Homework 2

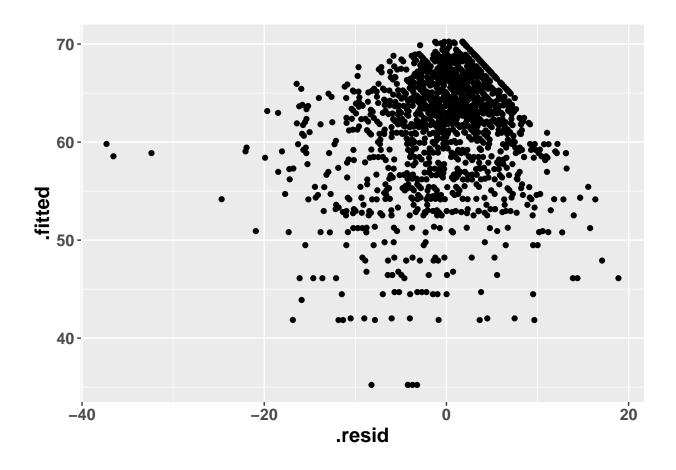
Brent Rappaport
September 12, 2017

Question 1

```
model1 <- lm(PPeerScale ~ age, data=data_long)</pre>
summary(model1)
##
## Call:
## lm(formula = PPeerScale ~ age, data = data_long)
##
## Residuals:
##
       Min
                1Q Median
                                 3Q
                                        Max
## -39.708 -4.254
                     2.502
                             6.677
                                    10.641
##
## Coefficients:
               Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) 61.01512
                           0.63836 95.581
                                              <2e-16 ***
               0.11097
                           0.07297
                                      1.521
                                               0.129
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 8.904 on 1680 degrees of freedom
     (766 observations deleted due to missingness)
## Multiple R-squared: 0.001375,
                                     Adjusted R-squared: 0.0007804
## F-statistic: 2.313 on 1 and 1680 DF, p-value: 0.1285
Average intercept is 61.02 and average slope is 0.11.
```

```
model2 <- lmer(PPeerScale ~ 1 + (1 | ID), data=data_long)</pre>
summary(model2)
## Linear mixed model fit by REML ['lmerMod']
## Formula: PPeerScale ~ 1 + (1 | ID)
##
      Data: data_long
##
## REML criterion at convergence: 11498.2
##
## Scaled residuals:
              1Q Median
                                3Q
                                       Max
## -6.0271 -0.4336 0.1407 0.5496 3.0506
##
## Random effects:
## Groups
                         Variance Std.Dev.
            Name
  ID
             (Intercept) 42.93
                                  6.552
##
```

```
38.33
## Residual
                                  6.191
## Number of obs: 1684, groups: ID, 302
##
## Fixed effects:
               Estimate Std. Error t value
## (Intercept) 61.7003
                          0.4114
# Calculate ICC
42.93/(42.93+38.33)
## [1] 0.5283042
model2.aug <- augment(model2, data_long)</pre>
## Warning: Deprecated: please use `purrr::possibly()` instead
## Warning in indices[which(stats::complete.cases(original))] =
## seq_len(nrow(x)): number of items to replace is not a multiple of
## replacement length
ggplot(model2.aug,aes(.resid,.fitted,group=ID)) +
  geom_point(alpha=1) +
  theme(text=element_text(lineheight=1, face="bold", size=15),
       legend.position="none")
```



```
model3 <- lmer(PPeerScale ~ age + (1 | ID), data=data_long)</pre>
summary(model3)
## Linear mixed model fit by REML ['lmerMod']
## Formula: PPeerScale ~ age + (1 | ID)
##
      Data: data_long
##
## REML criterion at convergence: 11476.2
##
## Scaled residuals:
       Min
            1Q Median
                                ЗQ
                                       Max
## -6.0643 -0.4159 0.1398 0.5449 3.1692
##
## Random effects:
   Groups
           Name
                         Variance Std.Dev.
             (Intercept) 43.02
                                  6.559
##
## Residual
                         38.00
                                  6.164
## Number of obs: 1682, groups: ID, 302
## Fixed effects:
              Estimate Std. Error t value
## (Intercept) 60.51297
                          0.59831 101.14
## age
               0.14905
                           0.05429
                                      2.75
```

```
##
## Correlation of Fixed Effects:
## (Intr)
## age -0.726
```

The esimates of the fixed effect for the Intercept doesn't change much, however now that age is being included as a fixed effect, we can get a sense of the average change over time in parent reported Peer scores. That is, for ever 1 year, participants increase an average of 0.15 in their score.

```
print(c(sigma(model2), sigma(model3)))
```

```
## [1] 6.190725 6.164179
```

The residual standard error (sigma here), doesn't change much between the two models, but does decrease when age is introduced as a fixed effect.

