# Linear mixed effect model- Birth rates data

HIERARCHICAL AND MIXED EFFECTS MODELS IN R



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### Chapter outline

- Random-effect intercept model
- Random-effect slope model

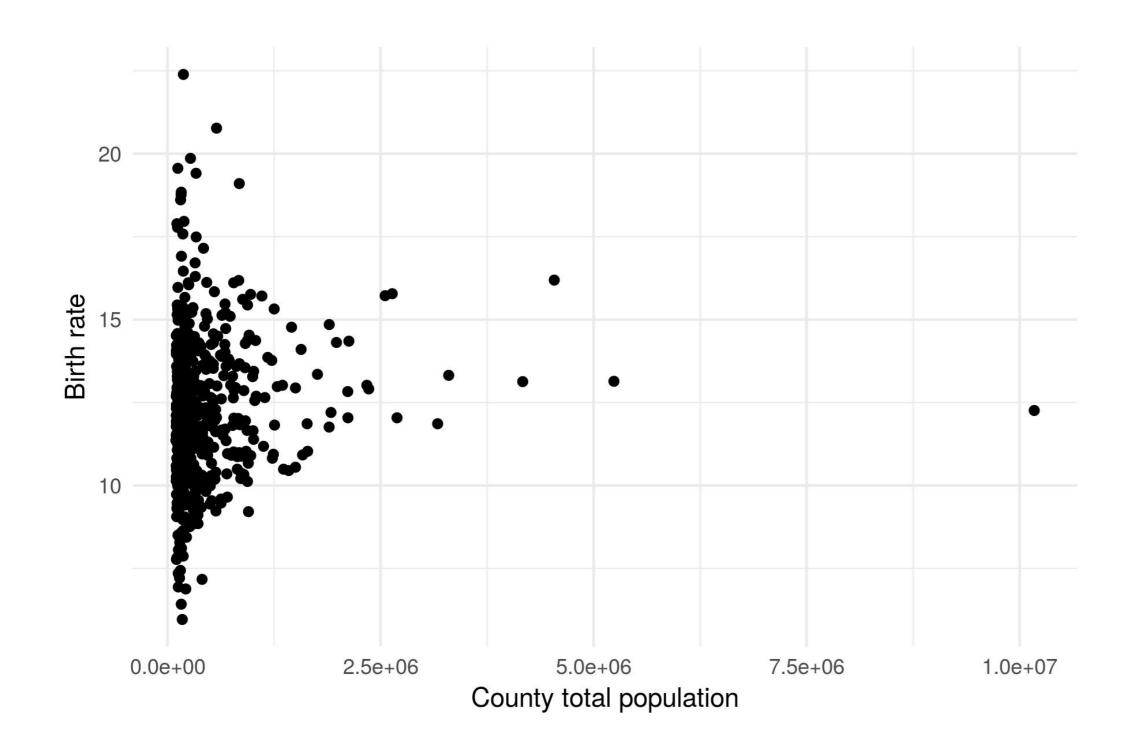
#### Birth rates data

- Small populations subject to stochasticity
- Random-effects one solution to this problem
- Birth rates one such variable



### How does a mothers age impact birth rate?

- Does a mother's age impact birth rate?
- Marketing and policy implications



### Imer syntax in R

```
library(lme4)
lmer(y ~ x + (Random-effect), data = my_data)
```

#### Random-effect syntax

- (1 | group): Random intercept with fixed mean
- (1 | g1/g2): Intercepts vary among g1 and g2 within g1
- (1 | g1) + (1 | g2) : Random intercepts for 2 variables
- x + (x | g): Correlated random slope and intercept
- x + (x | | g): Uncorrelated random slope and intercept
- See **Lme4** for additional details

# Let's practice!

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# Understanding and reporting the output of a lmer

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#### The model

#### **Print**

```
out # print(out) is what R is calling
```

```
Linear mixed model fit by REML ['lmerMod']
Formula: BirthRate ~ AverageAgeofMother + (AverageAgeofMother | State)
  Data: county_births_data
REML criterion at convergence: 2337.506
Random effects:
        Name
                Std.Dev. Corr
Groups
        (Intercept)
State
                    8.8744
         AverageAgeofMother 0.2912
                                 -0.99
Residual
                          1.6742
Number of obs: 578, groups: State, 50
Fixed Effects:
      (Intercept) AverageAgeofMother
          27.2204
                            -0.5235
```



### Summary

summary(out)

```
. . .
Scaled residuals:
    Min
             1Q Median
                                   Max
-2.8399 -0.5966 -0.1133 0.5228 5.1815
Random effects:
                            Variance Std.Dev. Corr
 Groups
         Name
         (Intercept)
                            78.75478 8.8744
State
         AverageAgeofMother 0.08482 0.2912 -0.99
Residual
                             2.80306 1.6742
Number of obs: 578, groups: State, 50
Fixed effects:
                  Estimate Std. Error t value
(Intercept)
                  27.22041
                              2.41279 11.282
AverageAgeofMother -0.52347
                              0.08302 -6.306
Correlation of Fixed Effects:
            (Intr)
AvrgAgfMthr -0.997
```



### Extracting fixed-effects estimates

fixef(out)

(Intercept) AverageAgeofMother 34.5756764 -0.7556129



#### Extracting fixed-effects confidence intervals

```
confint(out)
```

```
Computing profile confidence intervals ...

2.5 % 97.5 %

.sig01 0.9458700 1.612440

.sigma 1.6091447 1.815929

(Intercept) 24.0121843 31.146685

AverageAgeofMother -0.6605319 -0.411231
```



### **Extracting random-effects**

ranef(out)

```
$State
AK 1.03549148
AL -0.52500819
   0.48023356
AZ -1.04094123
   0.50530542
   0.09585582
CT -1.91638101
   0.96029531
DE -0.38938118
FL -1.87440508
GA 0.39776424
#...
```



## Reporting Imer output

- Know your audience
- Figures
- Table
- In-text



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# Statistical inference with Maryland crime data

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#### Maryland crime data

• Number of violent crimes per year by County

```
County Year Crime
ANNE ARUNDEL 2006 3167
BALTIMORE CITY 2006 10871
```

- Useful for policy/crime analysis or insurance
- Is the crime rate changing through time across counties?

## Null hypothesis test

- $H_0$ : No difference exists
- $H_a$ : A difference exists

#### P-values with Imer

```
library(lmerTest)
summary(lmer(...))
```



#### **ANOVA**

- Analysis of Variance (ANOVA)
- Compare variability of model with and without parameter
- lmer(response ~ (1 | group)) vs.
   lmer(response ~ predictor + (1 | group))

#### Summary

- Null hypothesis testing and ANOVAs
- Build and compare models
- High-level details, important assumptions covered in other DataCamp courses

# Let's practice!

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