

Analyseskript und zusätzliche Resultate zu:

Ben Jann und Sandra Hupka-Brunner: Warum werden Frauen so selten MINT-Fachkräfte? Zur Bedeutung der Differenz zwischen mathematischen Kompetenzen und Selbstkonzept

Software: Stata/MP 16.0

Zusatzmodule: coefplot, colspace, estout, fre, grstyle, moremata, oxaca, palettes

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1 Setup

```
. // Allgemein
. about
Stata/MP 16.0 for Mac (64-bit Intel)
Revision 23 Aug 2019
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    Serial number: 501606206203
    Licensed to: Ben Jann
                UniBe

. version 16.0
. clear all
. set linesize 110
. set type double
.
. // Grafikeinstellungen
. set scheme s2mono
. grstyle init
. grstyle set plain, hor grid compact
. grstyle set color mono
. grstyle set color mono, opacity(60): p#bar
. grstyle set linewidth thin: xyline
. grstyle set lpattern dash: xyline
. grstyle set legend 10, inside nobox
```

2 Skript für MINT-Klassifikation

```
program mint08
    args mint isco08
    quietly {
        generate byte `mint' = 0 if `isco08'<.
        // Informatik
        // 1.1*: Informatik (Informatik, Informatik- und Softwareingenieurwesen, Wirtschafts- und
        //      Betriebsinformatik)
        replace `mint' = 11 if `isco08'==2500 // Akademische und vergleichbare Fachkräfte in der Informations-
        // und Kommunikationstechnologie, onA
        replace `mint' = 11 if `isco08'==2511 // Systemanalytiker
        replace `mint' = 11 if `isco08'==2512 // Softwareentwickler
        replace `mint' = 11 if `isco08'==2513 // Web- und Multimediaentwickler
        replace `mint' = 11 if `isco08'==2514 // Anwendungsprogrammierer
        replace `mint' = 11 if `isco08'==2529 // Akademische und vergleichbare Fachkräfte für Datenbanken und
        // Netzwerke, anderweitig nicht genannt
        replace `mint' = 11 if `isco08'==3514 // Webmaster
        // Technik
        // 1.2*: Elektrotechnik (Elektrotechnik und Elektroingenieurwesen)
        replace `mint' = 12 if `isco08'==2151 // Ingenieure im Bereich Elektrotechnik
        replace `mint' = 12 if `isco08'==3113 // Elektrotechniker
        // 1.3*: Maschinenteknik (Maschinenteknik und Maschineningenieurwesen, Aviatik, Automobiltechnik)
        replace `mint' = 13 if `isco08'==2144 // Maschinenbauingenieure
        replace `mint' = 13 if `isco08'==3115 // Maschinenbautechniker
        // 1.4*: Mikrotechnik (Elektronik und Elektroingenieurwesen, Mikrotechnik und Mikrotechnikingenieurwesen,
        //      Systemtechnik, Mechatronik, Kommunikationssysteme, Telekommunikationstechnik)
        replace `mint' = 14 if `isco08'==2152 // Ingenieure im Bereich Elektronik
        replace `mint' = 14 if `isco08'==3114 // Techniker im Bereich Elektronik
        replace `mint' = 14 if `isco08'==3510 // Techniker für den Betrieb von Informations- und
        // Kommunikationstechnologie und für die Anwenderbetreuung, onA
        replace `mint' = 14 if `isco08'==3521 // Techniker für Rundfunk und audiovisuelle Medien
    }
```

```

replace `mint' = 14 if `isco08'==3522 // Telekommunikationstechniker
// 1.5*: Wirtschaftsingenieurwesen (Betriebs- und Produktionswissenschaften, Wirtschaftsingenieurwesen,
// Medieningenieurwesen)
replace `mint' = 15 if `isco08'==2141 // Wirtschafts- und Produktionsingenieure
replace `mint' = 15 if `isco08'==3122 // Produktionsleiter bei der Herstellung von Waren
// 1.6*: Anderes aus Technik & IT (Materialwissenschaften, Holztechnik, Werkstoffe- und
// Materialingenieurwesen, andere/fachrichtungsübergreifende Richtungen aus Technik & IT)
replace `mint' = 16 if `isco08'==2140 // Ingenieurwissenschaftler (ohne Elektrotechnik, Elektronik und
// Telekommunikation), onA
replace `mint' = 16 if `isco08'==2149 // Ingenieure, anderweitig nicht genannt
replace `mint' = 16 if `isco08'==3110 // Material- und ingenieurtechnische Fachkräfte, onA
replace `mint' = 16 if `isco08'==3118 // Technische Zeichner
replace `mint' = 16 if `isco08'==3119 // Material- und ingenieurtechnische Fachkräfte, anderweitig nicht
// genannt

// Bauwesen
// 2.1*: Bau (Bauingenieurwesen, Gebäudetechnik, Heizungs-, Lüftungs- und Klimaanlageingenieurwesen)
replace `mint' = 21 if `isco08'==2142 // Bauingenieure
replace `mint' = 21 if `isco08'==3112 // Bautechniker
replace `mint' = 21 if `isco08'==3120 // Produktionsleiter im Bergbau, bei der Herstellung von Waren und
// im Bau, onA
replace `mint' = 21 if `isco08'==3123 // Bauleiter
replace `mint' = 21 if `isco08'==3132 // Steuerer von Verbrennungs- und Wasserbehandlungsanlagen
// 2.2 : Planung und Vermessung (Geomatik, Geodäsie, Vermessungsingenieurwesen, Umweltingenieurwesen,
// Kulturechnik und Kulturingenieurwesen, Raum-, Landschafts-, Siedlungs- & Ortsplanung)
replace `mint' = 22 if `isco08'==2143 // Umweltschutzingenieure
replace `mint' = 22 if `isco08'==2160 // Architekten, Raum-, Stadt- und Verkehrsplaner,
// Vermessungsingenieure und Designer, onA
replace `mint' = 22 if `isco08'==2164 // Raum-, Stadt- und Verkehrsplaner
replace `mint' = 22 if `isco08'==2165 // Kartografen und Vermessungsingenieure
// 2.3 : Architektur (Architektur)
replace `mint' = 23 if `isco08'==2161 // Architekten
replace `mint' = 23 if `isco08'==2162 // Landschaftsarchitekten
// 2.4 : Anderes aus Bauwesen (Andere / fachrichtungsübergreifende Richtungen aus dem Bauwesen)
// Chemie & Life Sciences
// 3.1*: Chemie (Chemieingenieurwesen, Lebensmittelwissenschaften, Lebensmitteltechnologie,
// Verfahrenstechnik)
replace `mint' = 31 if `isco08'==2113 // Chemiker
replace `mint' = 31 if `isco08'==2145 // Chemieingenieure
replace `mint' = 31 if `isco08'==3111 // Chemo- und Physikotechniker
replace `mint' = 31 if `isco08'==3116 // Chemiebetriebs- und Verfahrenstechniker
// 3.2 : Biotechnologie (Biotechnologie, Life Sciences, Molecular Life Sciences, Life Science
// Technologies, Lebensmittelwissenschaften, Lebensmitteltechnologie)
replace `mint' = 32 if `isco08'==3141 // Biotechniker (ohne medizinische Fachberufe)
// 3.3 : Gesundheit (Pharmazie, Pharmatechnologie, Medizinaltechnik, Medizinaltechnologie)
replace `mint' = 33 if `isco08'==3212 // Medizintechniker im Bereich Labor und Pathologie
replace `mint' = 33 if `isco08'==3213 // Pharmazeutisch-technische Assistenten
replace `mint' = 33 if `isco08'==3214 // Medizinische und zahnmedizinische Prothetiktechniker
replace `mint' = 33 if `isco08'==3255 // Physiotherapeutische Techniker und Assistenten
// 3.4 : Anderes aus C&LS (Andere / fachrichtungsübergreifende Richtungen aus C&LS)
// Andere MINT
// 4.1 : Geografie (Geografie, Umwelt(natur)wissenschaften, andere/fachrichtungsübergreifende
// Richtungen aus den Geo- und Umweltwissenschaften)
replace `mint' = 41 if `isco08'==2114 // Geologen und Geophysiker
replace `mint' = 41 if `isco08'==2131 // Biologen, Botaniker, Zoologen und verwandte Berufe
replace `mint' = 41 if `isco08'==2133 // Umweltwissenschaftler
// 4.2*: Exakte Wissenschaften (Mathematik, Statistik, Physik, Astronomie,
// andere/fachrichtungsübergreifende exakte Wissenschaften)
replace `mint' = 42 if `isco08'==2111 // Physiker und Astronomen
replace `mint' = 42 if `isco08'==2112 // Meteorologen
replace `mint' = 42 if `isco08'==2120 // Mathematiker, Versicherungsmathematiker und Statistiker
// 4.3 : Andere MINT (Agronomie, Forstwirtschaft, andere/kategorieübergreifende MINT-Fachrichtungen)
replace `mint' = 43 if `isco08'==2132 // Agrar-, Forst- und Fischereiwissenschaftler und -berater
replace `mint' = 43 if `isco08'==3142 // Agrartechniker
replace `mint' = 43 if `isco08'==3143 // Forsttechniker
}
// labels
lab def `mint' ///
0 "nicht MINT" ///

```

```

11 "MINT* 1.1: Informatik" ///
12 "MINT* 1.2: Elektrotechnik" ///
13 "MINT* 1.3: Maschinentechnik" ///
14 "MINT* 1.4: Mikrotechnik" ///
15 "MINT* 1.5: Wirtschaftsingenieurwesen" ///
16 "MINT* 1.6: Anderes aus Technik & IT" ///
21 "MINT* 2.1: Bau" ///
22 "MINT 2.2: Planung und Vermessung" ///
23 "MINT 2.3: Architektur" ///
24 "MINT 2.4: Anderes aus Bauwesen" ///
31 "MINT* 3.1: Chemie" ///
32 "MINT 3.2: Biotechnologie" ///
33 "MINT 3.3: Gesundheit" ///
34 "MINT 3.4: Anderes aus C&LS" ///
41 "MINT 4.1: Geografie" ///
42 "MINT* 4.2: Exakte Wissenschaften" ///
43 "MINT 4.3: Andere MINT", replace
lab val `mint' `mint'
fre `mint'
end

```

3 Datenaufbereitung

```

. use ../Data/1004_UGK16_Data_E_v1.0.0.dta
( )

```

3.1 Variablen generieren

```

. // math score; using weighted likelihood estimates (WLE)
. su wlem

```

Variable	Obs	Mean	Std. Dev.	Min	Max
wlem	22,141	-.0825861	1.396275	-5.709	5.214

```

. // mathematical self-concept; using provided scale
. d matcon*

```

variable name	storage type	display format	value label	variable label
matcon1	int	%8.0g	MATCON1	Get good marks in mathematics (Self-concept in mathematics)
matcon2	int	%8.0g	MATCON2	Mathematics one of best subjects (Self-concept in mathematics)
matcon3	int	%8.0g	MATCON3	Have always done well in mathematics (Self-concept in mathematics)
matcon_fs	double	%12.0g	MATCON_F	Math self-concept

```

. gen double matcon = matcon_fs if matcon_fs>-900
(230 missing values generated)

```

```

. su matcon

```

Variable	Obs	Mean	Std. Dev.	Min	Max
matcon	22,193	-.0171532	.964152	-1.705292	1.615017

```

. // mathematical self-efficacy; using simple average of sub-scales
. d selfeff*

```

variable name	storage type	display format	value label	variable label
selfeff01	int	%8.0g	SELFEFF0	Confidence calculating a 30 percent discount (Specific self-efficacy: algebra)
selfeff02	int	%8.0g	V1234_A	Confidence calculating square metres (Specific self-efficacy: numeracy)
selfeff03	int	%8.0g	V1235_A	Confidence calculate petrol consumption car (Specific

```

selfeff04    int    %8.0g    V1236_A    self-efficacy: numeracy)
Confidence find distance between places (Specific self-efficacy:
numeracy)
selfeff05    int    %8.0g    V1237_A    Confidence solving an equation like 3x+5= 17 (Specific
self-efficacy: algebra)
selfeff06    int    %8.0g    V1238_A    Confidence solve equation 2(x+3)=(x+3)(x-3) (Specific
self-efficacy: algebra)
selfeff07    int    %8.0g    V1239_A    Confidence simplifying expression 2a(5a-3b)^2 (Specific
self-efficacy: algebra)
selfeff08    int    %8.0g    V1240_A    Confidence solving equation like 2x-3=4x+5 (Specific
self-efficacy: algebra)
selfeff09    int    %8.0g    V1241_A    Confidence applying the Pythagorean theorem (Specific
self-efficacy: geometry)
selfeff10    int    %8.0g    SELFEFF1    Confidence construct perpendicular bisector (Specific
self-efficacy: geometry)
selfeff11    int    %8.0g    V1243_A    Confidence calculate area of a parallelogram (Specific
self-efficacy: geometry)
selfeff12    int    %8.0g    V1244_A    Confidence constructing centroid of triangle (Specific
self-efficacy: geometry)
selfeff13    int    %8.0g    V1245_A    Confidence probability throw six twice (Specific self-efficacy:
probability)
selfeff14    int    %8.0g    V1246_A    Confidence probability win lottery (Specific self-efficacy:
probability)
selfeff15    int    %8.0g    V1247_A    Confidence probability sweets same colour (Specific
self-efficacy: probability)
selfeff16    int    %8.0g    V1248_A    Confidence probability pupil same birthday (Specific
self-efficacy: probability)
selfeffa_fs  double %12.0g    SELFEFFA    Specific self-efficacy: numeracy
selfeffb_fs  double %12.0g    SELFEFFB    Specific self-efficacy: algebra
selfeffc_fs  double %12.0g    SELFEFFC    Specific self-efficacy: geometry
selfeffd_fs  double %12.0g    SELFEFFD    Specific self-efficacy: probability

