

Regression

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HomeWork 3 - Regression

Linear Regression

Linear Regression is a Machine Learning algorithm based on the mathematics concept that allows us to predict one dependent target variable, based on one or more independent variables.

In very simple terms, if we look at the equation $y = mx + b$, With target y and independent given variable x , Linear Regression will estimate values m and b allowing us to plug them in and estimate target y for any x .

In this notebook, we will look at a data set describing various attributes of Anime and Manga, and build some linear models with them.

Data Source: <https://www.kaggle.com/datasets/hernan4444/anime-recommendation-database-2020>

Pros of Linear Regression

- Easy To implement and interpret
- Easy to identify Use cases for by spotting potential correlations in data
- Has some techniques to avoid over fitting

Cons of Linear Regression

- Assumes relationship between independent and dependent variables. Assumes there is a straight line.
- Does not provide a complete description of relationships among variables.

The below chunk does the following

- Imports and cleans data
- Splits into 80/20 train/test
- Explores the data
- Plots two informative graphs

```
data <- read.csv(file="Anime-Data.csv",header=TRUE)
data = dplyr::select(data, -c('Japanese.name', 'Score.1': 'Score.10'))
data <- na.omit(data)

set.seed(2)
library(caTools)
```

```
split <- sample.split(data,SplitRatio=0.8)
train <- subset(data, split==TRUE)
test <- subset(data,split==FALSE)

names(train)
```

```
## [1] "MAL_ID"      "Name"        "Score"       "Genres"
## [5] "English.name" "Type"        "Episodes"    "Aired"
## [9] "Premiered"    "Producers"    "Licensors"    "Studios"
## [13] "Source"       "Duration"     "Rating"      "Ranked"
## [17] "Popularity"   "Members"      "Favorites"    "Watching"
## [21] "Completed"    "On.Hold"      "Dropped"      "Plan.to.Watch"
```

```
head(train)
```

```
##      MAL_ID      Name Score
## 1         1      Cowboy Bebop 8.78
## 3         6      Trigun 8.24
## 4         7      Witch Hunter Robin 7.27
## 7        16 Hachimitsu to Clover 8.06
## 13        22 Tennis no Ouji-sama 7.90
## 15        24      School Rumble 7.94
##
##                                     Genres      English.name
## 1      Action, Adventure, Comedy, Drama, Sci-Fi, Space      Cowboy Bebop
## 3      Action, Sci-Fi, Adventure, Comedy, Drama, Shounen      Trigun
## 4      Action, Mystery, Police, Supernatural, Drama, Magic      Witch Hunter Robin
## 7      Comedy, Drama, Josei, Romance, Slice of Life      Honey and Clover
## 13      Action, Comedy, Sports, School, Shounen The Prince of Tennis
## 15      Comedy, Romance, School, Shounen      School Rumble
##      Type Episodes      Aired      Premiered
## 1      TV        26 Apr 3, 1998 to Apr 24, 1999 Spring 1998
## 3      TV        26 Apr 1, 1998 to Sep 30, 1998 Spring 1998
## 4      TV        26 Jul 2, 2002 to Dec 24, 2002 Summer 2002
## 7      TV        24 Apr 15, 2005 to Sep 27, 2005 Spring 2005
## 13     TV       178 Oct 10, 2001 to Mar 23, 2005 Fall 2001
## 15     TV        26 Oct 5, 2004 to Mar 29, 2005 Fall 2004
##
##                                     Producers
## 1                                     Bandai Visual
## 3                                     Victor Entertainment
## 4      TV Tokyo, Bandai Visual, Dentsu, Victor Entertainment
## 7                                     Genco, Fuji TV, Shueisha
## 13      Production I.G, Nihon Ad Systems
## 15 TV Tokyo, Sotsu, Marvelous, Starchild Records, Media Factory, DAX Production, Studio Jack
##
##      Licensors      Studios      Source      Duration
## 1      Funimation, Bandai Entertainment      Sunrise Original 24 min. per ep.
