152039
<b>Bayesian Information Criterion (BIC)</b>			152111
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker–Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.795

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.703
(Cronbach's alpha = .652)	
<b>McDonald's Omega</b>	.718
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	1.19
Factor 2	-.08
Factor 3	-.22

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
achmot2	0.54	.006	0.52	0.55
achmot4	0.62	.006	0.60	0.63
achmot6	0.86	.007	0.85	0.87

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
achmot2	3.0	0.8	1	4	22249
achmot4	2.8	0.8	1	4	22242
achmot6	2.6	0.9	1	4	22239

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

Indicators	Coef.	Cut1	Cut2	Cut3
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

[List of scales \(wave 0\)](#)

## Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	1286	18	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	12774	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			148710
<b>Bayesian Information Criterion (BIC)</b>			148782
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker–Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.792

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.648
(Cronbach's alpha = .589)	
<b>McDonald's Omega</b>	.690
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
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	0.86

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid 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.	Cut1	Cut2	Cut3
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

[List of scales \(wave 0\)](#)

Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	1767	18	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	28969	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			144091
<b>Bayesian Information Criterion (BIC)</b>			144163
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker–Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.865

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.848
(Cronbach's alpha = .796)	
<b>McDonald's Omega</b>	.850
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	1.81
Factor 2	-.10
Factor 3	-.14

**Standardized factor loadings**

Indicators *	Coef.	(SE)	[95% Conf. interval]	
insmot1	0.75	0.00	0.74	0.76
insmot2	0.79	0.00	0.78	0.80
insmot3	0.88	0.00	0.88	0.89

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

**Item descriptives**

Indicators *	Mean	Std. dev.	Min.	Max.	Valid Obs.
insmot1	2.8	0.9	1	4	22246
insmot2	2.9	0.9	1	4	22220
insmot3	3.1	0.9	1	4	22220

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

Indicators	Coef.	Cut1	Cut2	Cut3
insmot1	2.05	-3.82	-0.83	2.13
insmot2	2.35	-3.90	-1.28	1.70
insmot3	3.48	-6.32	-3.28	0.89

[List of scales \(wave 0\)](#)

## Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	347	18	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

## Model and Fit Statistics

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	44643	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			153979
<b>Bayesian Information Criterion (BIC)</b>			154051
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker–Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.924

## Reliability and Dimensionality

<b>Ordinal Cronbach's Alpha</b>	.906
(Cronbach's alpha = .864)	
<b>McDonald's Omega</b>	.907
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	2.19
Factor 2	-.07
Factor 3	-.11

## Standardized factor loadings

Indicators *	Coef.	(SE)	[95% Conf. interval]	
intrea1	0.86	.002	0.85	0.86
intrea2	0.94	.002	0.93	0.94
intrea3	0.83	.003	0.82	0.83

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

## Item descriptives

Indicators *	Mean	Std. dev.	Min.	Max.	Valid Obs.
intrea1	2.2	1.0	1	4	22180
intrea2	2.1	1.1	1	4	22178
intrea3	2.3	1.1	1	4	22165

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

Indicators	Coef.	Cut1	Cut2	Cut3
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

[List of scales \(wave 0\)](#)



## Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	732	18	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

## Model and Fit Statistics

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	15929	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			69317
<b>Bayesian Information Criterion (BIC)</b>			69383
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker–Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.884

## Reliability and Dimensionality

<b>Ordinal Cronbach's Alpha</b>	.855
(Cronbach's alpha = .797)	
<b>McDonald's Omega</b>	.860
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	1.88
Factor 2	-.09
Factor 3	-.13

## Standardized factor loadings

Indicators	Coef.	(SE)	[95% Conf. interval]	
ictmot2	0.69	.006	0.68	0.71
ictmot3	0.88	.004	0.87	0.89
ictmot4	0.87	.004	0.86	0.88

## Item descriptives

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
ictmot2	3.2	0.7	1	4	11068
ictmot3	2.4	1.0	1	4	11065
ictmot4	2.8	0.9	1	4	11060

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

Indicators	Coef.	Cut1	Cut2	Cut3
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

[List of scales \(wave 0\)](#)

## Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	408	18	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	1805	9	.000
Baseline vs. saturated	31076	15	.000
2) <b>Root mean squared error (RMSEA)</b>			.135
90% Confidence interval: lower bound			.130
90% Confidence interval: upper bound			.140
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			137195
<b>Bayesian Information Criterion (BIC)</b>			137326
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.942
Tucker–Lewis Index (TLI)			.904
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.041
Coefficient of determination (CD)			.888

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.875
(Cronbach's alpha = .836)	
<b>McDonald's Omega</b>	.876
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
	Adjusted eigenvalue
Factor 1	3.19
Factor 2	.14
Factor 3	-.01
Factor 4	-.05
Factor 5	-.13
Factor 6	-.14

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
intsubj1	0.84	.004	0.83	0.85
intsubj2	0.65	.006	0.64	0.66
intsubj3	0.75	.005	0.74	0.76
intsubj4	0.66	.006	0.65	0.67
intsubj5	0.69	.006	0.68	0.71
intsubj6	0.80	.004	0.80	0.81

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
intsubj1	2.5	0.9	1	4	10889
intsubj2	3.2	0.7	1	4	10922
intsubj3	2.9	0.8	1	4	10845
intsubj4	2.6	0.9	1	4	10842
intsubj5	2.8	0.8	1	4	10905
intsubj6	2.4	1.0	1	4	10853

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

Indicators	Coef.	Cut1	Cut2	Cut3
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

[List of scales \(wave 0\)](#)

Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	885	54	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

## Model and Fit Statistics

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	45	2	.000
Baseline vs. saturated	43936	6	.000
2) <b>Root mean squared error (RMSEA)</b>			.044
90% Confidence interval: lower bound			.034
90% Confidence interval: upper bound			.056
Probability RMSEA <= 0.05			.777
3) <b>Akaike's Information Criterion (AIC)</b>			72033
<b>Bayesian Information Criterion (BIC)</b>			72121
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.999
Tucker–Lewis Index (TLI)			.997
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.004
Coefficient of determination (CD)			.955

## Reliability and Dimensionality

<b>Ordinal Cronbach's Alpha</b>	.946
(Cronbach's alpha = .918)	
<b>McDonald's Omega</b>	.947
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	3.20
Factor 2	-.04
Factor 3	-.05
Factor 4	-.04

## Standardized factor loadings

Indicators	Coef.	(SE)	[95% Conf. interval]	
instrumot1	0.95	.001	0.94	0.95
instrumot2	0.93	.002	0.93	0.94
instrumot3	0.89	.002	0.88	0.89
instrumot4	0.85	.003	0.84	0.85

## Item descriptives

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
instrumot1	2.9	0.9	1	4	11018
instrumot2	2.9	0.9	1	4	11020
instrumot3	2.8	0.9	1	4	11030
instrumot4	2.9	0.9	1	4	11013

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

Indicators	Coef.	Cut1	Cut2	Cut3
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

[List of scales \(wave 0\)](#)

Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2
	387	14	.000

<b>Tests of measurement invariance</b>	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. invariant models**

Coefficient of determination	CD
Language: German	1.000
Language: French/ Italian	1.000

\* **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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	687	2	.000
Baseline vs. saturated	16452	6	.000
2) <b>Root mean squared error (RMSEA)</b>			.177
90% Confidence interval: lower bound			.166
90% Confidence interval: upper bound			.188
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			100910
<b>Bayesian Information Criterion (BIC)</b>			100998
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.958
Tucker–Lewis Index (TLI)			.875
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.038
Coefficient of determination (CD)			.844

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.820
(Cronbach's alpha = .764)	
<b>McDonald's Omega</b>	.826
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	2.06
Factor 2	.06
Factor 3	-.15
Factor 4	-.15

**Standardized factor loadings**

Indicators *	Coef.	(SE)	[95% Conf. interval]	
extreg2	0.76	.005	0.75	0.77
extreg3	0.81	.005	0.80	0.82
extreg4	0.58	.008	0.56	0.59
extreg5	0.78	.005	0.77	0.79

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
extreg2	1.9	0.9	1	4	10901
extreg3	2.0	0.9	1	4	10830
extreg4	2.4	0.9	1	4	10841
extreg5	1.8	0.9	1	4	10827

\* **Note:** Items Extreg1 and Extreg6 Excluded to Improve Scale Quality

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

Indicators	Coef.	Cut1	Cut2	Cut3
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

[List of scales \(wave 0\)](#)



## Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	222	28	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	584	5	.000
Baseline vs. saturated	28718	10	.000
2) <b>Root mean squared error (RMSEA)</b>			.103
90% Confidence interval: lower bound			.096
90% Confidence interval: upper bound			.110
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			97128
<b>Bayesian Information Criterion (BIC)</b>			97238
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.980
Tucker–Lewis Index (TLI)			.960
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.024
Coefficient of determination (CD)			.890

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.888
(Cronbach's alpha = .848)	
<b>McDonald's Omega</b>	.888
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
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**

Indicators	Mean	Std. dev.	Min.	Max.	Valid 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	Cut2	Cut3
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

[List of scales \(wave 0\)](#)

## Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	938	40	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	620	2	.000
Baseline vs. saturated	17637	6	.000
2) <b>Root mean squared error (RMSEA)</b>			.171
90% Confidence interval: lower bound			.159
90% Confidence interval: upper bound			.182
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			117025
<b>Bayesian Information Criterion (BIC)</b>			117112
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.965
Tucker–Lewis Index (TLI)			.895
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.040
Coefficient of determination (CD)			.865

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.834
(Cronbach's alpha = .804)	
<b>McDonald's Omega</b>	.837
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	2.16
Factor 2	.05
Factor 3	-.15
Factor 4	-.13

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
approxgoals1	0.74	.006	0.73	0.75
approxgoals2	0.84	.004	0.83	0.84
approxgoals3	0.57	.008	0.55	0.58
approxgoals4	0.84	.004	0.83	0.85