```

```

. gen double selfeff = (selfeffa_fs + selfeffb_fs + selfeffc_fs + selfeffd_fs)/4 ///
> if selfeffa_fs>-900 & selfeffb_fs>-900 & selfeffc_fs>-900 & selfeffd_fs>-900
(572 missing values generated)

```

```

. su selfeff

```

Variable	Obs	Mean	Std. Dev.	Min	Max
selfeff	21,851	-.048102	.7197313	-2.283963	1.474471

```

.
. // MINT classification of expected job at 30
. // (based on Gehrig et al. 2019)
. gen int isco08 = real(jobat30_isco08)
. replace isco08 = . if isco08<0
(5,985 real changes made, 5,985 to missing)
. mint08 MINT isco08

```

```

MINT

```

		Freq.	Percent	Valid	Cum.
Valid	0 nicht MINT	13846	61.75	84.23	84.23
	11 MINT* 1.1: Informatik	732	3.26	4.45	88.68
	12 MINT* 1.2: Elektrotechnik	65	0.29	0.40	89.08
	13 MINT* 1.3: Maschinentechnik	81	0.36	0.49	89.57
	14 MINT* 1.4: Mikrotechnik	55	0.25	0.33	89.91
	15 MINT* 1.5: Wirtschaftsingenieurwesen	7	0.03	0.04	89.95
	16 MINT* 1.6: Anderes aus Technik & IT	483	2.15	2.94	92.89
	21 MINT* 2.1: Bau	149	0.66	0.91	93.79
	22 MINT 2.2: Planung und Vermessung	76	0.34	0.46	94.26
	23 MINT 2.3: Architektur	465	2.07	2.83	97.09
	31 MINT* 3.1: Chemie	121	0.54	0.74	97.82
	32 MINT 3.2: Biotechnologie	12	0.05	0.07	97.90
	33 MINT 3.3: Gesundheit	130	0.58	0.79	98.69
	41 MINT 4.1: Geografie	104	0.46	0.63	99.32
	42 MINT* 4.2: Exakte Wissenschaften	80	0.36	0.49	99.81
	43 MINT 4.3: Andere MINT	32	0.14	0.19	100.00
	Total	16438	73.31	100.00	

Missing .	5985	26.69
Total	22423	100.00

```
. drop isco08
. gen byte MINT1 = inlist(MINT,11,12,13,14,15,16,21,31,42) if MINT<. // narrow (*)
(5,985 missing values generated)
. gen byte MINT2 = MINT>0 if MINT<. // wide
(5,985 missing values generated)
. su MINT1 MINT2
```

Variable	Obs	Mean	Std. Dev.	Min	Max
MINT1	16,438	.1078598	.3102127	0	1
MINT2	16,438	.1576834	.364455	0	1

```
. replace MINT = 99 if inlist(jobat30_isco08,"-950", "-957", "-960", "-997", "-998", "-999")
(5,455 real changes made)
. // not codable, houseman/housewife, not in list, refusal, don't know, no answer
. fre jobat30_isco08 if MINT==.
jobat30_isco08 — Expected job at 30 (ISCO-08 classification)
```

	Freq.	Percent	Valid	Cum.
Valid -903	84	15.85	15.85	15.85
-904	446	84.15	84.15	100.00
Total	530	100.00	100.00	

```
. // personal characteristics
. // - gender; using self-reported gender
. fre sex
sex — Gender (self-reported)
```

	Freq.	Percent	Valid	Cum.
Valid -999 No answer /missing [item]	12	0.05	0.05	0.05
-903 Not administered [item]	84	0.37	0.37	0.43
1 Female	10981	48.97	48.97	49.40
2 Male	11346	50.60	50.60	100.00
Total	22423	100.00	100.00	

```
. gen byte female = (sex==1) if sex>0
(96 missing values generated)
```

```
. // - immigration status
. fre immig
immig — Immigration status
```

	Freq.	Percent	Valid	Cum.
Valid -999 No answer /missing [item]	184	0.82	0.82	0.82
-904 Not administered [break-off]	3	0.01	0.01	0.83
-903 Not administered [item]	84	0.37	0.37	1.21
1 Native with 1+ Swiss parent	15863	70.74	70.74	71.95
2 Second generation	4191	18.69	18.69	90.64
3 First generation	2098	9.36	9.36	100.00
Total	22423	100.00	100.00	

```
. replace immig = . if immig<0
(271 real changes made, 271 to missing)
```

```
. // - absenteeism
. d truancy*
storage display value
```

variable name	type	format	label	variable label
truancy1	int	%8.0g	TRUANCY1	Skip whole school day (Absenteeism (truancy))
truancy2	int	%8.0g	TRUANCY2	Skip classes (Absenteeism (truancy))
truancy3	int	%8.0g	TRUANCY3	Late for school (Absenteeism (truancy))
truancy_fs	double	%12.0g	TRUANCY	Absenteeism (truancy)

```
. gen double truancy = truancy_fs if truancy_fs>-900
(169 missing values generated)
```

```
. su truancy
```

Variable	Obs	Mean	Std. Dev.	Min	Max
truancy	22,254	-.0215044	.6989421	-.4945608	2.845646

```
. // - received extra math coaching
. replace coachmath = . if coachmath<0
(207 real changes made, 207 to missing)
```

```
. fre coachmath
coachmath — Mathematics (Private tutoring / learning support: frequency)
```

		Freq.	Percent	Valid	Cum.
Valid	1 No, never	16852	75.15	75.86	75.86
	2 Yes, sometimes	3677	16.40	16.55	92.41
	3 Yes, regularly	1687	7.52	7.59	100.00
	Total	22216	99.08	100.00	
Missing	.	207	0.92		
Total		22423	100.00		

```
. // - ever repeated a grade
. gen byte repeat = inlist(graderep,2,3) if graderep>0
(134 missing values generated)
```

```
. fre repeat
repeat
```

		Freq.	Percent	Valid	Cum.
Valid	0	18259	81.43	81.92	81.92
	1	4030	17.97	18.08	100.00
	Total	22289	99.40	100.00	
Missing	.	134	0.60		
Total		22423	100.00		

```
. // - grades in math
. gen grademath = markmath if markmath>0
(964 missing values generated)
```

```
. su grademath
```

Variable	Obs	Mean	Std. Dev.	Min	Max
grademath	21,459	4.616175	.7223925	1	6

```
. // - school type
. gen byte gym = schoolgym==1 if schoolgym>-904 // assigning "not administered [item]" to zero
(97 missing values generated)
```

```
. su gym
```

Variable	Obs	Mean	Std. Dev.	Min	Max
gym	22,326	.2252083	.4177288	0	1

```
. // family resources
. // - number of books (transform to numeric scale?)
```

```
. fre books
books — Number of books
```

		Freq.	Percent	Valid	Cum.
Valid	-999 No answer /missing [item]	53	0.24	0.24	0.24
	-904 Not administered [break-off]	5	0.02	0.02	0.26
	-903 Not administered [item]	84	0.37	0.37	0.63
1	None	712	3.18	3.18	3.81
2	1-10 books	2742	12.23	12.23	16.04
3	11-50 books	5459	24.35	24.35	40.38
4	51-100 books	4750	21.18	21.18	61.57
5	101-250 books	3997	17.83	17.83	79.39
6	251-500 books	2625	11.71	11.71	91.10
7	More than 500 books	1996	8.90	8.90	100.00
	Total	22423	100.00	100.00	

```
. replace books = . if books<0
(142 real changes made, 142 to missing)
```

```
. // - highest occupational status
. d hisei88
```

variable name	storage type	display format	value label	variable label
hisei88	int	%8.0g	HISEI88	Highest occupational code parents (ISEI-88 status)

```
. gen int isei = hisei88 if hisei88>0
(884 missing values generated)
```

```
. su isei
```

Variable	Obs	Mean	Std. Dev.	Min	Max
isei	21,539	50.07271	16.08841	16	90

```
. // parents: labor market participation, MINT occupation, value of math
```

```
. // - mother works full 50% or more
```

```
. fre emplm
```

```
emplm — Employment status mother
```

		Freq.	Percent	Valid	Cum.
Valid	-999 No answer /missing [item]	64	0.29	0.29	0.29
	-903 Not administered [item]	287	1.28	1.28	1.57
1	Yes, full-time	5346	23.84	23.84	25.41
2	Yes, part-time (50% or more)	7665	34.18	34.18	59.59
3	Yes, part-time (less than 50%)	4500	20.07	20.07	79.66
4	No, not working but looking for a job	759	3.38	3.38	83.04
5	No, not working (e.g., homemaker, retiree, disability pensioner)	3802	16.96	16.96	100.00
	Total	22423	100.00	100.00	

```
. gen byte lfpm = inlist(emplm,1,2) if emplm>0
(351 missing values generated)
```

```
. su lfpm
```

Variable	Obs	Mean	Std. Dev.	Min	Max
lfpm	22,072	.5894799	.4919393	0	1

```
. // - MINT occupation
```

```
. gen int isco08 = real(occupm_isco08)
```

```
. replace isco08 = . if isco08<0
```

```
(2,929 real changes made, 2,929 to missing)
```

```
. mint08 MINTm isco08
```


MINTm

		Freq.	Percent	Valid	Cum.
Valid	0 nicht MINT	18780	83.75	96.34	96.34
	11 MINT* 1.1: Informatik	70	0.31	0.36	96.70
	12 MINT* 1.2: Elektrotechnik	4	0.02	0.02	96.72
	13 MINT* 1.3: Maschinentechnik	7	0.03	0.04	96.75
	14 MINT* 1.4: Mikrotechnik	11	0.05	0.06	96.81
	15 MINT* 1.5: Wirtschaftsingenieurwesen	11	0.05	0.06	96.87
	16 MINT* 1.6: Anderes aus Technik & IT	131	0.58	0.67	97.54
	21 MINT* 2.1: Bau	24	0.11	0.12	97.66
	22 MINT 2.2: Planung und Vermessung	13	0.06	0.07	97.73
	23 MINT 2.3: Architektur	64	0.29	0.33	98.06
	31 MINT* 3.1: Chemie	110	0.49	0.56	98.62
	32 MINT 3.2: Biotechnologie	11	0.05	0.06	98.68
	33 MINT 3.3: Gesundheit	208	0.93	1.07	99.74
	41 MINT 4.1: Geografie	35	0.16	0.18	99.92
	42 MINT* 4.2: Exakte Wissenschaften	11	0.05	0.06	99.98
	43 MINT 4.3: Andere MINT	4	0.02	0.02	100.00
	Total	19494	86.94	100.00	
Missing	.	2929	13.06		
Total		22423	100.00		

```

. replace MINTm = 0 if isco08>=. // no info => assume no MINT
(2,929 real changes made)
. drop isco08
. gen byte MINT1m = inlist(MINTm,11,12,13,14,15,16,21,31,42) if MINTm<. // narrow (*)
. gen byte MINT2m = MINTm>0 if MINTm<. // wide
. su MINT1m MINT2m

```

Variable	Obs	Mean	Std. Dev.	Min	Max
MINT1m	22,423	.0169023	.1289083	0	1
MINT2m	22,423	.0318423	.175584	0	1

```

. gen int isco08 = real(occupf_isco08)
. replace isco08 = . if isco08<0
(2,489 real changes made, 2,489 to missing)
. mint08 MINTf isco08
MINTf

