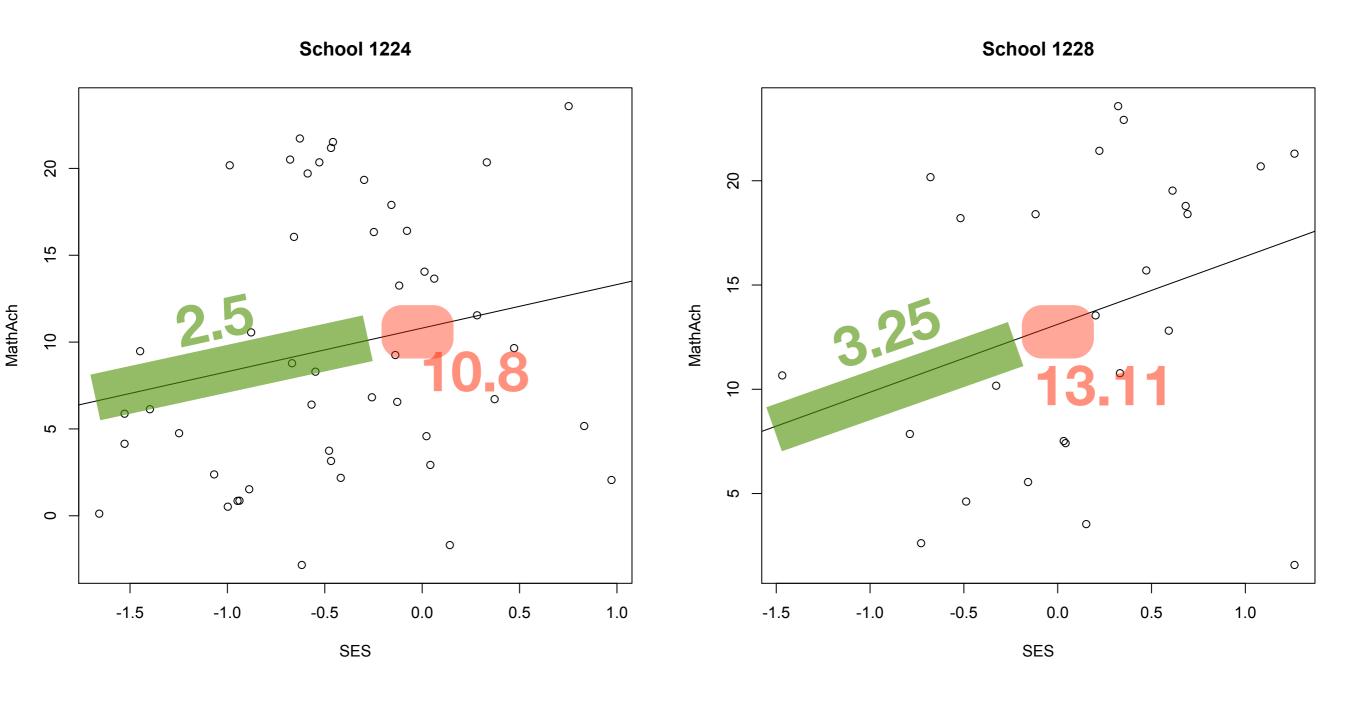
April 19th linear mixed effects models

```
> library(lme4)
Loading required package: Matrix
Loading required package: lattice
> library(nlme)
> MathAchieve[1:10,]
Grouped Data: MathAch ~ SES
                               School
                              SES MathAch MEANSES
   School Minority
                       Sex
                No Female -1.528
     1224
                                     5.876
                                            -0.428
1
     1224
                No Female -0.588
                                    19.708
                                            -0.428
2
     1224
                      Male -0.528
                                    20.349
                                            -0.428
3
                No
     1224
                      Male -0.668
                                    8.781
                                            -0.428
4
                No
5
     1224
                      Male -0.158
                                    17.898
                                            -0.428
                No
6
     1224
                                    4.583
                No
                      Male 0.022
                                            -0.428
     1224
7
                No Female -0.618
                                    -2.832
                                            -0.428
8
     1224
                      Male -0.998
                                     0.523
                                            -0.428
                No
9
     1224
                No Female -0.888
                                     1.527
                                            -0.428
                      Male -0.458
10
     1224
                                    21.521
                                            -0.428
                No
```

Complete pooling

```
> lm0 <- lm(MathAch ~ SES, data=MathAchieve)</pre>
> summary(lm0)
Call:
lm(formula = MathAch ~ SES, data = MathAchieve)
Residuals:
    Min 10 Median
                                3Q Max
-19.4382 \quad -4.7580 \quad 0.2334 \quad 5.0649 \quad 15.9007
Coefficients:
           Estimate Std. Error t value Pr(>|t|)
(Intercept) 12.74740 0.07569 168.42 <2e-16 ***
SES 3.18387 0.09712 32.78 <2e-16 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 6.416 on 7183 degrees of freedom
Multiple R-squared: 0.1301, Adjusted R-squared: 0.13
F-statistic: 1075 on 1 and 7183 DF, p-value: < 2.2e-16
```

