

Analyzing a
Longitudinal
Study on
Cognitive
Decline Using
Mixed-Effects
Models

Jeremy Faria

Introduction

Health and
Retirement
Study

Longitudinal
Data

Mixed-Effects
Model

Exploratory
Data Analysis

Next Steps

Analyzing a Longitudinal Study on Cognitive Decline Using Mixed-Effects Models

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In Canada less than 1% of citizens ages 65-69 have dementia.
This increases to 25% for ages 85 and older.

There are modifiable and non-modifiable factors that lead to cognitive decline. This research is aimed to:

- Encourage people to change their lifestyle
- Get tested sooner if they feel at risk

Health and Retirement study (HRS)

- A multi-purpose longitudinal survey from 1992-2022 on United States citizens ages 50 and over

Cohort Title	Birth Years	Year Added into Survey
HRS Cohort	1931-1941	1992
AHEAD Cohort	1924 or earlier	1993
Child of Depression	1924-1930	1998
War Baby	1942-1947	1998
Early Baby Boomer	1948-1953	2004
Mid Baby Boomer	1954-1959	2010
Late Baby Boomer	1960-1965	2016
Early Generation X	1966-1971	2022

Table: Cohorts of HRS Survey

Step 1: Select primary sampling units (PSU)



Step 2: Sample smaller areas in PSU's

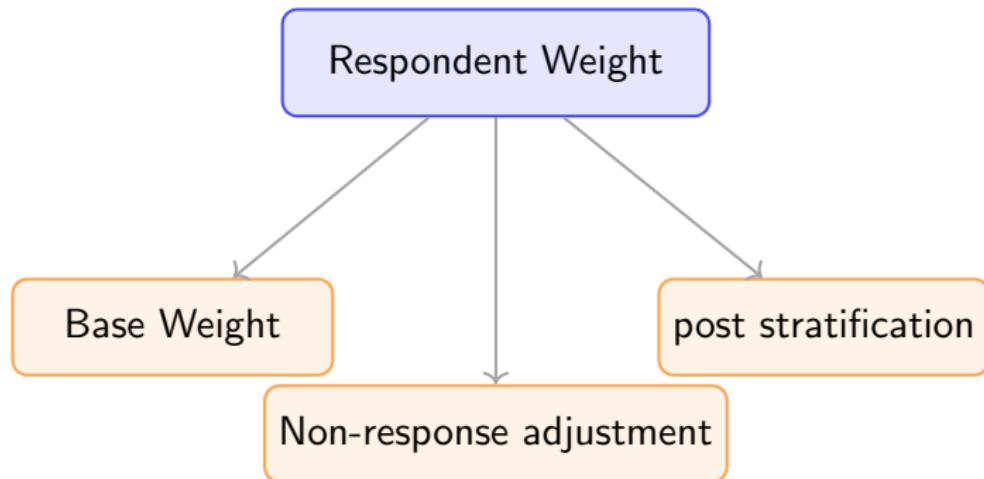


Step 3: A cluster of homes is selected



Step 4: Age-eligible individuals are selected

Longitudinal analyzes use the latest available weight of the respondent.



Cognition on the 27 point scale

The HRS used a clinician approved measurement on cognition using these tests:

- **10 points** for immediate noun recall
- **10 points** for delayed noun recall
- **5 points** for counting backwards from 100
- **2 points** for counting backwards from 20

Although not a formal diagnosis, clinicians believed the scores could be categorized into these groups:

- **12-27 points** - normal cognition
- **7-11 points** - cognitive impairment without dementia
- **0-6 points** - dementia

Longitudinal Studies

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- Able to track cognitive decline **overtime** through **repeated observations**
- Easier to find **cause and effect** relationships