## 3      Funimation, Geneon Entertainment USA      Madhouse      Manga 24 min. per ep.
## 4      Funimation, Bandai Entertainment      Sunrise Original 25 min. per ep.
## 7      VIZ Media, Discotek Media      J.C.Staff      Manga 23 min. per ep.
## 13      VIZ Media      Trans Arts      Manga 22 min. per ep.
## 15      Funimation Studio Comet      Manga 23 min. per ep.
##
##      Rating Ranked Popularity Members Favorites Watching
## 1 R - 17+ (violence & profanity)      28      39 1251960      61971      105808
```

```
## 3      PG-13 - Teens 13 or older    266      201 558913      12944      29113
## 4      PG-13 - Teens 13 or older   2481      1467  94683        587        4300
## 7      PG-13 - Teens 13 or older    468      687 214499        4101       11909
## 13     PG-13 - Teens 13 or older    675     1039 141832        3124       11235
## 15     PG-13 - Teens 13 or older    625      514 275464        5137       12277
##      Completed On.Hold Dropped Plan.to.Watch
## 1      718161    71513    26678      329800
## 3      343492    25465    13925      146918
## 4      46165     5121     5378      33719
## 7      81145     11901    11026      98518
## 13     76881     12905    12516      28295
## 15     157789    12856    13491      79051
```

```
summary(train)
```

```
##      MAL_ID      Name      Score      Genres
## Min. : 1.0 Length:976 Min. :4.720 Length:976
## 1st Qu.: 956.8 Class :character 1st Qu.:6.840 Class :character
## Median : 6105.5 Mode :character Median :7.260 Mode :character
## Mean :10356.0 Mean :7.276
## 3rd Qu.:18548.0 3rd Qu.:7.702
## Max. :32214.0 Max. :9.190
## English.name      Type      Episodes      Aired
## Length:976 Length:976 Min. : 3.00 Length:976
## Class :character Class :character 1st Qu.: 12.00 Class :character
## Mode :character Mode :character Median : 13.00 Mode :character
## Mean : 24.27
## 3rd Qu.: 25.00
## Max. :500.00
## Premiered      Producers      Licensors      Studios
## Length:976 Length:976 Length:976 Length:976
## Class :character Class :character Class :character Class :character
## Mode :character Mode :character Mode :character Mode :character
##
##
## Source      Duration      Rating      Ranked
## Length:976 Length:976 Length:976 Min. : 1
## Class :character Class :character Class :character 1st Qu.: 1012
## Mode :character Mode :character Mode :character Median : 2520
## Mean : 2907
## 3rd Qu.: 4286
## Max. :10803
## Popularity      Members      Favorites      Watching
## Min. : 1.0 Min. : 763 Min. : 0.0 Min. : 23
## 1st Qu.: 512.5 1st Qu.: 43240 1st Qu.: 140.5 1st Qu.: 1969
## Median : 1258.5 Median : 114871 Median : 560.5 Median : 6126
## Mean : 1779.3 Mean : 239974 Mean : 4486.2 Mean : 12994
## 3rd Qu.: 2438.8 3rd Qu.: 278572 3rd Qu.: 2186.5 3rd Qu.: 13484
## Max. :11065.0 Max. :2589552 Max. :183914.0 Max. :362124
## Completed      On.Hold      Dropped      Plan.to.Watch
## Min. : 300 Min. : 27 Min. : 94 Min. : 146
## 1st Qu.: 22721 1st Qu.: 1697 1st Qu.: 1858 1st Qu.: 12417
## Median : 62792 Median : 4255 Median : 5485 Median : 30107
```

```
## Mean : 160033 Mean : 7827 Mean : 8782 Mean : 50338
## 3rd Qu.: 174432 3rd Qu.: 8623 3rd Qu.: 11333 3rd Qu.: 65370
## Max. :2182587 Max. :109707 Max. :148408 Max. :425531
```

