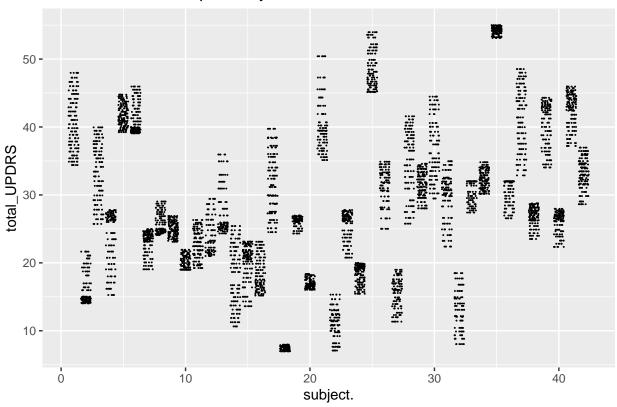
project_option2

```
categorical_col = c(1,2,3)
full_data = read.csv("D:/R/data_467/data467_project/parkinsons_updrs.data")
data = full_data[-categorical_col]
for (i in 1:length(data)){
 data[,i] = scale(data[,i], center = TRUE, scale = TRUE)
subject_data = full_data
head(data)
     test_time motor_UPDRS total_UPDRS
##
                                        Jitter... Jitter.Abs. Jitter.RAP
## 1 -1.6319513
                0.8491244
                            0.5027024 0.08289818 -0.2842180 0.3274246
## 2 -1.5005486 0.8796314
                            0.5490563 -0.56074568 -0.7566586 -0.5337008
## 3 -1.3692936 0.9101384 0.5953167 -0.23892375 -0.5393359 -0.3000125
## 4 -1.2576661
                 ## 5 -1.1080747
                 0.9706604
                            0.6874638 -0.49851492 -0.6638379 -0.6585480
## 6 -0.9769133
                 1.0011674
                            0.7337243 -0.46651053 -0.5871358 -0.5753165
##
    Jitter.PPQ5 Jitter.DDP
                             Shimmer Shimmer.dB. Shimmer.APQ3 Shimmer.APQ5
## 1 -0.0286346 0.3284775 -0.3245661
                                     -0.3516122 -0.2096910
                                                               -0.4233205
## 2 -0.4761714 -0.5347790 -0.5339707
                                      -0.5731071
                                                  -0.5451114
                                                               -0.5655438
     -0.3207395 -0.2989574 -0.6690579
                                     -0.5644210
                                                  -0.7415288
                                                               -0.7023662
## 4 -0.1706673 -0.3448413 -0.4236559
                                      0.0696623
                                                  -0.4605008
                                                               -0.4497248
    -0.5297687 -0.6596258 -0.6582200
                                      -0.5861362
                                                  -0.7830786
                                                               -0.6513579
## 6 -0.4520527 -0.5753275 -0.4553956 -0.4211008
                                                               -0.4065177
                                                  -0.5360460
    Shimmer.APQ11 Shimmer.DDA
                                     NHR
                                                  HNR
                                                            RPDE
## 1
       -0.5434195 -0.2096865 -0.29869541 -0.009203976 -1.21396224 -1.478374
## 2
       -0.5299101 -0.5451066 -0.35193510 1.282540523 -1.05502898 -1.247774
## 3
       -0.6454902 -0.7415238 -0.19935262 0.318684263 -0.78479294 -1.540008
## 4
       -0.3928152 -0.4607479 -0.07174823 0.644475136 -0.53644116 -1.062024
## 5
       -0.4648651 -0.7833254 -0.34334103 1.036216523 -0.68913591 -1.297843
## 6
       -0.2427112 -0.5357894 -0.37997892 0.295147154 -0.01963576 -1.139737
##
## 1 -0.6506028
## 2 -1.2184810
## 3 -0.1032714
## 4 1.2369692
## 5 -0.2839301
## 6 -0.2687386
#TOTAL UPDRS
ggplot(full_data, aes(subject., total_UPDRS)) +
 geom_jitter(size = 0.01)+
 ggtitle('Total UPDRS Score per Subject')
```

