

## Agenda

1. Adding student-level predictors
2. Adding class-level predictors
3. Random intercepts in R

# Intercept-only model

**Partial pooling  
(random effects)**

$$S_{ik} \sim \text{Norm}(\mu_{ik}, \sigma)$$

$$\mu_{ik} = a_k$$

$$a_k \sim \text{Norm}(\gamma, \phi)$$

$$\sigma \sim \text{Unif}(0, 100)$$

$$\gamma \sim \text{Norm}(500, 100)$$

$$\eta \sim \text{Unif}(0, 100)$$

# Accounting for student race

**Number of participants by race/ethnicity**

<i>White</i>	4222
<i>Black</i>	2126
<i>Asian</i>	19
<i>Hispanic</i>	9
<i>Native American</i>	4
<i>Other</i>	11
<b><i>Total</i></b>	<b>6391</b>

**Number of classes by experimental condition**

<i>Small</i>	122
<i>Large</i>	114
<i>Large + Aide</i>	98
<b><i>Total</i></b>	<b>334</b>

# Accounting for student race

## But first, some notation

$$S_{ik} \sim \text{Norm}(\mu_{ik}, \sigma)$$

$$\mu_{ik} = \beta_{0k} + \beta_1 B_i$$

$\beta_{0k}$  is the intercept for classroom  $k$

$$\beta_{0k} \sim \text{Norm}(\gamma_0, \phi_0)$$

Coefficient  $\beta_1$  measures difference in test scores for Black and white students

Subscripts on  $\gamma$  and  $\phi$  to remind us which coefficient they refer to ( $\beta_{0k}$ )

# Accounting for student race

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$$\beta_{0k} \sim \text{Norm}(\gamma_0, \phi_0)$$

$$S_{ik} \sim \text{Norm}(\mu_{ik}, \sigma)$$

$$\mu_{ik} = \beta_{0k} + \beta_1 B_i$$

$$\beta_{0k} = \gamma_0 + \eta_{0k}$$

$$\eta_{0k} \sim \text{Norm}(0, \phi_0)$$

Equivalent ways to describe the same distribution for  $\beta_{0k}$ .

# Accounting for student race

## Full model

$$S_{ik} \sim \text{Norm}(\mu_{ik}, \sigma)$$

$$\begin{aligned} \mu_{ik} = & \beta_{0k} + \beta_1 \text{Black}_i + \beta_2 \text{Asian}_i \\ & + \beta_3 \text{Hispanic}_i + \beta_4 \text{NativeAmerican}_i \\ & + \beta_5 \text{Other}_i \end{aligned}$$