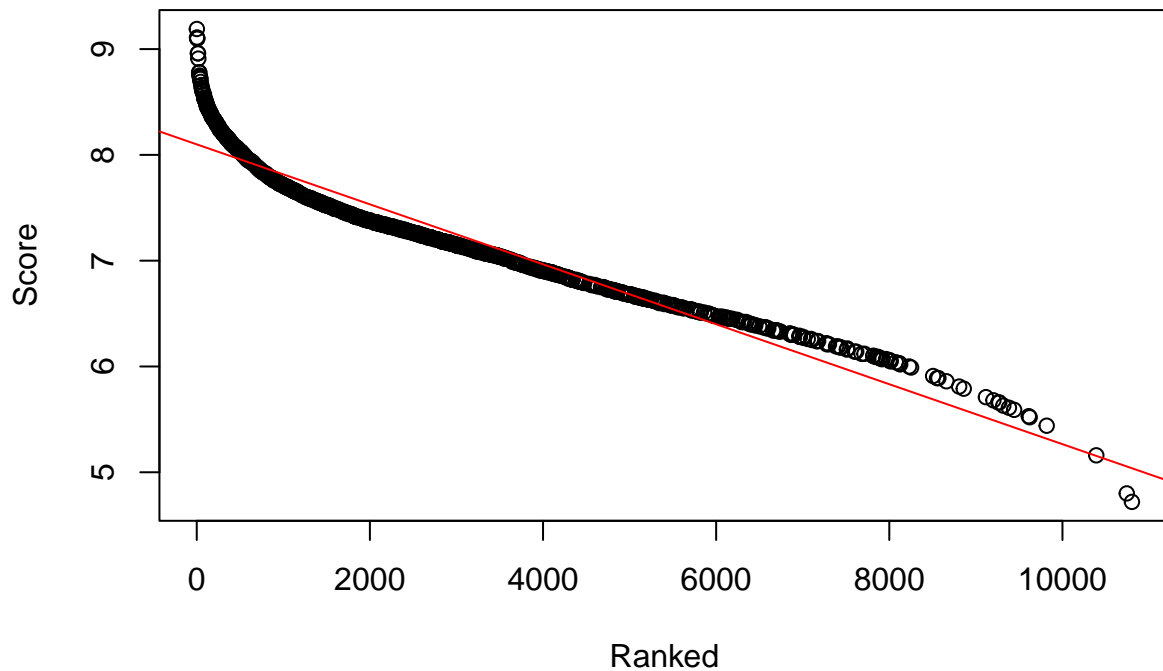
```
mean(train$Score)
```

```
## [1] 7.275809
```

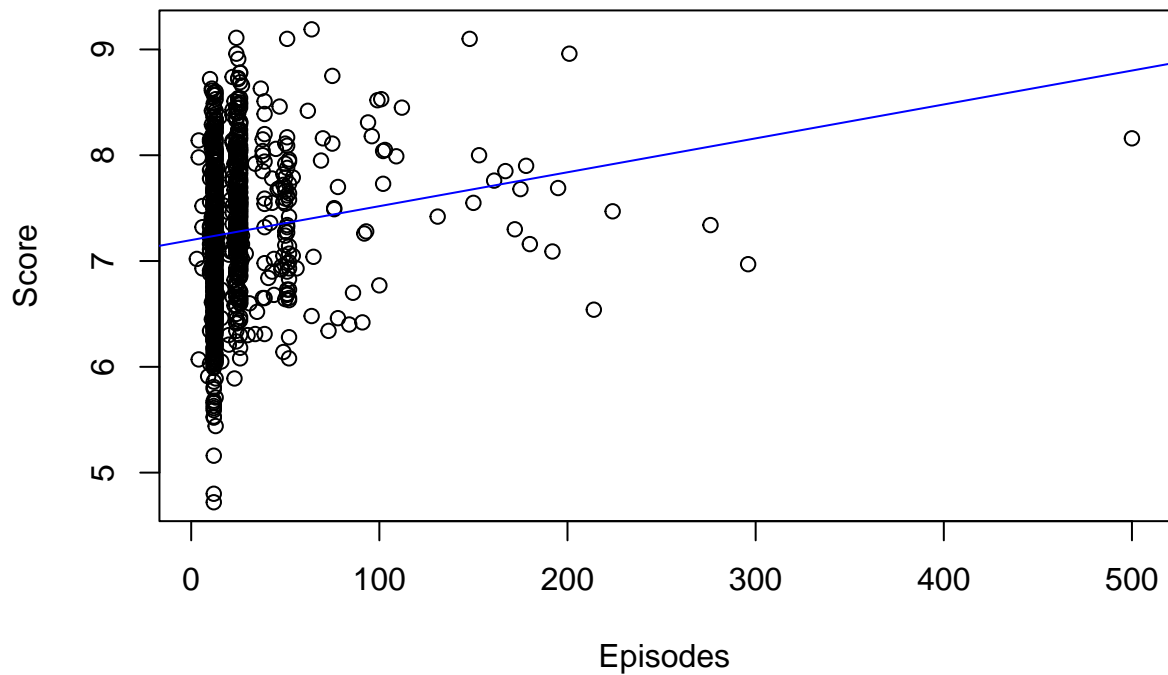
```
median(train$Score)
```

```
## [1] 7.26
```

```
plot(train$Score~train$Ranked,xlab="Ranked",ylab="Score")
abline(lm(train$Score~train$Ranked),col="red")
```



```
plot(train$Score~train$Episodes,xlab="Episodes",ylab="Score")
abline(lm(train$Score~train$Episodes),col="blue")
```