Total UPDRS Score per Subject



```
#MOTOR UPDRS
ggplot(full_data, aes(subject., motor_UPDRS)) +
  geom_jitter(size = 0.01)+
  ggtitle('Motor UPDRS Score per Subject')
```

Motor UPDRS Score per Subject

Jitter.RAP

Jitter.DDP

Shimmer

Jitter.PPQ5

-13.80989

-0.02475

13.88979

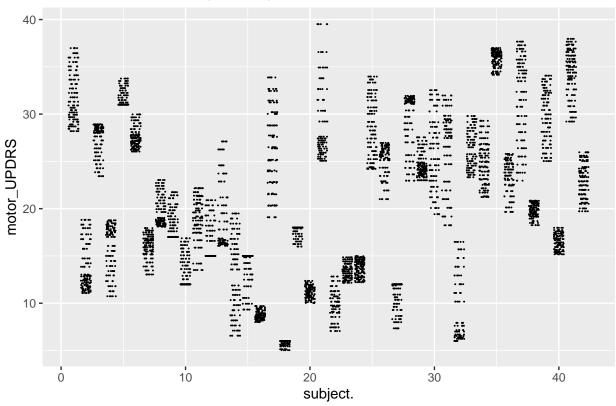
0.28441

14.19452

14.19602

0.16238

0.06819



```
set.seed(7)
tot_df = data[-2] #Removing motor_UPDRS
#FULL MODEL
full_model = lm(total_UPDRS ~.-1, data = tot_df)
summary(full_model)
##
## Call:
  lm(formula = total_UPDRS ~ . - 1, data = tot_df)
## Residuals:
##
       Min
                1Q Median
                                 3Q
                                        Max
  -2.6178 -0.6973 -0.1820 0.7077
                                     2.8463
##
## Coefficients:
                  Estimate Std. Error t value Pr(>|t|)
##
## test_time
                   0.08829
                              0.01240
                                         7.121 1.20e-12 ***
## Jitter...
                   0.06863
                               0.11639
                                         0.590 0.555426
## Jitter.Abs.
                  -0.10601
                              0.03220
                                        -3.293 0.000999 ***
```

-0.973 0.330642

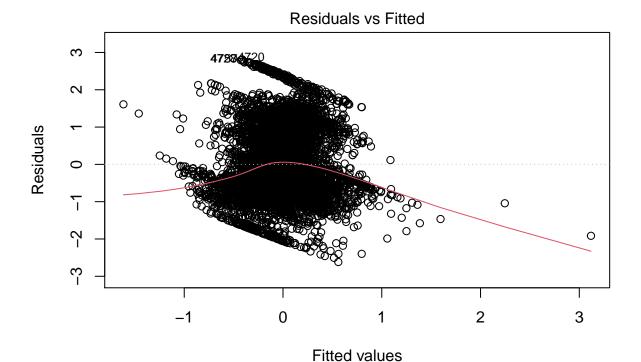
-0.363 0.716636

0.978 0.327903

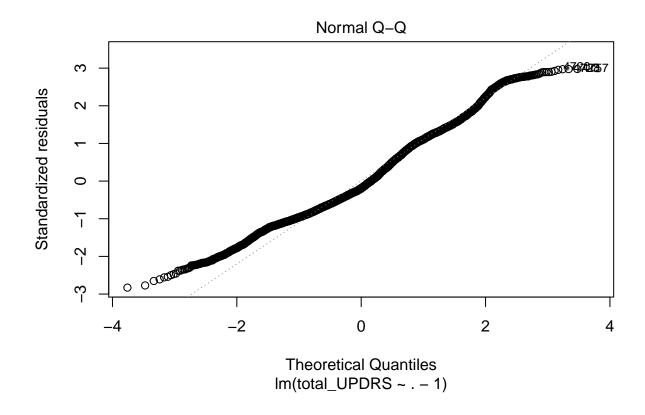
1.751 0.079917 .