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
approxgoals1	2.8	1.2	1	5	10608
approxgoals2	2.5	1.2	1	5	10478
approxgoals3	3.3	1.1	1	5	10596
approxgoals4	2.7	1.2	1	5	10474

[List of scales \(wave 0\)](#)

## Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	370	28	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	396	2	.000
Baseline vs. saturated	16559	6	.000
2) <b>Root mean squared error (RMSEA)</b>			.136
90% Confidence interval: lower bound			.125
90% Confidence interval: upper bound			.147
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			113590
<b>Bayesian Information Criterion (BIC)</b>			113677
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.976
Tucker–Lewis Index (TLI)			.929
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.028
Coefficient of determination (CD)			.841

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.839
(Cronbach's alpha = .808)	
<b>McDonald's Omega</b>	.839
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	2.15
Factor 2	-.01
Factor 3	-.15
Factor 4	-.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

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid 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

[List of scales \(wave 0\)](#)

Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	887	28	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	370	2	.000
Baseline vs. saturated	9625	6	.000
2) <b>Root mean squared error (RMSEA)</b>			.131
90% Confidence interval: lower bound			.120
90% Confidence interval: upper bound			.143
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			122140
<b>Bayesian Information Criterion (BIC)</b>			122227
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.962
Tucker–Lewis Index (TLI)			.885
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.033
Coefficient of determination (CD)			.761

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.747
(Cronbach's alpha = .712)	
<b>McDonald's Omega</b>	.750
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	1.59
Factor 2	-.02
Factor 3	-.09
Factor 4	-.22

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
avoidwork1	0.53	.009	0.51	0.54
avoidwork2	0.70	.007	0.68	0.71
avoidwork3	0.67	.008	0.66	0.69
avoidwork4	0.71	.007	0.70	0.72

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
avoidwork1	2.9	1.1	1	5	10615
avoidwork2	3.1	1.1	1	5	10483
avoidwork3	3.2	1.2	1	5	10599
avoidwork4	3.1	1.1	1	5	10480

[List of scales \(wave 0\)](#)



## Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	611	28	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	550	2	.000
Baseline vs. saturated	20651	6	.000
2) <b>Root mean squared error (RMSEA)</b>			.160
90% Confidence interval: lower bound			.149
90% Confidence interval: upper bound			.172
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			117023
<b>Bayesian Information Criterion (BIC)</b>			117111
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.973
Tucker–Lewis Index (TLI)			.920
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.027
Coefficient of determination (CD)			.877

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.866
(Cronbach's alpha = .830)	
<b>McDonald's Omega</b>	.867
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	2.37
Factor 2	.01
Factor 3	-.09
Factor 4	-.14

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
avoidblame1	0.73	.005	0.72	0.74
avoidblame2	0.75	.005	0.74	0.76
avoidblame3	0.86	.004	0.85	0.87
avoidblame4	0.81	.005	0.80	0.81

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
avoidblame1	2.6	1.2	1	5	10594
avoidblame2	2.6	1.3	1	5	10496
avoidblame3	2.5	1.2	1	5	10604
avoidblame4	2.3	1.1	1	5	10509

[List of scales \(wave 0\)](#)

Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	378	28	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	20015	20	.000
Baseline vs. saturated	64288	28	.000
2) <b>Root mean squared error (RMSEA)</b>			.250
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			329588
<b>Bayesian Information Criterion (BIC)</b>			329772
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.689
Tucker-Lewis Index (TLI)			.564
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.147
Coefficient of determination (CD)			.887

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.859
(Cronbach's alpha = .820)	
<b>McDonald's Omega</b>	.852
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: Retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
factor 1	3.56
factor 2	1.12
factor 3	.07
factor 4	-.05
factor 5	-.09
factor 6	-.10
factor 7	-.12
factor 8	-.13

**Standardized factor loadings**

Indicators *	Coef.	(SE)	[95% Conf. interval]	
sele1	0.63	0.01	0.62	0.64
sele2	0.51	0.01	0.49	0.52
sele3	0.44	0.01	0.43	0.46
sele4	0.49	0.01	0.48	0.51
seld1	0.85	0.00	0.84	0.85
seld3	0.75	0.00	0.74	0.75
seld4	0.65	0.01	0.64	0.66
seld5	0.80	0.00	0.79	0.81

\* **Note:** Reversed categories for all seld-items

**Item descriptives**

Indicators *	Mean	Std. dev.	Min.	Max.	Valid Obs.
sele1	4.0	0.9	1	5	15991
sele2	4.1	0.8	1	5	15961
sele3	3.9	0.8	1	5	15957
sele4	3.8	1.0	1	5	15946
seld1	3.8	1.2	1	5	15972
seld3	3.2	1.2	1	5	15953
seld4	3.2	1.3	1	5	15902
seld5	4.0	1.2	1	5	15943

[List of scales \(wave 0\)](#)

Tests and Indices of Factorial Invariance across ...

Equality of the

variance-covariance matrices across ...

Survey languages		
chi2	df	p > chi2
5550	88	.000

Survey settings		
chi2	df	p > chi2
693	44	.000

Survey modes		
chi2	df	p > chi2
136	44	.000

Tests of measurement invariance across ...

Survey languages		
chi2	df	p > chi2
85	14	.000
3216	14	.000
415	14	.000

Survey settings		
chi2	df	p > chi2
27	7	.000
618	7	.000
205	7	.000

Survey modes		
chi2	df	p > chi2
38	7	.000
42	7	.000
25	7	.001

Configural factor similarity across ...

Tucker's congruence coefficient

Survey languages		TCC
German vs. French		.999
French vs. Italian		.998
Italian vs. German		.996

Survey settings		TCC
classroom vs. unproctored		.999

Survey modes		TCC
web vs. PAP		.991

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

Coefficient of determination

Survey languages		CD
German		1.000
French		.994
Italian		.989

Survey settings		CD
classroom		1.000
unproctored		.998

Survey modes		CD
web		1.000
PAP		.985

Factor score descriptives

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	17789	14	.000
Baseline vs. saturated	55337	21	.000
2) <b>Root mean squared error (RMSEA)</b>			.282
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			283054
<b>Bayesian Information Criterion (BIC)</b>			283215
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.679
Tucker-Lewis Index (TLI)			.518
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.133
Coefficient of determination (CD)			.860

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.852
(Cronbach's alpha = .809)	
<b>McDonald's Omega</b>	.852
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: Retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
factor 1	3.24
factor 2	.97
factor 3	-.01
factor 4	-.06
factor 5	-.11
factor 6	-.12
factor 7	-.14

**Standardized factor loadings**

Indicators *	Coef.	(SE)	[95% Conf. interval]	
sele1	0.77	0.00	0.76	0.78
sele2	0.72	0.01	0.71	0.73
sele3	0.67	0.01	0.66	0.68
sele4	0.68	0.01	0.67	0.69
seld1	0.66	0.01	0.64	0.67
seld3	0.56	0.01	0.55	0.58
seld5	0.63	0.01	0.62	0.64

\* **Note:** Reversed categories for all seld-items

**Item descriptives**

Indicators *	Mean	Std. dev.	Min.	Max.	Valid Obs.
sele1	4.0	0.9	1	5	15991
sele2	4.1	0.8	1	5	15961
sele3	3.9	0.8	1	5	15957
sele4	3.8	1.0	1	5	15946
seld1	3.8	1.2	1	5	15972
seld3	3.2	1.2	1	5	15953
seld5	4.0	1.2	1	5	15943

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

Indicators	Coef.	Cut1	Cut2	Cut3
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[List of scales \(wave 0\)](#)

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

## Equality of the

## variance-covariance matrices across ...

## Survey languages

chi2	df	p > chi2
4643	70	.000

## Survey settings

chi2	df	p > chi2
628	35	.000

## Survey modes

chi2	df	p > chi2
125	35	.000

## Tests of measurement invariance across ...

## Survey languages

chi2	df	p > chi2
------	----	----------

Metric invariance (equal factor loadings)

130	12	.000
-----	----	------

Strong invariance (plus equal intercepts)

1838	12	.000
------	----	------

Strict invariance (plus equal error variances)

320	12	.000
-----	----	------

## Survey settings

chi2	df	p > chi2
------	----	----------

Metric invariance (equal factor loadings)

40	6	.000
----	---	------

Strong invariance (plus equal intercepts)

589	6	.000
-----	---	------

Strict invariance (plus equal error variances)

142	6	.000
-----	---	------

## Survey modes

chi2	df	p > chi2
------	----	----------

Metric invariance (equal factor loadings)

12	6	.069
----	---	------

Strong invariance (plus equal intercepts)

52	6	.000
----	---	------

Strict invariance (plus equal error variances)

15	6	.017
----	---	------

## Configural factor similarity across ...

## Survey languages

Tucker's congruence coefficient

	TCC
German vs. French	.996
French vs. Italian	.983
Italian vs. German	.966

## Survey settings

Tucker's congruence coefficient

	TCC
classroom vs. unproctored	.996

## Survey modes

Tucker's congruence coefficient

	TCC
web vs. PAP	.996

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

Coefficient of determination

## Survey languages

CD

German	.999
French	.997
Italian	.826

## Survey settings

CD

classroom	.999
unproctored	.991

## Survey modes

CD

web	1.000
PAP	.997

## 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	329	2	.000
Baseline vs. saturated	26567	6	.000
2) <b>Root mean squared error (RMSEA)</b>			.101
90% Confidence interval: lower bound			.092
90% Confidence interval: upper bound			.110
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			140371
<b>Bayesian Information Criterion (BIC)</b>			140463
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.988
Tucker-Lewis Index (TLI)			.963
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.018
Coefficient of determination (CD)			.856

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.848
(Cronbach's alpha = .801)	
<b>McDonald's Omega</b>	.849
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: Retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
factor 1	2.21
factor 2	-.06
factor 3	-.07
factor 4	-.15

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
sele1	0.72	0.00	0.71	0.73
sele2	0.83	0.00	0.82	0.83
sele3	0.78	0.00	0.78	0.79
sele4	0.72	0.00	0.71	0.73

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
sele1	4.0	0.9	1	5	15991
sele2	4.1	0.8	1	5	15961
sele3	3.9	0.8	1	5	15957
sele4	3.8	1.0	1	5	15946

[List of scales \(wave 0\)](#)



Tests and Indices of Factorial Invariance across ...

