HRS testing

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% Table created by stargazer v.5.2.2 by Marek Hlavac, Harvard University. E-mail: hlavac at fas.harvard.edu % Date and time: Fri, Feb 22, 2019 - 5:35:55 PM

Table 1: Black only

	cognitive			
	Crude	**Weight**	**Crude**	**Weight**
	Model 1	**Model 2**	**Model 3**	**Model 4**
as.factor(female)1	0.179***	0.175***	0.139***	0.136***
	(0.040)	(0.040)	(0.033)	(0.033)
as.factor(cohort_group)1930	−0.132**	−0.127**	−0.275***	−0.267***
	(0.059)	(0.062)	(0.050)	(0.052)
as.factor(cohort_group)1940	−0.094	−0.088	−0.366***	−0.359***
	(0.066)	(0.069)	(0.057)	(0.059)
as.factor(cohort_group)1950	0.003	0.056	−0.428**	−0.362**
	(0.216)	(0.212)	(0.176)	(0.173)
edu_years			0.431***	0.435***
			(0.016)	(0.016)
wealth			0.017*	0.021**
			(0.009)	(0.009)
log_income			0.039***	0.033***
			(0.009)	(0.009)
age_poly_1	−0.456***	−0.455***	−0.442***	−0.439***
	(0.015)	(0.017)	(0.015)	(0.017)
age_poly_2	−0.082***	−0.081***	−0.077***	−0.076***
	(0.012)	(0.014)	(0.012)	(0.014)
Constant	−0.067	−0.065	0.112**	0.109**
	(0.058)	(0.061)	(0.049)	(0.051)
N	10585	10111	10577	10103
Log Likelihood	−11333.480	−11333.390	−10989.900	−10989.550
AIC	22694.960	22694.780	22013.800	22013.100
BIC	22796.700	22795.880	22137.330	22135.850

***p < .01; **p < .05; *p < .1

```
## Computing p-values via Wald-statistics approximation (treating t as Wald z).

## Caution! ICC for random-slope-intercept models usually not meaningful. Use `adjusted = TRUE` to use the mean ran

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```

cognitive

cognitive

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Predictors

Estimates

CI

p

Estimates

CI

p

Estimates

CI

p

Estimates

CI

p

(Intercept)

-0.07

-0.18 – 0.05

0.249

-0.07

-0.18 – 0.05

0.281

0.11

0.02 – 0.21

0.022

0.11

0.01 – 0.21

0.033

as factor(female)1

0.18

0.10 – 0.26

<0.001

0.18

0.10 – 0.25

<0.001

0.14

0.07 – 0.20

<0.001

0.14

0.07 – 0.20

<0.001

as factor(cohortgroup)1930

-0.13

-0.25 – -0.02

0.025

-0.13

-0.25 – -0.01

0.040

-0.28

-0.37 – -0.18

<0.001

-0.27

-0.37 – -0.16

<0.001

as factor(cohortgroup)1940

-0.09

-0.22 – 0.04

0.159

-0.09

-0.22 – 0.05

0.204

-0.37

-0.48 – -0.25

<0.001

-0.36

-0.48 – -0.24

<0.001

as factor(cohortgroup)1950

0.00

-0.42 – 0.43

0.990

0.06

-0.36 – 0.47

0.790
-0.43
-0.77 – -0.08
0.015
-0.36
-0.70 – -0.02
0.036
age poly 1
-0.46
-0.49 – -0.43
<0.001
-0.46
-0.49 – -0.42
<0.001
-0.44
-0.47 – -0.41
<0.001
-0.44
-0.47 – -0.41
<0.001
age poly 2
-0.08
-0.11 – -0.06
<0.001
-0.08
-0.11 – -0.05
<0.001
-0.08
-0.10 – -0.05
<0.001
-0.08
-0.10 – -0.05
<0.001
edu years
0.43
0.40 – 0.46
<0.001
0.44

0.40 – 0.47

<0.001

wealth

0.02

-0.00 – 0.03

0.056

0.02

0.00 – 0.04

0.015

log income

0.04

0.02 – 0.06

<0.001

0.03

0.02 – 0.05

<0.001

Random Effects

2

0.33

0.07

0.33

0.07

00

0.46 id_factor

0.45 id_factor

0.26 id_factor

0.25 id_factor

11

0.08 id_factor.age_poly_1

0.08 id_factor.age_poly_1

0.07 id_factor.age_poly_1

0.07 id_factor.age_poly_1

0.01 id_factor.age_poly_2

0.04 id_factor.age_poly_2

0.01 id_factor.age_poly_2

0.04 id_factor.age_poly_2

01

-0.14 id_factor.age_poly_1

-0.11 id_factor.age_poly_1

-0.13 id_factor.age_poly_1

-0.14 id_factor.age_poly_1

-0.80 id_factor.age_poly_2

-0.46 id_factor.age_poly_2

-0.85 id_factor.age_poly_2

-0.57 id_factor.age_poly_2

ICC

0.58 id_factor

0.87 id_factor

0.45 id_factor

0.79 id_factor

Observations

10585

10111

10577

10103

Marginal R2 / Conditional R2

0.137 / 0.674

0.181 / 0.910

0.332 / 0.671

0.446 / 0.909