Final Project STATS-101A

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Change Data

```
movies <- read.csv("movies.csv")
attach(movies)
movies <- movies[!is.na(score) & !is.na(budget) & !is.na(gross) & !is.na(runtime), ]

## Transform budget and gross to millions of dollars
movies$budget <- movies$budget / 1e9
movies$gross <- movies$gross / 1e9</pre>
```

Summaries

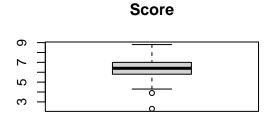
```
attach(movies)

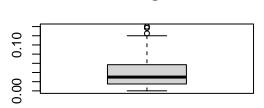
## The following objects are masked from movies (pos = 3):

##

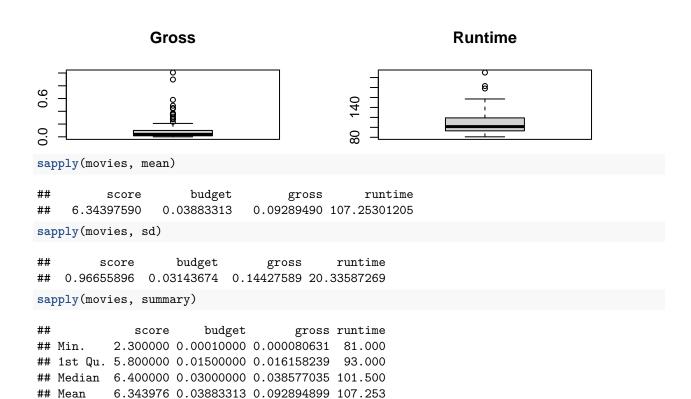
## budget, gross, runtime, score

par(mfrow = c(2, 2))
boxplot(score, main = "Score")
boxplot(budget, main = "Budget")
boxplot(gross, main = "Gross")
boxplot(runtime, main = "Runtime")
```





Budget



First Model

3rd Qu. 6.975000 0.05600000 0.099609142 119.000

8.800000 0.14000000 1.006968171 210.000

pairs(movies)

```
0.00 0.04 0.08 0.12
                                                                120
                                                                    160 200
        score
                                                                               2
                                                       0
                          budget
0.00
                                              gross
                                                                               0.4
200
                                                       0
140
                                                               runtime
     3 4 5 6 7 8 9
                                               0.4
                                                     8.0
                                         0.0
movies_model <- lm(score ~ budget + gross + runtime)</pre>
summary(movies_model)
##
## Call:
## lm(formula = score ~ budget + gross + runtime)
##
## Residuals:
##
      Min
                1Q Median
## -3.9243 -0.4411 0.0526 0.5306 1.6546
## Coefficients:
                Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) 4.134388
                           0.344458 12.003 < 2e-16 ***
## budget
               -9.920431
                           2.528044 -3.924 0.000128 ***
## gross
                                    4.441 1.65e-05 ***
                2.451998
                           0.552143
## runtime
                0.022070
                           0.003335
                                    6.617 5.11e-10 ***
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.8001 on 162 degrees of freedom
## Multiple R-squared: 0.3272, Adjusted R-squared: 0.3147
## F-statistic: 26.26 on 3 and 162 DF, p-value: 6.771e-14
anova(movies_model)
## Analysis of Variance Table
##
## Response: score
              Df Sum Sq Mean Sq F value
                                            Pr(>F)
## budget
               1
                 0.584 0.5838 0.9118
                                            0.3411
```

```
## gross 1 21.823 21.8233 34.0882 2.798e-08 ***

## runtime 1 28.029 28.0291 43.7817 5.105e-10 ***

## Residuals 162 103.713 0.6402

## ---

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

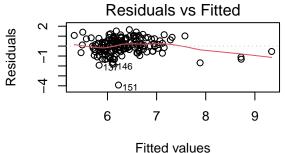
par(mfrow = c(2, 2))

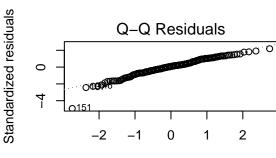
plot(movies_model)

Residuals vs Fitted

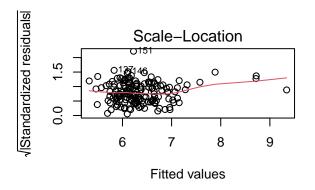
Q-Q Residuals

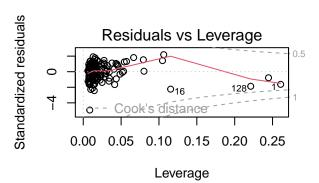
Q-Q Residuals
```