This chunk will

- Build a simple linear model of the data
- outputs the summary

```
lm1 <- lm(Score~Ranked,data=train)
summary(lm1)
```

```
##
## Call:
## lm(formula = Score ~ Ranked, data = train)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.31804 -0.12011 -0.06364  0.07201  1.09077
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  8.100e+00  9.165e-03   883.7  <2e-16 ***
## Ranked      -2.834e-04  2.493e-06  -113.7  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1753 on 974 degrees of freedom
```

```
## Multiple R-squared:  0.9299, Adjusted R-squared:  0.9298
## F-statistic: 1.292e+04 on 1 and 974 DF,  p-value: < 2.2e-16
```

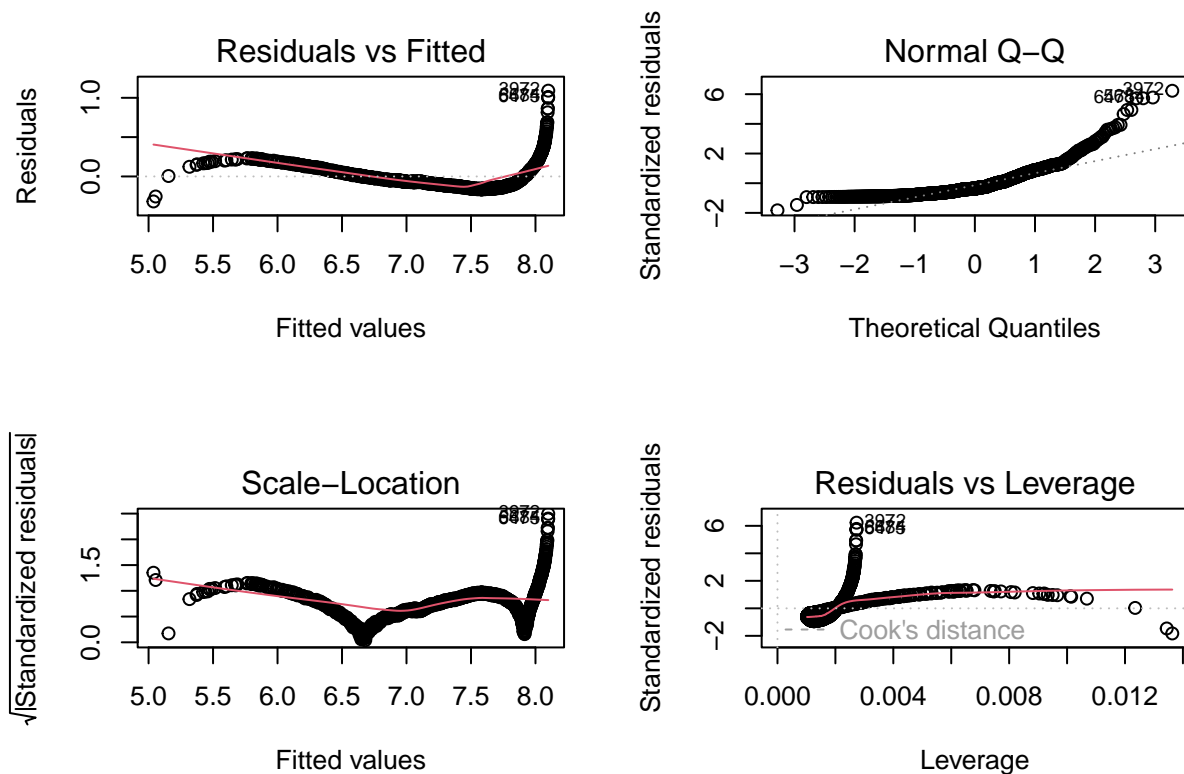
Summary of Simple Linear Model

The summary function in R has outputted a number of things.

