## Lesson 6, Part 2: Linear Mixed Effects Models

## This Lesson's Goals

Learn about other methods for LMEMs

Update our LMEMs in R
Summarise results in an R Markdown document

But, in the ANOVA we got rid of our baseline issue, can I do that with an LMEM?

## Math (Part 1)

## $y_{i}=a+b_{1} x_{1 i}+b_{2} x_{2 i}+b_{3} x_{1 i} x_{2 i}+e_{i}$

 Percentage of Votes for Incumbentby Country in Civil War and Party of Incumbent

- Democrat • Republican



## $y_{i}=a+b_{1} x_{1 i}+b_{2} x_{2 i}+b_{3} x_{1 i} x_{2 i}+e_{i}$

 Percentage of Votes for Incumbentby Country in Civil War and Party of Incumbent
$\begin{array}{cc}\text { - Democrat } \\ 0 & \text { Republican } \\ 1\end{array}$


## $y_{i}=a+b_{1} x_{1 i}+b_{2} x_{2 i}+b_{3} x_{1 i} x_{2 i}+e_{i}$

 Percentage of Votes for Incumbentby Country in Civil War and Party of Incumbent


# $y_{i}=a+b_{1} x_{1 i}+b_{2} x_{2 i}+b_{3} x_{1 i} x_{2 i}+e_{i}$ <br> Percentage of Votes for Incumbent by Country in Civil War and Party of Incumbent 





# $y_{i}=a+b_{1} x_{1 i}+b_{2} x_{2 i}+b_{3} x_{1 i} x_{2 i}+e_{i}$ <br> Percentage of Votes for Incumbent <br> by Country in Civil War and Party of Incumbent 



R Code (Part 1)

## lme4

$$
y_{i}=a+a_{s}+a_{y}+\left(b_{s 1}+b_{1}\right) x_{1 i}+\left(b_{y 1}+b_{2}\right) x_{2 i}+b_{3} x_{1 i} x_{2 i}+e_{i}
$$

lmer(perc_votes incumbent ~
incumbent party_contrast * civil_war_contrast + (1+incumbent_party_contrast|state) + (1+civil_war_contrast|year))

Fixed effects:

## (Intercept)

incumbent_party_contrast
civil_war_contrast
incumbent_party_contrast:civil_war_contrast 18.39637

Estimate Std. Error t value
$52.04864 \quad 3.1134616 .717$
$2.92546 \quad 6.37541 \quad 0.459$
$0.04273 \quad 2.77575 \quad 0.015$

## LMEM with dummy coding

Fixed effects:
(Intercept)
incumbent_partyRepublican
civil_warConfederacy
incumbent_partyRepublican:civil_warConfederacy

Estimate Std. Error t value

| 55.164 | 5.591 | 9.866 |
| ---: | ---: | ---: |
| -6.273 | 8.012 | -0.783 |
| -9.155 | 4.183 | -2.189 |
| 18.396 | 6.189 | 2.972 |

## LMEM with contrast coding

Fixed effects:
(Intercept)
incumbent_party_contrast civil_war_contrast
incumbent_party_contrast:civil_war_contrast
Estimate Std. Error $t$ value

| 52.04864 | 3.11346 | 16.717 |
| ---: | ---: | ---: |
| 2.92546 | 6.37541 | 0.459 |
| 0.04273 | 2.77575 | 0.015 |
| 18.39637 | 6.18901 | 2.972 |

How do I get p-values out of this?

## Math (Part 2)

very good fit

not as good fit


R Code (Part 2)
$y_{i}=a+a_{s}+a_{y}+\left(b_{s 1}+b_{1}\right) x_{1 i}+\left(b_{y 1}+b_{2}\right) x_{2 i}+b_{3} x_{1 i} x_{2 i}+e_{i}$ lmer (perc_votes_incumbent ~ incumbent_party_contrast * civil_war_contrast + (1+incumbent_party_contrast|state) + (1+civil_war_contrast|year), REML = F)

$$
\begin{gathered}
y_{i}=a+a_{s}+a_{y}+\left(b_{s 1}+b_{1}\right) x_{1 i}+\left(b_{y 1}+b_{2}\right) x_{2 i}+b_{3} x_{1 i} x_{2 i} \\
-b_{1} x_{1 i}+e_{i}
\end{gathered}
$$

lmer (perc_votes_incumbent ~ incumbent_party_contras ${ }^{\text {F }}$ * civil_war_contrast - incumbent_party_contrast + (1+incumbent_party_contrast|state)