Covariance Structure

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- $\vec{y} = [y_{11}, y_{1,2}, \dots, y_{1,n}, y_{2,1}, \dots, y_{m,n}]^T$ where y_{ij} is observation j for subject i

$$\text{var}(\vec{y}) = \begin{bmatrix} V_1 & 0_{nxn} & 0_{nxn} & \cdots & 0_{nxn} \\ 0_{nxn} & V_2 & 0_{nxn} & \cdots & 0_{nxn} \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ 0_{nxn} & 0_{nxn} & \cdots & \cdots & V_m \end{bmatrix}_{(m*n) \times (m*n)}$$

$$\text{where } V_i = \begin{bmatrix} \sigma_{i1}^2 & \rho_{12}\sigma_{i1}\sigma_{i2} & \cdots & \rho_{1n}\sigma_{i1}\sigma_{in} \\ \rho_{12}\sigma_{i2}\sigma_{i1} & \sigma_{i2}^2 & \cdots & \rho_{2n}\sigma_{i2}\sigma_{in} \\ \vdots & \vdots & \ddots & \vdots \\ \rho_{1n}\sigma_{in}\sigma_{i1} & \rho_{2n}\sigma_{in}\sigma_{i2} & \cdots & \sigma_{in}^2 \end{bmatrix}_{nxn}$$

- Violates the independence between observations assumption in linear regression

Linear Mixed-Effects Model

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For observation j of individual i the linear mixed-effects model can be written as:

$$y_{ij} = \beta_0 + \beta_1 x_{1ij} + \dots + \beta_p x_{pij} + \gamma_{0i} + \gamma_{1i} z_{1ij} + \dots + \gamma_{ki} z_{kij} + \epsilon_{ij}$$

- **Fixed effects β** are population based inferences
- **Random effects γ_i** are individual deviations from fixed effects for individual i

Random Effects : $\gamma_{0i} + \gamma_{1i}z_{1ij} + \dots + \gamma_{ki}z_{kij}$

$$\gamma \sim MVN \left(\vec{0}_{k \times 1}, \begin{bmatrix} \sigma_{\gamma_0}^2 & \sigma_{\gamma_0}\sigma_{\gamma_1} & \dots & \dots \\ \sigma_{\gamma_0}\sigma_{\gamma_1} & \sigma_{\gamma_1}^2 & \dots & \dots \\ \vdots & \vdots & \ddots & \vdots \\ \vdots & \vdots & \ddots & \sigma_{\gamma_k}^2 \end{bmatrix} \right)$$

- Random intercepts **very common**
- Random slope for time **somewhat common**
- Random slope for other covariates **not as common**

Random intercepts model correlation by:

$$Corr(y_{ij}, y_{ik}) = \frac{\sigma_{\gamma_0}^2}{\sigma_{\gamma_0}^2 + \sigma^2}$$