- A **formula** that shows that we are modelling the score as a function of Rank
- **Residuals** that show the difference between what the model predicted and the actual value of **y**
- **Coefficients**
- The **Estimates** where the intercept tells us the value when all other features are 0. For the other features, the estimates give us the expected change in the response due to a unit change in the feature.
- **Standard Error** which allows us to construct marginal confidence intervals for the estimate of that particular feature.
- **t-value** which tells us about how far our estimated parameter is from the hypothesized 0 value.
- The **p-value** for the individual coefficient, which is the level of marginal significance within a statistical hypothesis test, representing the probability of the occurrence of a given event.
- The Residual Standard Error which gives the standard deviation of the residuals, and tells us about how large the prediction error is.
- **Multiple and Adjusted R²** which tell us what proportion of the variance is explained by our model
- **F-Stats** Which is the ratio of two variances

The next chunk will output various residuals plots

```
par(mfrow=c(2,2))
plot(lm1)
```



Explanation of Residual plot for the Simple Linear Model

For starters, the residuals max is relatively low. This can be verified by the fact that the original plot of the linear model showed a very strong linear trend. Also, in a residual plot, we want there to be a spread of values above and below the line that are close to even. In this residual plot however, many fall exactly on that line which shows an almost direct causation relationship between ranking and Score, which implies that Score is directly derived from the Rank.

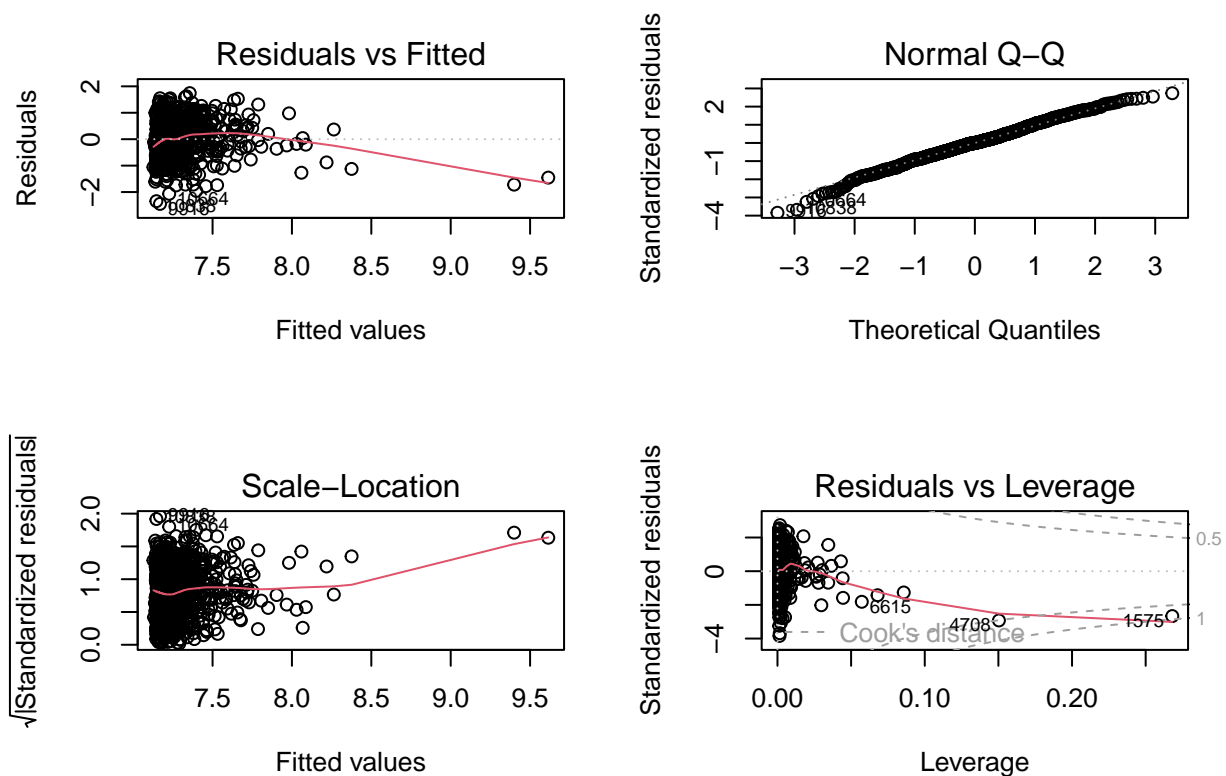
Next, we create a multiple linear model using Number of Episodes, Rank, and target Score

```
lm2 <- lm(Score~Dropped+Episodes,data=train)
summary(lm2)

##
## Call:
## lm(formula = Score ~ Dropped + Episodes, data = train)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.44889 -0.40969  0.00062  0.40777  1.74373
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
##
```

```
## (Intercept) 7.118e+00 2.809e-02 253.346 < 2e-16 ***
## Dropped     1.342e-05 1.849e-06 7.256 8.13e-13 ***
## Episodes    1.659e-03 6.854e-04 2.421 0.0157 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.6378 on 973 degrees of freedom
## Multiple R-squared:  0.0732, Adjusted R-squared:  0.0713
## F-statistic: 38.43 on 2 and 973 DF,  p-value: < 2.2e-16
```

```
par(mfrow=c(2,2))
plot(lm2)
```