```

		Freq.	Percent	Valid	Cum.
Valid	0 nicht MINT	17141	76.44	85.99	85.99
	11 MINT* 1.1: Informatik	784	3.50	3.93	89.92
	12 MINT* 1.2: Elektrotechnik	130	0.58	0.65	90.57
	13 MINT* 1.3: Maschinentechnik	144	0.64	0.72	91.30
	14 MINT* 1.4: Mikrotechnik	66	0.29	0.33	91.63
	15 MINT* 1.5: Wirtschaftsingenieurwesen	57	0.25	0.29	91.91
	16 MINT* 1.6: Anderes aus Technik & IT	512	2.28	2.57	94.48
	21 MINT* 2.1: Bau	529	2.36	2.65	97.14
	22 MINT 2.2: Planung und Vermessung	59	0.26	0.30	97.43
	23 MINT 2.3: Architektur	193	0.86	0.97	98.40
	31 MINT* 3.1: Chemie	128	0.57	0.64	99.04
	32 MINT 3.2: Biotechnologie	2	0.01	0.01	99.05
	33 MINT 3.3: Gesundheit	53	0.24	0.27	99.32
	41 MINT 4.1: Geografie	50	0.22	0.25	99.57
	42 MINT* 4.2: Exakte Wissenschaften	34	0.15	0.17	99.74
	43 MINT 4.3: Andere MINT	52	0.23	0.26	100.00
	Total	19934	88.90	100.00	
Missing	.	2489	11.10		
Total		22423	100.00		

```

. replace MINTf = 0 if isco08>=. // no info => assume no MINT
(2,489 real changes made)

```

```

. drop isco08
. gen byte MINT1f = inlist(MINTf,11,12,13,14,15,16,21,31,42) if MINTf<. // narrow (*)
. gen byte MINT2f = MINTf>0 if MINTf<. // wide
. su MINT1f MINT2f

```

Variable	Obs	Mean	Std. Dev.	Min	Max
MINT1f	22,423	.1063194	.3082529	0	1
MINT2f	22,423	.1245596	.3302262	0	1

```

. // - parents' math norms (only half of sample)
. d socnorm*

```

variable name	storage type	display format	value label	variable label
socnormsm1	int	%8.0g	SOCNORMS	Important to learn math (Mother's social norms about mathematics)
socnormsm2	int	%8.0g	V1173_A	Important for career to learn math (Mother's social norms about mathematics)
socnormsm3	int	%8.0g	V1174_A	Likes math (Mother's social norms about mathematics)
socnormsm_fs	double	%12.0g	V1175_A	Mother's social norms about mathematics
socnormsf1	int	%8.0g	V1176_A	Important to learn math (Father's social norms about mathematics)
socnormsf2	int	%8.0g	V1177_A	Important for career to learn math (Father's social norms about mathematics)
socnormsf3	int	%8.0g	V1178_A	Likes math (Father's social norms about mathematics)
socnormsf_fs	double	%12.0g	V1179_A	Father's social norms about mathematics

```

. gen double socnormsm = socnormsm_fs if socnormsm_fs>-900
(11,576 missing values generated)

```

```

. gen double socnormsf = socnormsf_fs if socnormsf_fs>-900
(11,836 missing values generated)

```

```

. su socnormsm socnormsf

```

Variable	Obs	Mean	Std. Dev.	Min	Max
socnormsm	10,847	.0727577	.873796	-2.337935	1.409555
socnormsf	10,587	.115669	.8854172	-2.400167	1.224326

```

. d expect*

```

variable name	storage type	display format	value label	variable label
expectm1	int	%8.0g	EXPECTM1	Important to do well in science (Achievement expectations mother)
expectm2	int	%8.0g	EXPECTM2	Important to do well in mathematics (Achievement expectations mother)
expectm3	int	%8.0g	EXPECTM3	Important to do well in test language (Achievement expectations mother)
expectm4	float	%9.0g	EXPECTM4	Expected marks in mathematics (Achievement expectations mother)
expectm4_org	double	%12.0g	V1184_A	Expected marks in mathematics (Unedited original variable expectm4)
expectm_fs	double	%12.0g	EXPECTM	Achievement expectations mother
expectf1	int	%8.0g	EXPECTF1	Important to do well in science (Achievement expectations father)
expectf2	int	%8.0g	EXPECTF2	Important to do well in mathematics (Achievement expectations father)
expectf3	int	%8.0g	EXPECTF3	Important to do well in test language (Achievement expectations father)
expectf4	float	%9.0g	EXPECTF4	Expected marks in mathematics (Achievement expectations father)
expectf4_org	double	%12.0g	V1190_A	Expected marks in mathematics (Unedited original variable expectf4)
expectf_fs	double	%12.0g	EXPECTF	Achievement expectations father
expectp_fs	double	%12.0g	EXPECTP	Achievement expectations parents

```

. gen byte mathimpm = expectm2==4 if expectm2>0
(11,561 missing values generated)

```

```

. gen byte mathimpf = expectf2==4 if expectf2>0
(11,855 missing values generated)

```

```
. su mathimpm mathimpf // very important to do well in math
```

Variable	Obs	Mean	Std. Dev.	Min	Max
mathimpm	10,862	.4705395	.4991543	0	1
mathimpf	10,568	.5293338	.4991624	0	1

3.2 Selektion vollständiger Beobachtungen

```
. fre female
```

```
female
```

		Freq.	Percent	Valid	Cum.
Valid	0	11346	50.60	50.82	50.82
	1	10981	48.97	49.18	100.00
	Total	22327	99.57	100.00	
Missing	.	96	0.43		
Total		22423	100.00		

```
. drop if female>=.
```

```
(96 observations deleted)
```

```
. fre wlem, t(2)
```

```
wlem — Weighted likelihood estimates (WLE)
```

		Freq.	Percent	Valid	Cum.
Valid	-5.709	1	0.00	0.00	0.00
	-5.497	1	0.00	0.00	0.01
	:	:	:	:	:
	5.211	1	0.00	0.00	100.00
	5.214	1	0.00	0.00	100.00
	Total	22045	98.74	100.00	
Missing	.	282	1.26		
Total		22327	100.00		

```
. drop if wlem>=.
```

```
(282 observations deleted)
```

```
. fre MINT
```

```
MINT
```

		Freq.	Percent	Valid	Cum.
Valid	0 nicht MINT	13679	62.05	63.31	63.31
	11 MINT* 1.1: Informatik	724	3.28	3.35	66.67
	12 MINT* 1.2: Elektrotechnik	64	0.29	0.30	66.96
	13 MINT* 1.3: Maschinenteknik	79	0.36	0.37	67.33
	14 MINT* 1.4: Mikrotechnik	55	0.25	0.25	67.58
	15 MINT* 1.5: Wirtschaftsingenieurwesen	7	0.03	0.03	67.61
	16 MINT* 1.6: Anderes aus Technik & IT	476	2.16	2.20	69.82
	21 MINT* 2.1: Bau	148	0.67	0.69	70.50
	22 MINT 2.2: Planung und Vermessung	75	0.34	0.35	70.85
	23 MINT 2.3: Architektur	461	2.09	2.13	72.98
	31 MINT* 3.1: Chemie	118	0.54	0.55	73.53
	32 MINT 3.2: Biotechnologie	12	0.05	0.06	73.58
	33 MINT 3.3: Gesundheit	125	0.57	0.58	74.16
	41 MINT 4.1: Geografie	104	0.47	0.48	74.64
	42 MINT* 4.2: Exakte Wissenschaften	79	0.36	0.37	75.01
	43 MINT 4.3: Andere MINT	32	0.15	0.15	75.16
	99	5367	24.35	24.84	100.00
	Total	21605	98.00	100.00	
Missing	.	440	2.00		
Total		22045	100.00		

```

. drop if MINT>=.
(440 observations deleted)

. count if matcon>=.
107

. drop if matcon>=.
(107 observations deleted)

. count if selfeff>=.
217

. drop if selfeff>=.
(217 observations deleted)

```

3.3 MINT-Verteilung

```

. fre MINT [aw=smp_w_nrastubw] if female==0, include
MINT

```

		Freq.	Percent	Valid	Cum.
Valid	0 nicht MINT	6056.908	56.40	56.40	56.40
	11 MINT* 1.1: Informatik	649.8262	6.05	6.05	62.45
	12 MINT* 1.2: Elektrotechnik	59.54559	0.55	0.55	63.00
	13 MINT* 1.3: Maschinentechnik	85.17598	0.79	0.79	63.79
	14 MINT* 1.4: Mikrotechnik	50.1191	0.47	0.47	64.26
	15 MINT* 1.5: Wirtschaftsingenieurwesen	1.326018	0.01	0.01	64.27
	16 MINT* 1.6: Anderes aus Technik & IT	416.4263	3.88	3.88	68.15
	21 MINT* 2.1: Bau	119.9866	1.12	1.12	69.27
	22 MINT 2.2: Planung und Vermessung	44.74996	0.42	0.42	69.68
	23 MINT 2.3: Architektur	326.228	3.04	3.04	72.72
	24 MINT 2.4: Anderes aus Bauwesen	0	0.00	0.00	72.72
	31 MINT* 3.1: Chemie	80.76886	0.75	0.75	73.47
	32 MINT 3.2: Biotechnologie	4.080289	0.04	0.04	73.51
	33 MINT 3.3: Gesundheit	9.502926	0.09	0.09	73.60
	34 MINT 3.4: Anderes aus C&LS	0	0.00	0.00	73.60
	41 MINT 4.1: Geografie	52.73258	0.49	0.49	74.09
	42 MINT* 4.2: Exakte Wissenschaften	63.37185	0.59	0.59	74.68
	43 MINT 4.3: Andere MINT	25.91312	0.24	0.24	74.92
	99	2693.339	25.08	25.08	100.00
	Total	10740	100.00	100.00	

```

. fre MINT [aw=smp_w_nrastubw] if female==1, include
MINT

```

		Freq.	Percent	Valid	Cum.
Valid	0 nicht MINT	7266.417	68.93	68.93	68.93
	11 MINT* 1.1: Informatik	72.60006	0.69	0.69	69.62
	12 MINT* 1.2: Elektrotechnik	3.487517	0.03	0.03	69.66
	13 MINT* 1.3: Maschinentechnik	2.221513	0.02	0.02	69.68
	14 MINT* 1.4: Mikrotechnik	.5966027	0.01	0.01	69.68
	15 MINT* 1.5: Wirtschaftsingenieurwesen	3.562429	0.03	0.03	69.72
	16 MINT* 1.6: Anderes aus Technik & IT	62.57383	0.59	0.59	70.31
	21 MINT* 2.1: Bau	6.486903	0.06	0.06	70.37
	22 MINT 2.2: Planung und Vermessung	30.71171	0.29	0.29	70.66
	23 MINT 2.3: Architektur	210.9861	2.00	2.00	72.67
	24 MINT 2.4: Anderes aus Bauwesen	0	0.00	0.00	72.67
	31 MINT* 3.1: Chemie	33.57419	0.32	0.32	72.98
	32 MINT 3.2: Biotechnologie	5.831783	0.06	0.06	73.04
	33 MINT 3.3: Gesundheit	100.7292	0.96	0.96	73.99
	34 MINT 3.4: Anderes aus C&LS	0	0.00	0.00	73.99
	41 MINT 4.1: Geografie	60.47487	0.57	0.57	74.57
	42 MINT* 4.2: Exakte Wissenschaften	26.7383	0.25	0.25	74.82
	43 MINT 4.3: Andere MINT	3.427072	0.03	0.03	74.85
	99	2650.581	25.15	25.15	100.00

Total	10541	100.00	100.00
-------	-------	--------	--------

3.4 Arbeitsdatensatz speichern

```
. // - drop observations for which MINT is not determined
. drop if MINT==99
(5,165 observations deleted)

.
. // - re-standardize key variables
. center wlem matcon selfeff [pw=smp_w_nrastubw], standard inplace
(modified variables: wlem matcon selfeff)

.
. // - sample design
. assert id_school!=""
. assert smp_w_nrastubw>0 & smp_w_nrastubw<.

.
. // - variables to keep
. local vlist ///
> id_student wlem matcon selfeff MINT MINT1 MINT2 female immigr truancy coachmath repeat grademath ///
> gym books isei lfpm MINTm MINT1m MINT2m MINTf MINT1f MINT2f socnormsm socnormsf mathimpm mathimpf ///
> id_school id_region smp_w_nrastubw smp_w_nrasturw*
. keep `vlist'
. order `vlist'
. compress
variable immigr was int now byte
variable coachmath was int now byte
variable books was int now byte
variable isei was int now byte
(64,464 bytes saved)
. save workingdata, replace
file workingdata.dta saved
```

4 Analysen

```
. use workingdata, clear
( )
```

4.1 Surveydesign

```
. // A) linearization (just for quick approximation)
.
. // strata not available in official release; M.Werner provided
. // instructions how to generate the strata; however, strategy here is to use
. // schools as clusters and ignore strata for linearization estimates and then
. // at the end use BRR for the final estimates
.
. svyset id_school [pw=smp_w_nrastubw]
      pweight: smp_w_nrastubw
      VCE: linearized
Single unit: missing
Strata 1: <one>
      SU 1: id_school
      FPC 1: <zero>
.
. // B) BRR (for final results)
.
. svyset [pw=smp_w_nrastubw], brr(smp_w_nrasturw*) fay(.5) vce(brr)
```

```

pweight: smp_w_nrastubw
VCE: brr
MSE: off
brrweight: smp_w_nrasturw1 .. smp_w_nrasturw120
fay: .5
Single unit: missing
Strata 1: <one>
SU 1: <observations>
FPC 1: <zero>

```

4.2 Beruf mit 30 im Bereich der MINT-Fachkräfte

```

. svy: mean MINT1 MINT2, over(female) citype(logit)
(running mean on estimation sample)

```

BRR replications (120)

```

-----|----- 1 -----|----- 2 -----|----- 3 -----|----- 4 -----|----- 5
..... 50
..... 100
.....