Age as a Covariate

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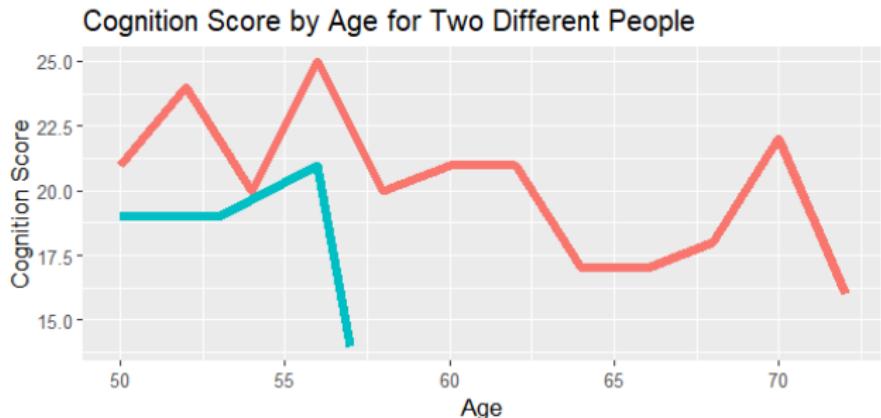
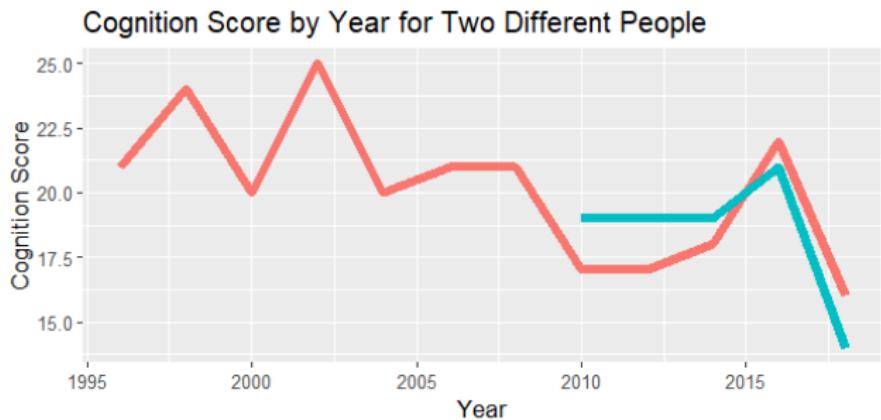
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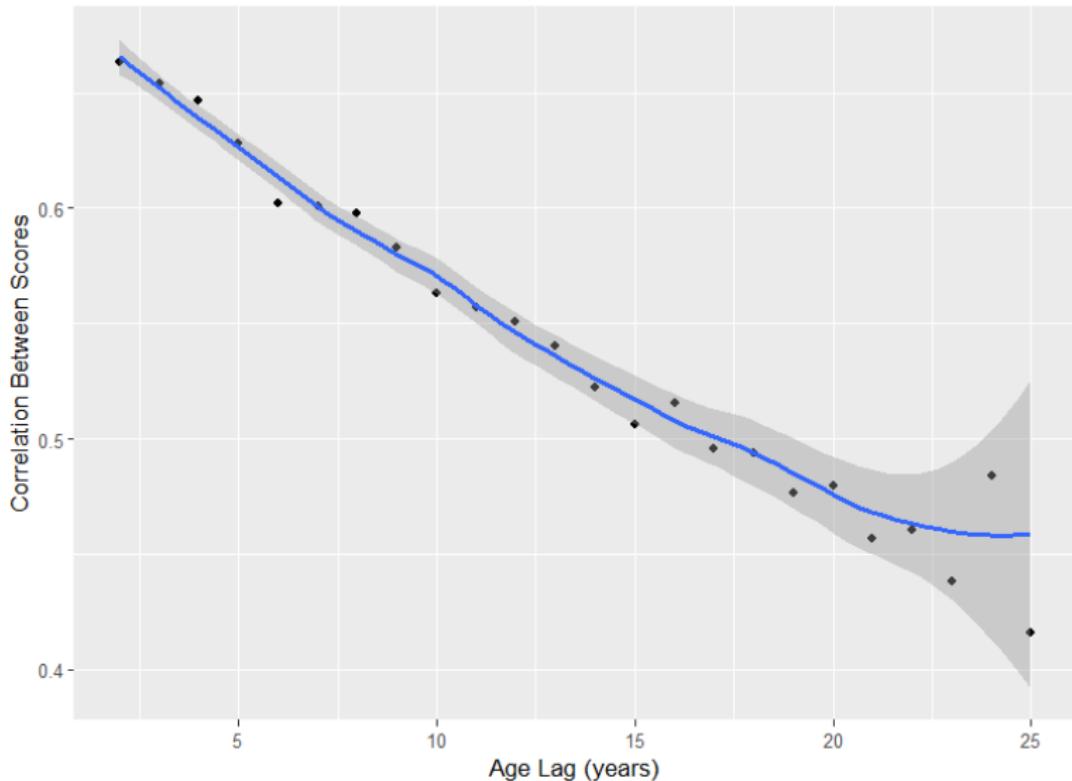
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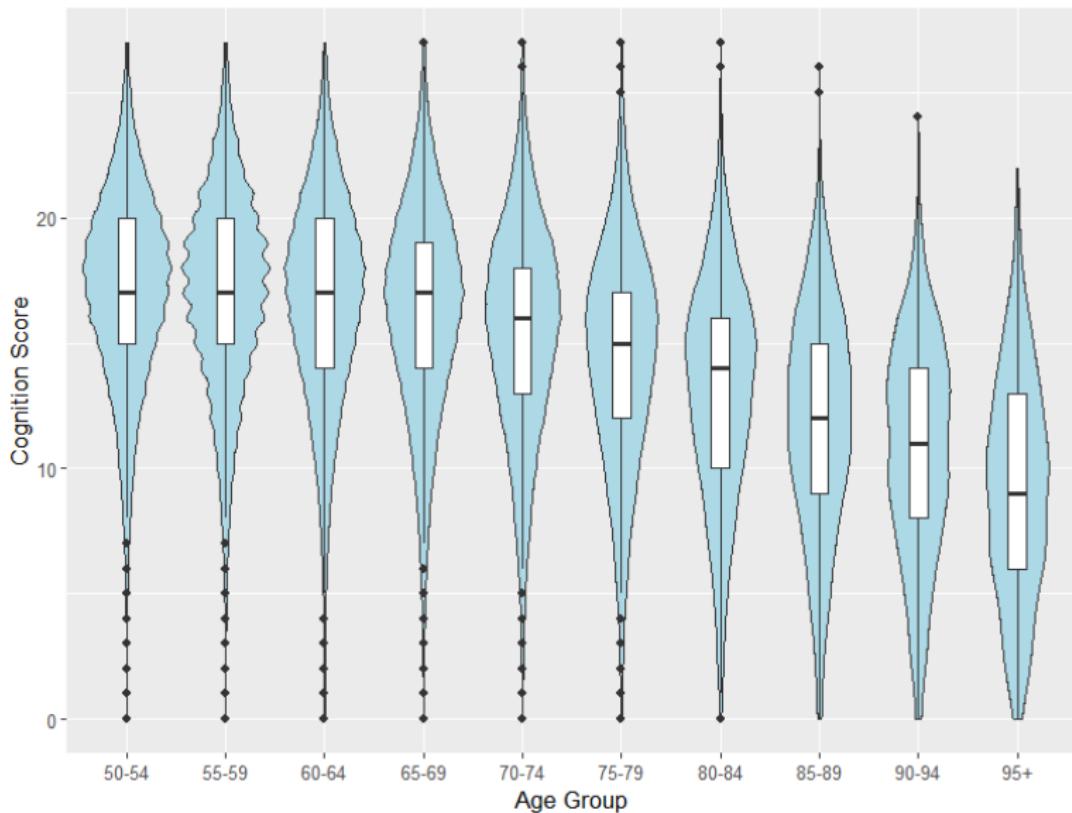
Correlation Structure of HRS data

