

# 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

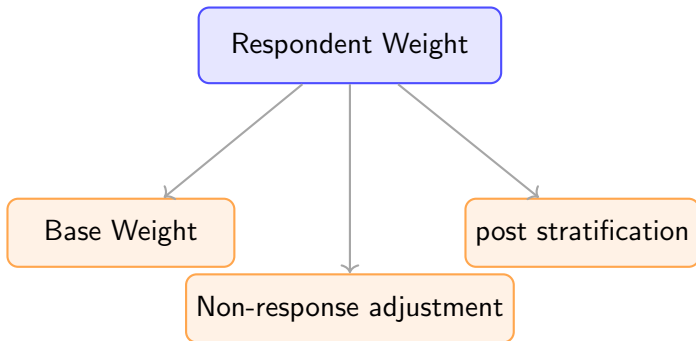


Step 3: A cluster of homes is selected



Step 4: Age-eligible individuals are selected

Longitudinal analyzes use the latest available weight of the respondent.



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

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

# Covariance Structure

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

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

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

- Violates the independence between observations assumption in linear regression



# Linear Mixed-Effects Model

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  $\beta$  are population based inferences
- Random effects  $\gamma_i$  are individual deviations from fixed effects for individual  $i$

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

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

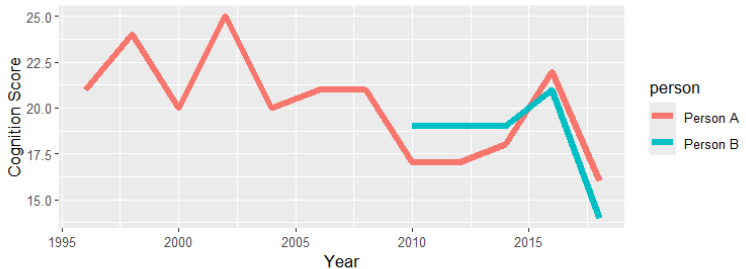
- Random intercepts **very common**
- Random slope for time **somewhat common**
- Random slope for other covariates **not as common**

Random intercepts model correlation by:

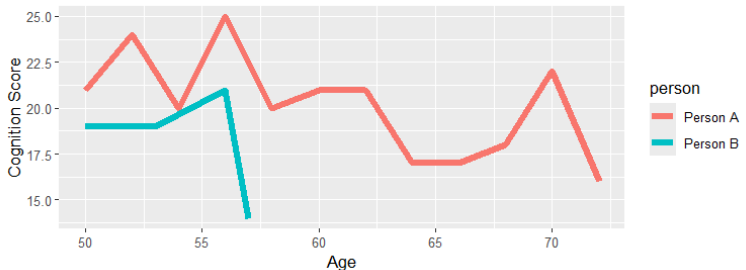
$$\text{Corr}(y_{ij}, y_{ik}) = \frac{\sigma_{\gamma 0}^2}{\sigma_{\gamma 0}^2 + \sigma^2}$$