No pooling



arate models $Y_{ij} = \beta_{oj}^{\text{level for school}} + \beta_{1j}(X_{ij} - \bar{X}_{.j}) + \epsilon_{ij}$ 160 separate models

Random effects

$$Y_{ij} = \beta_{oj} + \beta_{1j}(X_{ij} - \bar{X}_{.j}) + \epsilon_{ij}$$

dependent on something else

- > attach(MathAchSchool)
- > MathScores <- merge(MathAchieve, MathAchSchool, by="School")</pre>
- > head(MathScores)

	School	Minority	Sex	SES	MathAch	MEANSES.x	Size	Sector	PRACAD	DISCLIM	HIMINTY	MEANSES.y
1	1224	No	Female	-1.528	5.876	-0.428	842	Public	0.35	1.597	0	-0.428
2	1224	No	Female	-0.588	19.708	-0.428	842	Public	0.35	1.597	0	-0.428
3	1224	No	Male	-0.528	20.349	-0.428	842	Public	0.35	1.597	0	-0.428
4	1224	No	Male	-0.668	8.781	-0.428	842	Public	0.35	1.597	0	-0.428
5	1224	No	Male	-0.158	17.898	-0.428	842	Public	0.35	1.597	0	-0.428
6	1224	No	Male	0.022	4.583	-0.428	842	Public	0.35	1.597	0	-0.428

```
> attach(MathAchSchool)
> MathScores <- merge(MathAchieve, MathAchSchool, by="School")</pre>
> head(MathScores)
  School Minority
                              SES MathAch MEANSES.x Size Sector PRACAD DISCLIM HIMINTY MEANSES.y
                      Sex
                                                      842 Public
    1224
               No Female -1.528
                                    5.876
                                              -0.428
                                                                    0.35
                                                                            1.597
                                                                                              -0.428
1
                                                                                         0
                                                      842 Public
    1224
               No Female -0.588
                                   19.708
                                              -0.428
                                                                    0.35
                                                                            1.597
                                                                                              -0.428
2
                                                      842 Public
    1224
                     Male -0.528
                                   20.349
                                              -0.428
                                                                    0.35
                                                                            1.597
                                                                                              -0.428
3
                No
                     Male -0.668
                                    8.781
                                                      842 Public
    1224
                                              -0.428
                                                                    0.35
                                                                            1.597
                                                                                              -0.428
               No
4
    1224
                     Male -0.158
                                   17.898
                                              -0.428
                                                      842 Public
                                                                    0.35
                                                                            1.597
                                                                                              -0.428
5
               No
                                                      842 Public
    1224
               No
                     Male 0.022
                                    4.583
                                              -0.428
                                                                    0.35
                                                                            1.597
                                                                                              -0.428
```

vs Catholic

- > attach(MathAchSchool)
 > MathScores <- merge(MathAchieve, MathAchSchool, by="School")
 > head(MathScores)
- School Minority Sex SES MathAch MEANSES.x Size Sector PRACAD DISCLIM HIMINTY MEANSES.y 842 Public 1224 No Female -1.5285.876 -0.4280.35 1.597 -0.4281 0 19.708 842 Public No Female -0.588-0.428 1224 -0.4280.35 1.597 2 1224 Male -0.528 20.349 -0.428 842 Public -0.4280.35 1.597 No 842 Public 1224 Male -0.668 8.781 -0.4280.35 1.597 -0.428 4 No 1224 Male -0.158 17.898 -0.428 842 Public 0.35 1.597 -0.428 5 No 842 Public 1224 Male 0.022 4.583 -0.4280.35 1.597 -0.428No

Catholicism indicator, W_j

vs Catholic

- > attach(MathAchSchool)
- > MathScores <- merge(MathAchieve, MathAchSchool, by="School")</pre>
- > head(MathScores)

	School	-				MEANSES.x						-
1	1224	No	Female	-1.528	5.876	-0.428	842	Public	0.35	1.597	0	-0.428
2	1224	No	Female	-0.588	19.708	-0.428	842	Public	0.35	1.597	0	-0.428
3	1224	No	Male	-0.528	20.349	-0.428	842	Public	0.35	1.597	0	-0.428
4	1224	No	Male	-0.668	8.781	-0.428	842	Public	0.35	1.597	0	-0.428
5	1224	No	Male	-0.158	17.898	-0.428	842	Public	0.35	1.597	0	-0.428
6	1224	No	Male	0.022	4.583	-0.428	842	Public	0.35	1.597	0	-0.428

