

STATS 419 Survey of Multivariate Analysis

Week 03 Assignment 02_datasets Revisited

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0.1 Get Everything Set Up

```
library(devtools)
github.path = 'https://raw.githubusercontent.com/JoshuaPickel/WSU_STATS419_FALL2020/';
source_url(paste0(github.path,"master/functions/libraries.R"));
source_url(paste0(github.path,"master/functions/functions-imdb.R"));
```

1 Matrix

Create the “rotate matrix” functions as described in lectures. Apply to the example “myMatrix”.

```
source_url(paste0(github.path,"master/functions/functions-matrix.R"))
# Create myMatrix
myMatrix = matrix ( c (
1, 0, 2,
0, 3, 0,
4, 0, 5
), nrow=3,
byrow=T);
```

1.1 Matrix 90 Degrees

```
# Rotating clockwise.
rotateMatrix90(myMatrix)
```

```
##      [,1] [,2] [,3]
## [1,]    4    0    1
## [2,]    0    3    0
## [3,]    5    0    2
```

1.2 Matrix 180 Degrees

```
# Rotating Clockwise  
rotateMatrix180(myMatrix)
```

```
##      [,1] [,2] [,3]  
## [1,]    5    0    4  
## [2,]    0    3    0  
## [3,]    2    0    1
```

1.3 Matrix 270 Degrees

```
# Rotating Clockwise  
rotateMatrix270(myMatrix)
```

```
##      [,1] [,2] [,3]  
## [1,]    2    0    5  
## [2,]    0    3    0  
## [3,]    1    0    4
```

2 IRIS Plot

Recreate the graphic for the IRIS Data Set using R. Same titles, same scales, same colors.

2.1 Plot

```
library(datasets)  
plot(iris[1:4], pch = 21, cex = 1, col = 'black', bg = c("red", "green3",  
"blue")[iris$Species], main = 'Iris Data (red=sestosa, green=versicolor, blue=virginica)', gap.axis = 1.5)
```