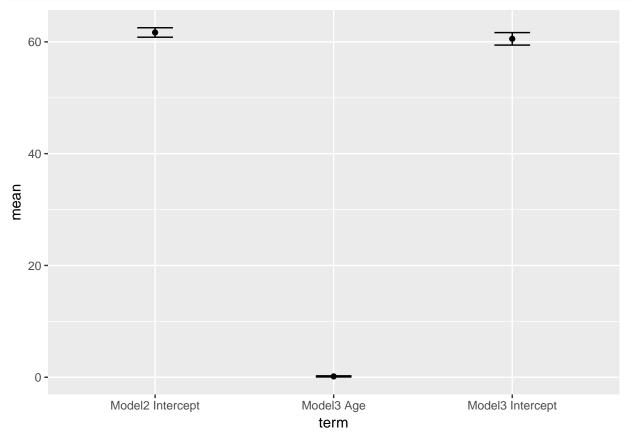
```
plot3_table <- FEsim(model2)
plot3_table[2:3,] <- FEsim(model3)</pre>
```

```
## Warning in `[<-.factor`(`*tmp*`, iseq, value = structure(1:2, .Label =
## c("(Intercept)", : invalid factor level, NA generated

plot3_table[1] <- c("Model2 Intercept", "Model3 Intercept", "Model3 Age")

plot3_table$ci <- 1.96*plot3_table$sd

ggplot(plot3_table,aes(x=term,y=mean)) +
    geom_point() +
    geom_errorbar(aes(ymin=mean-ci,ymax=mean+ci),width=0.2)</pre>
```



Question 4

```
model4 <- lmer(PPeerScale ~ age + (age | ID), data=data_long)</pre>
print(c(sigma(model3), sigma(model4)))
## [1] 6.164179 5.511522
anova(model3,model4)
## refitting model(s) with ML (instead of REML)
## Data: data_long
## Models:
## model3: PPeerScale ~ age + (1 | ID)
## model4: PPeerScale ~ age + (age | ID)
                   BIC logLik deviance Chisq Chi Df Pr(>Chisq)
         Df
              AIC
## model3 4 11480 11502 -5736.1
                                   11472
## model4 6 11386 11418 -5686.8
                                   11374 98.722
                                                     2 < 2.2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

Adding the random slope improves fit, according to the AIC, BIC, and likelihood ratio test, while also considerably lowering the residual standard error (from 6.1641795 to 5.5115224).

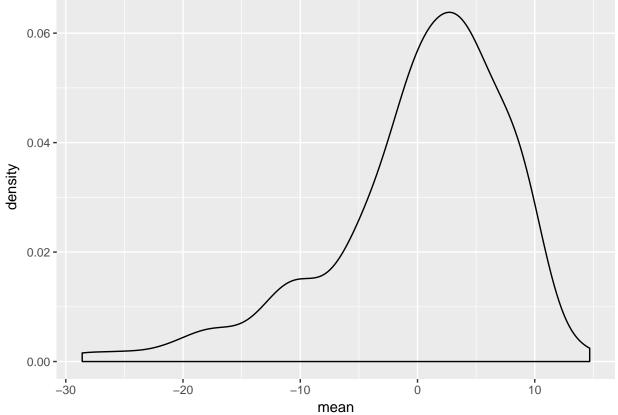
```
summary(model4)
```

```
## Linear mixed model fit by REML ['lmerMod']
## Formula: PPeerScale ~ age + (age | ID)
##
     Data: data_long
## REML criterion at convergence: 11376.8
##
## Scaled residuals:
      Min
               1Q Median
                                3Q
                                       Max
## -6.8222 -0.3835 0.1106 0.4980
                                    2.7892
##
## Random effects:
                         Variance Std.Dev. Corr
   Groups
            Name
##
             (Intercept) 88.0332 9.3826
##
                          0.8576 0.9261
                                           -0.72
                         30.3769 5.5115
## Number of obs: 1682, groups: ID, 302
## Fixed effects:
               Estimate Std. Error t value
## (Intercept) 60.58395
                           0.69693
                                     86.93
## age
                0.13846
                           0.07578
                                      1.83
##
## Correlation of Fixed Effects:
##
       (Intr)
## age -0.812
```

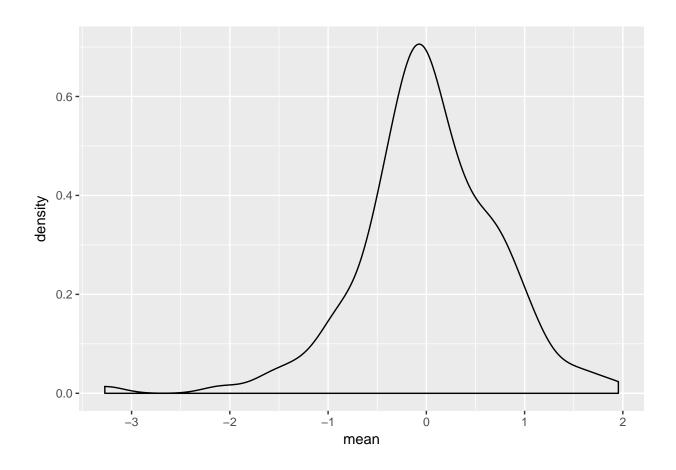
The correlation means that there is a negative correlation between the intercept and the slope, such that the lower the intercept, the higher the slope will be.