Catholicism indicator, W_j

vs Catholic

$$\beta_{oj} = \gamma_{00} + \gamma_{01} W_j + \epsilon_{oj}$$

$$\beta_{1j} = \gamma_{10} + \gamma_{11} W_j + \epsilon_{1j}$$

Tell Ime4 what to do

```
separate random effects for intercept \beta_{oj} and slope \beta_{1j}
```

```
> lme1.fm <- lmer(MathAch ~ SES + Sector + (1+ SES | School), MathScores)</pre>
```

```
> install.packages("arm")
> library(arm)
> display(lme1.fm)
lmer(formula = MathAch ~ SES + Sector + (1 + SES | School), data = MathScores)
             coef.est coef.se
(Intercept) 14.01
                       0.26
SES
              2.39
                       0.12
SectorPublic -2.54
                       0.34
Error terms:
 Groups
          Name
                      Std.Dev. Corr
 School
          (Intercept) 1.99 ———
                      0.66 - 0.55
          SES
 Residual
                      6.07
number of obs: 7185, groups: School, 160
AIC = 46615.9, DIC = 46593
deviance = 46597.4
```

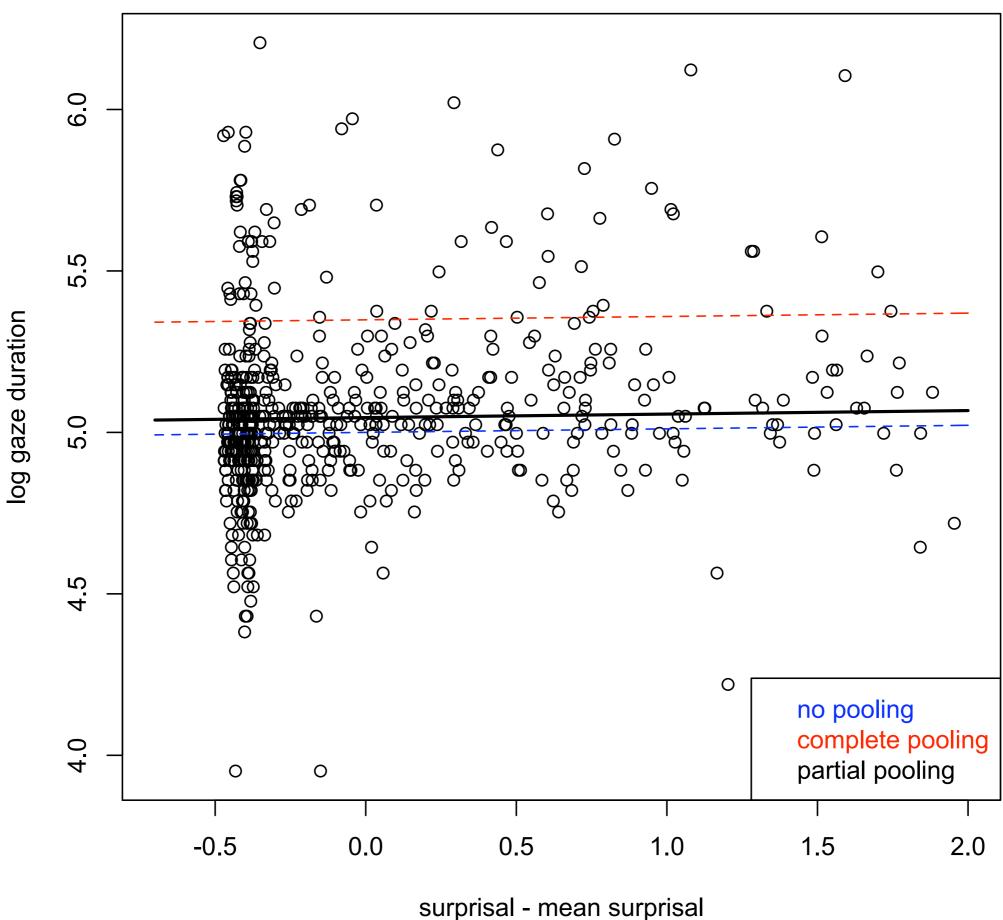
Kim, Kobele, Runner & Hale, to appear sometime in 2012

		Match	Mismatch
	No ellipsis	(1-a)	(2-a)
Voice	T-11::-	(1-b) (1-a)	(2-b) (2-a)
	Ellipsis	(1-b)	(2-b)
C +	No ellipsis	(3)	(4-a) (4-b)
Category	Ellipsis	(3)	(4-a) $(4-b)$