```

```

Survey: Mean estimation          Number of obs   =    16,116
                                Population size = 57,724.632
                                Replications   =      120
                                Design df      =      119

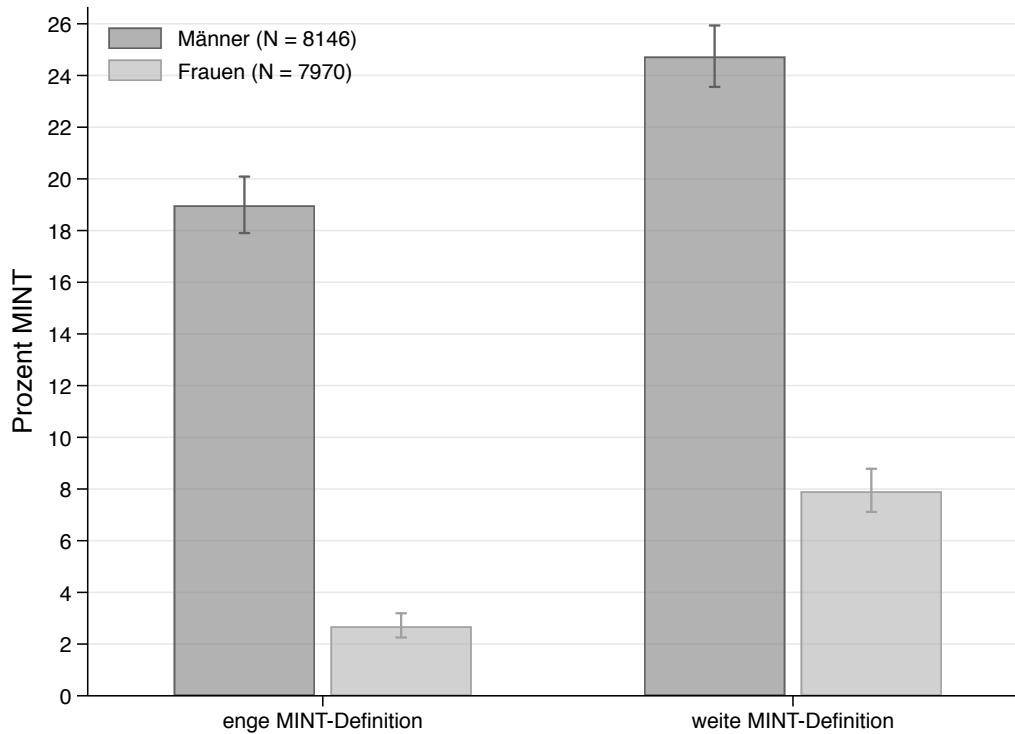
```

	BRR			
	Mean	Std. Err.	[95% Conf. Interval]	
c.MINT1@female				
0	.1897118	.0055188	.1790243	.2008812
1	.0268479	.002356	.0225567	.0319288
c.MINT2@female				
0	.2472769	.0059983	.2355921	.2593447
1	.0790835	.0042101	.0711386	.0878318

```

. local Nm = e1(e_N),1,1)
. local Nf = e1(e_N),1,2)
. coefplot (, keep(*@0.female) label("Männer (N = `Nm')") rename(*@0.female = "")) ///
>    (, keep(*@1.female) label("Frauen (N = `Nf')") rename(*@1.female = "")) ///
>    , vertical recast(bar) base(0) barw(0.3) plotr(m(b=0)) cirecast(rcap) citop citype(logit) ///
>    rescale(100) yti("Prozent MINT") ylabel(0(2)26) ///
>    coefl(c.MINT1 = "enge MINT-Definition" c.MINT2 = "weite MINT-Definition")

```



4.3 Generieren von Rangdaten bzw. Rangdifferenzen

```

. gen double r_wlem = .
(16,116 missing values generated)
. mata: st_store(., "r_wlem", mm_ranks(st_data(., "wlem"), st_data(., "smp_w_nrastubw"), 3, 1, 1))
. gen double r_matcon = .
(16,116 missing values generated)
. mata: st_store(., "r_matcon", mm_ranks(st_data(., "matcon"), st_data(., "smp_w_nrastubw"), 3, 1, 1))
. gen double r_selfeff = .
(16,116 missing values generated)
. mata: st_store(., "r_selfeff", mm_ranks(st_data(., "selfeff"), st_data(., "smp_w_nrastubw"), 3, 1, 1))
. gen double d_matcon = r_matcon - r_wlem
. gen double d_selfeff = r_selfeff - r_wlem

```

4.4 Korrelation zwischen Leistungsdaten und Selbsteinschätzung

```

. corr wlem matcon selfeff [aw=smp_w_nrastubw]
(sum of wgt is 57,724.6319285608)
(obs=16,116)

```

	wlem	matcon	selfeff
wlem	1.0000		
matcon	0.3575	1.0000	
selfeff	0.5776	0.4825	1.0000

```

. corr r_wlem r_matcon r_selfeff [aw=smp_w_nrastubw]
(sum of wgt is 57,724.6319285608)
(obs=16,116)

```

	r_wlem	r_matcon	r_self-f
r_wlem	1.0000		
r_matcon	0.3575	1.0000	
r_self-f	0.5776	0.4825	1.0000

r_wlem	1.0000			
r_matcon	0.3473	1.0000		
r_selfeff	0.5965	0.4755	1.0000	

4.5 Unter- bzw. Überschätzung der eigenen Kompetenzen

```
. su d_matcon [aw=smp_w_nrastubw]
Variable | Obs Weight Mean Std. Dev. Min Max
-----+-----
d_matcon | 16,116 57724.6319 -9.16e-16 .3293434 -.9101894 .9337547
. gen abs = abs(d_matcon)
. su abs [aw=smp_w_nrastubw]
Variable | Obs Weight Mean Std. Dev. Min Max
-----+-----
abs | 16,116 57724.6319 .2614751 .2002338 .0000104 .9337547
. drop abs
. svy: regress d_matcon i.female
(running regress on estimation sample)
BRR replications (120)
-----+-----
| 1 | 2 | 3 | 4 | 5
|-----|-----|-----|-----|-----
| ..... 50
| ..... 100
| .....
Survey: Linear regression
Number of obs = 16,116
Population size = 57,724.632
Replications = 120
Design df = 119
F( 1, 119) = 310.19
Prob > F = 0.0000
R-squared = 0.0293
```

d_matcon	BRR		t	P> t	[95% Conf. Interval]	
	Coef.	Std. Err.				
1.female	-.1127828	.0064036	-17.61	0.000	-.1254626	-.100103
_cons	.0563989	.0054893	10.27	0.000	.0455295	.0672683

```
. xtile q_matcon = d_matcon [pw=smp_w_nrastubw], nq(3)
. svy: proportion q_matcon, over(female)
(running proportion on estimation sample)
BRR replications (120)
-----+-----
| 1 | 2 | 3 | 4 | 5
|-----|-----|-----|-----|-----
| ..... 50
| ..... 100
| .....
Survey: Proportion estimation
Number of obs = 16,116
Population size = 57,724.632
Replications = 120
Design df = 119
```

	BRR		Normal	
	Proportion	Std. Err.	[95% Conf. Interval]	
q_matcon@female				
1 0	.2687109	.0064988	.2558427	.2815791
1 1	.3980009	.0088371	.3805025	.4154993
2 0	.3362307	.0069069	.3225544	.3499071
2 1	.3307742	.0066666	.3175736	.3439747
3 0	.3950584	.008173	.378875	.4112417
3 1	.2712249	.0069757	.2574124	.2850375


```

. est sto matcon
. drop q_matcon
. su d_selfeff [aw=smp_w_nrastubw]

```

Variable	Obs	Weight	Mean	Std. Dev.	Min	Max
d_selfeff	16,116	57724.6319	2.12e-16	.2593433	-.894361	.9796289

```

. gen abs = abs(d_selfeff)
. su abs [aw=smp_w_nrastubw]

```

Variable	Obs	Weight	Mean	Std. Dev.	Min	Max
abs	16,116	57724.6319	.1992255	.1660291	2.76e-06	.9796289

```

. drop abs
. svy: regress d_selfeff i.female
(running regress on estimation sample)
BRR replications (120)

```

Variable	1	2	3	4	5
d_selfeff

```

..... 50
..... 100
.....
Survey: Linear regression

```

Number of obs	=	16,116
Population size	=	57,724.632
Replications	=	120
Design df	=	119
F(1, 119)	=	226.63
Prob > F	=	0.0000
R-squared	=	0.0186

```


```

d_selfeff	BRR				[95% Conf. Interval]	
	Coef.	Std. Err.	t	P> t		
1.female	-.0707105	.0046971	-15.05	0.000	-.0800111	-.0614098
_cons	.0353599	.0038106	9.28	0.000	.0278145	.0429054

```

. xtile q_selfeff = d_selfeff [pw=smp_w_nrastubw], nq(3)
. svy: proportion q_selfeff, over(female)
(running proportion on estimation sample)
BRR replications (120)

```

Variable	1	2	3	4	5
q_selfeff

```

..... 50
..... 100
.....
Survey: Proportion estimation

```

Number of obs	=	16,116
Population size	=	57,724.632
Replications	=	120
Design df	=	119

```


```

	BRR		Normal	
	Proportion	Std. Err.	[95% Conf. Interval]	
q_selfeff@female				
1 0	.2780048	.0065847	.2649664	.2910431
1 1	.3886636	.0073076	.3741938	.4031333
2 0	.3412319	.0069502	.3274697	.354994
2 1	.3256262	.00585	.3140427	.3372098
3 0	.3807634	.0064523	.3679872	.3935395
3 1	.2857102	.0064721	.2728948	.2985256

```

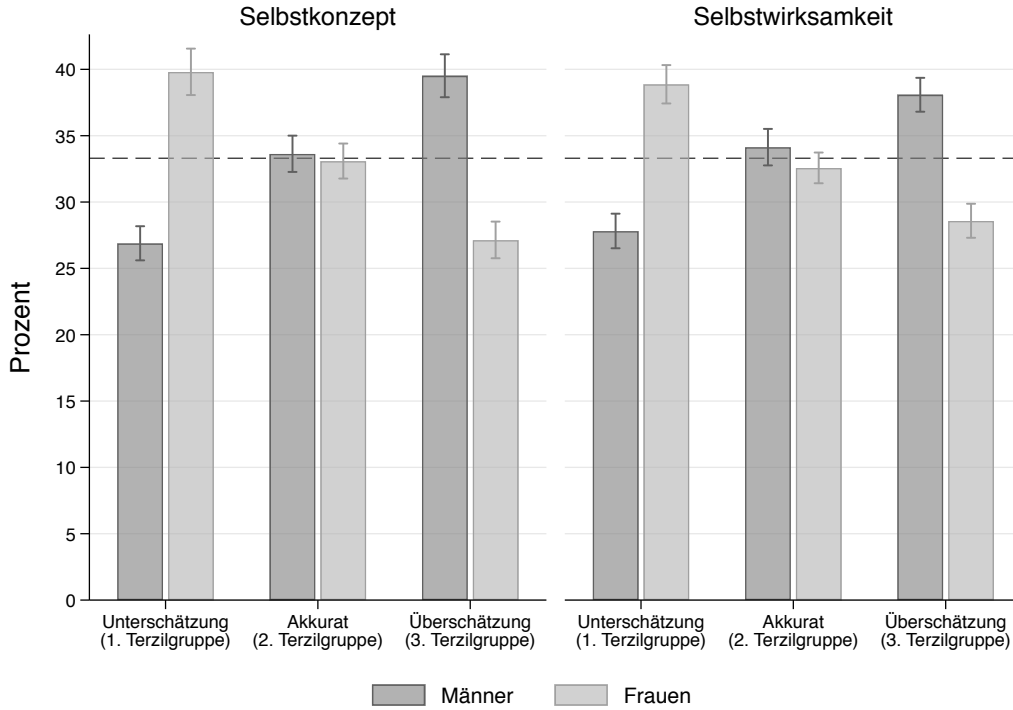
. est sto selfeff
. drop q_selfeff
. coefplot (matcon, keep(*@0.female) label("Männer") rename(*_matcon@0.female = "")) ///
> (matcon, keep(*@1.female) label("Frauen") rename(*_matcon@1.female = "")) ///