# Age as a Covariate

## Cognition Score by Year for Two Different People

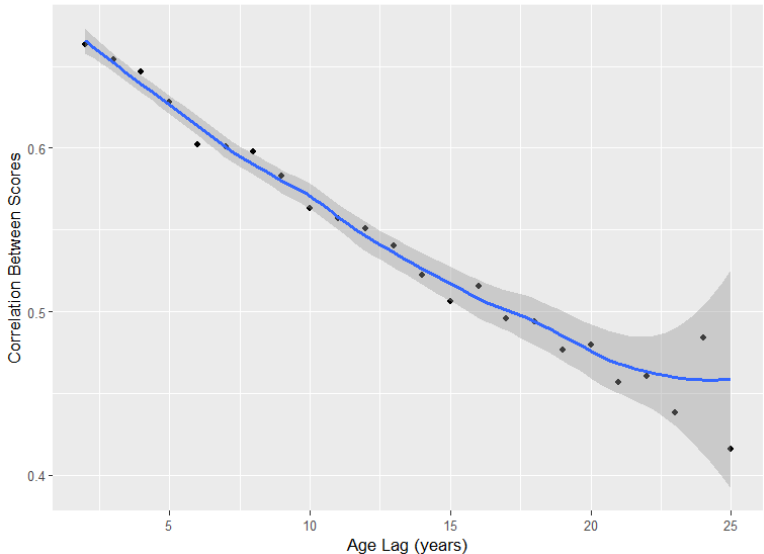


## Cognition Score by Age for Two Different People



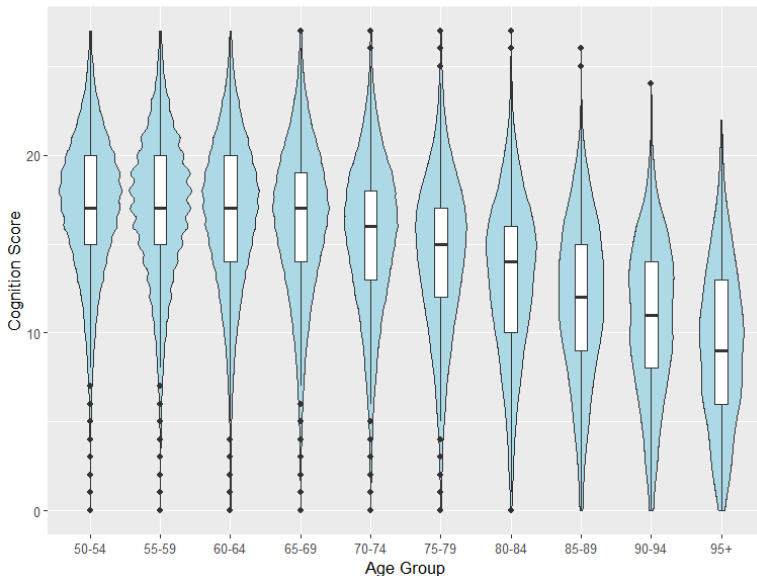
# Correlation Structure of HRS data

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



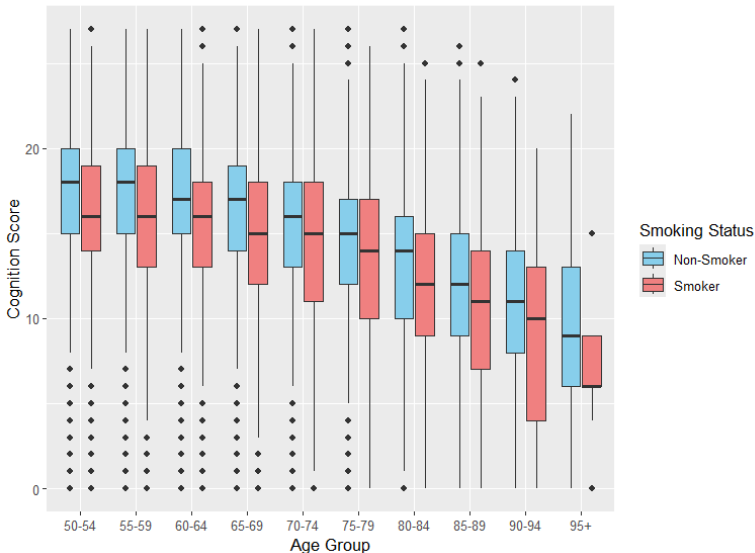
# Distribution of Cognition Scores in Age Groups