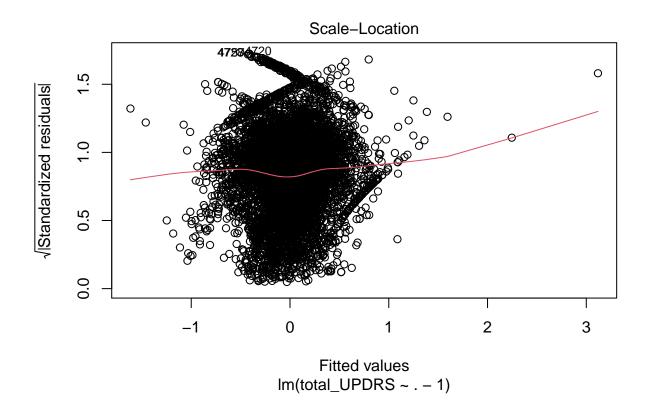
```
-0.10831
                              0.10810 -1.002 0.316407
## Shimmer.dB.
## Shimmer.APQ3
                -25.52941
                             60.42833 -0.422 0.672694
## Shimmer.APQ5
                  -0.20346
                              0.08931
                                       -2.278 0.022755 *
## Shimmer.APQ11
                  0.20057
                              0.04799
                                        4.179 2.97e-05 ***
## Shimmer.DDA
                  25.32614
                             60.42825
                                       0.419 0.675150
## NHR
                  -0.23069
                              0.03502
                                       -6.587 4.87e-11 ***
## HNR
                  -0.24593
                              0.02859
                                       -8.602 < 2e-16 ***
## RPDE
                                        3.082 0.002068 **
                   0.05478
                              0.01778
## DFA
                  -0.27176
                              0.01561 -17.414 < 2e-16 ***
## PPE
                              0.02588
                                        6.937 4.43e-12 ***
                   0.17956
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 0.9456 on 5858 degrees of freedom
## Multiple R-squared: 0.1083, Adjusted R-squared: 0.1057
## F-statistic: 41.85 on 17 and 5858 DF, p-value: < 2.2e-16
```

plot(full_model)

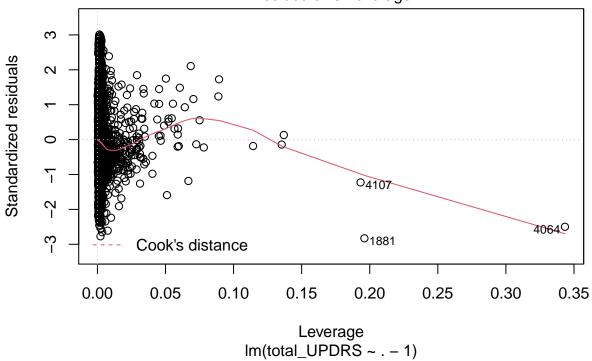


Im(total_UPDRS ~ . - 1)