Lag Correlation of Cognitive Score by Age Difference
With Weighted LOESS Smoothing



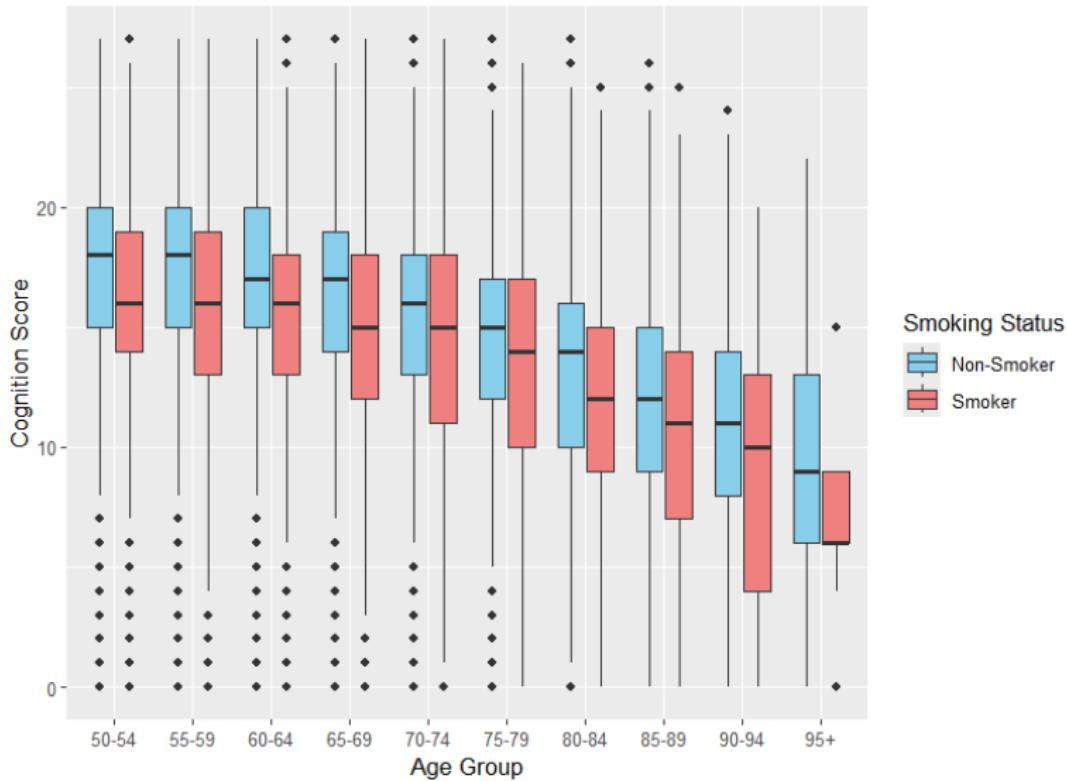
Distribution of Cognition Scores in Age Groups