$$\beta_{0k} = \gamma_0 + \eta_{0k}$$

$$\eta_{0k} \sim \text{Norm}(0, \phi_0)$$

$$\beta_1, \dots, \beta_5 \sim \text{Norm}(0, 50)$$

$$\sigma \sim \text{Unif}(0, 100)$$

$$\gamma_0 \sim \text{Norm}(500, 100)$$

$$\phi_0 \sim \text{Unif}(0, 100)$$

# Accounting for student race

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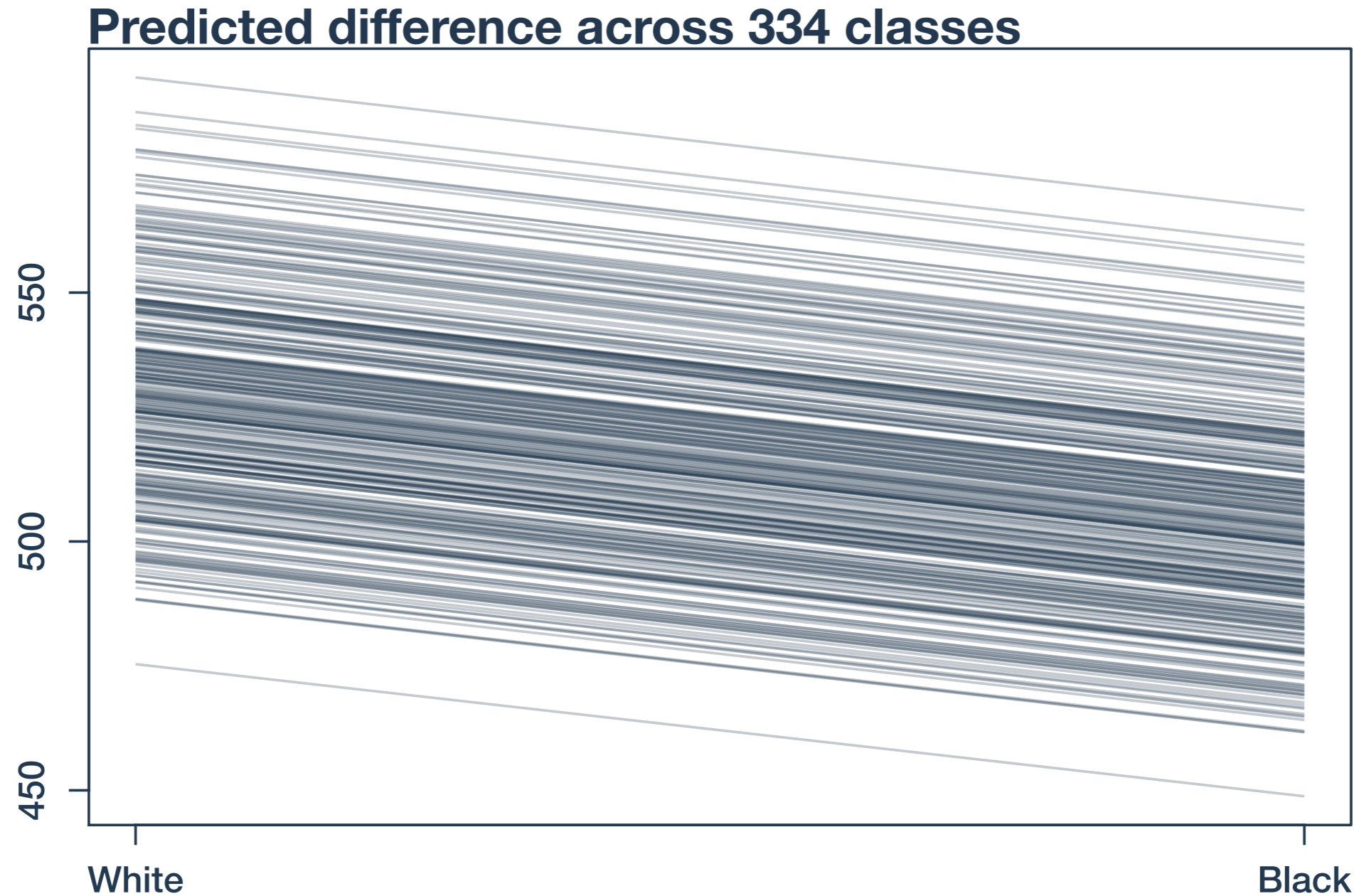
$$\sigma \sim \text{Unif}(0, 100)$$

$$\gamma_0 \sim \text{Norm}(500, 100)$$

$$\phi_0 \sim \text{Unif}(0, 100)$$

	<i>Mean</i>	<i>90% credible interval</i>	
<b><math>\gamma_0</math></b>	530.65	528.21	533.29
<b><math>\beta_1</math></b>	-26.78	-29.52	-23.17
<b><math>\beta_2</math></b>	6.70	-9.03	25.07
<b><math>\beta_3</math></b>	16.39	-10.53	41.08
<b><math>\beta_4</math></b>	-13.10	-49.60	20.18
<b><math>\beta_5</math></b>	14.81	-9.17	33.92
<b><math>\sigma</math></b>	47.03	46.37	47.75
<b><math>\phi_0</math></b>	23.98	22.01	25.83

# Accounting for student race



$$\gamma_0 = 530.65$$

$$\beta_1 = -26.78$$

$$\phi_0 = 23.98$$



# Comparing to pooled model

$$S_{ik} \sim \text{Norm}(\mu_{ik}, \sigma)$$

$$\begin{aligned} \mu_{ik} = & \beta_0 + \beta_1 \text{Black}_i + \beta_2 \text{Asian}_i \\ & + \beta_3 \text{Hispanic}_i + \beta_4 \text{NativeAmerican}_i \\ & + \beta_5 \text{Other}_i \end{aligned}$$

$$\beta_0, \dots, \beta_5 \sim \text{Norm}(0, 50)$$

$$\sigma \sim \text{Unif}(0, 100)$$

	<i>Mean</i>	<i>90% credible interval</i>	
<b><math>\beta_0</math></b>	533.00	531.59	534.29
<b><math>\beta_1</math></b>	-36.86	-39.11	-34.71
<b><math>\beta_2</math></b>	12.61	-7.84	29.98
<b><math>\beta_3</math></b>	7.61	-21.58	31.06
<b><math>\beta_4</math></b>	-28.47	-66.14	8.46
<b><math>\beta_5</math></b>	13.83	-10.70	35.93
<b><math>\sigma</math></b>	52.37	51.49	53.07