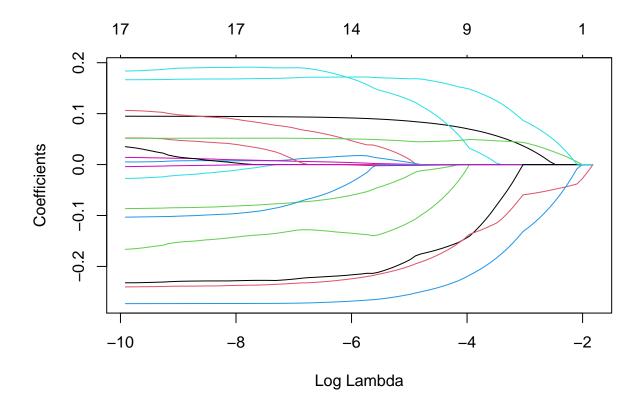




Residuals vs Leverage

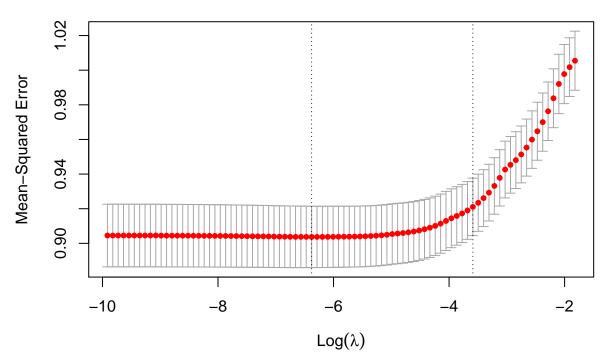


```
#CROSS VALIDATION
tot_split = initial_split(tot_df , prop = 0.7, strata = 'total_UPDRS')
tot_train = training(tot_split); tot_test = testing(tot_split)
tot_train_y = tot_train$total_UPDRS
tot_train_x = model.matrix(lm(total_UPDRS ~ .-1, data = tot_train))
# #RIDGE
\# tot_ridge = glmnet(x = tot_train_x, y = tot_train_y, alpha = 0)
# plot(tot_ridge, xvar = 'lambda')
# tot_ridge_cv = cv.glmnet(x = tot_train_x, y = tot_train_y, alpha = 0)
# plot(tot ridge cv)
# ridge_coef = coef(tot_ridge, tot_ridge_cv$lambda.1se)
# ridge_coef_df = as.data.frame(as.matrix(ridge_coef))
# colnames(ridge_coef_df) = c('slopes')
# ridge_coef_df$labels = rownames(ridge_coef_df)
\# ggplot(data = ridge\_coef\_df, mapping = aes(x = slopes, y = labels)) + geom\_point()+ggtitle('Ridge')
# ridge_coef_df
#LASSO
tot_lasso = glmnet(x = tot_train_x, y = tot_train_y, alpha = 1)
plot(tot_lasso, xvar = 'lambda')
```



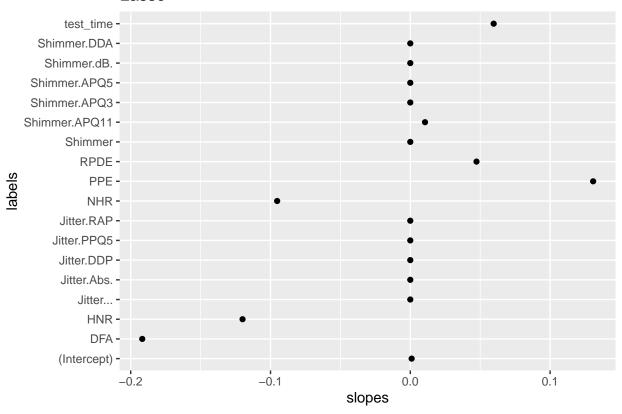
```
tot_lasso_cv = cv.glmnet(x = tot_train_x, y = tot_train_y, alpha = 1)
plot(tot_lasso_cv)
```





```
lasso_coef = coef(tot_lasso, tot_lasso_cv$lambda.1se)
lasso_coef_df = as.data.frame(as.matrix(lasso_coef))
colnames(lasso_coef_df) = c('slopes')
lasso_coef_df$labels = rownames(lasso_coef_df)
ggplot(data = lasso_coef_df, mapping = aes(x = slopes, y = labels)) + geom_point() +ggtitle('Lasso')
```