Equality of the

variance-covariance matrices across ...

Survey languages

chi2	df	p > chi2
1803	28	.000

Survey settings

chi2	df	p > chi2
346	14	.000

Survey modes

chi2	df	p > chi2
35	14	.002

Tests of measurement invariance across ...

Survey languages

chi2	df	p > chi2
21	6	.002
1214	6	.000
216	6	.000

Survey settings

chi2	df	p > chi2
11	3	.013
140	3	.000
123	3	.000

Survey modes

chi2	df	p > chi2
1	3	.769
8	3	.052
10	3	.017

Metric invariance (equal factor loadings)

Strong invariance (plus equal intercepts)

Strict invariance (plus equal error variances)

Configural factor similarity across ...

Survey languages

	TCC
German vs. French	1.000
French vs. Italian	.998
Italian vs. German	.997

Survey settings

	TCC
classroom vs. unproctored	1.000

Survey modes

	TCC
web vs. PAP	1.000

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

Survey languages

	CD
German	1.000
French	.998
Italian	.992

Survey settings

	CD
classroom	1.000
unproctored	1.000

Survey modes

	CD
web	1.000
PAP	1.000

Factor score descriptives

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	712	2	.000
Baseline vs. saturated	31810	6	.000
2) <b>Root mean squared error (RMSEA)</b>			.149
90% Confidence interval: lower bound			.140
90% Confidence interval: upper bound			.158
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			175983
<b>Bayesian Information Criterion (BIC)</b>			176075
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.978
Tucker-Lewis Index (TLI)			.933
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.028
Coefficient of determination (CD)			.887

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.866
(Cronbach's alpha = .824)	
<b>McDonald's Omega</b>	.868
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: Retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
factor 1	2.39
factor 2	.02
factor 3	-.13
factor 4	-.12

**Standardized factor loadings**

Indicators *	Coef.	(SE)	[95% Conf. interval]	
seld1	0.88	0.00	0.88	0.89
seld3	0.79	0.00	0.78	0.80
seld4	0.67	0.01	0.66	0.68
seld5	0.80	0.00	0.80	0.81

\* **Note:** Reversed Item Categories

**Item descriptives**

Indicators *	Mean	Std. dev.	Min.	Max.	Valid Obs.
seld1	3.8	1.2	1	5	15972
seld3	3.2	1.2	1	5	15953
seld4	3.2	1.3	1	5	15902
seld5	4.0	1.2	1	5	15943

[List of scales \(wave 0\)](#)

Tests and Indices of Factorial Invariance across ...

Equality of the

variance-covariance matrices across ...

Survey languages

chi2	df	p > chi2
4554	28	.000

Survey settings

chi2	df	p > chi2
140	14	.000

Survey modes

chi2	df	p > chi2
59	14	.000

Tests of measurement invariance across ...

Survey languages

chi2	df	p > chi2
107	6	.000
2496	6	.000
355	6	.000

Survey settings

chi2	df	p > chi2
4	3	.235
86	3	.000
1	3	.707

Survey modes

chi2	df	p > chi2
7	3	.064
27	3	.000
7	3	.089

Metric invariance (equal factor loadings)

Strong invariance (plus equal intercepts)

Strict invariance (plus equal error variances)

Configural factor similarity across ...

Survey languages

	TCC
German vs. French	.997
French vs. Italian	1.000
Italian vs. German	.998

Survey settings

	TCC
classroom vs. unproctored	1.000

Survey modes

	TCC
web vs. PAP	.998

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

Survey languages

	CD
German	1.000
French	.990
Italian	.980

Survey settings

	CD
classroom	1.000
unproctored	1.000

Survey modes

	CD
web	1.000
PAP	.999

Factor score descriptives

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	23184	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			130616
<b>Bayesian Information Criterion (BIC)</b>			130685
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker-Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.885

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.863
(Cronbach's alpha = .816)	
<b>McDonald's Omega</b>	.865
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: Retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
factor 1	1.90
factor 2	-.08
factor 3	-.14

**Standardized factor loadings**

Indicators *	Coef.	(SE)	[95% Conf. interval]	
seld1	0.91	0.00	0.90	0.91
seld3	0.80	0.00	0.79	0.80
seld5	0.77	0.00	0.76	0.78

\* **Note:** Reversed Item Categories

**Item descriptives**

Indicators *	Mean	Std. dev.	Min.	Max.	Valid Obs.
seld1	3.8	1.2	1	5	15972
seld3	3.2	1.2	1	5	15953
seld5	4.0	1.2	1	5	15943

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

Indicators	Coef.	Cut1	Cut2	Cut3
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[List of scales \(wave 0\)](#)

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

## Equality of the

## variance-covariance matrices across ...

## Survey languages

chi2	df	p > chi2
2872	18	.000

## Survey settings

chi2	df	p > chi2
104	9	.000

## Survey modes

chi2	df	p > chi2
53	9	.000

## Tests of measurement invariance across ...

## Survey languages

chi2	df	p > chi2
61	4	.000
1218	4	.000
27	4	.000

## Survey settings

chi2	df	p > chi2
1	2	.749
62	2	.000
1	2	.511

## Survey modes

chi2	df	p > chi2
6	2	.061
26	2	.000
5	2	.087

Metric invariance (equal factor loadings)

Strong invariance (plus equal intercepts)

Strict invariance (plus equal error variances)

## Configural factor similarity across ...

## Survey languages

	TCC
German vs. French	.998
French vs. Italian	1.000
Italian vs. German	.998

## Survey settings

	TCC
classroom vs. unproctored	1.000

## Survey modes

	TCC
web vs. PAP	.998

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

Coefficient of determination

## Survey languages

	CD
German	.999
French	.989
Italian	.980

## Survey settings

	CD
classroom	1.000
unproctored	1.000

## Survey modes

	CD
web	1.000
PAP	.997

## 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	63	2	.000
Baseline vs. saturated	23581	6	.000
2) <b>Root mean squared error (RMSEA)</b>			.044
90% Confidence interval: lower bound			.035
90% Confidence interval: upper bound			.053
Probability RMSEA <= 0.05			.847
3) <b>Akaike's Information Criterion (AIC)</b>			104477
<b>Bayesian Information Criterion (BIC)</b>			104569
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.997
Tucker-Lewis Index (TLI)			.992
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.009
Coefficient of determination (CD)			.836

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.835
(Cronbach's alpha = .772)	
<b>McDonald's Omega</b>	.835
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: Retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
factor 1	2.10
factor 2	-.08
factor 3	-.12
factor 4	-.13

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
seef1	0.73	0.00	0.72	0.74
seef2	0.77	0.00	0.76	0.78
seef3	0.76	0.00	0.75	0.77
seef4	0.73	0.00	0.72	0.74

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
seef1	3.1	0.6	1	4	15941
seef2	3.1	0.7	1	4	15928
seef3	2.8	0.7	1	4	15916
seef4	3.0	0.7	1	4	15923

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

Indicators	Coef.	Cut1	Cut2	Cut3
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

[List of scales \(wave 0\)](#)

Tests and Indices of Factorial Invariance across ...

Equality of the

variance-covariance matrices across ...

Survey languages		
chi2	df	p > chi2
1049	28	.000

Survey settings		
chi2	df	p > chi2
104	14	.000

Survey modes		
chi2	df	p > chi2
24	14	.044

Tests of measurement invariance across ...

	chi2	df	p > chi2
Metric invariance (equal factor loadings)	47	6	.000
Strong invariance (plus equal intercepts)	448	6	.000
Strict invariance (plus equal error variances)	230	6	.000

Survey languages		
chi2	df	p > chi2
1	3	.763
10	3	.018
12	3	.008

Survey settings		
chi2	df	p > chi2
4	3	.252
2	3	.652
4	3	.303

Survey modes		
chi2	df	p > chi2
1	3	.763
10	3	.018
12	3	.008

Configural factor similarity across ...

Tucker's congruence coefficient

Survey languages		TCC
German vs. French		.998
French vs. Italian		.995
Italian vs. German		.996

Survey settings		TCC
classroom vs. unproctored		1.000

Survey modes		TCC
web vs. PAP		.999

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

Coefficient of determination

Survey languages		CD
German		1.000
French		.997
Italian		.993

Survey settings		CD
classroom		1.000
unproctored		1.000

Survey modes		CD
web		1.000
PAP		.999

Factor score descriptives

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	32752	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			179405
<b>Bayesian Information Criterion (BIC)</b>			179477
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker–Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.874

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.868
(Cronbach's alpha = .836)	
<b>McDonald's Omega</b>	.869
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	1.92
Factor 2	-.11
Factor 3	-.13

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
acaself1	0.81	.003	0.80	0.81
acaself2	0.87	.003	0.87	0.88
acaself3	0.81	.003	0.80	0.81

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
acaself1	4.7	1.1	1	6	22256
acaself2	4.1	1.2	1	6	22248
acaself3	4.3	1.2	1	6	22252

[List of scales \(wave 0\)](#)



## Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	774	18	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)

[List of scales \(wave 0\)](#)

## Model and Fit Statistics

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	31794	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			111791
<b>Bayesian Information Criterion (BIC)</b>			111863
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker–Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.884

## Reliability and Dimensionality

<b>Ordinal Cronbach's Alpha</b>	.856
(Cronbach's alpha = .795)	
<b>McDonald's Omega</b>	.860
Test of (one-)dimensionality (parallel analysis)	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	1.89
Factor 2	-.08
Factor 3	-.14

## Standardized factor loadings

Indicators *	Coef.	(SE)	[95% Conf. interval]	
scacad1	0.70	.004	0.70	0.71
scacad2	0.89	.003	0.89	0.90
scacad3	0.85	.003	0.84	0.86