# Comparing to pooled model

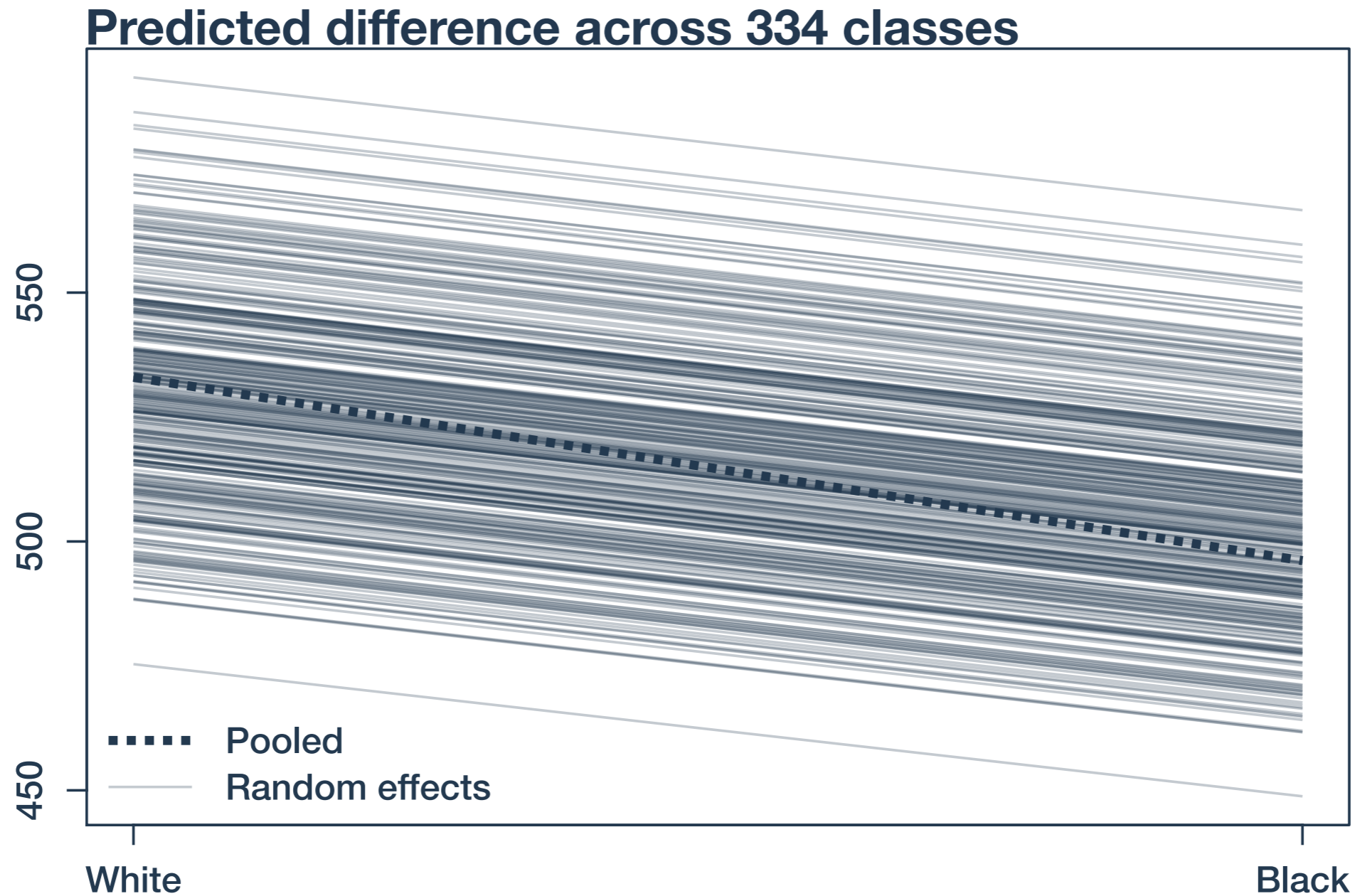
## Random intercept

	<i>Mean</i>	<i>90% credible interval</i>	
$\gamma_0$	530.65	528.21	533.29
$\beta_1$	-26.78	-29.52	-23.17
$\beta_2$	6.70	-9.03	25.07
$\beta_3$	16.39	-10.53	41.08
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## Pooled