Lasso



lasso_coef_df

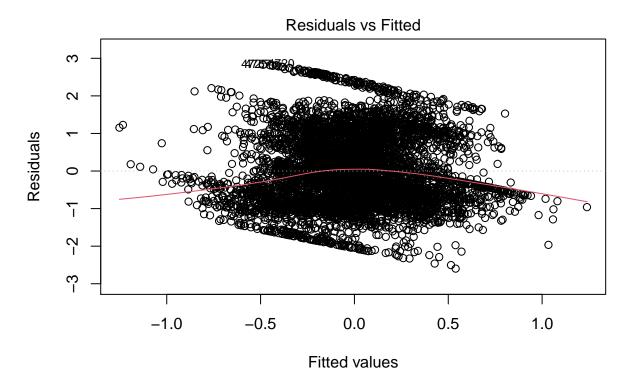
```
slopes
                                       labels
##
## (Intercept)
                  0.0009700013
                                  (Intercept)
## test_time
                  0.0596532999
                                    test_time
## Jitter...
                  0.000000000
                                    Jitter...
## Jitter.Abs.
                  0.000000000
                                  Jitter.Abs.
## Jitter.RAP
                  0.000000000
                                   Jitter.RAP
## Jitter.PPQ5
                  0.000000000
                                  Jitter.PPQ5
## Jitter.DDP
                  0.000000000
                                   Jitter.DDP
## Shimmer
                  0.000000000
                                      Shimmer
## Shimmer.dB.
                  0.000000000
                                  Shimmer.dB.
## Shimmer.APQ3
                  0.000000000
                                 Shimmer.APQ3
## Shimmer.APQ5
                                 Shimmer.APQ5
                  0.000000000
## Shimmer.APQ11
                  0.0105209293 Shimmer.APQ11
## Shimmer.DDA
                                  Shimmer.DDA
                  0.000000000
## NHR
                 -0.0952388334
                                          NHR
## HNR
                 -0.1200046925
                                          HNR
## RPDE
                  0.0473549429
                                         RPDE
## DFA
                 -0.1917464936
                                          DFA
## PPE
                  0.1308031817
                                          PPE
```

red_model = lm(total_UPDRS~ test_time+ Shimmer.APQ11 + RPDE + PPE + NHR + HNR + DFA -1, data = tot_df)
summary(red_model)

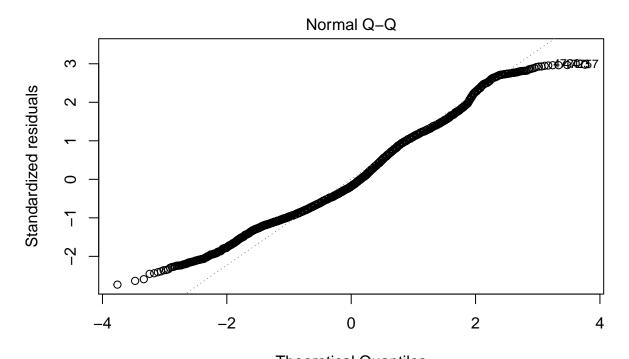
##

```
## Call:
## lm(formula = total_UPDRS ~ test_time + Shimmer.APQ11 + RPDE +
      PPE + NHR + HNR + DFA - 1, data = tot_df)
##
## Residuals:
##
      Min
                1Q Median
                                3Q
                                       Max
  -2.5971 -0.7127 -0.1682 0.7171
##
## Coefficients:
##
                 Estimate Std. Error t value Pr(>|t|)
## test_time
                 0.08513
                             0.01243
                                       6.848 8.24e-12 ***
                             0.02152
                                       2.454 0.014156 *
## Shimmer.APQ11 0.05281
## RPDE
                             0.01676
                                       3.382 0.000725 ***
                  0.05668
## PPE
                  0.19601
                             0.02053
                                       9.547
                                             < 2e-16 ***
## NHR
                 -0.23690
                             0.01981 -11.959 < 2e-16 ***
## HNR
                 -0.17864
                             0.02678 -6.670 2.79e-11 ***
## DFA
                 -0.26997
                             0.01451 -18.601 < 2e-16 ***
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.9505 on 5868 degrees of freedom
## Multiple R-squared: 0.09745,
                                    Adjusted R-squared: 0.09638
## F-statistic: 90.51 on 7 and 5868 DF, p-value: < 2.2e-16
```