```
model4_re.sim <- REsim(model4)
p1.gg <- model4_re.sim %>%
    filter(term == c("(Intercept)", "age"))

# Intercept
ggplot(p1.gg[p1.gg$term %in% "(Intercept)",], aes(mean)) +
geom_density()
```



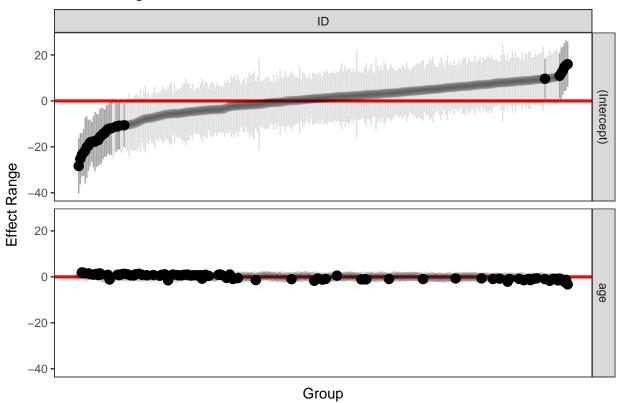
```
# Age
ggplot(p1.gg[p1.gg$term %in% "age",], aes(mean)) +
geom_density()
```



Question 7

plotREsim(model4_re.sim)

Effect Ranges



No, there are no obvious outliers in terms of unusually large standard errors around the intercept or slopes.

```
PT <- predictInterval(merMod=model4, newdata=data_long, level=0.9, n.sims=100, stat="mean",
                      include.resid.var=TRUE)
## Warning: executing %dopar% sequentially: no parallel backend registered
  Warning:
                 The following levels of ID from newdata
   -- 86 -- are not in the model data.
##
        Currently, predictions for these values are based only on the
   fixed coefficients and the observation-level error.
##
##
                 The following levels of ID from newdata
  Warning:
##
   -- 163, 165 -- are not in the model data.
        Currently, predictions for these values are based only on the
##
##
   fixed coefficients and the observation-level error.
                 The following levels of ID from newdata
## Warning:
##
   -- 196 -- are not in the model data.
        Currently, predictions for these values are based only on the
##
   fixed coefficients and the observation-level error.
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
```

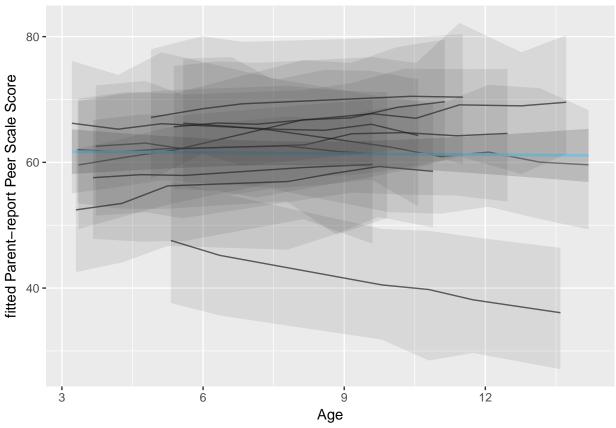
```
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
   Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
   Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
```

```
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
   Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
   Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
```

```
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
   Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
   Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
```

```
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
## Warning in rnorm(N, yhat[, x], sigmahat[x]): NAs produced
PT_plot <- data_long; PT_plot[,15:17] <- PT
#example graph using ONLY FIRST 10 participants and fitted scores with confidence bands
ggplot(PT_plot[1:80,],aes(age,fit,group=ID)) +
  geom_line(alpha=0.6) +
  stat_smooth(aes(group=1),method="lm", color="sky blue") +
  geom_ribbon(aes(ymin=lwr,ymax=upr),linetype=2,alpha=0.1) +
  labs(x="Age",y="fitted Parent-report Peer Scale Score")
```

- ## Warning: Removed 13 rows containing non-finite values (stat_smooth).
- ## Warning: Removed 13 rows containing missing values (geom_path).



```
#graphed using ALL participants and fitted scores with confidence bands
ggplot(PT_plot,aes(age,fit,group=ID)) +
  geom_line(alpha=0.6) +
  stat_smooth(aes(group=1),method="lm", color="sky blue") +
  geom_ribbon(aes(ymin=lwr,ymax=upr),linetype=2,alpha=0.1) +
  labs(x="Age",y="fitted Parent-report Peer Scale Score")
```

- ## Warning: Removed 685 rows containing non-finite values (stat_smooth).
- $\mbox{\tt \#\#}$ Warning: Removed 685 rows containing missing values (geom_path).