Weighted Distribution of Cognition Scores Across Age Groups



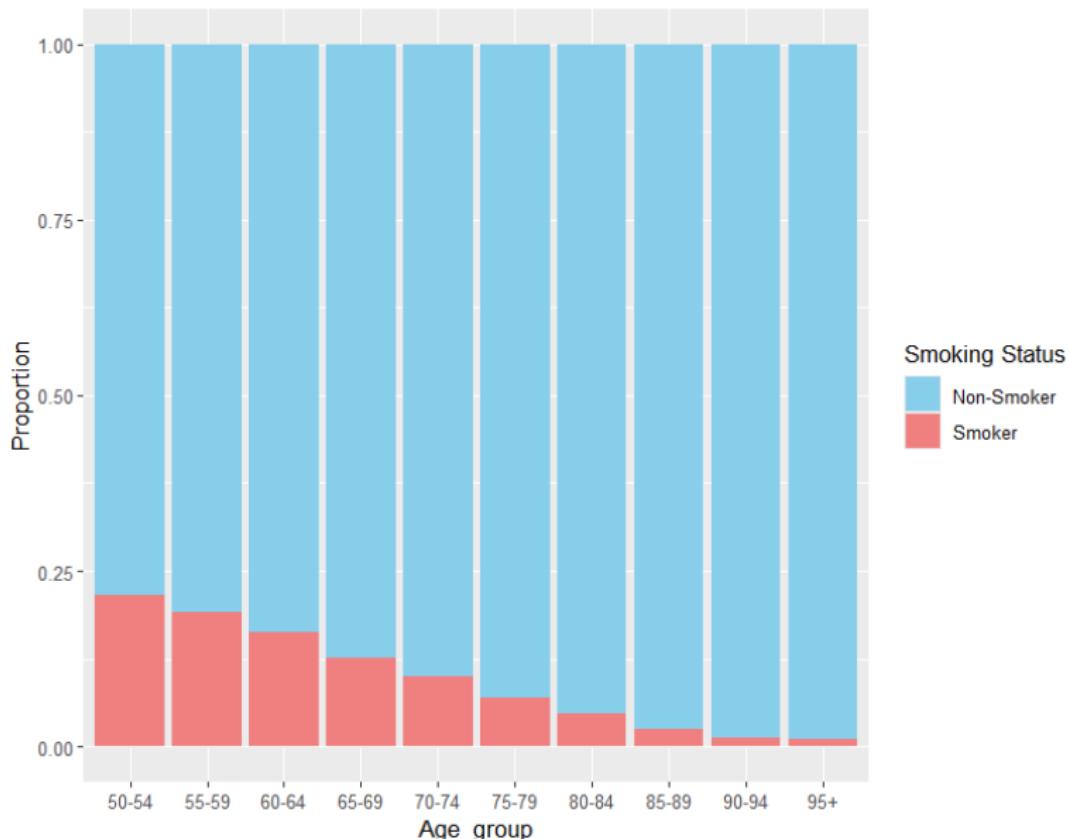
Smoking as a Covariate

Weighted Distribution of Cognition Score Across Age Groups
Between Smokers and Non-Smokers