```

```

> || (selfeff, keep(*@0.female) label("Männer") rename(*_selfeff@0.female = "")) ///
> (selfeff, keep(*@1.female) label("Frauen") rename(*_selfeff@1.female = "")) ///
> || , vertical recast(bar) base(0) barw(0.3) plotr(m(b=0)) cirecast(rcap) citop citype(logit) ///
> rescale(100) yti(Prozent) ylab(0(5)40) byopt(legend(pos(6))) legend(row(1)) yline(33.3) ///
> bylabels("Selbstkonzept" "Selbstwirksamkeit") ///
> coeflab(1.* = `""Unterschätzung" "(1. Terzilgruppe)""' ///
> 2.* = `""Akkurat" "(2. Terzilgruppe)""' ///
> 3.* = `""Überschätzung" "(3. Terzilgruppe)""')

```



4.6 Effekt der Unter- bzw. Überschätzung der eigenen Kompetenzen auf die beruflichen Vorstellungen

```

. svy: logit MINT1 r_wlem d_matcon
(running logit on estimation sample)
BRR replications (120)
-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|
| 1 | 2 | 3 | 4 | 5 |
..... 50
..... 100
.....

Survey: Logistic regression
Number of obs = 16,116
Population size = 57,724.632
Replications = 120
Design df = 119
F( 2, 118) = 229.29
Prob > F = 0.0000

```

MINT1	BRR			t	P> t	[95% Conf. Interval]	
	Coef.	Std. Err.					
r_wlem	3.278281	.1529654	21.43	0.000	2.975394	3.581168	
d_matcon	2.087413	.1268366	16.46	0.000	1.836264	2.338562	
_cons	-3.982272	.1067144	-37.32	0.000	-4.193578	-3.770967	

```

. margins, dydx(*) post
Average marginal effects      Number of obs   =   16,116
Model VCE      : BRR
Expression    : Pr(MINT1), predict()
dy/dx w.r.t. : r_wlem d_matcon

```

	Delta-method		t	P> t	[95% Conf. Interval]	
	dy/dx	Std. Err.				
r_wlem	.2985849	.013565	22.01	0.000	.271725	.3254449
d_matcon	.190121	.0112973	16.83	0.000	.1677512	.2124909

```

. est sto MINT1_matcon
. svy: logit MINT1 r_wlem d_selfeff
(running logit on estimation sample)

```

```

BRR replications (120)
-----|-----|-----|-----|-----|
| 1 | 2 | 3 | 4 | 5 |
..... 50
..... 100
.....

```

```

Survey: Logistic regression      Number of obs   =   16,116
Population size = 57,724.632
Replications     =   120
Design df       =   119
F( 2, 118)     =   163.91
Prob > F       =   0.0000

```

MINT1	BRR		t	P> t	[95% Conf. Interval]	
	Coef.	Std. Err.				
r_wlem	2.620817	.1442057	18.17	0.000	2.335275	2.906359
d_selfeff	1.630809	.1478314	11.03	0.000	1.338088	1.92353
_cons	-3.593769	.0997776	-36.02	0.000	-3.791339	-3.396199

```

. margins, dydx(*) post
Average marginal effects      Number of obs   =   16,116
Model VCE      : BRR
Expression    : Pr(MINT1), predict()
dy/dx w.r.t. : r_wlem d_selfeff

```

	Delta-method		t	P> t	[95% Conf. Interval]	
	dy/dx	Std. Err.				
r_wlem	.2427915	.0132902	18.27	0.000	.2164755	.2691075
d_selfeff	.1510776	.0135778	11.13	0.000	.1241922	.177963

```

. est sto MINT1_selfeff
. svy: logit MINT2 r_wlem d_matcon
(running logit on estimation sample)

```

```

BRR replications (120)
-----|-----|-----|-----|-----|
| 1 | 2 | 3 | 4 | 5 |
..... 50
..... 100
.....

```

```

Survey: Logistic regression      Number of obs   =   16,116
Population size = 57,724.632
Replications     =   120
Design df       =   119
F( 2, 118)     =   260.78
Prob > F       =   0.0000

```

MINT2	BRR			t	P> t	[95% Conf. Interval]	
	Coef.	Std. Err.					
r_wlem	2.999314	.1308563	22.92	0.000	2.740206	3.258423	
d_matcon	1.516505	.1031237	14.71	0.000	1.31231	1.7207	
_cons	-3.297974	.0883136	-37.34	0.000	-3.472844	-3.123104	

```
. margins, dydx(*) post
Average marginal effects      Number of obs   =   16,116
Model VCE      : BRR
Expression     : Pr(MINT2), predict()
dy/dx w.r.t.  : r_wlem d_matcon
```

	Delta-method			t	P> t	[95% Conf. Interval]	
	dy/dx	Std. Err.					
r_wlem	.3829688	.0152487	25.11	0.000	.3527748	.4131627	
d_matcon	.1936356	.0122869	15.76	0.000	.1693063	.2179649	

```
. est sto MINT2_matcon
. svy: logit MINT2 r_wlem d_selfeff
(running logit on estimation sample)
BRR replications (120)
-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|
..... 1 ..... 2 ..... 3 ..... 4 ..... 5
..... 50
..... 100
.....
Survey: Logistic regression      Number of obs   =   16,116
                                Population size   = 57,724.632
                                Replications      =   120
                                Design df        =   119
                                F( 2, 118)       =   210.75
                                Prob > F         =   0.0000
```

MINT2	BRR			t	P> t	[95% Conf. Interval]	
	Coef.	Std. Err.					
r_wlem	2.597146	.1259712	20.62	0.000	2.347711	2.846582	
d_selfeff	1.367145	.1217421	11.23	0.000	1.126084	1.608207	
_cons	-3.079471	.0832887	-36.97	0.000	-3.244391	-2.914551	

```
. margins, dydx(*) post
Average marginal effects      Number of obs   =   16,116
Model VCE      : BRR
Expression     : Pr(MINT2), predict()
dy/dx w.r.t.  : r_wlem d_selfeff
```

	Delta-method			t	P> t	[95% Conf. Interval]	
	dy/dx	Std. Err.					
r_wlem	.335073	.0157214	21.31	0.000	.303943	.366203	
d_selfeff	.1763834	.0154796	11.39	0.000	.1457323	.2070345	

```
. est sto MINT2_selfeff
. esttab MINT1_matcon MINT1_selfeff MINT2_matcon MINT2_selfeff, nostar nomti t(1)
```

	(1)	(2)	(3)	(4)
r_wlem	0.299 (22.0)	0.243 (18.3)	0.383 (25.1)	0.335 (21.3)
d_matcon	0.190 (16.8)		0.194 (15.8)	

d_selfeff		0.151 (11.1)		0.176 (11.4)
N	16116	16116	16116	16116

t statistics in parentheses

4.7 Zerlegung des MINT-Gender-Gaps

```
. svy: regress r_wlem i.female // gender difference in wlem
(running regress on estimation sample)
```

BRR replications (120)

```
-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|
| 1 | 2 | 3 | 4 | 5
..... 50
..... 100
.....
```

```
Survey: Linear regression      Number of obs   =   16,116
                               Population size    = 57,724.632
                               Replications       =     120
                               Design df         =     119
                               F( 1, 119)        =     90.64
                               Prob > F         =     0.0000
                               R-squared         =     0.0080
```

r_wlem	Coef.	BRR Std. Err.	t	P> t	[95% Conf. Interval]	
1.female	-.0517707	.0054377	-9.52	0.000	-.0625379	-.0410035
_cons	.5258888	.004625	113.71	0.000	.5167309	.5350468

```
.
. oaxaca MINT1 r_wlem d_matcon d_selfeff, by(female) svy pooled logit
(running oaxaca on estimation sample)
```

BRR replications (120)

```
-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|
| 1 | 2 | 3 | 4 | 5
..... 50
..... 100
.....
```

```
Blinder-Oaxaca decomposition      Number of obs   =   16,116
                                   Population size    = 57,724.632
                                   Replications       =     120
                                   Design df         =     119
```

MINT1	Coef.	BRR Std. Err.	t	P> t	[95% Conf. Interval]	
overall						
group_1	.1897118	.0055188	34.38	0.000	.178784	.2006396
group_2	.0268479	.002356	11.40	0.000	.0221828	.031513
difference	.1628639	.0056631	28.76	0.000	.1516503	.1740775
explained	.0427123	.0030597	13.96	0.000	.0366537	.0487708
unexplained	.1201516	.0055212	21.76	0.000	.1092191	.1310842
explained						
r_wlem	.019193	.0021285	9.02	0.000	.0149782	.0234077
d_matcon	.0188075	.0022212	8.47	0.000	.0144092	.0232058
d_selfeff	.0047118	.001478	3.19	0.002	.0017852	.0076385
unexplained						
r_wlem	-.0253094	.0139398	-1.82	0.072	-.0529116	.0022928
d_matcon	.0023267	.0009656	2.41	0.018	.0004146	.0042388
d_selfeff	.0002428	.0007478	0.32	0.746	-.0012378	.0017235
_cons	.1428915	.0130211	10.97	0.000	.1171083	.1686746

```

. nlcom (r_wlem: _b[explained:r_wlem] / _b[overall:difference]*100) ///
> (d_matcon: _b[explained:d_matcon] / _b[overall:difference]*100) ///
> (d_selfeff: _b[explained:d_selfeff] / _b[overall:difference]*100) ///
> (d_X: (_b[explained:d_matcon]/*
> */+_b[explained:d_selfeff]) / _b[overall:difference]*100) ///
> (tot: (_b[explained:r_wlem]/*
> */+_b[explained:d_matcon]/*
> */+_b[explained:d_selfeff]) / _b[overall:difference]*100) ///
> , post
r_wlem: _b[explained:r_wlem] / _b[overall:difference]*100
d_matcon: _b[explained:d_matcon] / _b[overall:difference]*100
d_selfeff: _b[explained:d_selfeff] / _b[overall:difference]*100
d_X: (_b[explained:d_matcon]+_b[explained:d_selfeff]) / _b[overall:difference]*100
tot: (_b[explained:r_wlem]+_b[explained:d_matcon]+_b[explained:d_selfeff]) / _b[overall:difference]*
> 100

```

MINT1	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
r_wlem	11.78466	1.235813	9.54	0.000	9.362511	14.20681
d_matcon	11.54797	1.38939	8.31	0.000	8.824814	14.27112
d_selfeff	2.893107	.9070403	3.19	0.001	1.115341	4.670874
d_X	14.44108	1.512225	9.55	0.000	11.47717	17.40498
tot	26.22574	1.810657	14.48	0.000	22.67691	29.77456

```

. est sto MINT1
. oxaca MINT2 r_wlem d_matcon d_selfeff, by(female) svy pooled logit
(running oxaca on estimation sample)

```

BRR replications (120)

```

-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|
      1       2       3       4       5
..... 50
..... 100
.....

```

```

Blinder-Oaxaca decomposition      Number of obs      =      16,116
                                   Population size      = 57,724.632
                                   Replications          =       120
                                   Design df              =       119

```

MINT2	BRR		t	P> t	[95% Conf. Interval]	
	Coef.	Std. Err.				
overall						
group_1	.2472769	.0059983	41.22	0.000	.2353997	.2591541
group_2	.0790835	.0042101	18.78	0.000	.0707472	.0874199
difference	.1681934	.0069722	24.12	0.000	.1543878	.181999
explained	.0462278	.0030723	15.05	0.000	.0401443	.0523112
unexplained	.1219656	.0063586	19.18	0.000	.109375	.1345563
explained						
r_wlem	.0232603	.0025056	9.28	0.000	.0182989	.0282216
d_matcon	.0161478	.0021709	7.44	0.000	.0118491	.0204464
d_selfeff	.0068197	.0016208	4.21	0.000	.0036103	.0100291
unexplained						
r_wlem	.0268525	.0183063	1.47	0.145	-.0093958	.0631008
d_matcon	.0002447	.0005869	0.42	0.677	-.0009174	.0014068
d_selfeff	.0002775	.0005434	0.51	0.611	-.0007985	.0013534
_cons	.094591	.0172012	5.50	0.000	.0605309	.1286511

```

. nlcom (r_wlem: _b[explained:r_wlem] / _b[overall:difference]*100) ///
> (d_matcon: _b[explained:d_matcon] / _b[overall:difference]*100) ///
> (d_selfeff: _b[explained:d_selfeff] / _b[overall:difference]*100) ///
> (d_X: (_b[explained:d_matcon]/*
> */+_b[explained:d_selfeff]) / _b[overall:difference]*100) ///
> (tot: (_b[explained:r_wlem]/*
> */+_b[explained:d_matcon]/*
> */+_b[explained:d_selfeff]) / _b[overall:difference]*100) ///
>