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# Accounting for student race



$$\text{Random effects} \left| \begin{array}{l} \gamma_0 = 530.65 \\ \beta_1 = -26.78 \\ \phi_0 = 23.98 \end{array} \right.$$

$$\text{Pooled} \left| \begin{array}{l} \beta_0 = 533.00 \\ \beta_1 = -36.86 \end{array} \right.$$

# Accounting for class size

**Number of  
participants by  
race/ethnicity**

<i>White</i>	4222
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# Accounting for class size

## Full model

$$S_{ik} \sim \text{Norm}(\mu_{ik}, \sigma)$$

$$\begin{aligned} \mu_{ik} = & \beta_{0k} + \beta_1 \text{Black}_i + \beta_2 \text{Asian}_i \\ & + \beta_3 \text{Hispanic}_i + \beta_4 \text{NativeAmerican}_i \\ & + \beta_5 \text{Other}_i \end{aligned}$$

$$\beta_{0k} = \gamma_0 + \gamma_1 \text{Small}_k + \eta_{0k}$$

$$\eta_{0k} \sim \text{Norm}(0, \phi_0)$$

$$\beta_1, \dots, \beta_5 \sim \text{Norm}(0, 50)$$

$$\sigma \sim \text{Unif}(0, 100)$$

$$\gamma_0 \sim \text{Norm}(500, 100)$$

$$\gamma_1 \sim \text{Norm}(0, 50)$$

$$\phi_0 \sim \text{Unif}(0, 100)$$

Coefficient  $\gamma_1$  measures average difference in test score for classes in the “small” experimental condition.

# Accounting for class size

**Note** For computational efficiency and other pragmatic reasons, second-level terms like  $\gamma_1$  sometimes need to be implemented as first-level components.

In random-intercept models, this is a trivial matter, but it gets more complex with models we will cover soon.

$$S_{ik} \sim \text{Norm}(\mu_{ik}, \sigma)$$
$$\mu_{ik} = \beta_{0k} + \beta_1 \text{Black}_i + \beta_2 \text{Asian}_i$$
$$+ \beta_3 \text{Hispanic}_i + \beta_4 \text{NativeAmerican}_i$$
$$+ \beta_5 \text{Other}_i$$

$$\beta_{0k} = \gamma_0 + \gamma_1 \text{Small}_k + \eta_{0k}$$
$$\eta_{0k} \sim \text{Norm}(0, \phi_0)$$

Think about  $\gamma_1$  as a second-level variable.

Mathematically,  $\gamma_1$  can be included at the first level.

$$S_{ik} \sim \text{Norm}(\mu_{ik}, \sigma)$$
$$\mu_{ik} = \beta_{0k} + \gamma_1 \text{Small}_k + \beta_1 \text{Black}_i + \beta_2 \text{Asian}_i$$
$$+ \beta_3 \text{Hispanic}_i + \beta_4 \text{NativeAmerican}_i$$
$$+ \beta_5 \text{Other}_i$$

$$\beta_{0k} = \gamma_0 + \eta_{0k}$$
$$\eta_{0k} \sim \text{Norm}(0, \phi_0)$$

# Accounting for class size

$$S_{ik} \sim \text{Norm}(\mu_{ik}, \sigma)$$

$$\begin{aligned} \mu_{ik} = & \beta_{0k} + \beta_1 \text{Black}_i + \beta_2 \text{Asian}_i \\ & + \beta_3 \text{Hispanic}_i + \beta_4 \text{NativeAmerican}_i \\ & + \beta_5 \text{Other}_i \end{aligned}$$

$$\beta_{0k} = \gamma_0 + \gamma_1 \text{Small}_k + \eta_{0k}$$

$$\eta_{0k} \sim \text{Norm}(0, \phi_0)$$

	<i>Mean</i>	<i>90% credible interval</i>	
<b><math>\gamma_0</math></b>	526.15	523.34	529.04
<b><math>\gamma_1</math></b>	12.37	8.26	16.82
<b><math>\beta_1</math></b>	-26.70	-29.88	-23.51
<b><math>\beta_2</math></b>	7.34	-8.59	24.19
<b><math>\beta_3</math></b>	16.37	-6.66	42.92
<b><math>\beta_4</math></b>	-14.17	-48.66	23.74
<b><math>\beta_5</math></b>	14.83	-8.74	35.25
<b><math>\sigma</math></b>	47.03	46.33	47.69
<b><math>\phi_0</math></b>	23.32	21.32	25.31
<b><math>\beta_{1k}</math></b>	⋮	⋮	⋮

# Accounting for class size