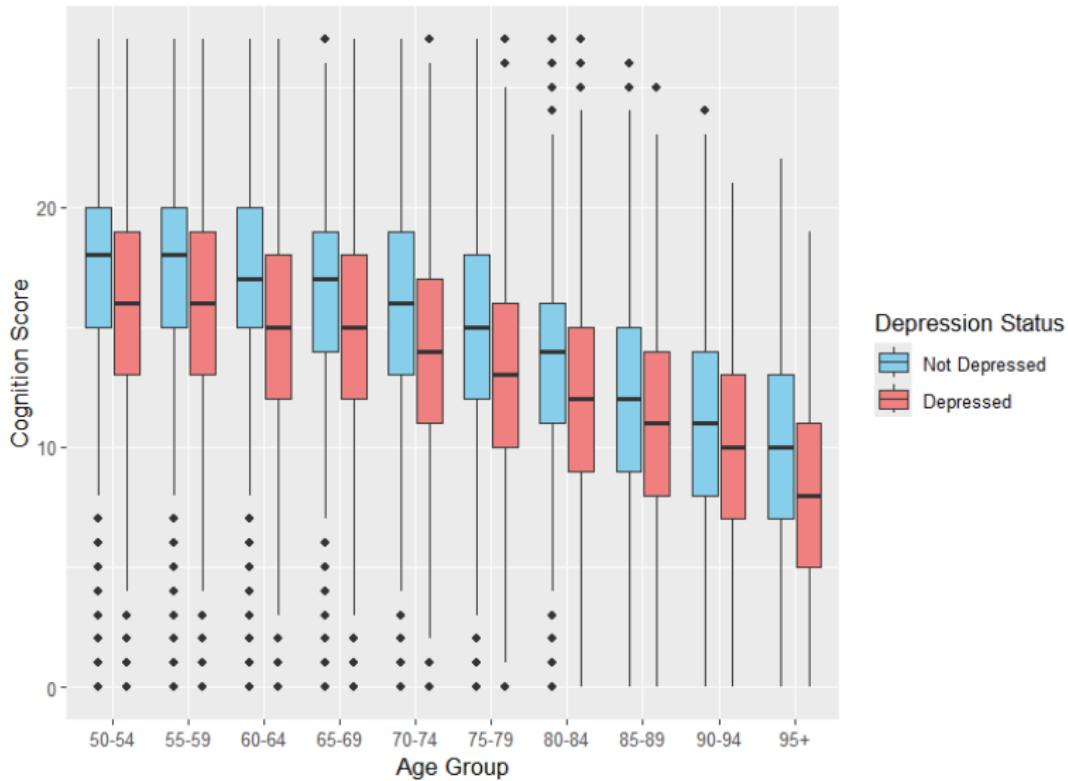
Smoking as a Covariate

Weighted Distribution of Smokers for each Age Group



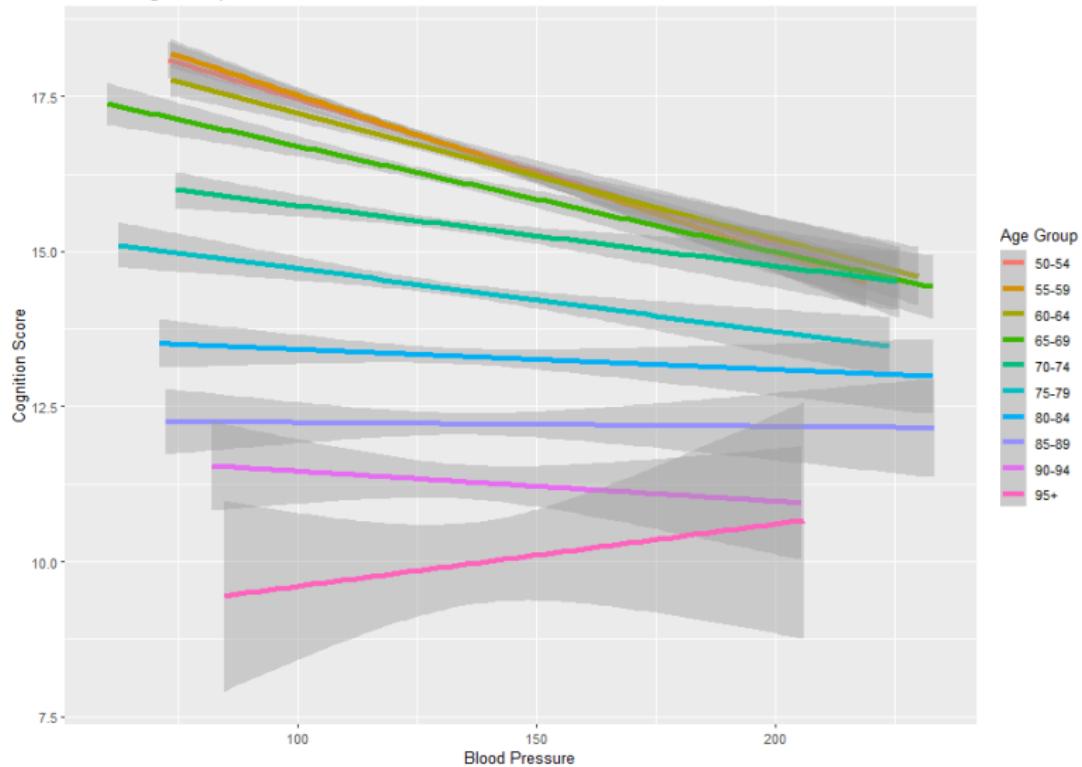
Depression as a Covariate

Weighted Distribution of Cognition Score Across Age Groups
On Depression Status