```

```

> , post
    r_wlem:  _b[explained:r_wlem]      / _b[overall:difference]*100
    d_matcon: _b[explained:d_matcon]   / _b[overall:difference]*100
    d_selfeff: _b[explained:d_selfeff] / _b[overall:difference]*100
    d_X:      (_b[explained:d_matcon]+_b[explained:d_selfeff]) / _b[overall:difference]*100
    tot:      (_b[explained:r_wlem]+_b[explained:d_matcon]+_b[explained:d_selfeff]) / _b[overall:difference]*
> 100

```

MINT2	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
r_wlem	13.82948	1.345725	10.28	0.000	11.19191	16.46705
d_matcon	9.600714	1.316369	7.29	0.000	7.020678	12.18075
d_selfeff	4.054686	.9793435	4.14	0.000	2.135208	5.974164
d_X	13.6554	1.489326	9.17	0.000	10.73637	16.57442
tot	27.48488	1.709676	16.08	0.000	24.13398	30.83578

```

. est sto MINT2
. esttab MINT1 MINT2, b(1) nostar ci wide

```

	(1) MINT1		(2) MINT2	
r_wlem	11.8	[9.4,14.2]	13.8	[11.2,16.5]
d_matcon	11.5	[8.8,14.3]	9.6	[7.0,12.2]
d_selfeff	2.9	[1.1,4.7]	4.1	[2.1,6.0]
d_X	14.4	[11.5,17.4]	13.7	[10.7,16.6]
tot	26.2	[22.7,29.8]	27.5	[24.1,30.8]
N	16116		16116	

95% confidence intervals in brackets

5 Zusatzanalysen

5.1 Zerlegung des MINT-Gender-Gaps: mit Kontrollvariablen

```

. // Models
. svy: logit MINT1 i.female r_wlem d_matcon d_selfeff ///
> i.MINT1m i.MINT1f i.coachmath grademath i.gym i.repeat truancy i.immig isei books
(running logit on estimation sample)

```

BRR replications (120)

```

-----| 1 |-----| 2 |-----| 3 |-----| 4 |-----| 5
..... 50
..... 100
.....

```

```

Survey: Logistic regression          Number of obs   =   15,356
                                   Population size    = 55,142.607
                                   Replications        =     120
                                   Design df           =     119
                                   F( 16, 104)         =     57.02
                                   Prob > F           =     0.0000

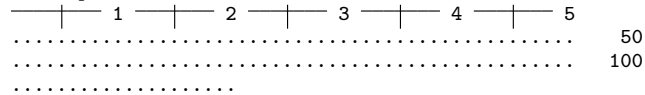
```

MINT1	Coef.	BRR Std. Err.	t	P> t	[95% Conf. Interval]	
1.female	-1.890756	.0940562	-20.10	0.000	-2.076997	-1.704515
r_wlem	2.964869	.224299	13.22	0.000	2.520734	3.409003
d_matcon	1.270647	.1693687	7.50	0.000	.9352798	1.606014
d_selfeff	.5956355	.167028	3.57	0.001	.2649035	.9263675
1.MINT1m	.1392663	.2101745	0.66	0.509	-.2769003	.5554329
1.MINT1f	.5259487	.1011923	5.20	0.000	.3255779	.7263196

coachmath						
Yes, sometimes	.2582722	.111959	2.31	0.023	.0365822	.4799621
Yes, regularly	.0977643	.1626013	0.60	0.549	-.2242026	.4197311
grademath						
1.gym	-.3674725	.1090914	-3.37	0.001	-.5834843	-.1514607
1.repeat	-.2624633	.114341	-2.30	0.023	-.48887	-.0360567
truancy	-.0106802	.0577671	-0.18	0.854	-.1250648	.1037045
immig						
Second generation	.1075048	.1070067	1.00	0.317	-.1043792	.3193889
First generation	.2724158	.1250581	2.18	0.031	.0247883	.5200432
isei						
books	-.0232627	.0248419	-0.94	0.351	-.0724521	.0259267
_cons	-3.530666	.321409	-10.98	0.000	-4.167088	-2.894244

```
. svy: logit MINT1 i.female##(c.r_wlem c.d_matcon c.d_selfeff ///
> i.MINT1m i.MINT1f i.coachmath c.grademath i.gym i.repeat c.truancy i.immig c.isei c.books) ///
> , vsquish nofvlabel
(running logit on estimation sample)
```

BRR replications (120)



Survey: Logistic regression

```
Number of obs      =    15,356
Population size     = 55,142.607
Replications        =     120
Design df           =     119
F( 31, 89)          =     32.03
Prob > F             =     0.0000
```

MINT1	BRR			P> t	[95% Conf. Interval]	
	Coef.	Std. Err.	t			
1.female	-3.101625	1.133679	-2.74	0.007	-5.346422	-.856828
r_wlem	2.86433	.2271822	12.61	0.000	2.414487	3.314174
d_matcon	1.142186	.1793539	6.37	0.000	.7870473	1.497325
d_selfeff	.5940445	.1846715	3.22	0.002	.2283765	.9597125
1.MINT1m	.027572	.2342378	0.12	0.906	-.4362423	.4913863
1.MINT1f	.532217	.105431	5.05	0.000	.3234532	.7409809
coachmath						
2	.254535	.1266734	2.01	0.047	.0037089	.505361
3	.1549268	.1775587	0.87	0.385	-.1966572	.5065108
grademath						
1.gym	-.3901835	.1115252	-3.50	0.001	-.6110145	-.1693526
1.repeat	-.2938601	.1089598	-2.70	0.008	-.5096113	-.0781089
truancy	-.0017462	.0617436	-0.03	0.977	-.1240046	.1205123
immig						
2	.1177583	.1161805	1.01	0.313	-.1122908	.3478074
3	.2204295	.1372261	1.61	0.111	-.0512919	.4921508
isei						
books	-.0303378	.0268276	-1.13	0.260	-.0834592	.0227836
female#c.r_wlem						
1	.3606759	.6808303	0.53	0.597	-.9874361	1.708788
female#c.d_matcon						
1	.5876639	.5260568	1.12	0.266	-.4539811	1.629309
female#c.d_selfeff						
1	.0118821	.4964869	0.02	0.981	-.9712115	.9949757
female#MINT1m						
1 1	.5758281	.5381686	1.07	0.287	-.4897994	1.641456
female#MINT1f						
1 1	-.0544456	.24535	-0.22	0.825	-.5402632	.431372
female#coachmath						
1 2	-.0032211	.3014397	-0.01	0.991	-.6001018	.5936595
1 3	-.2788959	.5493647	-0.51	0.613	-1.366693	.8089011

female#c.grademath							
1	.1814424	.2318737	0.78	0.435	-.2776906	.6405754	
female#gym							
1 1	.131365	.2383554	0.55	0.583	-.3406024	.6033324	
female#repeat							
1 1	.2348674	.3172023	0.74	0.460	-.3932248	.8629597	
female#c.truancy							
1	-.0577649	.1847257	-0.31	0.755	-.4235403	.3080104	
female#immig							
1 2	-.0403102	.263916	-0.15	0.879	-.5628903	.4822699	
1 3	.2356895	.3495173	0.67	0.501	-.4563897	.9277686	
female#c.isei							
1	-.0030373	.0073944	-0.41	0.682	-.017679	.0116044	
female#c.books							
1	.0442957	.077471	0.57	0.569	-.1091046	.1976961	
_cons	-3.349347	.3507496	-9.55	0.000	-4.043866	-2.654828	

```
. // => no gender differences
. svy: logit MINT2 i.female r_wlem d_matcon d_selfeff ///
> i.MINT2m i.MINT2f i.coachmath grademath i.gym i.repeat truancy i.immig isei books
(running logit on estimation sample)
```

BRR replications (120)

1	2	3	4	5
.....	50			
.....	100			
.....				

Survey: Logistic regression

Number of obs	=	15,356
Population size	=	55,142.607
Replications	=	120
Design df	=	119
F(16, 104)	=	48.09
Prob > F	=	0.0000

MINT2	BRR			P> t	[95% Conf. Interval]	
	Coef.	Std. Err.	t			
1.female	-1.133372	.0666469	-17.01	0.000	-1.26534	-1.001405
r_wlem	2.818291	.1826906	15.43	0.000	2.456546	3.180037
d_matcon	.9327338	.1467158	6.36	0.000	.6422218	1.223246
d_selfeff	.6192376	.1352406	4.58	0.000	.3514477	.8870276
1.MINT2m	.1371658	.1520022	0.90	0.369	-.1638138	.4381454
1.MINT2f	.4641995	.0844554	5.50	0.000	.2969695	.6314296
coachmath						
Yes, sometimes	.3108424	.096294	3.23	0.002	.1201706	.5015142
Yes, regularly	.0177666	.1302127	0.14	0.892	-.2400675	.2756008
grademath	-.0171054	.0580699	-0.29	0.769	-.1320896	.0978788
1.gym	-.1935358	.0825494	-2.34	0.021	-.3569918	-.0300798
1.repeat	-.2206166	.105119	-2.10	0.038	-.4287627	-.0124705
truancy	-.0066254	.0414572	-0.16	0.873	-.0887148	.075464
immig						
Second generation	.2144136	.0820014	2.61	0.010	.0520427	.3767845
First generation	.4267538	.1038913	4.11	0.000	.2210388	.6324689
isei	.0044384	.0022216	2.00	0.048	.0000394	.0088374
books	.0207998	.0184249	1.13	0.261	-.0156833	.0572829
_cons	-3.131029	.2675729	-11.70	0.000	-3.66085	-2.601208

```
. svy: logit MINT2 i.female##(c.r_wlem c.d_matcon c.d_selfeff ///
> i.MINT2m i.MINT2f i.coachmath c.grademath i.gym i.repeat c.truancy i.immig c.isei c.books) ///
> , vsquish nofvlabel
(running logit on estimation sample)
```

BRR replications (120)

1	2	3	4	5
---	---	---	---	---

```

..... 50
..... 100
.....

```

```

Survey: Logistic regression      Number of obs   =    15,356
                                Population size   = 55,142.607
                                Replications     =     120
                                Design df        =     119
                                F( 31, 89)       =     27.65
                                Prob > F        =     0.0000

```

MINT2	BRR		t	P> t	[95% Conf. Interval]	
	Coef.	Std. Err.				
1.female	-.2014918	.596291	-0.34	0.736	-1.382208	.979224
r_wlem	2.924859	.2094169	13.97	0.000	2.510192	3.339525
d_matcon	.9203881	.175127	5.26	0.000	.5736192	1.267157
d_selfeff	.5939928	.1688231	3.52	0.001	.2597062	.9282794
1.MINT2m	.1140747	.1856105	0.61	0.540	-.2534527	.4816021
1.MINT2f	.5435623	.0989168	5.50	0.000	.347697	.7394275
coachmath						
2	.3042449	.1124217	2.71	0.008	.0816387	.526851
3	.2211996	.1518476	1.46	0.148	-.0794738	.5218731
grademath	.0287354	.0741722	0.39	0.699	-.118133	.1756038
1.gym	-.2153487	.0956471	-2.25	0.026	-.4047395	-.025958
1.repeat	-.306283	.1105147	-2.77	0.006	-.5251132	-.0874528
truancy	-.0052584	.0540413	-0.10	0.923	-.1122656	.1017487
immig						
2	.2512428	.1069171	2.35	0.020	.0395363	.4629493
3	.3354932	.122812	2.73	0.007	.0923131	.5786734
isei	.006424	.0025747	2.50	0.014	.0013258	.0115223
books	.0054445	.0242505	0.22	0.823	-.0425738	.0534628
female#c.r_wlem						
1	-.3797809	.4013929	-0.95	0.346	-1.174579	.4150172
female#c.d_matcon						
1	.0774616	.3066007	0.25	0.801	-.5296385	.6845616
female#c.d_selfeff						
1	.0396639	.3259377	0.12	0.903	-.6057252	.6850531
female#MINT2m						
1 1	.0348092	.3325683	0.10	0.917	-.6237092	.6933277
female#MINT2f						
1 1	-.2279333	.1695534	-1.34	0.181	-.563666	.1077993
female#coachmath						
1 2	.0033853	.1875599	0.02	0.986	-.368002	.3747726
1 3	-.5630747	.3433144	-1.64	0.104	-1.242871	.1167222
female#c.grademath						
1	-.1367772	.1214166	-1.13	0.262	-.3771941	.1036398
female#gym						
1 1	.0949502	.1749139	0.54	0.588	-.2513968	.4412973
female#repeat						
1 1	.3215846	.1820703	1.77	0.080	-.0389328	.6821019
female#c.truancy						
1	.0058051	.1108333	0.05	0.958	-.2136559	.2252661
female#immig						
1 2	-.1026198	.1890875	-0.54	0.588	-.4770319	.2717923
1 3	.2382752	.2447508	0.97	0.332	-.2463558	.7229062
female#c.isei						
1	-.006119	.0046492	-1.32	0.191	-.0153248	.0030868
female#c.books						
1	.0617888	.0469695	1.32	0.191	-.0312155	.1547931
_cons	-3.456211	.3472139	-9.95	0.000	-4.143729	-2.768693

```

. // => no gender differences
. // Decompositions
. xi: oaxaca MINT1 r_wlem d_matcon d_selfeff ///
> i.MINT1m i.MINT1f i.coachmath grademath i.gym i.repeat truancy i.immig isei books ///
> , by(female) svy pooled logit
i.MINT1m _IMINT1m_0-1 (naturally coded; _IMINT1m_0 omitted)