Weighted Distribution of Cognition Scores Across Age Groups

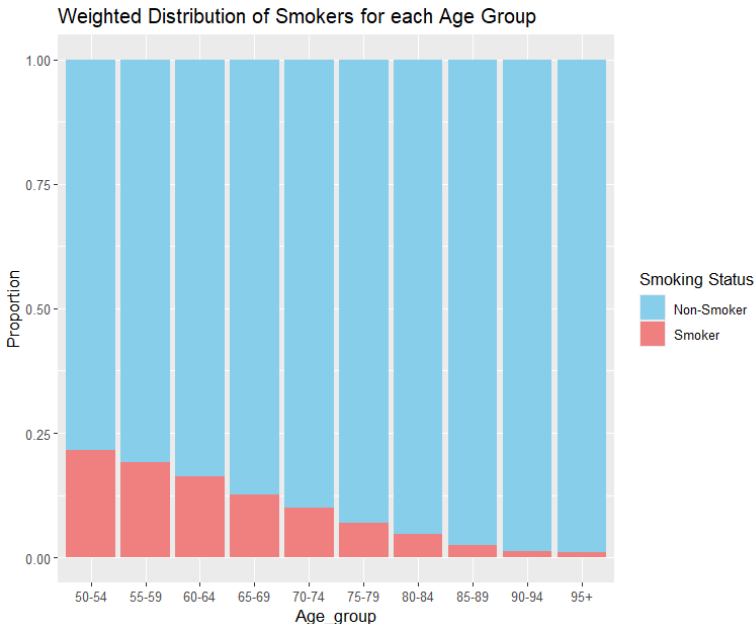


# Smoking as a Covariate

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

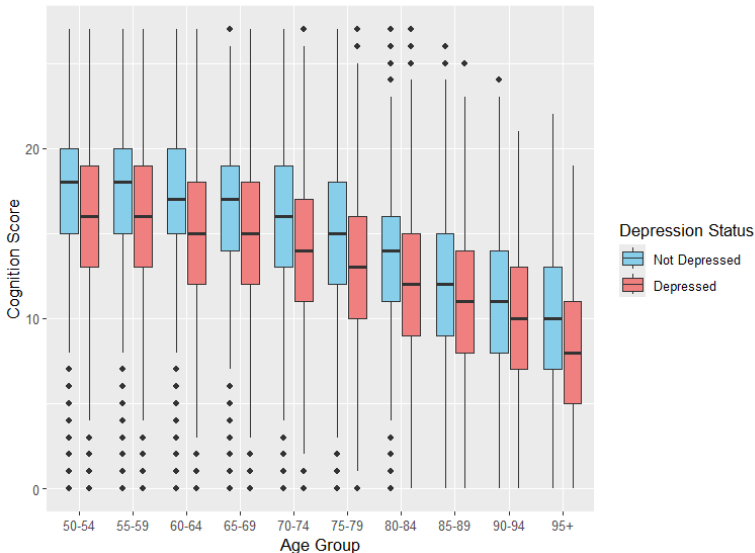


# Smoking as a Covariate



# Depression as a Covariate

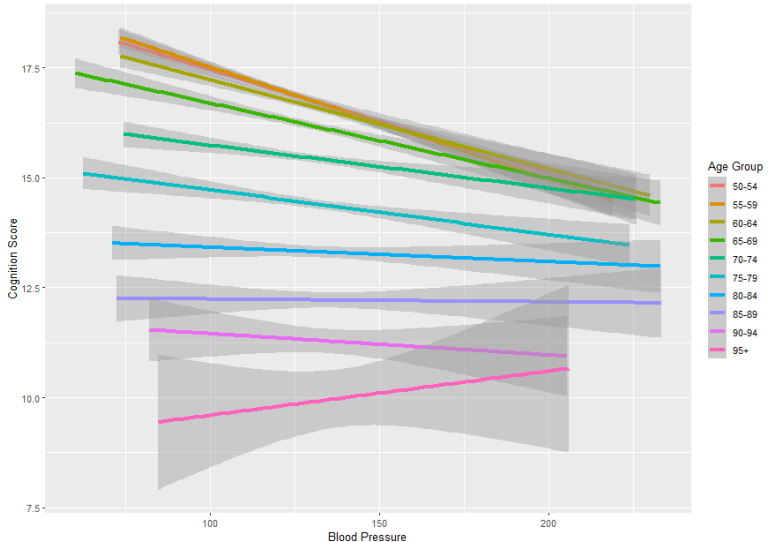
Weighted Distribution of Cognition Score Across Age Groups  
On Depression Status





# Systolic Blood Pressure as a Covariate

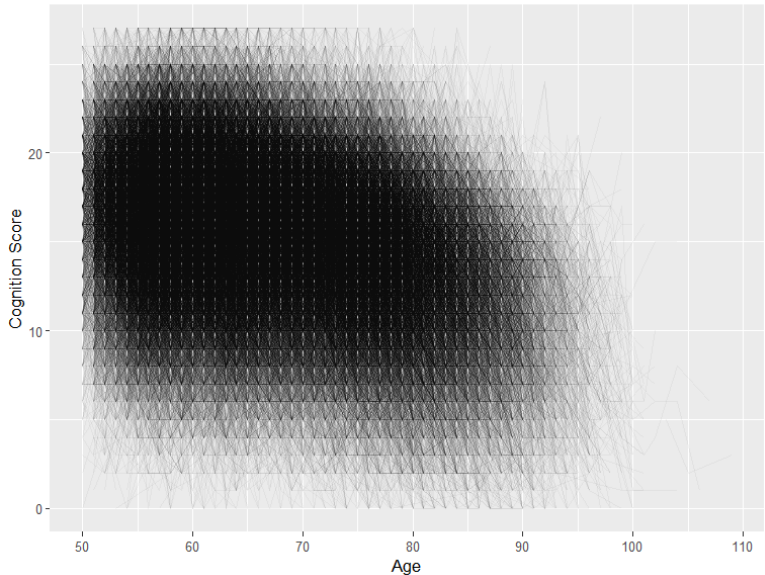
Weighted Regression Lines on Cognitions Scores By Blood Pressure Across Age Groups



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

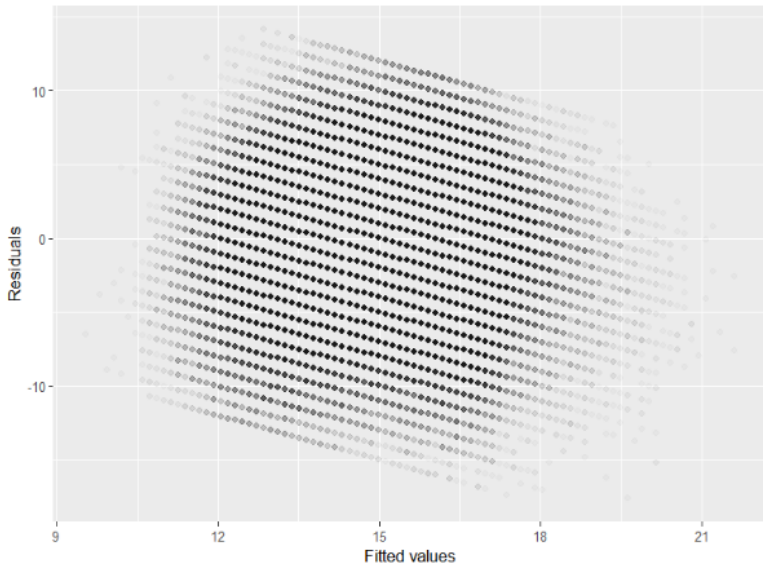
# Appendix - Individual Cognition Scores Trajectories

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



Analyzing a  
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Jeremy Faria

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Exploratory  
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Next Steps

# Appendix - Restricted Maximum Likelihood Estimation

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- 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:**

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

**Conditional:**

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

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