\* **Note:** Replication of 'Scacad'-Scale from TREE1 / PISA2000

## Item descriptives

Indicators *	Mean	Std. dev.	Min.	Max.	Valid Obs.
scacad1	2.9	0.7	1	4	22202
scacad2	2.9	0.7	1	4	22175
scacad3	2.9	0.7	1	4	22168

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

Indicators	Coef.	Cut1	Cut2	Cut3
scacad1	1.87	-4.54	-1.94	2.37
scacad2	3.96	-7.57	-2.86	3.92
scacad3	3.05	-6.36	-2.61	3.41

[List of scales \(wave 0\)](#)

## Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	1571	18	.000

<b>Tests of measurement invariance</b>	chi2	df	p > chi2
Metric invariance (equal factor loadings)	76	4	.000
Strong invariance (plus equal intercepts)	768	4	.000
Strict invariance (plus equal error variances)	427	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	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

## Model and Fit Statistics

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	32226	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			128063
<b>Bayesian Information Criterion (BIC)</b>			128135
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker–Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.888

## Reliability and Dimensionality

<b>Ordinal Cronbach's Alpha</b>	.856
(Cronbach's alpha = .795)	
<b>McDonald's Omega</b>	.861
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	1.90
Factor 2	-.08
Factor 3	-.14

## Standardized factor loadings

Indicators *	Coef.	(SE)	[95% Conf. interval]	
scverb1 **	0.70	0.00	0.69	0.70
scverb2	0.90	0.00	0.89	0.90
scverb3	0.86	0.00	0.85	0.86

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

\*\* **Note:** Reversed Categories for Item Scverb1

## Item descriptives

Indicators *	Mean	Std. dev.	Min.	Max.	Valid Obs.
scverb1 **	3.2	0.8	1	4	22196
scverb2	2.8	0.8	1	4	22173
scverb3	2.9	0.8	1	4	22171

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

Indicators	Coef.	Cut1	Cut2	Cut3
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

[List of scales \(wave 0\)](#)

## Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	621	18	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	57824	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			134733
<b>Bayesian Information Criterion (BIC)</b>			134805
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker–Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.980

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.927
(Cronbach's alpha = .888)	
<b>McDonald's Omega</b>	.930
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
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

\* **Note:** Replication of 'Matcon'-Scale from TREE1 / PISA2000

**Item descriptives**

Indicators *	Mean	Std. dev.	Min.	Max.	Valid 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.	Cut1	Cut2	Cut3
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

[List of scales \(wave 0\)](#)

## Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2
	937	18	.000

<b>Tests of measurement invariance</b>	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. invariant models**

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

\* **Note:** Language-specific models do not converge and the related invariance tests and indices may not be calculated unless the error variance of item matcon2 is constrained to zero.

**Factor score descriptives**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

## Model and Fit Statistics

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	20861	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			68148
<b>Bayesian Information Criterion (BIC)</b>			68214
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker–Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.912

## Reliability and Dimensionality

<b>Ordinal Cronbach's Alpha</b>	.896
(Cronbach's alpha = .849)	
<b>McDonald's Omega</b>	.898
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	2.12
Factor 2	-.08
Factor 3	-.10

## Standardized factor loadings

Indicators	Coef.	(SE)	[95% Conf. interval]	
ictmot6	0.78	.004	0.77	0.79
ictmot7	0.90	.003	0.89	0.90
ictmot8	0.91	.003	0.90	0.91

## Item descriptives

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
ictmot6	2.9	0.9	1	4	11064
ictmot7	2.2	0.9	1	4	11057
ictmot8	2.4	0.9	1	4	11058

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

Indicators	Coef.	Cut1	Cut2	Cut3
ictmot6	2.31	-4.15	-1.43	1.80
ictmot7	3.82	-2.56	1.99	5.06
ictmot8	4.06	-3.74	0.04	4.72

[List of scales \(wave 0\)](#)



## Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	628	18	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	536	2	.000
Baseline vs. saturated	36814	6	.000
2) <b>Root mean squared error (RMSEA)</b>			.110
90% Confidence interval: lower bound			.103
90% Confidence interval: upper bound			.118
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			196455
<b>Bayesian Information Criterion (BIC)</b>			196551
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.985
Tucker–Lewis Index (TLI)			.956
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.020
Coefficient of determination (CD)			.854

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.851
(Cronbach's alpha = .831)	
<b>McDonald's Omega</b>	.852
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	2.23
Factor 2	-.05
Factor 3	-.08
Factor 4	-.16

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
selfeffo1	0.77	.004	0.76	0.77
selfeffo2	0.77	.004	0.76	0.78
selfeffo3	0.80	.003	0.79	0.81
selfeffo4	0.73	.004	0.72	0.74

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
selfeffo1	3.3	0.9	1	4	21801
selfeffo2	3.0	0.9	1	4	21827
selfeffo3	2.8	0.9	1	4	10734
selfeffo4	2.7	0.9	1	4	10755

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

Indicators	Coef.	Cut1	Cut2	Cut3
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

[List of scales \(wave 0\)](#)

## Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	651	28	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	3889	2	.000
Baseline vs. saturated	92426	6	.000
2) <b>Root mean squared error (RMSEA)</b>			.298
90% Confidence interval: lower bound			.290
90% Confidence interval: upper bound			.306
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			147967
<b>Bayesian Information Criterion (BIC)</b>			148063
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.958
Tucker–Lewis Index (TLI)			.874
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.026
Coefficient of determination (CD)			.957

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.947
(Cronbach's alpha = .926)	
<b>McDonald's Omega</b>	.948
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	3.24
Factor 2	.07
Factor 3	-.06
Factor 4	-.06

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
selfeffo5	0.86	.002	0.86	0.87
selfeffo6	0.95	.001	0.95	0.96
selfeffo7	0.88	.002	0.88	0.89
selfeffo8	0.92	.001	0.92	0.93

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
selfeffo5	3.3	0.9	1	4	21809
selfeffo6	3.0	1.0	1	4	21794
selfeffo7	2.8	1.0	1	4	10747
selfeffo8	3.2	0.9	1	4	10730

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

Indicators	Coef.	Cut1	Cut2	Cut3
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

[List of scales \(wave 0\)](#)

Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	506	28	.000

<b>Tests of measurement invariance</b>	chi2	df	p > chi2
Metric invariance (equal factor loadings)	17	6	.010
Strong invariance (plus equal intercepts)	116	6	.000
Strict invariance (plus equal error variances)	238	6	.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	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	229	2	.000
Baseline vs. saturated	30977	6	.000
2) <b>Root mean squared error (RMSEA)</b>			.072
90% Confidence interval: lower bound			.064
90% Confidence interval: upper bound			.080
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			203347
<b>Bayesian Information Criterion (BIC)</b>			203443
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.993
Tucker–Lewis Index (TLI)			.978
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.015
Coefficient of determination (CD)			.836

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.823
(Cronbach's alpha = .803)	
<b>McDonald's Omega</b>	.825
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	2.05
Factor 2	-.07
Factor 3	-.09
Factor 4	-.16

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
selfeff09	0.81	.004	0.80	0.81
selfeff10	0.76	.004	0.75	0.76
selfeff11	0.75	.004	0.74	0.75
selfeff12	0.63	.005	0.62	0.64

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
selfeff09	3.3	0.9	1	4	10752
selfeff10	3.2	0.9	1	4	21783
selfeff11	3.0	1.0	1	4	21802
selfeff12	2.6	0.9	1	4	10751

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

Indicators	Coef.	Cut1	Cut2	Cut3
selfeff09	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

[List of scales \(wave 0\)](#)

Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	3499	28	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	1326	2	.000
Baseline vs. saturated	63299	6	.000
2) <b>Root mean squared error (RMSEA)</b>			.174
90% Confidence interval: lower bound			.166
90% Confidence interval: upper bound			.182
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			178726
<b>Bayesian Information Criterion (BIC)</b>			178821
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.979
Tucker–Lewis Index (TLI)			.937
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.022
Coefficient of determination (CD)			.919

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.917
(Cronbach's alpha = .907)	
<b>McDonald's Omega</b>	.917
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	2.86
Factor 2	.01
Factor 3	-.09
Factor 4	-.10

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
selfeff13	0.87	.002	0.86	0.87
selfeff14	0.84	.002	0.83	0.84
selfeff15	0.89	.002	0.88	0.89
selfeff16	0.83	.003	0.83	0.84

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
selfeff13	2.7	1.0	1	4	21778
selfeff14	2.6	1.0	1	4	10754
selfeff15	2.8	0.9	1	4	21776
selfeff16	2.5	0.9	1	4	10751

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

Indicators	Coef.	Cut1	Cut2	Cut3
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

[List of scales \(wave 0\)](#)



Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	118	28	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

## Model and Fit Statistics

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	1904	5	.000
Baseline vs. saturated	37885	10	.000
2) <b>Root mean squared error (RMSEA)</b>			.186
90% Confidence interval: lower bound			.179
90% Confidence interval: upper bound			.193
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			114426
<b>Bayesian Information Criterion (BIC)</b>			114535
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.950
Tucker–Lewis Index (TLI)			.900
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.035
Coefficient of determination (CD)			.916

## Reliability and Dimensionality

<b>Ordinal Cronbach's Alpha</b>	.914
(Cronbach's alpha = .877)	
<b>McDonald's Omega</b>	.914
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	3.35
Factor 2	.10
Factor 3	-.03
Factor 4	-.10
Factor 5	-.12

## Standardized factor loadings

Indicators	Coef.	(SE)	[95% Conf. interval]	
anxmath1	0.83	.004	0.82	0.84
anxmath2	0.79	.004	0.79	0.80
anxmath3	0.84	.004	0.83	0.85
anxmath4	0.80	.004	0.79	0.81
anxmath5	0.86	.003	0.85	0.86

## Item descriptives

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
anxmath1	2.4	1.0	1	4	10999
anxmath2	1.9	0.9	1	4	10996
anxmath3	1.8	0.9	1	4	10992
anxmath4	2.5	1.0	1	4	10995
anxmath5	2.1	1.0	1	4	10994