```



```

>      */+_b[explained:d_matcon]/*
>      */+_b[explained:d_selfeff]) / _b[overall:difference]*100) ///
>      (rest:      (_b[overall:explained]-_b[explained:r_wlem]/*
>      */-_b[explained:d_matcon]/*
>      */-_b[explained:d_selfeff]) / _b[overall:difference]*100) ///
>      , post
>      r_wlem:  _b[explained:r_wlem]      / _b[overall:difference]*100
>      d_matcon:  _b[explained:d_matcon] / _b[overall:difference]*100
>      d_selfeff:  _b[explained:d_matcon] / _b[overall:difference]*100
>      d_X:      (_b[explained:d_matcon]+_b[explained:d_selfeff]) / _b[overall:difference]*100
>      tot:      (_b[explained:r_wlem]+_b[explained:d_matcon]+_b[explained:d_selfeff]) / _b[overall:difference]*
> 100
>      rest:      (_b[overall:explained]-_b[explained:r_wlem]-_b[explained:d_matcon]-_b[explained:d_selfeff]) / _
> b[overall:difference]*100

```

MINT1	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
r_wlem	13.37827	1.466252	9.12	0.000	10.50447	16.25207
d_matcon	10.91702	1.58474	6.89	0.000	7.810987	14.02305
d_selfeff	3.218247	.9078803	3.54	0.000	1.438834	4.99766
d_X	14.13527	1.663775	8.50	0.000	10.87433	17.39621
tot	27.51354	2.220725	12.39	0.000	23.161	31.86608
rest	-.1491199	1.416581	-0.11	0.916	-2.925569	2.627329

```

. est sto MINT1
. xi: oaxaca MINT2 r_wlem d_matcon d_selfeff ///
> i.MINT2m i.MINT2f i.coachmath grademath i.gym i.repeat truancy i.immig isei books ///
> , by(female) svy pooled logit
i.MINT2m      _IMINT2m_0-1      (naturally coded; _IMINT2m_0 omitted)
i.MINT2f      _IMINT2f_0-1      (naturally coded; _IMINT2f_0 omitted)
i.coachmath   _Icoachmath_1-3   (naturally coded; _Icoachmath_1 omitted)
i.gym         _Igy_0-1          (naturally coded; _Igy_0 omitted)
i.repeat      _Irepeat_0-1      (naturally coded; _Irepeat_0 omitted)
i.immig       _Immig_1-3       (naturally coded; _Immig_1 omitted)
(running oaxaca on estimation sample)

```

```

BRR replications (120)
-----|-----|-----|-----|-----|
| 1 | 2 | 3 | 4 | 5 |
..... 50
..... 100
.....

```

```

Blinder-Oaxaca decomposition      Number of obs      =      16,116
                                   Population size      =      57,724.632
                                   Replications          =      120
                                   Design df            =      119

```

MINT2	BRR			t	P> t	[95% Conf. Interval]	
	Coef.	Std. Err.					
overall							
group_1	.2495778	.0061534	40.56	0.000	.2373935	.261762	
group_2	.0800784	.0042856	18.69	0.000	.0715926	.0885642	
difference	.1694994	.0071019	23.87	0.000	.1554368	.1835619	
explained	.0468527	.0037672	12.44	0.000	.0393932	.0543122	
unexplained	.1226467	.006774	18.11	0.000	.1092334	.1360599	
explained							
r_wlem	.0261858	.0029354	8.92	0.000	.0203735	.0319981	
d_matcon	.0165014	.0027049	6.10	0.000	.0111454	.0218575	
d_selfeff	.0068894	.0015983	4.31	0.000	.0037247	.0100541	
_IMINT2m_1	.0001043	.0001368	0.76	0.447	-.0001666	.0003753	
_IMINT2f_1	-.0003519	.0005985	-0.59	0.558	-.001537	.0008332	
_Icoachmath_2	-.0018858	.0008437	-2.24	0.027	-.0035563	-.0002152	
_Icoachmath_3	-.0001619	.0011934	-0.14	0.892	-.0025249	.0022011	
grademath	-.000409	.0013852	-0.30	0.768	-.0031518	.0023338	
_Igy_1	.0013913	.0006418	2.17	0.032	.0001205	.0026621	
_Irepeat_1	-.0008771	.0005222	-1.68	0.096	-.0019112	.0001569	

truancy	-.0000397	.0002519	-0.16	0.875	-.0005384	.0004591
_limmig_2	-.0003873	.0003411	-1.14	0.258	-.0010626	.0002881
_limmig_3	-.0004611	.0004177	-1.10	0.272	-.0012881	.000366
isei	.0005123	.0003835	1.34	0.184	-.0002471	.0012716
books	-.0001581	.0001611	-0.98	0.329	-.0004771	.000161
<hr/>						
unexplained						
r_wlem	.0213865	.0229989	0.93	0.354	-.0241536	.0669266
d_matcon	.0003685	.0008396	0.44	0.662	-.0012941	.0020311
d_selfeff	-.0000288	.0006375	-0.05	0.964	-.0012911	.0012335
_IMINT2m_1	-.0001437	.0013057	-0.11	0.913	-.002729	.0024417
_IMINT2f_1	.0036086	.0027146	1.33	0.186	-.0017666	.0089839
_lcoachmath_2	-.0000451	.0037482	-0.01	0.990	-.0074669	.0073767
_lcoachmath_3	.0060463	.0035804	1.69	0.094	-.0010433	.0131359
grademath	.0713113	.06336	1.13	0.263	-.0541478	.1967703
_Igy_1	-.0027863	.0050669	-0.55	0.583	-.0128192	.0072467
_Irepeat_1	-.0061655	.0035562	-1.73	0.086	-.0132072	.0008762
truancy	.0000401	.000538	0.07	0.941	-.0010252	.0011054
_limmig_2	.0024112	.0044754	0.54	0.591	-.0064506	.011273
_limmig_3	-.0024168	.0024729	-0.98	0.330	-.0073133	.0024798
isei	.0355297	.0271709	1.31	0.194	-.0182714	.0893308
books	-.0294199	.022415	-1.31	0.192	-.0738037	.014964
_cons	.0229504	.067371	0.34	0.734	-.110451	.1563518

```

.nlcom (r_wlem: _b[explained:r_wlem] / _b[overall:difference]*100) ///
> (d_matcon: _b[explained:d_matcon] / _b[overall:difference]*100) ///
> (d_selfeff: _b[explained:d_selfeff] / _b[overall:difference]*100) ///
> (d_X: (_b[explained:d_matcon]/*
> */+_b[explained:d_selfeff]) / _b[overall:difference]*100) ///
> (tot: (_b[explained:r_wlem]/*
> */+_b[explained:d_matcon]/*
> */+_b[explained:d_selfeff]) / _b[overall:difference]*100) ///
> (rest: (_b[overall:explained]-_b[explained:r_wlem]/*
> */-_b[explained:d_matcon]/*
> */-_b[explained:d_selfeff]) / _b[overall:difference]*100) ///
> , post
r_wlem: _b[explained:r_wlem] / _b[overall:difference]*100
d_matcon: _b[explained:d_matcon] / _b[overall:difference]*100
d_selfeff: _b[explained:d_selfeff] / _b[overall:difference]*100
d_X: (_b[explained:d_matcon]+_b[explained:d_selfeff]) / _b[overall:difference]*100
tot: (_b[explained:r_wlem]+_b[explained:d_matcon]+_b[explained:d_selfeff]) / _b[overall:difference]*
> 100
rest: (_b[overall:explained]-_b[explained:r_wlem]-_b[explained:d_matcon]-_b[explained:d_selfeff]) / _
> b[overall:difference]*100

```

MINT2	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
r_wlem	15.44889	1.647902	9.37	0.000	12.21906	18.67872
d_matcon	9.735403	1.672073	5.82	0.000	6.4582	13.0126
d_selfeff	4.064555	.9498163	4.28	0.000	2.20295	5.926161
d_X	13.79996	1.870131	7.38	0.000	10.13457	17.46535
tot	29.24885	2.542842	11.50	0.000	24.26497	34.23273
rest	-1.607033	1.535613	-1.05	0.295	-4.616779	1.402713

```

. est sto MINT2
. esttab MINT1 MINT2, b(1) nostar ci wide

```

	(1) MINT1		(2) MINT2	
r_wlem	13.4	[10.5,16.3]	15.4	[12.2,18.7]
d_matcon	10.9	[7.8,14.0]	9.7	[6.5,13.0]
d_selfeff	3.2	[1.4,5.0]	4.1	[2.2,5.9]
d_X	14.1	[10.9,17.4]	13.8	[10.1,17.5]
tot	27.5	[23.2,31.9]	29.2	[24.3,34.2]
rest	-0.1	[-2.9,2.6]	-1.6	[-4.6,1.4]

N 16116 16116

95% confidence intervals in brackets

5.2 Erweiterte Vorhersagemodelle für Leistung und Fehleinschätzung

```
. svy: regress r_wlem i.female i.MINT2m i.MINT2f i.lfp socnormsm socnormsf i.mathimpm i.mathimpf ///
> i.coachmath grademath i.gym i.repeat truancy i.immig isei books
(running regress on estimation sample)
```

BRR replications (120)

```
-----|-----|-----|-----|-----|-----|
      1       2       3       4       5
..... 50
..... 100
.....
```

```
Survey: Linear regression      Number of obs   =      7,139
                               Population size    = 23,509.267
                               Replications        =      120
                               Design df           =      119
                               F( 18, 102)         =    252.71
                               Prob > F           =      0.0000
                               R-squared           =      0.3947
```

r_wlem	Coef.	BRR Std. Err.	t	P> t	[95% Conf. Interval]	
1.female	-.0499838	.0069326	-7.21	0.000	-.063711	-.0362566
1.MINT2m	.0321407	.0181431	1.77	0.079	-.0037844	.0680658
1.MINT2f	.0179158	.0104143	1.72	0.088	-.0027056	.0385372
1.lfpm	-.0177321	.0068491	-2.59	0.011	-.031294	-.0041702
socnormsm	-.0020825	.0066213	-0.31	0.754	-.0151933	.0110284
socnormsf	.0121621	.006466	1.88	0.062	-.0006413	.0249655
1.mathimpm	-.0098547	.0085479	-1.15	0.251	-.0267803	.007071
1.mathimpf	-.0097829	.0072866	-1.34	0.182	-.024211	.0046453
coachmath						
Yes, sometimes	-.0722052	.0095524	-7.56	0.000	-.0911198	-.0532905
Yes, regularly	-.0723973	.0112132	-6.46	0.000	-.0946006	-.050194
grademath	.0965207	.0048654	19.84	0.000	.0868867	.1061546
1.gym	.2071322	.0085591	24.20	0.000	.1901844	.2240799
1.repeat	-.0940179	.0096213	-9.77	0.000	-.113069	-.0749668
truancy	-.037504	.0060447	-6.20	0.000	-.049473	-.0255349
immig						
Second generation	-.0686523	.0103975	-6.60	0.000	-.0892405	-.0480642
First generation	-.0337398	.0124014	-2.72	0.007	-.0582959	-.0091837
isei	.0014597	.000253	5.77	0.000	.0009587	.0019607
books	.0276396	.0028977	9.54	0.000	.0219019	.0333773
_cons	-.0827051	.0269081	-3.07	0.003	-.1359859	-.0294244

```
. svy: regress r_wlem i.female##(i.MINT2m i.MINT2f i.lfp c.socnormsm c.socnormsf i.mathimpm i.mathimpf ///
> i.coachmath c.grademath i.gym i.repeat c.truancy i.immig c.isei c.books) ///
> , vsquish nofvlabel
(running regress on estimation sample)
```

BRR replications (120)

```
-----|-----|-----|-----|-----|
      1       2       3       4       5
..... 50
..... 100
.....
```

```
Survey: Linear regression      Number of obs   =      7,139
                               Population size    = 23,509.267
                               Replications        =      120
```