Next, we create a second multiple linear model using Popularity, Rank, and target Score

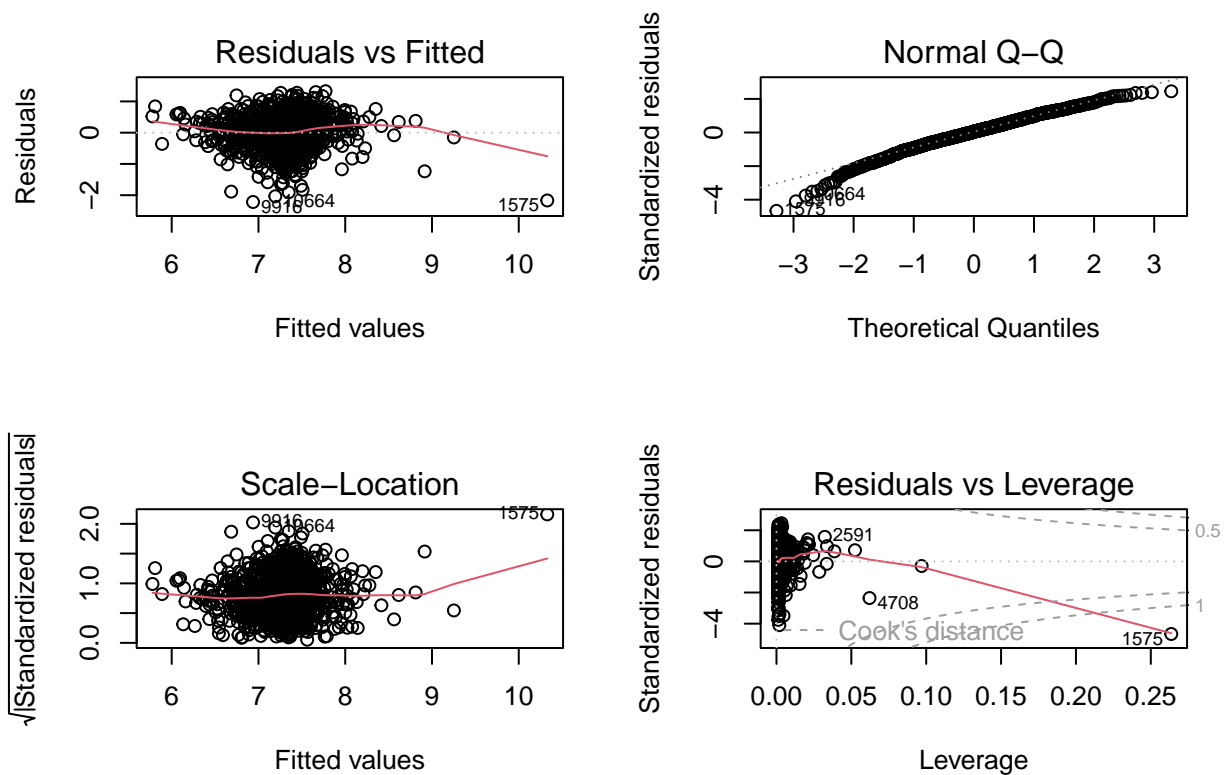
```
lm3 <- lm(Score~Watching+Popularity,data=train)
summary(lm3)
```

```
##
## Call:
## lm(formula = Score ~ Watching + Popularity, data = train)
##
## Residuals:
```



```
##      Min      1Q   Median      3Q      Max
## -2.21746 -0.31713  0.02945  0.36631  1.32679
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  7.439e+00  3.118e-02 238.621  <2e-16 ***
## Watching     7.987e-06  8.173e-07   9.773  <2e-16 ***
## Popularity   -1.501e-04  1.099e-05 -13.661  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.5419 on 973 degrees of freedom
## Multiple R-squared:  0.3311, Adjusted R-squared:  0.3297
## F-statistic: 240.8 on 2 and 973 DF,  p-value: < 2.2e-16
```

```
par(mfrow=c(2,2))
plot(lm3)
```