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

Indicators	Coef.	Cut1	Cut2	Cut3
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

[List of scales \(wave 0\)](#)

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

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	1137	40	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	689	2	.000
Baseline vs. saturated	20215	6	.000
2) <b>Root mean squared error (RMSEA)</b>			.178
90% Confidence interval: lower bound			.167
90% Confidence interval: upper bound			.189
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			125128
<b>Bayesian Information Criterion (BIC)</b>			125216
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.966
Tucker–Lewis Index (TLI)			.898
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.032
Coefficient of determination (CD)			.863

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.863
(Cronbach's alpha = .831)	
<b>McDonald's Omega</b>	.863
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	2.34
Factor 2	.02
Factor 3	-.11
Factor 4	-.15

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
boredom1	0.78	.005	0.77	0.79
boredom2	0.78	.005	0.77	0.79
boredom3	0.80	.005	0.79	0.81
boredom4	0.77	.005	0.76	0.78

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
boredom1	2.9	1.3	1	5	10877
boredom2	2.6	1.2	1	5	10834
boredom3	2.5	1.3	1	5	10813
boredom4	3.0	1.3	1	5	10877

[List of scales \(wave 0\)](#)

## Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	815	28	.000

<b>Tests of measurement invariance</b>	chi2	df	p > chi2
Metric invariance (equal factor loadings)	15	6	.022
Strong invariance (plus equal intercepts)	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**

Variable name	Mean	Std. 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)

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	79	2	.000
Baseline vs. saturated	27251	6	.000
2) <b>Root mean squared error (RMSEA)</b>			.059
90% Confidence interval: lower bound			.049
90% Confidence interval: upper bound			.071
Probability RMSEA <= 0.05			.073
3) <b>Akaike's Information Criterion (AIC)</b>			120644
<b>Bayesian Information Criterion (BIC)</b>			120732
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.997
Tucker–Lewis Index (TLI)			.992
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.010
Coefficient of determination (CD)			.915

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.895
(Cronbach's alpha = .865)	
<b>McDonald's Omega</b>	.897
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	2.66
Factor 2	-.05
Factor 3	-.08
Factor 4	-.09

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
anger1	0.70	.005	0.69	0.71
anger2	0.89	.003	0.89	0.90
anger3	0.89	.003	0.88	0.89
anger4	0.82	.004	0.82	0.83

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
anger1	2.6	1.2	1	5	10891
anger2	2.4	1.3	1	5	10815
anger3	2.5	1.3	1	5	10810
anger4	2.5	1.4	1	5	10869

[List of scales \(wave 0\)](#)

Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	1045	28	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	191	2	.000
Baseline vs. saturated	23069	6	.000
2) <b>Root mean squared error (RMSEA)</b>			.093
90% Confidence interval: lower bound			.082
90% Confidence interval: upper bound			.104
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			114281
<b>Bayesian Information Criterion (BIC)</b>			114369
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.992
Tucker–Lewis Index (TLI)			.975
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.014
Coefficient of determination (CD)			.892

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.877
(Cronbach's alpha = .845)	
<b>McDonald's Omega</b>	.879
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
	Adjusted eigenvalue
Factor 1	2.47
Factor 2	-.04
Factor 3	-.09
Factor 4	-.11

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
enjoymath1	0.86	.004	0.86	0.87
enjoymath2	0.86	.004	0.86	0.87
enjoymath3	0.73	.005	0.72	0.74
enjoymath4	0.75	.005	0.74	0.76

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
enjoymath1	2.5	1.2	1	5	10880
enjoymath2	2.5	1.2	1	5	10830
enjoymath3	2.3	1.2	1	5	10882
enjoymath4	2.3	1.1	1	5	10823

[List of scales \(wave 0\)](#)



## Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	333	28	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	18182	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			168695
<b>Bayesian Information Criterion (BIC)</b>			168767
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker–Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.825

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.767
(Cronbach's alpha = .731)	
<b>McDonald's Omega</b>	.775
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	1.43
Factor 2	-.09
Factor 3	-.20

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
persev1	0.67	.005	0.66	0.68
persev2	0.87	.005	0.86	0.88
persev3	0.64	.005	0.63	0.65

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
persev1	3.5	0.9	1	5	22268
persev2	3.4	1.0	1	5	22269
persev3	2.9	1.0	1	5	22265

[List of scales \(wave 0\)](#)

## Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	2678	18	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)

[List of scales \(wave 0\)](#)

**Composite descriptives**

Variable name	Mean	Std. dev.	Min.	Max.	Obs.
effper_comp	2.8	0.8	1	4	22265
Share of cases with imputed missing 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

\* **Note:** Original items from TREE1 / PISA2000

[List of scales \(wave 0\)](#)

Composite descriptives		Mean	Std. dev.	Min.	Max.	Obs.
	Variable name					
<b>Big five: extraversion</b>	big5_e_comp	3.3	0.9	1	5	15915
<b>Big five: agreeableness</b>	big5_a_comp	3.5	0.7	1	5	15915
<b>Big five: conscientiousness</b>	big5_c_comp	3.2	0.8	1	5	15915
<b>Big five: neuroticism</b>	big5_n_comp	2.9	0.9	1	5	15915
<b>Big five: openness</b>	big5_o_comp	3.3	0.9	1	5	15915
Share of cases with imputed missing values:		1.4%				

Item descriptives		Mean	Std. dev.	Min.	Max.	Valid obs.
	Indicators					
<b>Big five: extraversion</b>	bigfive1	3.1	1.1	1	5	15890 *
	bigfive6	3.6	1.0	1	5	15851
<b>Big five: agreeableness</b>	bigfive2	3.2	1.1	1	5	15879
	bigfive7	3.3	1.0	1	5	15854 *
	bigfive11	3.8	1.0	1	5	15838
<b>Big five: conscientiousness</b>	bigfive3	2.8	1.1	1	5	15863 *
	bigfive8	3.6	0.9	1	5	15854
<b>Big five: neuroticism</b>	bigfive4	2.8	1.1	1	5	15875 *
	bigfive9	3.0	1.1	1	5	15869
<b>Big five: openness</b>	bigfive5	3.0	1.4	1	5	15875 *
	bigfive10	3.7	1.1	1	5	15864

\* Item category order reversed for composit calculation (see Rammstedt et al., 2007)

[List of scales \(wave 0\)](#)

Composite descriptives		Mean	Std. dev.	Min.	Max.	Obs.
	Variable name					
<b>Internal locus of control</b>						
	loci_comp	4.0	0.7	1	5	15833
<b>External locus of control</b>						
	loce_comp	2.5	0.9	1	5	15833
Share of cases with imputed missing values:		0.6%				

Item descriptives		Mean	Std. dev.	Min.	Max.	Valid Obs.
	Indicators					
<b>Internal locus of control</b>						
	loci1	3.9	0.9	1	5	15811
	loci2	4.2	0.8	1	5	15812
<b>External locus of control</b>						
	loce1	2.3	1.1	1	5	15793
	loce2	2.6	1.1	1	5	15777

[List of scales \(wave 0\)](#)



**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	6673	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			96617
<b>Bayesian Information Criterion (BIC)</b>			96686
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker-Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.668

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.655
(Cronbach's alpha = .560)	
<b>McDonald's Omega</b>	.658
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: Retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue *	
factor 1	.96
factor 2	-.14
factor 3	-.20
* No component with an adjusted eigenvalue ≥ 1	

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
vawe1	0.70	0.01	0.68	0.71
vawe2	0.62	0.01	0.60	0.63
vawe4	0.56	0.01	0.54	0.58

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
vawe1	3.2	0.7	1	4	16066
vawe2	3.7	0.6	1	4	16064
vawe4	2.9	0.9	1	4	16065

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

Indicators	Coef.	Cut1	Cut2	Cut3
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

[List of scales \(wave 0\)](#)



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

## Equality of the

## variance-covariance matrices across ...

## Survey languages

chi2	df	p > chi2
273	18	.000

## Survey settings

chi2	df	p > chi2
237	9	.000

## Survey modes

chi2	df	p > chi2
19	9	.026

## Tests of measurement invariance across ...

## Survey languages

chi2	df	p > chi2
12	4	.016
86	4	.000
90	4	.000

## Survey settings

chi2	df	p > chi2
7	2	.033
21	2	.000
6	2	.050

## Survey modes

chi2	df	p > chi2
1	2	.629
0	2	.815
6	2	.043

Metric invariance (equal factor loadings)

Strong invariance (plus equal intercepts)

Strict invariance (plus equal error variances)

## Configural factor similarity across ...

## Survey languages

	TCC
German vs. French	.997
French vs. Italian	.988
Italian vs. German	.997

## Survey settings

	TCC
classroom vs. unproctored	.999

## Survey modes

	TCC
web vs. PAP	.997

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

## Coefficient of determination

## Survey languages

	CD
German	1.000
French	.994
Italian	.977

## Survey settings

	CD
classroom	1.000
unproctored	.995

## Survey modes

	CD
web	1.000
PAP	.988

## 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)

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	14560	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			80533
<b>Bayesian Information Criterion (BIC)</b>			80602
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker-Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.818

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.789
(Cronbach's alpha = .705)	
<b>McDonald's Omega</b>	.793
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: Retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
factor 1	1.52
factor 2	-.11
factor 3	-.18

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
vawi1	0.72	0.01	0.71	0.73
vawi2	0.85	0.01	0.84	0.86
vawi5	0.67	0.01	0.66	0.68

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
vawi1	3.2	0.7	1	4	16078
vawi2	3.5	0.6	1	4	16071
vawi5	3.5	0.6	1	4	16065

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

Indicators	Coef.	Cut1	Cut2	Cut3
vawi1	1.83	-5.30	-2.78	0.95
vawi2	3.18	-8.88	-6.16	-0.70
vawi5	1.64	-5.46	-3.70	-0.35

[List of scales \(wave 0\)](#)

Tests and Indices of Factorial Invariance across ...

Equality of the

variance-covariance matrices across ...