Theoretical Quantiles

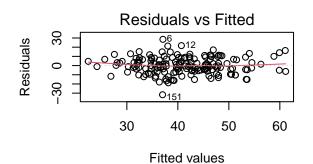


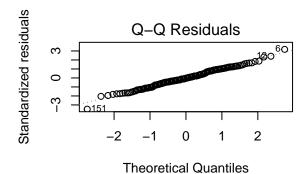


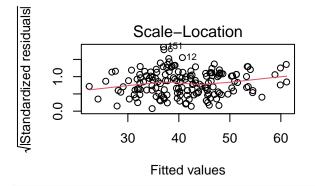
Transformed Model

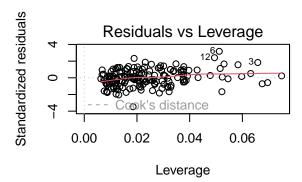
```
library(car)
## Loading required package: carData
summary(powerTransform(cbind(score, budget, gross, runtime) ~ 1))
## bcPower Transformations to Multinormality
           Est Power Rounded Pwr Wald Lwr Bnd Wald Upr Bnd
##
              2.0112
                            2.00
                                        1.3745
                                                     2.6478
## score
                                                     0.4530
## budget
              0.3495
                            0.33
                                        0.2459
                                                     0.2382
   gross
              0.1690
                            0.17
                                        0.0997
## runtime
             -1.8979
                           -2.00
                                       -2.6836
                                                    -1.1123
##
## Likelihood ratio test that transformation parameters are equal to 0
   (all log transformations)
                                     LRT df
                                                  pval
## LR test, lambda = (0 0 0 0) 151.8116 4 < 2.22e-16
## Likelihood ratio test that no transformations are needed
```

```
##
                                   LRT df
## LR test, lambda = (1 1 1 1) 507.9174 4 < 2.22e-16
t score <- score^2</pre>
t_budget <- budget^.33
t_gross <- gross^0.17
t_runtime <- runtime^(-2)</pre>
t_movies_model <- lm(t_score ~ t_budget + t_gross + t_runtime)</pre>
summary(t_movies_model)
##
## Call:
## lm(formula = t_score ~ t_budget + t_gross + t_runtime)
## Residuals:
      Min
##
               1Q Median
                               ЗQ
                                      Max
## -31.713 -5.575 -0.319
                            6.920 28.538
## Coefficients:
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) 5.999e+01 4.552e+00 13.180 < 2e-16 ***
## t_budget -6.105e+01 1.011e+01 -6.037 1.03e-08 ***
              3.669e+01 6.746e+00 5.438 1.95e-07 ***
## t_gross
## t_runtime -2.239e+05 2.660e+04 -8.419 1.92e-14 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 9.228 on 162 degrees of freedom
## Multiple R-squared: 0.4152, Adjusted R-squared: 0.4043
## F-statistic: 38.33 on 3 and 162 DF, p-value: < 2.2e-16
par(mfrow = c(2, 2))
plot(t_movies_model)
```









anova(t_movies_model)

```
## Analysis of Variance Table
##
## Response: t_score
             Df
                 Sum Sq Mean Sq F value
## t_budget
                   30.1
                           30.1 0.3534
                                           0.553
              1
                 3728.1 3728.1 43.7754 5.118e-10 ***
## t_gross
              1
## t_runtime
                 6035.8 6035.8 70.8735 1.919e-14 ***
              1
## Residuals 162 13796.4
                           85.2
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

FALSE

FALSE

Variable Selection

t_budget

```
library(leaps)

vif(t_movies_model)

## t_budget t_gross t_runtime

## 1.938831 1.940866 1.112464

X <- cbind(t_budget, t_gross, t_runtime)
b <- regsubsets(as.matrix(X), t_score)

summary(b)

## Subset selection object

## 3 Variables (and intercept)

## Forced in Forced out</pre>
```

```
## t_gross
                  FALSE
                              FALSE
                  FALSE
                              FALSE.
## t_runtime
## 1 subsets of each size up to 3
## Selection Algorithm: exhaustive
            t_budget t_gross t_runtime
## 1 (1)""
                               "*"
                      11 11
                               "*"
## 2 (1) "*"
## 3 (1) "*"
                      "*"
                               "*"
Optimal Models: 1 predictor: score^(2) ~ runtime^(-2) 2 predictors: score^(2) ~ budget^(0.33) + runtime^(-2)
3 predictors: score^{(2)} \sim budget^{(0.33)} + gross^{(0.17)} + runtime^{(-2)}
om1 <- lm(t_score ~ t_runtime)</pre>
om2 <- lm(t_score ~ t_budget + t_runtime)</pre>
om3 <- lm(t_score ~ t_budget + t_gross + t_runtime)
Radj2_vect <- summary(b)$adjr2</pre>
om_list <- list(om1, om2, om3)</pre>
AIC_vect <- numeric(3)
AICc_vect <- numeric(3)
BIC_vect <- numeric(3)
for (i in seq_len(3)) {
  AIC_vect[i] <- extractAIC(om_list[[i]])[2]
}
for (i in seq_len(3)) {
  AICc_vect[i] <- extractAIC(om_list[[i]])[2] + 2 * (i + 2) * (i + 3) / (nrow(movies) - i - 1)
}
for (i in seq_len(3)) {
  BIC_vect[i] <- extractAIC(om_list[[i]], k = log(nrow(movies)))[2]</pre>
data.frame("Size" = 1:3, "Radj2" = Radj2_vect, "AIC" = AIC_vect, "AICc" = AICc_vect, "BIC" = BIC_vect)
##
     Size
               Radj2
                           AIC
                                   AICc
                                              BIC
## 1
        1 0.2656625 774.5293 774.6756 780.7533
## 2
        2 0.2999124 767.5853 767.8307 776.9213
        3 0.4043372 741.7497 742.1200 754.1976
R<sup>2</sup>, AIC, AICc, and BIC all suggest the model with 3 predictors.
backAIC <- step(t_movies_model, direction = "backward", data = movies)</pre>
## Start: AIC=741.75
## t_score ~ t_budget + t_gross + t_runtime
##
                Df Sum of Sq
                               RSS
                                        AIC
## <none>
                              13796 741.75
## - t_gross
                      2518.7 16315 767.59
                 1
## - t_budget
                 1
                      3104.1 16901 773.44
## - t_runtime 1
                      6035.8 19832 799.99
Back AIC suggests the model with 3 predictors.
backBIC <- step(t_movies_model, direction = "backward", data = movies,</pre>
                 k = log(nrow(movies)))
```