- (1) Voice match
 - a. Active-Active: Jill betrayed Abby, and Matt did betray Abby, too.
 - b. Passive-Passive: Abby was betrayed by Jill, and Matt was betrayed by Jill, too.
- (2) Voice mismatch
 - a. Active-Passive: Jill betrayed Abby, and Matt was betrayed by Jill, too.
 - b. Passive-Active: Abby was betrayed by Jill, and Matt did betray Abby, too.
- (3) Category match
 - a. VP-VP: Everyone criticized Roy, but Kate didn't criticize Roy
 - b. VP-VP: The report criticized Roy, but Kate didn't criticize Roy
- (4) Category mismatch
 - a. Noun-VP: The criticism of Roy was harsh, but Kate didn't criticize Roy
 - b. Adjective-VP: The report was critical of Roy, but Kate didn't criticize Roy

Three random effects

```
NV.glmm <- lmer(LogeNorm ~ Ellipsis + Canonical + NPVP + Ellipsis:NPVP + (1 | Subject) + (1 | Item) + (1 | Trial), data=NV, family="gaussian")
```



CoNLL '05 data

```
vnum, verb, sense, args, A0loc, A1loc, A0size, A1size, A2size, A3size, N0size, N1size
1, take, 01, (N0)(A0)(V)(A1)(AM)(N1), -1, 1, 2, 2, 0, 0, 17, 25
                                                                 amount of stuff
0, \text{say}, 01, (A0)(V)(A1), -1, 1, 4, 14, 0, 0, 0, 0
1, expect, 01, (N0)(A0)(V)(A1), -1, 1, 1, 12, 0, 0, 5, 0
0, sell, 01, (A0)(AM)(V)(A1)(AM), -2, 1, 4, 2, 0, 0, 0, 0
                                                                 before the
0, \text{say}, 01, (A0)(V)(A1), -1, 1, 9, 18, 0, 0, 0, 0
0, increase, 01, (A0)(V)(A1)(A4)(A3), -1, 1, 3, 2, 0, 5, 0, 0
                                                                 clause of interest
1, \text{make}, 01, (A0)(V)(A1), -1, 1, 7, 8, 0, 0, 0, 0
0, \text{make}, 01, (A0)(V)(A1)(AM), -1, 1, 1, 2, 0, 0, 0
3, report, 01, (N0)(A0)(V)(A1), -1, 1, 3, 7, 0, 0, 24, 0
2, report, 01, (N0)(A0)(V)(A1)(AM), -1, 1, 1, 3, 0, 0, 6, 0
4, get, 03, (N0) (A1) (V) (A2) (AM), 0, -1, 0, 1, 3, 0, 40, 0
0, pay, 01, (N0)(A0)(AM)(V)(A1)(A2)(N1), -2, 1, 2, 1, 8, 0, 1, 30
1, report, 01, (N0)(A0)(V)(A1)(AM), -1, 1, 7, 8, 0, 0, 5, 0
3, call, 11, (N0)(A0)(A0)(AM)(V)(A1)(N1), -2, 1, 2, 2, 0, 0, 15, 23
4, \text{say}, 01, (A1)(V)(A0), 1, -1, 18, 24, 0, 0, 0, 0
2, \text{say}, 01, (A1)(A0)(V), -1, -2, 3, 30, 0, 0, 0, 0
1, \text{say}, 01, (A1)(A0)(V), -1, -2, 2, 11, 0, 0, 0, 0
3, \text{say}, 01, (N0)(A0)(V)(A1), -1, 1, 1, 33, 0, 0, 14, 0
4, have, 03, (N0)(A1)(A0)(AM)(V)(A1), -2, 1, 4, 29, 0, 0, 12, 0
1, \text{say}, 01, (A1)(A0)(V), -1, -2, 3, 19, 0, 0, 0, 0
2, \text{say}, 01, (A1)(A0)(V), -1, -2, 2, 15, 0, 0, 0, 0
```

Fit & compare models

```
vbarg <- read.csv("vbarg.txt")</pre>
vbarg$a0 <- log(vbarg$A0size+1)</pre>
vbarg$n0 <- log(vbarg$N0size+1)</pre>
a0n0.lme \leftarrow lmer(a0 \sim n0 + (1|verb),data=vbarg)
a0n0.lme2 <- lmer(a0 \sim n0 + (1+n0|verb),data=vbarg)
anova(a0n0.lme,a0n0.lme2)
Data: vbarq
Models:
a0n0.lme: a0 \sim n0 + (1 | verb)
a0n0.lme2: a0 \sim n0 + (1 + n0 | verb)
          Df AIC BIC logLik Chisq Chi Df Pr(>Chisq)
a0n0.lme 4 58398 58431 -29195
a0n0.lme2 6 58083 58133 -29036 318.37 2 < 2.2e-16 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Next class....



Categorical Data Analysis: Away from ANOVAs (transformation or not) and towards Logit Mixed Models. Journal of Memory and Language, 59 (4): 434-446.



§7.4