Survey languages

chi2	df	p > chi2
376	18	.000

Survey settings

chi2	df	p > chi2
413	9	.000

Survey modes

chi2	df	p > chi2
32	9	.000

Tests of measurement invariance across ...

Survey languages

chi2	df	p > chi2
2	4	.727
179	4	.000
81	4	.000

Survey settings

chi2	df	p > chi2
5	2	.075
109	2	.000
3	2	.236

Survey modes

chi2	df	p > chi2
24	2	.000
1	2	.760
5	2	.070

Metric invariance (equal factor loadings)

Strong invariance (plus equal intercepts)

Strict invariance (plus equal error variances)

Configural factor similarity across ...

Survey languages

	TCC
German vs. French	1.000
French vs. Italian	1.000
Italian vs. German	1.000

Survey settings

	TCC
classroom vs. unproctored	.999

Survey modes

	TCC
web vs. PAP	.985

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

Survey languages

	CD
German	1.000
French	1.000
Italian	1.000

Survey settings

	CD
classroom	1.000
unproctored	.999

Survey modes

	CD
web	.999
PAP	.962

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)					

[List of scales \(wave 0\)](#)

Composite descriptives		Mean	Std. dev.	Min.	Max.	Obs.
Variable name						
vafa_comp		3.1	0.8	1	4	16075
Share of cases with imputed missing values:		0.2%				

Item descriptives		Mean	Std. dev.	Min.	Max.	Valid obs.
Indicators						
vafa1		3.3	0.8	1	4	16064
vafa2		3.0	0.9	1	4	16051

[List of scales \(wave 0\)](#)



**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	1110	5	.000
Baseline vs. saturated	13955	10	.000
2) <b>Root mean squared error (RMSEA)</b>			.208
90% Confidence interval: lower bound			.198
90% Confidence interval: upper bound			.218
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			57850
<b>Bayesian Information Criterion (BIC)</b>			57948
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.921
Tucker-Lewis Index (TLI)			.841
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.050
Coefficient of determination (CD)			.893

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.880
(Cronbach's alpha = .844)	
<b>McDonald's Omega</b>	.881
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: Retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
factor 1	2.91
factor 2	.18
factor 3	-.03
factor 4	-.13
factor 5	-.11

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
posl1	0.72	0.01	0.70	0.74
posl2	0.84	0.01	0.83	0.85
posl3	0.78	0.01	0.76	0.79
posl5	0.67	0.01	0.65	0.69
posl6	0.85	0.01	0.84	0.86

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
posl1	5.0	0.9	1	6	5106
posl2	5.4	0.9	1	6	5107
posl3	4.8	1.0	1	6	5106
posl5	4.6	1.1	1	6	5108
posl6	5.0	1.1	1	6	5103

[List of scales \(wave 0\)](#)

Tests and Indices of Factorial Invariance across ...

Equality of the

variance-covariance matrices across ...

Survey languages

chi2	df	p >	chi2
933	40	.000	

Survey settings

chi2	df	p >	chi2
/		/	

Survey modes

chi2	df	p >	chi2
146	20	.000	

Tests of measurement invariance across ...

Survey languages

chi2	df	p >	chi2
9	8	.385	
311	8	.000	
282	8	.000	

Survey settings

chi2	df	p >	chi2
/		/	
/		/	
/		/	

Survey modes

chi2	df	p >	chi2
17	4	.002	
7	4	.113	
20	4	.001	

Metric invariance (equal factor loadings)

Strong invariance (plus equal intercepts)

Strict invariance (plus equal error variances)

Configural factor similarity across ...

Survey languages

	TCC
German vs. French	.999
French vs. Italian	.998
Italian vs. German	1.000

Survey settings

	TCC
classroom vs. unproctored	/

Survey modes

	TCC
web vs. PAP	.998

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

Coefficient of determination

Survey languages

	CD
German	1.000
French	1.000
Italian	.999

Survey settings

	CD
classroom unproctored	/

Survey modes

	CD
web PAP	1.000 .999

Factor score descriptives

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	129	2	.000
Baseline vs. saturated	14527	6	.000
2) <b>Root mean squared error (RMSEA)</b>			.076
90% Confidence interval: lower bound			.065
90% Confidence interval: upper bound			.087
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			145766
<b>Bayesian Information Criterion (BIC)</b>			145853
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.991
Tucker–Lewis Index (TLI)			.974
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.016
Coefficient of determination (CD)			.832

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.807
(Cronbach's alpha = .779)	
<b>McDonald's Omega</b>	.811
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	1.94
Factor 2	-.04
Factor 3	-.11
Factor 4	-.15

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
realref1	0.61	.007	0.60	0.63
realref2	0.65	.007	0.64	0.66
realref3	0.80	.005	0.79	0.81
realref4	0.80	.005	0.79	0.81

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
realref1	3.8	1.5	1	6	11042
realref2	3.9	1.4	1	6	10995
realref3	3.7	1.5	1	6	10984
realref4	4.1	1.5	1	6	11035

[List of scales \(wave 0\)](#)



Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	388	28	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	132	2	.000
Baseline vs. saturated	19790	6	.000
2) <b>Root mean squared error (RMSEA)</b>			.076
90% Confidence interval: lower bound			.066
90% Confidence interval: upper bound			.088
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			143687
<b>Bayesian Information Criterion (BIC)</b>			143775
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.993
Tucker–Lewis Index (TLI)			.980
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.013
Coefficient of determination (CD)			.867

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.858
(Cronbach's alpha = .836)	
<b>McDonald's Omega</b>	.859
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
	Adjusted eigenvalue
Factor 1	2.30
Factor 2	-.06
Factor 3	-.09
Factor 4	-.13

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
disclearn1	0.73	.005	0.72	0.74
disclearn2	0.84	.004	0.83	0.85
disclearn3	0.81	.004	0.80	0.82
disclearn4	0.72	.005	0.71	0.74

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
disclearn1	3.5	1.6	1	6	11049
disclearn2	3.5	1.5	1	6	10986
disclearn3	3.6	1.5	1	6	11002
disclearn4	3.7	1.5	1	6	11006

[List of scales \(wave 0\)](#)

## Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	712	28	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	5090	9	.000
Baseline vs. saturated	36459	15	.000
2) <b>Root mean squared error (RMSEA)</b>			.226
90% Confidence interval: lower bound			.221
90% Confidence interval: upper bound			.231
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			211536
<b>Bayesian Information Criterion (BIC)</b>			211668
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.861
Tucker–Lewis Index (TLI)			.768
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.096
Coefficient of determination (CD)			.912

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.869
(Cronbach's alpha = .849)	
<b>McDonald's Omega</b>	.865
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
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. interval]	
comlearn1	0.54	.007	0.52	0.55
comlearn2	0.51	.008	0.50	0.53
comlearn3	0.62	.006	0.61	0.64
soclearn1	0.83	.004	0.83	0.84
soclearn2	0.88	.003	0.87	0.89
soclearn3	0.87	.003	0.87	0.88

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid 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

[List of scales \(wave 0\)](#)

## Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	580	54	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. dev.	Min.	Max.	Obs.
socomlearn_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)

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	21585	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			100479
<b>Bayesian Information Criterion (BIC)</b>			100545
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker–Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.914

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.904
(Cronbach's alpha = .882)	
<b>McDonald's Omega</b>	.905
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	2.16
Factor 2	-.07
Factor 3	-.11

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
soclearn1	0.85	.003	0.84	0.86
soclearn2	0.92	.003	0.92	0.93
soclearn3	0.84	.004	0.84	0.85

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid 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

[List of scales \(wave 0\)](#)

Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	142	18	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	9617	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			111136
<b>Bayesian Information Criterion (BIC)</b>			111202
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker–Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.816

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.782
(Cronbach's alpha = .751)	
<b>McDonald's Omega</b>	.786
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
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**

Indicators	Mean	Std. dev.	Min.	Max.	Valid 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

[List of scales \(wave 0\)](#)



Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	261	18	.000

<b>Tests of measurement invariance</b>	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**

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

**Factor score descriptives**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	4517	20	.000
Baseline vs. saturated	29913	28	.000
2) <b>Root mean squared error (RMSEA)</b>			.143
90% Confidence interval: lower bound			.139
90% Confidence interval: upper bound			.146
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			286311
<b>Bayesian Information Criterion (BIC)</b>			286487
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.850
Tucker–Lewis Index (TLI)			.789
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.066
Coefficient of determination (CD)			.848

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.841
(Cronbach's alpha = .818)	
<b>McDonald's Omega</b>	.842
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
	Adjusted eigenvalue
Factor 1	3.18
Factor 2	.36
Factor 3	.21
Factor 4	.05
Factor 5	-.10
Factor 6	-.14
Factor 7	-.14
Factor 8	-.20

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
instrlearn1	0.65	.006	0.64	0.67
instrlearn2	0.65	.007	0.63	0.66
instrlearn3	0.48	.008	0.47	0.50
instrlearn4	0.70	.006	0.69	0.71
replearn1	0.67	.006	0.66	0.68
replearn2	0.59	.007	0.58	0.61
replearn3	0.60	.007	0.59	0.62
replearn4	0.70	.006	0.69	0.71

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
instrlearn1	4.6	1.4	1	6	11031
instrlearn2	3.8	1.4	1	6	11001
instrlearn3	3.3	1.5	1	6	10993
instrlearn4	4.6	1.4	1	6	11052
replearn1	4.4	1.4	1	6	11041
replearn2	4.3	1.3	1	6	10990
replearn3	3.6	1.4	1	6	10991
replearn4	4.3	1.4	1	6	11010

[List of scales \(wave 0\)](#)

Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	4066	88	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	605	2	.000
Baseline vs. saturated	9077	6	.000
2) <b>Root mean squared error (RMSEA)</b>			.165
90% Confidence interval: lower bound			.154
90% Confidence interval: upper bound			.176
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			147556
<b>Bayesian Information Criterion (BIC)</b>			147643
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.934
Tucker–Lewis Index (TLI)			.801
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.045
Coefficient of determination (CD)			.741