Systolic Blood Pressure as a Covariate

Weighted Regression Lines on Cognition Scores By Blood Pressure
Across Age Groups



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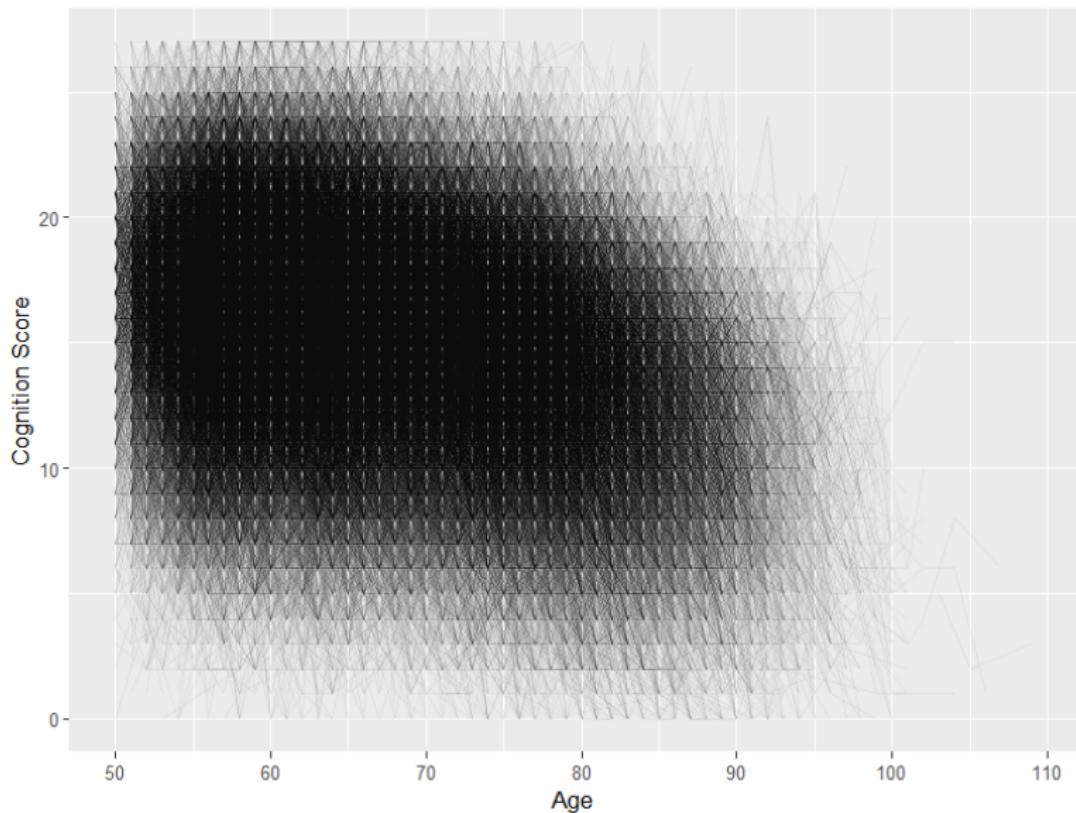
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Next Steps