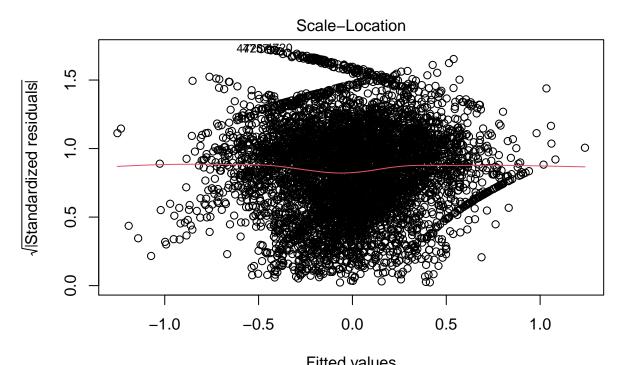
plot(red_model)



Im(total_UPDRS ~ test_time + Shimmer.APQ11 + RPDE + PPE + NHR + HNR + DFA

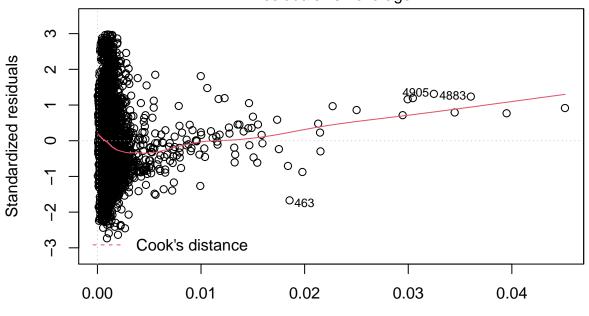


Theoretical Quantiles
Im(total_UPDRS ~ test_time + Shimmer.APQ11 + RPDE + PPE + NHR + HNR + DFA



Fitted values
Im(total_UPDRS ~ test_time + Shimmer.APQ11 + RPDE + PPE + NHR + HNR + DFA

Residuals vs Leverage



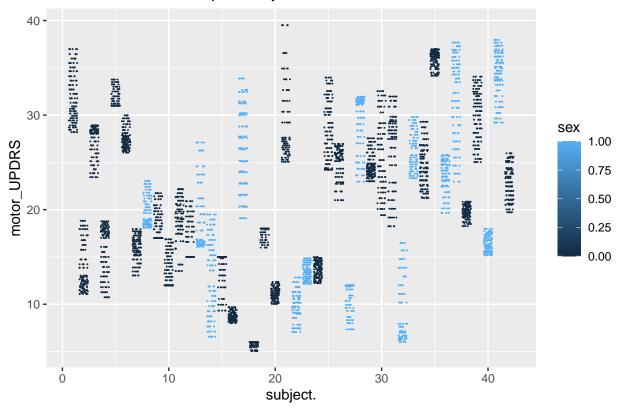
Leverage Im(total_UPDRS ~ test_time + Shimmer.APQ11 + RPDE + PPE + NHR + HNR + DFA ·

```
anova1 = anova(red_model, full_model)
anova1
## Analysis of Variance Table
##
## Model 1: total_UPDRS ~ test_time + Shimmer.APQ11 + RPDE + PPE + NHR +
##
      HNR + DFA - 1
## Model 2: total_UPDRS ~ (test_time + Jitter... + Jitter.Abs. + Jitter.RAP +
       Jitter.PPQ5 + Jitter.DDP + Shimmer + Shimmer.dB. + Shimmer.APQ3 +
##
      Shimmer.APQ5 + Shimmer.APQ11 + Shimmer.DDA + NHR + HNR +
##
##
      RPDE + DFA + PPE) - 1
##
    Res.Df
               RSS Df Sum of Sq
                                          Pr(>F)
## 1
      5868 5301.6
      5858 5237.9 10
                         63.642 7.1176 3.113e-11 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
qf(0.95, anova1$Df[2], anova1$Res.Df[1])
## [1] 1.832312
dt(2.454,5686)
```