```
## Start: AIC=754.2
## t_score ~ t_budget + t_gross + t_runtime
##
##
              Df Sum of Sq RSS
                                   AIC
## <none>
                            13796 754.20
## - t gross
                     2518.7 16315 776.92
               1
## - t budget
                     3104.1 16901 782.77
               1
## - t_runtime 1
                    6035.8 19832 809.33
Back BIC suggests the model with 3 predictors.
mint <- lm(score ~ 1, data = movies)</pre>
forwardAIC <- step(mint, scope = list(lower = ~ 1, upper = ~ t_budget + t_gross + t_runtime),</pre>
                direction = "forward", data = movies)
## Start: AIC=-10.3
## score ~ 1
##
##
              Df Sum of Sq
                             RSS
                                       AIC
## + t_runtime 1 39.422 114.73 -57.325
                    9.911 144.24 -19.326
## + t_gross
                1
## <none>
                           154.15 -10.295
## + t budget 1
                     0.174 153.97 -8.483
## Step: AIC=-57.33
## score ~ t_runtime
##
##
             Df Sum of Sq
                             RSS
                                      AIC
## + t budget 1 5.5191 109.21 -63.509
## + t_gross 1
                   1.8720 112.86 -58.056
                          114.73 -57.325
## <none>
##
## Step: AIC=-63.51
## score ~ t_runtime + t_budget
##
                            RSS
            Df Sum of Sq
                                     AIC
## + t_gross 1 15.128 94.08 -86.262
                         109.21 -63.509
## <none>
## Step: AIC=-86.26
## score ~ t_runtime + t_budget + t_gross
Forward AIC suggests the model with 3 predictors
forwardBIC <- step(mint, scope = list(lower = ~ 1, upper = ~ t_budget + t_gross + t_runtime),</pre>
                   direction = "forward", data = movies, k = log(nrow(movies)))
## Start: AIC=-7.18
## score ~ 1
##
##
              Df Sum of Sq
                              RSS
                                       AIC
## + t_runtime 1
                     39.422 114.73 -51.101
## + t_gross
                     9.911 144.24 -13.102
## <none>
                            154.15 -7.183
## + t_budget 1 0.174 153.97 -2.259
##
```

```
## Step: AIC=-51.1
## score ~ t_runtime
##
##
              Df Sum of Sq
                             RSS
                                      ATC
## + t_budget 1
                 5.5191 109.21 -54.173
## <none>
                           114.73 -51.101
                    1.8720 112.86 -48.720
## + t_gross
              1
##
## Step: AIC=-54.17
## score ~ t_runtime + t_budget
                             RSS
            Df Sum of Sq
##
## + t_gross 1 15.128 94.08 -73.814
                          109.21 -54.173
## <none>
##
## Step: AIC=-73.81
## score ~ t_runtime + t_budget + t_gross
Forward BIC suggests the model with 3 predictors.
```

Final Model

```
t movies model
##
## Call:
## lm(formula = t_score ~ t_budget + t_gross + t_runtime)
## Coefficients:
## (Intercept)
                  t_budget
                                           t_runtime
                                t_gross
                                          -223906.15
##
        59.99
                    -61.05
                                  36.69
summary(t_movies_model)
##
## Call:
## lm(formula = t_score ~ t_budget + t_gross + t_runtime)
##
## Residuals:
      Min
               10 Median
                               3Q
                                      Max
## -31.713 -5.575 -0.319
                            6.920
                                   28.538
##
## Coefficients:
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) 5.999e+01 4.552e+00 13.180 < 2e-16 ***
              -6.105e+01 1.011e+01 -6.037 1.03e-08 ***
## t budget
## t_gross
               3.669e+01 6.746e+00 5.438 1.95e-07 ***
## t_runtime
             -2.239e+05 2.660e+04 -8.419 1.92e-14 ***
## ---
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```

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