- Set up models
- Estimation and hypothesis testing
- Model selection
- Diagnostics

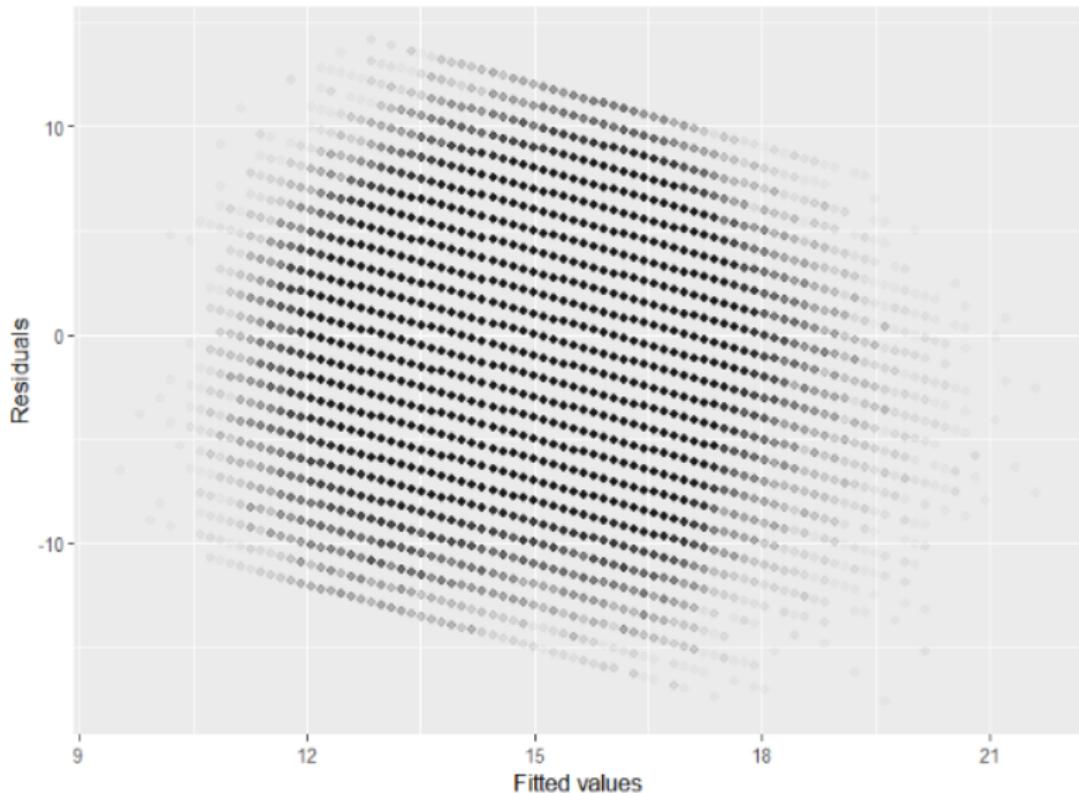
Appendix - Individual Cognition Scores Trajectories

Cognition Scores Across Ages for Each Individual



Appendix - Linear regression on Longitudinal Data

Residual Plot with Density Visualization
of The OLS Model of Cognition Score by Age



Appendix - Restricted Maximum Likelihood Estimation

- Weighted Least Squares and Maximum Likelihood underestimate variance and covariance parameters
- REML provides unbiased estimates for variance and covariance

MLE:

$$\hat{\sigma}(V_0) = \frac{RSS(V_0)}{nm}$$

REML:

$$\hat{\sigma}(V_0) = \frac{RSS(V_0)}{nm - p}$$

Appendix - Model Evaluation

- **Model comparisons:** AIC, BIC, Likelihood ratio test (random effects), marginal R^2 (fixed effects), conditional R^2

Marginal:

$$R_c^2 = \frac{Var(X\beta)}{Var(X\beta) + Var(Z\gamma) + Var(\epsilon)}$$

Conditional:

$$R_c^2 = \frac{Var(X\beta) + Var(Z\gamma)}{Var(X\beta) + Var(Z\gamma) + Var(\epsilon)}$$

- **Visually:** residual plots, violin plots of residuals of different times, QQ plots for residuals and random effects