[1] 0.01966379

```
grouped_subject = subject_data %>%
  group_by(subject.)
head(grouped_subject)
## # A tibble: 6 x 22
## # Groups:
               subject. [1]
     subject.
                      sex test_time motor_UPDRS total_UPDRS Jitter... Jitter.Abs.
                age
        <int> <int> <int>
                              <dbl>
                                          <dbl>
                                                      <dbl>
                                                                 <dbl>
## 1
                72
                               5.64
                                           28.2
                                                       34.4
                                                               0.00662
                                                                         0.0000338
           1
                        0
## 2
           1
                              12.7
                                           28.4
                                                        34.9
                                                              0.003
                                                                         0.0000168
                72
                        0
## 3
           1
                72
                        0
                              19.7
                                           28.7
                                                        35.4
                                                              0.00481
                                                                         0.0000246
## 4
           1
                72
                        0
                              25.6
                                           28.9
                                                        35.8
                                                              0.00528
                                                                         0.0000266
## 5
                 72
                                           29.2
                                                       36.4
                                                               0.00335
            1
                        0
                              33.6
                                                                         0.0000201
## 6
            1
                 72
                        0
                              40.7
                                           29.4
                                                        36.9
                                                               0.00353
                                                                         0.0000229
## # ... with 14 more variables: Jitter.RAP <dbl>, Jitter.PPQ5 <dbl>,
       Jitter.DDP <dbl>, Shimmer <dbl>, Shimmer.dB. <dbl>, Shimmer.APQ3 <dbl>,
       Shimmer.APQ5 <dbl>, Shimmer.APQ11 <dbl>, Shimmer.DDA <dbl>, NHR <dbl>,
## #
       HNR <dbl>, RPDE <dbl>, DFA <dbl>, PPE <dbl>
gs_lmer = lmer(formula = total_UPDRS ~ sex + (1 | subject.), data = grouped_subject)
summary(gs_lmer)
## Linear mixed model fit by REML ['lmerMod']
## Formula: total_UPDRS ~ sex + (1 | subject.)
##
     Data: grouped_subject
##
## REML criterion at convergence: 28948.8
##
## Scaled residuals:
                1Q Median
                                3Q
## -3.2925 -0.5928 0.0396 0.5703 3.7142
##
## Random effects:
## Groups
            Name
                         Variance Std.Dev.
## subject. (Intercept) 110.064 10.491
                                   2.769
## Residual
                           7.666
## Number of obs: 5875, groups: subject., 42
## Fixed effects:
               Estimate Std. Error t value
                             1.983 14.84
## (Intercept)
                29.431
## sex
                 -2.679
                             3.435
                                     -0.78
##
## Correlation of Fixed Effects:
       (Intr)
## sex -0.577
ggplot(grouped_subject, aes(subject., motor_UPDRS, color = sex)) +
  geom_jitter(size = 0.01)+
 ggtitle('Motor UPDRS Score per Subject')
```

Motor UPDRS Score per Subject



```
gs_lm = lm(total_UPDRS ~ sex, data = grouped_subject)
gs_me = lmer(total_UPDRS ~ (1 | subject.), data = grouped_subject)
anova(gs_lmer, gs_me)
## refitting model(s) with ML (instead of REML)
## Data: grouped_subject
## Models:
## gs_me: total_UPDRS ~ (1 | subject.)
## gs_lmer: total_UPDRS ~ sex + (1 | subject.)
                       BIC logLik deviance Chisq Df Pr(>Chisq)
                  AIC
##
           npar
              3 28963 28983 -14478
                                      28957
## gs_me
              4 28964 28991 -14478
                                      28956 0.634
                                                          0.4259
## gs_lmer
anova(gs_lmer, gs_lm)
```

```
## refitting model(s) with ML (instead of REML)

## Data: grouped_subject

## Models:

## gs_lm: total_UPDRS ~ sex

## gs_lmer: total_UPDRS ~ sex + (1 | subject.)

## npar AIC BIC logLik deviance Chisq Df Pr(>Chisq)

## gs_lm 3 44473 44493 -22234 44467
```

```
## gs_lmer 4 28964 28991 -14478 28956 15511 1 < 2.2e-16 ***
```

Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1