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.723
(Cronbach's alpha = .683)	
<b>McDonald's Omega</b>	.727
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	1.48
Factor 2	.05
Factor 3	-.12
Factor 4	-.22

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
instrlearn1	0.66	.008	0.65	0.68
instrlearn2	0.68	.008	0.67	0.70
instrlearn3	0.49	.009	0.47	0.51
instrlearn4	0.69	.008	0.67	0.70

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
instrlearn1	4.6	1.4	1	6	11031
instrlearn2	3.8	1.4	1	6	11001
instrlearn3	3.3	1.5	1	6	10993
instrlearn4	4.6	1.4	1	6	11052

[List of scales \(wave 0\)](#)

## Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	2118	28	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	24	2	.000
Baseline vs. saturated	9920	6	.000
2) <b>Root mean squared error (RMSEA)</b>			.032
90% Confidence interval: lower bound			.021
90% Confidence interval: upper bound			.043
Probability RMSEA <= 0.05			.996
3) <b>Akaike's Information Criterion (AIC)</b>			145662
<b>Bayesian Information Criterion (BIC)</b>			145750
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.998
Tucker–Lewis Index (TLI)			.993
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.008
Coefficient of determination (CD)			.774

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.745
(Cronbach's alpha = .713)	
<b>McDonald's Omega</b>	.751
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
	Adjusted eigenvalue
Factor 1	1.58
Factor 2	-.08
Factor 3	-.10
Factor 4	-.16

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
replearn1	0.76	.007	0.75	0.78
replearn2	0.71	.007	0.70	0.72
replearn3	0.49	.009	0.48	0.51
replearn4	0.64	.007	0.63	0.66

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
replearn1	4.4	1.4	1	6	11041
replearn2	4.3	1.3	1	6	10990
replearn3	3.6	1.4	1	6	10991
replearn4	4.3	1.4	1	6	11010

[List of scales \(wave 0\)](#)

## Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	1353	28	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	2443	9	.000
Baseline vs. saturated	31459	15	.000
2) <b>Root mean squared error (RMSEA)</b>			.157
90% Confidence interval: lower bound			.152
90% Confidence interval: upper bound			.162
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			185422
<b>Bayesian Information Criterion (BIC)</b>			185553
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.923
Tucker–Lewis Index (TLI)			.871
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.050
Coefficient of determination (CD)			.879

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.878
(Cronbach's alpha = .854)	
<b>McDonald's Omega</b>	.878
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
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]	
formasp1	0.71	.006	0.70	0.73
formasp2	0.72	.005	0.71	0.73
formasp3	0.75	.005	0.74	0.76
systasp1	0.74	.005	0.73	0.75
systasp2	0.76	.005	0.75	0.77
systasp3	0.75	.005	0.74	0.76

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid 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

[List of scales \(wave 0\)](#)



## Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	478	54	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	12550	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			92905
<b>Bayesian Information Criterion (BIC)</b>			92970
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker–Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.833

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.832
(Cronbach's alpha = .792)	
<b>McDonald's Omega</b>	.832
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	1.70
Factor 2	-.13
Factor 3	-.14

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
systasp1	0.76	.006	0.75	0.78
systasp2	0.81	.005	0.79	0.82
systasp3	0.80	.005	0.79	0.81

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
systasp1	5.0	1.2	1	6	10967
systasp2	4.7	1.2	1	6	10925
systasp3	4.7	1.2	1	6	10975

[List of scales \(wave 0\)](#)

## Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	210	18	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	11712	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			97123
<b>Bayesian Information Criterion (BIC)</b>			97189
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker–Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.822

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.821
(Cronbach's alpha = .791)	
<b>McDonald's Omega</b>	.821
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	1.65
Factor 2	-.14
Factor 3	-.14

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
formasp1	0.78	.006	0.77	0.79
formasp2	0.79	.006	0.78	0.80
formasp3	0.77	.006	0.76	0.78

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid 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

[List of scales \(wave 0\)](#)

## Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	193	18	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. dev.	Min.	Max.	Obs.
formas <sub>p</sub> _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)					

[List of scales \(wave 0\)](#)

## Model and Fit Statistics

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	12713	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			100471
<b>Bayesian Information Criterion (BIC)</b>			100537
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker–Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.843

## Reliability and Dimensionality

<b>Ordinal Cronbach's Alpha</b>	.832
(Cronbach's alpha = .806)	
<b>McDonald's Omega</b>	.833
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	1.72
Factor 2	-.11
Factor 3	-.16

## Standardized factor loadings

Indicators	Coef.	(SE)	[95% Conf. interval]	
schemasp1	0.76	.006	0.75	0.77
schemasp2	0.76	.006	0.75	0.77
schemasp3	0.85	.005	0.84	0.86

## Item descriptives

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
schemasp1	3.9	1.4	1	6	10967
schemasp2	4.0	1.3	1	6	10926
schemasp3	3.7	1.4	1	6	10927

[List of scales \(wave 0\)](#)

## Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	313	18	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

## Model and Fit Statistics

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	316	2	.000
Baseline vs. saturated	20302	6	.000
2) <b>Root mean squared error (RMSEA)</b>			.119
90% Confidence interval: lower bound			.109
90% Confidence interval: upper bound			.131
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			129471
<b>Bayesian Information Criterion (BIC)</b>			129559
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.985
Tucker–Lewis Index (TLI)			.954
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.021
Coefficient of determination (CD)			.866

## Reliability and Dimensionality

<b>Ordinal Cronbach's Alpha</b>	.863
(Cronbach's alpha = .839)	
<b>McDonald's Omega</b>	.864
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	2.33
Factor 2	-.03
Factor 3	-.11
Factor 4	-.13

## Standardized factor loadings

Indicators	Coef.	(SE)	[95% Conf. interval]	
applyasp1	0.80	.005	0.79	0.81
applyasp2	0.79	.005	0.78	0.80
applyasp3	0.73	.005	0.72	0.74
applyasp4	0.81	.005	0.80	0.82

## Item descriptives

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
applyasp1	4.2	1.3	1	6	10982
applyasp2	4.6	1.3	1	6	10933
applyasp3	3.9	1.4	1	6	10958
applyasp4	4.3	1.3	1	6	10924

[List of scales \(wave 0\)](#)



## Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	498	28	.000

<b>Tests of measurement invariance</b>	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)	53	6	.000

**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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

## Model and Fit Statistics

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	5636	20	.000
Baseline vs. saturated	38613	28	.000
2) <b>Root mean squared error (RMSEA)</b>			.164
90% Confidence interval: lower bound			.160
90% Confidence interval: upper bound			.167
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			176245
<b>Bayesian Information Criterion (BIC)</b>			176419
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.854
Tucker–Lewis Index (TLI)			.796
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.078
Coefficient of determination (CD)			.894

## Reliability and Dimensionality

<b>Ordinal Cronbach's Alpha</b>	.873
(Cronbach's alpha = .844)	
<b>McDonald's Omega</b>	.872
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	3.74
Factor 2	.52
Factor 3	.15
Factor 4	-.03
Factor 5	-.07
Factor 6	-.13
Factor 7	-.14
Factor 8	-.14

## Standardized factor loadings

Indicators	Coef.	(SE)	[95% Conf. interval]	
cogself1	0.83	.004	0.82	0.83
cogself2	0.50	.008	0.48	0.51
cogself3	0.56	.007	0.54	0.57
cogself4	0.75	.005	0.74	0.76
cogself5	0.82	.004	0.81	0.83
cogself6	0.66	.006	0.64	0.67
cogself7	0.62	.007	0.61	0.63
cogself8	0.67	.006	0.66	0.68

## Item descriptives

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
cogself1	2.8	0.9	1	4	10443
cogself2	2.6	0.8	1	4	10290
cogself3	2.7	0.9	1	4	10324
cogself4	2.9	0.8	1	4	10423
cogself5	2.8	0.9	1	4	10428
cogself6	2.9	0.8	1	4	10432
cogself7	2.7	0.8	1	4	10271
cogself8	2.7	0.8	1	4	10278

## Parameters of Generalized Structural Equation Model (Ordinal Logit Link)

Indicators	Coef.	Cut1	Cut2	Cut3
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

[List of scales \(wave 0\)](#)

## Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	943	88	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	332	2	.000
Baseline vs. saturated	19997	6	.000
2) <b>Root mean squared error (RMSEA)</b>			.125
90% Confidence interval: lower bound			.114
90% Confidence interval: upper bound			.137
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			85451
<b>Bayesian Information Criterion (BIC)</b>			85538
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.984
Tucker–Lewis Index (TLI)			.951
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.023
Coefficient of determination (CD)			.878

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.864
(Cronbach's alpha = .825)	
<b>McDonald's Omega</b>	.865
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	2.37
Factor 2	-.02
Factor 3	-.08
Factor 4	-.15

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
cogself1	0.83	.004	0.83	0.84
cogself4	0.75	.005	0.74	0.76
cogself5	0.86	.004	0.85	0.86
cogself6	0.69	.006	0.68	0.71

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
cogself1	2.8	0.9	1	4	10443
cogself4	2.9	0.8	1	4	10423
cogself5	2.8	0.9	1	4	10428
cogself6	2.9	0.8	1	4	10432

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

Indicators	Coef.	Cut1	Cut2	Cut3
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

[List of scales \(wave 0\)](#)

Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	351	28	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	1037	2	.000
Baseline vs. saturated	12679	6	.000
2) <b>Root mean squared error (RMSEA)</b>			.224
90% Confidence interval: lower bound			.212
90% Confidence interval: upper bound			.235
Probability RMSEA <= 0.05			.000
3) <b>Akaike's Information Criterion (AIC)</b>			90475
<b>Bayesian Information Criterion (BIC)</b>			90562
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.918
Tucker–Lewis Index (TLI)			.755
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.061
Coefficient of determination (CD)			.816

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.788
(Cronbach's alpha = .743)	
<b>McDonald's Omega</b>	.787
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	1.84
Factor 2	.12
Factor 3	-.18
Factor 4	-.17