+ (1+civil_war_contrast|year), REML = F)

$$
y_{i}=a+a_{s}+a_{y}+\left(b_{s 1}+b_{1}\right) x_{1 i}+\left(b_{y 1}+b_{2}\right) x_{2 i}+b_{3} x_{1 i} x_{2 i}+e_{i}
$$

$y_{i}=a+a_{s}+a_{y}+\left(b_{s 1}+b_{1}\right) x_{1 i}+\left(b_{y 1}+b_{2}\right) x_{2 i}+b_{3} x_{1 i} x_{2 i}$ $-b_{1} X_{1 i}+e_{i}$
anova(m1, m2_party)


## LMEM summary

Fixed effects:
(Intercept)
incumbent_party_contrast civil_war_contrast

Estimate Std. Error t value

| 52.04934 | 2.70659 | 19.231 |
| ---: | ---: | ---: |
| 2.92405 | 5.57867 | 0.524 |
| 0.04413 | 2.43834 | 0.018 |
| 18.39355 | 5.57298 | 3.300 |

model comparison for incumbent party

|  | Df | AIC | BIC | logLik | deviance | Chisq Chi | Df $\operatorname{Pr}(>$ Chisq) |  |
| :--- | :--- | ---: | ---: | ---: | ---: | ---: | ---: | ---: |
| m2_party | 10 | 1153.5 | 1185.1 | -566.74 | 1133.5 |  |  |  |
| m1 | 11 | 1155.2 | 1190.0 | -566.61 | 1133.2 | 0.2706 | 1 | 0.6029 |

model comparison for civil war
Df AIC BIC logLik deviance Chisq Chi Df Pr(>Chisq)
m3_country $101153.21184 .9-566.61 \quad 1133.2$
$\begin{array}{llllllll}\mathrm{m} 1 \quad 11 & 1155.2 & 1190.0 & -566.61 & 1133.2 & 3 \mathrm{e}-04 & 1 & 0.98\end{array} \quad \begin{array}{llll} & & & \\ \text { model comparison for incumbent party x civil war }\end{array}$

|  | Df | AIC | BIC | logLik deviance Chisq Chi | Df | $\operatorname{Pr}(>C h i s q)$ |  |
| :--- | :--- | :--- | ---: | ---: | ---: | ---: | ---: |
| m4_partyxcountry | 10 | 1160.8 | 1192.4 | -570.39 | 1140.8 |  |  |
| m1 | 11 | 1155.2 | 1190.0 | -566.61 | 1133.2 | 7.567 | 1 |

## Lab

## Data set: Stroop Task

Congruency: Are responses to incongruent trials less accurate and slower than to congruent trials?

Experiment half: Are responses more accurate and faster in the second half of the experiment then the first half of the experiment?

Congruency x Experiment half: Is there an interaction between these variables?

```
accuracy (logistic)
logit }\mp@subsup{\textrm{p}}{\textrm{i}}{}=\mathrm{ accuracy
x1 = congruency
x2 = experiment half
r1 = participant
r2 = item
```

reaction times (linear)
$y_{i}=$ reaction times
x1 = congruency
x2 = experiment half
r1 = participant
r2 = item

## dplyr

data_accuracy_stats = data_accuracy_clean

## dplyr

data_accuracy_stats = data_accuracy_clean \%>\% mutate(congruency_contrast =

## dplyr

data_accuracy_stats = data_accuracy_clean \%>\%
mutate(congruency_contrast =


## dplyr

data_accuracy_stats = data_accuracy_clean \%>\% mutate(congruency_contrast =

call original
variable

## dplyr

data_accuracy_stats = data_accuracy_clean \%>\% mutate(congruency_contrast =


## dplyr

data_accuracy_stats = data_accuracy_clean \%>\%
mutate(congruency_contrast =


## dplyr

data_accuracy_stats = data_accuracy_clean \%>\%
mutate(congruency_contrast =