Design df = 119
 F(35, 85) = 134.38
 Prob > F = 0.0000
 R-squared = 0.3971

r_wlem	BRR		t	P> t	[95% Conf. Interval]	
	Coef.	Std. Err.				
1.female	-.0895104	.0551671	-1.62	0.107	-.1987468	.0197259
1.MINT2m	.0340244	.0282703	1.20	0.231	-.0219536	.0900024
1.MINT2f	.018508	.015114	1.22	0.223	-.0114193	.0484352
1.lfpm	-.0233647	.0098518	-2.37	0.019	-.0428722	-.0038571
socnormsm	-.0028797	.0098611	-0.29	0.771	-.0224058	.0166463
socnormsf	.0131168	.0097616	1.34	0.182	-.0062121	.0324456
1.mathimpm	-.0177764	.0123739	-1.44	0.153	-.0422779	.0067251
1.mathimpf	-.0016555	.012764	-0.13	0.897	-.0269294	.0236184
coachmath						
2	-.0828671	.0148418	-5.58	0.000	-.1122554	-.0534789
3	-.0682583	.0211709	-3.22	0.002	-.1101788	-.0263378
grademath	.0888503	.0073672	12.06	0.000	.0742624	.1034381
1.gym	.206231	.0133558	15.44	0.000	.1797852	.2326769
1.repeat	-.0993661	.0136121	-7.30	0.000	-.1263194	-.0724128
truancy	-.0380804	.0085242	-4.47	0.000	-.0549592	-.0212016
immig						
2	-.0537269	.0143609	-3.74	0.000	-.0821628	-.0252909
3	-.0376813	.0231214	-1.63	0.106	-.083464	.0081014
isei	.0021459	.0003474	6.18	0.000	.001458	.0028338
books	.02396	.0038149	6.28	0.000	.0164061	.0315139
female#MINT2m						
1 1	-.0080346	.0352226	-0.23	0.820	-.077779	.0617097
female#MINT2f						
1 1	-.000807	.0183924	-0.04	0.965	-.0372259	.0356119
female#lfpm						
1 1	.0110981	.0132015	0.84	0.402	-.0150421	.0372383
female#c.socnormsm						
1	.0010402	.0131214	0.08	0.937	-.0249415	.0270218
female#c.socnormsf						
1	-.0016106	.0118923	-0.14	0.892	-.0251585	.0219373
female#mathimpm						
1 1	.0158897	.0188777	0.84	0.402	-.0214901	.0532694
female#mathimpf						
1 1	-.0159707	.0191116	-0.84	0.405	-.0538137	.0218723
female#coachmath						
1 2	.0209824	.0172097	1.22	0.225	-.0130944	.0550592
1 3	-.0015969	.0253112	-0.06	0.950	-.0517157	.0485219
female#c.grademath						
1	.0157118	.0091625	1.71	0.089	-.0024309	.0338545
female#gym						
1 1	.0017629	.0174659	0.10	0.920	-.0328214	.0363471
female#repeat						
1 1	.0119884	.0212315	0.56	0.573	-.030052	.0540288
female#c.truancy						
1	.0024241	.0117362	0.21	0.837	-.0208147	.0256629
female#immig						
1 2	-.028457	.0172782	-1.65	0.102	-.0626696	.0057555
1 3	.0045831	.0294063	0.16	0.876	-.0536444	.0628106
female#c.isei						
1	-.0013756	.0004591	-3.00	0.003	-.0022847	-.0004665
female#c.books						
1	.0076298	.0053717	1.42	0.158	-.0030068	.0182664
_cons	-.0639187	.0412056	-1.55	0.124	-.1455098	.0176725

```
. // => gender difference only in isei
. svy: regress d_matcon i.female i.MINT2m i.MINT2f i.lfpm socnormsm socnormsf i.mathimpm i.mathimpf ///
> i.coachmath grademath i.gym i.repeat truancy i.immig isei books
(running regress on estimation sample)
BRR replications (120)
```


socnormsf	.0040217	.0110716	0.36	0.717	-.0179012	.0259446
1.mathimpm	.0324186	.0152592	2.12	0.036	.0022039	.0626333
1.mathimpf	.0166275	.0156356	1.06	0.290	-.0143326	.0475876
coachmath						
2	-.0013652	.0180864	-0.08	0.940	-.037178	.0344477
3	-.0451301	.031328	-1.44	0.152	-.1071627	.0169026
grademath	.1408964	.0098662	14.28	0.000	.1213603	.1604325
1.gym	-.1876522	.0150915	-12.43	0.000	-.2175349	-.1577694
1.repeat	.1004826	.0146094	6.88	0.000	.0715544	.1294107
truancy	.0397348	.0096377	4.12	0.000	.0206511	.0588185
immig						
2	.0633282	.013684	4.63	0.000	.0362326	.0904238
3	.0604702	.0240933	2.51	0.013	.0127631	.1081772
isei	-.0025504	.0004338	-5.88	0.000	-.0034094	-.0016913
books	-.0252932	.0040862	-6.19	0.000	-.0333842	-.0172021
female#MINT2m						
1 1	.0513199	.0446471	1.15	0.253	-.0370857	.1397255
female#MINT2f						
1 1	-.0172825	.0226628	-0.76	0.447	-.0621572	.0275922
female#lfpm						
1 1	-.0219791	.0145332	-1.51	0.133	-.0507562	.0067981
female#c.socnormsm						
1	.0099165	.0150629	0.66	0.512	-.0199096	.0397426
female#c.socnormsf						
1	-.0233328	.0152569	-1.53	0.129	-.053543	.0068775
female#mathimpm						
1 1	.0010155	.0226391	0.04	0.964	-.0438122	.0458433
female#mathimpf						
1 1	.0160561	.0231587	0.69	0.489	-.0298005	.0619127
female#coachmath						
1 2	.0039011	.0214196	0.18	0.856	-.0385118	.0463139
1 3	.0440366	.039522	1.11	0.267	-.034221	.1222942
female#c.grademath						
1	.0055232	.0119804	0.46	0.646	-.0181993	.0292456
female#gym						
1 1	.0017831	.0187994	0.09	0.925	-.0354416	.0390079
female#repeat						
1 1	.0042842	.0224639	0.19	0.849	-.0401967	.0487651
female#c.truancy						
1	-.0111927	.0124686	-0.90	0.371	-.0358817	.0134963
female#immig						
1 2	.014511	.0185399	0.78	0.435	-.0221999	.0512219
1 3	.0083429	.0294529	0.28	0.777	-.0499767	.0666625
female#c.isei						
1	.0023097	.0005518	4.19	0.000	.001217	.0034024
female#c.books						
1	-.0122973	.0056253	-2.19	0.031	-.023436	-.0011586
_cons	-.4221085	.0558519	-7.56	0.000	-.5327008	-.3115162

```

. // => gender difference in isei and books
. svy: regress d_selfeff i.female i.MINT2m i.MINT2f i.lfp socnormsm socnormsf i.mathimpm i.mathimpf ///
> i.coachmath grademath i.gym i.repeat truancy i.immig isei books
(running regress on estimation sample)

```

BRR replications (120)

```

-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|
| 1 | 2 | 3 | 4 | 5
..... 50
..... 100
.....

```

Survey: Linear regression

```

Number of obs   =    7,139
Population size  =  23,509.267
Replications    =    120
Design df       =    119
F( 18, 102)    =    31.13
Prob > F        =    0.0000
R-squared       =    0.1030

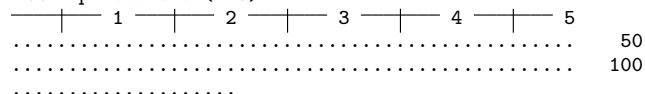
```

BRR

d_selfeff	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
1.female	-.0401152	.00777	-5.16	0.000	-.0555006	-.0247298
1.MINT2m	-.0205242	.0222314	-0.92	0.358	-.0645447	.0234963
1.MINT2f	-.0067126	.0106388	-0.63	0.529	-.0277784	.0143532
1.lfpm	.0223495	.0074334	3.01	0.003	.0076307	.0370684
socnormsm	.016575	.0075954	2.18	0.031	.0015354	.0316145
socnormsf	.0135423	.0064499	2.10	0.038	.0007709	.0263138
1.mathimpm	.0289999	.0097927	2.96	0.004	.0096094	.0483904
1.mathimpf	.0233923	.0093135	2.51	0.013	.0049506	.0418339
coachmath						
Yes, sometimes	.0284986	.0094979	3.00	0.003	.0096917	.0473054
Yes, regularly	.0201315	.0130477	1.54	0.126	-.0057042	.0459672
grademath						
1.gym	-.0616124	.0094903	-6.49	0.000	-.0804043	-.0428206
1.repeat	.0547194	.0097246	5.63	0.000	.0354637	.073975
truancy	.0187829	.0062487	3.01	0.003	.0064098	.031156
immig						
Second generation	.0644164	.0110452	5.83	0.000	.0425458	.0862871
First generation	.0710866	.0143008	4.97	0.000	.0427696	.0994036
isei	-.0003589	.0002611	-1.37	0.172	-.000876	.0001581
books	-.0028823	.0032738	-0.88	0.380	-.0093648	.0036001
_cons	-.1500036	.0345252	-4.34	0.000	-.2183669	-.0816404

```
. svy: regress d_selfeff i.female##(i.MINT2m i.MINT2f i.lfpm c.socnormsm c.socnormsf i.mathimpm i.mathimpf ///
> i.coachmath c.grademath i.gym i.repeat c.truancy i.immig c.isei c.books) ///
> , vsquish nofvlabel
(running regress on estimation sample)
```

BRR replications (120)



Survey: Linear regression

```
Number of obs      =      7,139
Population size     = 23,509.267
Replications        =      120
Design df           =      119
F( 35, 85)         =      17.03
Prob > F            =      0.0000
R-squared           =      0.1076
```

d_selfeff	Coef.	BRR Std. Err.	t	P> t	[95% Conf. Interval]	
1.female	.0582595	.068314	0.85	0.395	-.0770089	.193528
1.MINT2m	-.0826649	.0334918	-2.47	0.015	-.148982	-.0163478
1.MINT2f	-.0036507	.0150802	-0.24	0.809	-.033511	.0262095
1.lfpm	.0309815	.0114278	2.71	0.008	.0083533	.0536096
socnormsm	.0254018	.0108416	2.34	0.021	.0039344	.0468692
socnormsf	.0195217	.010553	1.85	0.067	-.0013744	.0404178
1.mathimpm	.0306904	.0142781	2.15	0.034	.0024182	.0589625
1.mathimpf	.0124199	.0163334	0.76	0.449	-.0199219	.0447616
coachmath						
2	.0227242	.0164685	1.38	0.170	-.0098852	.0553336
3	.0249747	.0255936	0.98	0.331	-.0257033	.0756526
grademath	.0338933	.0088978	3.81	0.000	.0162748	.0515118
1.gym	-.0505517	.0155521	-3.25	0.001	-.0813464	-.0197571
1.repeat	.0543526	.0144603	3.76	0.000	.0257197	.0829855
truancy	.0230336	.0095228	2.42	0.017	.0041775	.0418897
immig						
2	.0556759	.0166849	3.34	0.001	.0226381	.0887137
3	.0659195	.0211341	3.12	0.002	.0240718	.1077672
isei	-.0003417	.00037	-0.92	0.358	-.0010743	.000391

books	-0.0004426	.0044434	-0.10	0.921	-.009241	.0083558
female#MINT2m						
1 1	.119081	.047031	2.53	0.013	.025955	.212207
female#MINT2f						
1 1	-.0019682	.0248696	-0.08	0.937	-.0512125	.0472761
female#lfpf						
1 1	-.015649	.0150155	-1.04	0.299	-.0453812	.0140833
female#c.socnormsm						
1	-.0162936	.0154593	-1.05	0.294	-.0469045	.0143174
female#c.socnormsf						
1	-.0130586	.0138016	-0.95	0.346	-.0403871	.01427
female#mathimpf						
1 1	-.0033729	.0192821	-0.17	0.861	-.0415533	.0348075
female#mathimpf						
1 1	.0223684	.0210987	1.06	0.291	-.0194091	.064146
female#coachmath						
1 2	.008072	.0235108	0.34	0.732	-.0384817	.0546258
1 3	-.0034108	.032003	-0.11	0.915	-.06678	.0599584
female#c.grademath						
1	-.0169602	.0113027	-1.50	0.136	-.0393407	.0054202
female#gym						
1 1	-.0182385	.0221715	-0.82	0.412	-.0621403	.0256634
female#repeat						
1 1	.0007283	.0203225	0.04	0.971	-.0395122	.0409688
female#c.truancy						
1	-.0085129	.0120709	-0.71	0.482	-.0324144	.0153886
female#immig						
1 2	.0151642	.0213527	0.71	0.479	-.0271163	.0574446
1 3	.0076704	.0289472	0.26	0.791	-.049648	.0649888
female#c.isei						
1	-.0000132	.0005569	-0.02	0.981	-.0011159	.0010894
female#c.books						
1	-.0053864	.0057596	-0.94	0.352	-.016791	.0060181
_cons	-.2027562	.0505835	-4.01	0.000	-.3029166	-.1025958

. // => gender difference in MINT2m (such that effect for females is zero)