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
cogself2	0.60	.008	0.59	0.62
cogself3	0.58	.008	0.56	0.59
cogself7	0.76	.006	0.75	0.78
cogself8	0.81	.006	0.80	0.82

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
cogself2	2.6	0.8	1	4	10290
cogself3	2.7	0.9	1	4	10324
cogself7	2.7	0.8	1	4	10271
cogself8	2.7	0.8	1	4	10278

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

Indicators	Coef.	Cut1	Cut2	Cut3
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
cogself8	2.37	-4.28	-1.14	3.01

[List of scales \(wave 0\)](#)

Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	402	28	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	16993	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			63509
<b>Bayesian Information Criterion (BIC)</b>			63574
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker–Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.892

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.882
(Cronbach's alpha = .842)	
<b>McDonald's Omega</b>	.883
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	2.02
Factor 2	-.09
Factor 3	-.12

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
classman1	0.79	.005	0.78	0.80
classman2	0.85	.004	0.84	0.85
classman3	0.90	.004	0.89	0.90

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
classman1	2.4	0.9	1	4	10313
classman2	2.4	0.9	1	4	10295
classman3	2.3	0.9	1	4	10272

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

Indicators	Coef.	Cut1	Cut2	Cut3
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

[List of scales \(wave 0\)](#)



## Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	267	18	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

## Model and Fit Statistics

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	121	5	.000
Baseline vs. saturated	42736	10	.000
2) <b>Root mean squared error (RMSEA)</b>			.047
90% Confidence interval: lower bound			.040
90% Confidence interval: upper bound			.055
Probability RMSEA <= 0.05			.730
3) <b>Akaike's Information Criterion (AIC)</b>			94824
<b>Bayesian Information Criterion (BIC)</b>			94932
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			.997
Tucker–Lewis Index (TLI)			.995
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.007
Coefficient of determination (CD)			.936

## Reliability and Dimensionality

<b>Ordinal Cronbach's Alpha</b>	.935
(Cronbach's alpha = .907)	
<b>McDonald's Omega</b>	.935
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	3.63
Factor 2	-.04
Factor 3	-.06
Factor 4	-.05
Factor 5	-.06

## Standardized factor loadings

Indicators	Coef.	(SE)	[95% Conf. interval]	
indsup1	0.86	.003	0.85	0.86
indsup2	0.89	.003	0.88	0.89
indsup3	0.87	.003	0.87	0.88
indsup4	0.87	.003	0.86	0.87
indsup5	0.82	.004	0.81	0.83

## Item descriptives

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
indsup1	2.7	0.9	1	4	10434
indsup2	3.0	0.9	1	4	10436
indsup3	2.8	0.9	1	4	10464
indsup4	2.8	0.9	1	4	10439
indsup5	2.9	0.9	1	4	10423

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

Indicators	Coef.	Cut1	Cut2	Cut3
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

[List of scales \(wave 0\)](#)

Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	515	40	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	9348	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			71991
<b>Bayesian Information Criterion (BIC)</b>			72056
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker–Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.829

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.765
(Cronbach's alpha = .712)	
<b>McDonald's Omega</b>	.780
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	1.47
Factor 2	-.08
Factor 3	-.18

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
instqual1	0.80	.007	0.79	0.82
instqual2	0.85	.007	0.84	0.87
instqual3	0.53	.008	0.51	0.54

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
instqual1	2.8	0.9	1	4	10426
instqual2	2.8	0.8	1	4	10285
instqual3	2.6	0.9	1	4	10266

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

Indicators	Coef.	Cut1	Cut2	Cut3
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

[List of scales \(wave 0\)](#)

Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	432	18	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	11000	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			76347
<b>Bayesian Information Criterion (BIC)</b>			76413
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker–Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.834

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.806
(Cronbach's alpha = .757)	
<b>McDonald's Omega</b>	.810
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	1.60
Factor 2	-.10
Factor 3	-.17

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
intsit1	0.75	.006	0.73	0.76
intsit2	0.68	.007	0.67	0.70
intsit3	0.86	.006	0.85	0.87

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
intsit1	2.6	0.9	1	4	10891
intsit2	2.3	0.9	1	4	10836
intsit3	2.4	0.9	1	4	10897

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

Indicators	Coef.	Cut1	Cut2	Cut3
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

[List of scales \(wave 0\)](#)

## Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	801	18	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	10030	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			72281
<b>Bayesian Information Criterion (BIC)</b>			72346
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker–Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.809

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.799
(Cronbach's alpha = .748)	
<b>McDonald's Omega</b>	.800
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	1.55
Factor 2	-.13
Factor 3	-.17

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
persuppauto1	0.74	.006	0.73	0.76
persuppauto2	0.82	.006	0.81	0.83
persuppauto3	0.70	.007	0.69	0.72

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
persuppauto1	2.7	0.9	1	4	10665
persuppauto2	2.9	0.9	1	4	10627
persuppauto3	3.0	0.8	1	4	10655

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

Indicators	Coef.	Cut1	Cut2	Cut3
persuppauto1	2.02	-3.46	-0.78	2.16
persuppauto2	2.67	-4.43	-1.76	2.02
persuppauto3	1.88	-4.13	-1.81	1.12

[List of scales \(wave 0\)](#)



Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	229	18	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

## Model and Fit Statistics

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	19504	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			61112
<b>Bayesian Information Criterion (BIC)</b>			61178
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker–Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.951

## Reliability and Dimensionality

<b>Ordinal Cronbach's Alpha</b>	.888
(Cronbach's alpha = .842)	
<b>McDonald's Omega</b>	.892
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted 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

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
persuppcomp1	2.9	0.8	1	4	10639
persuppcomp2	2.7	0.9	1	4	10639
persuppcomp3	3.0	0.8	1	4	10645

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

Indicators	Coef.	Cut1	Cut2	Cut3
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

[List of scales \(wave 0\)](#)

## Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	281	18	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. dev.	Min.	Max.	Obs.
persupcomp_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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	15653	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			69393
<b>Bayesian Information Criterion (BIC)</b>			69459
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker–Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.886

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.858
(Cronbach's alpha = .814)	
<b>McDonald's Omega</b>	.862
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	1.90
Factor 2	-.08
Factor 3	-.13

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
persocincl1	0.89	.004	0.88	0.89
persocincl2	0.70	.006	0.69	0.71
persocincl3	0.87	.004	0.86	0.88

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
persocincl1	2.7	0.9	1	4	10635
persocincl2	2.7	0.9	1	4	10640
persocincl3	2.4	0.9	1	4	10632

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

Indicators	Coef.	Cut1	Cut2	Cut3
persocincl1	3.65	-4.81	-1.00	3.37
persocincl2	1.82	-3.18	-0.78	2.11
persocincl3	3.34	-2.89	0.28	4.36

[List of scales \(wave 0\)](#)

## Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	1205	18	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

**Model and Fit Statistics**

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	19804	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			53455
<b>Bayesian Information Criterion (BIC)</b>			53521
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker–Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.946

**Reliability and Dimensionality**

<b>Ordinal Cronbach's Alpha</b>	.834
(Cronbach's alpha = .776)	
<b>McDonald's Omega</b>	.859
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	1.94
Factor 2	-.02
Factor 3	-.08

**Standardized factor loadings**

Indicators	Coef.	(SE)	[95% Conf. interval]	
apprmath1	0.92	.004	0.92	0.93
apprmath2	0.96	.004	0.95	0.97
apprmath3	0.53	.007	0.51	0.54

**Item descriptives**

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
apprmath1	2.0	0.7	1	4	10778
apprmath2	2.0	0.7	1	4	10775
apprmath3	2.7	0.8	1	4	10776

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

Indicators	Coef.	Cut1	Cut2	Cut3
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

[List of scales \(wave 0\)](#)

## Tests and Indices of Factorial Invariance across Survey Languages

Equality of variance-covariance matrices	chi2	df	p > chi2
	320	9	.000

<b>Tests of measurement invariance</b>	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. invariant models**

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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

## Model and Fit Statistics

1) <b>Likelihood-ratio tests</b>	chi2	df	p > chi2
Model vs. saturated	0	0	
Baseline vs. saturated	30122	3	.000
2) <b>Root mean squared error (RMSEA)</b>			.000
90% Confidence interval: lower bound			.000
90% Confidence interval: upper bound			.000
Probability RMSEA <= 0.05			1.000
3) <b>Akaike's Information Criterion (AIC)</b>			84033
<b>Bayesian Information Criterion (BIC)</b>			84105
4) <b>Baseline comparison</b>			
Comparative Fit Index (CFI)			1.000
Tucker–Lewis Index (TLI)			1.000
5) <b>Size of residuals</b>			
Stand. root mean squared residual (SRMR)			.000
Coefficient of determination (CD)			.923

## Reliability and Dimensionality

<b>Ordinal Cronbach's Alpha</b>	.819
(Cronbach's alpha = .648)	
<b>McDonald's Omega</b>	.837
<b>Test of (one-)dimensionality (parallel analysis)</b>	
Criterion: retain factors with adj. eigenvalue > 0	
Adjusted eigenvalue	
Factor 1	1.77
Factor 2	-.03
Factor 3	-.14

## Standardized factor loadings

Indicators	Coef.	(SE)	[95% Conf. interval]	
truancy1	0.84	.004	0.83	0.85
truancy2 *	0.95	.004	0.94	0.96
truancy3 *	0.56	.005	0.55	0.57

\* **Note:** Original item from TREE1 / PISA2000

## Item descriptives

Indicators	Mean	Std. dev.	Min.	Max.	Valid Obs.
truancy1	1.1	0.4	1	4	22242
truancy2 *	1.2	0.5	1	4	22245
truancy3 *	1.5	0.8	1	4	22251

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

Indicators	Coef.	Cut1	Cut2	Cut3
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 \(wave 0\)](#)



Tests and Indices of Factorial Invariance across Survey Languages

<b>Equality of variance-covariance matrices</b>	chi2	df	p > chi2
	2001	18	.000

<b>Tests of measurement invariance</b>	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**

Variable name	Mean	Std. 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)					

[List of scales \(wave 0\)](#)

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