Next, let us use our models and make predictions

```
pred1 <- predict(lm1,newdata=test)
pred2 <- predict(lm2,newdata=test)
pred3 <- predict(lm3,newdata=test)
```

```

cor1 <- cor(pred1,test$Score)
mse1 <- mean((pred1 - test$Score)^2)
rmse1 <- sqrt(mse1)
print(paste("Correlation of the first model: ", cor1))

## [1] "Correlation of the first model:  0.971267531662069"

print(paste("mse of the first model: ", mse1))

## [1] "mse of the first model:  0.0217860311525616"

print(paste("rmse of the first model: ", rmse1))

## [1] "rmse of the first model:  0.14760091853563"

cor2 <- cor(pred2,test$Score)
mse2 <- mean((pred2 - test$Score)^2)
rmse2 <- sqrt(mse2)
print(paste("Correlation of the second model: ", cor2))

## [1] "Correlation of the second model:  0.23700889131067"

print(paste("mse of the second model: ", mse2))

## [1] "mse of the second model:  0.356885250337707"

print(paste("rmse of the second model: ", rmse2))

## [1] "rmse of the second model:  0.597398736471468"

cor3 <- cor(pred3,test$Score)
mse3 <- mean((pred3 - test$Score)^2)
rmse3 <- sqrt(mse3)
print(paste("Correlation of the third model: ", cor3))

## [1] "Correlation of the third model:  0.567895691688818"

print(paste("mse of the third model: ", mse3))

## [1] "mse of the third model:  0.249705718964579"

print(paste("rmse of the third model: ", rmse3))

## [1] "rmse of the third model:  0.499705632312244"

```

Analysis of models

The first model showed Score as a function of Rank. The second model showed Score as a function of number of existing episodes + how many people dropped the show. The third model showed Score as a function of user rated popularity + individuals currently watching the show.

Ultimately, the simple linear model performed best out of the 3. Despite the latter two providing a multi-varied analysis using multiple attributes, the single best predictor was proven to be the ranking of each anime, as the plots and correlation showed that rank directly influences the score it received.

The first simple model had the greatest correlation sitting at 0.97, as compared to the second with 0.23 and third, 0.56. We want this number to be as close to -1/1 as possible.

The rmse of the first was also the least, which shows we are less off using the first model than the other two. The first model was off by an average of 0.14, while the other two were off by 0.59 and 0.49 respectively.

Ultimately, I believe the reason for this was simply the chosen attributes. It was interesting though how the third model performed, which showed a semi-decent correlation with popularity, people currently watching, and Score.