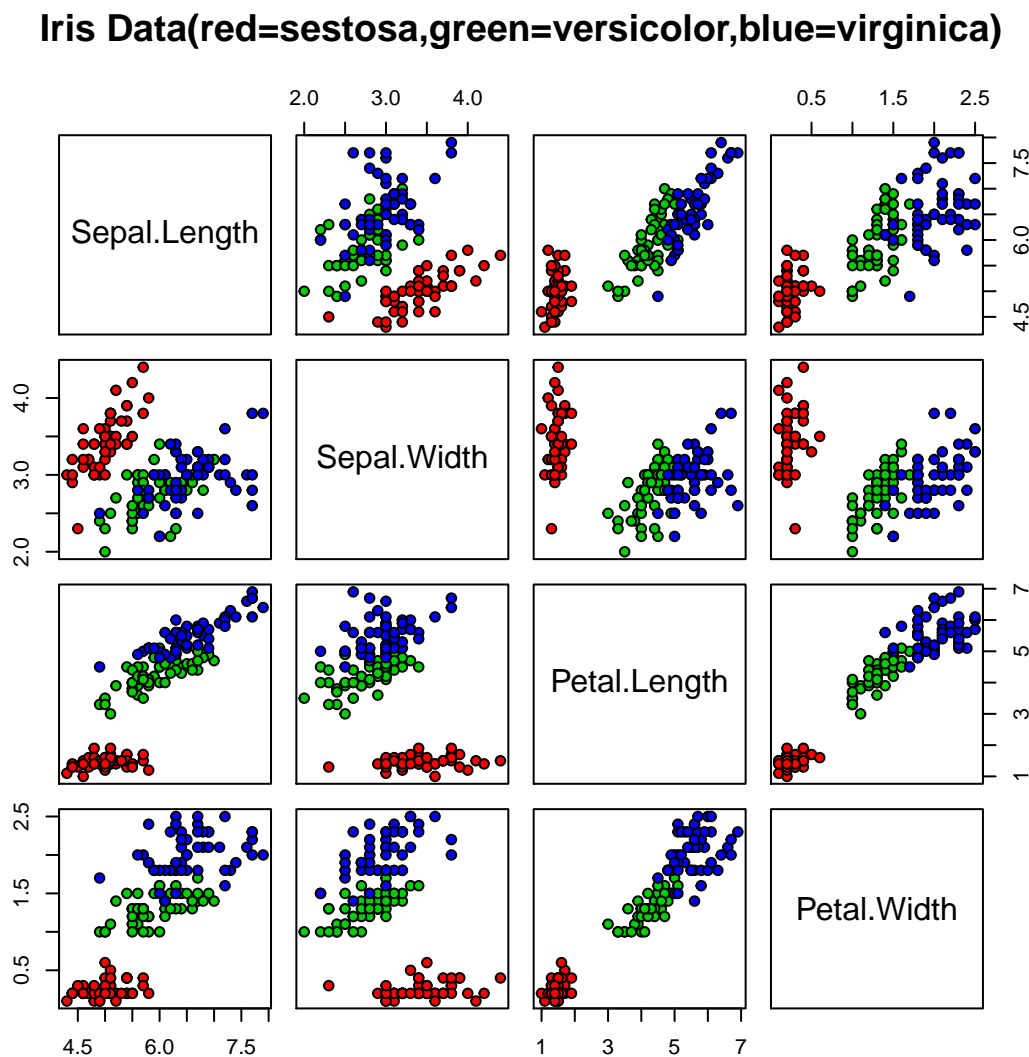


Figure 1: Plot From Plotting IRIS Data

3 IRIS Summary

Right 2-3 sentences concisely defining the IRIS Data Set. Maybe search KAGGLE for a nice template. Be certain the final writeup are your own sentences (make certain you modify what you find, make it your own, but also cite where you got your ideas from). NOTE: Watch the video, Figure 8 has a +5 EASTER EGG.

3.1 Response

The Iris data set was created in 1936 by Edgar Anderson and is a multivariate data set that containing four different measurements of three different Iris flower species. The data set contains 50 records for each of the different flower species; Setosa, Virginica, and Versicolor recording measurements for petal length, petal width, sepal width and sepal length, all recorded in centimeters.

4 Personality Data

Import “personality-raw.txt” into R. Remove the V00 column. Create two new columns from the current column “date_test”: year and week. Stack Overflow may help: <https://stackoverflow.com/questions/22439540/how-to-get-week-numbers-from-dates> ... Sort the new data frame by YEAR, WEEK so the newest tests are first ... The newest tests (e.g., 2020 or 2019) are at the top of the data frame. Then remove duplicates using the unique function based on the column “md5_email”. Save the data frame in the same “pipe-delimited format” (| is a pipe) with the headers. You will keep the new data frame as “personality-clean.txt” for future work (you will not upload it at this time). In the homework, for this tasks, report how many records your raw dataset had and how many records your clean dataset has.

4.1 Clean Personality Data

```
source_url(paste0(github.path,"master/functions/functions-cleanup.R"))
files = paste0(github.path,"master/datasets/personality-raw.txt")
personality_raw = read.csv(files,header = T, sep = '|')
personality_clean = removeColumn(personality_raw,'V00')
personality_clean = convertDates(personality_clean,personality_clean$date_test)
personality_clean = removeDuplicates(personality_clean,personality_clean$md5_email)
head(personality_clean)
```

```
##              md5_email      date_test V01 V02 V03 V04 V05 V06
## 838 b62c73cdaf59e0a13de495b84030734e 4/6/2020 12:57 3.4 4.2 2.6 4.2 2.6 2.6
## 837 1358d38e6898b1a0e5940f8b99ba2325 12/1/2019 22:12 3.4 3.4 3.4 4.2 4.2 4.2
## 835 f529455e4400e76f323f8c68154e194b 5/6/2019 4:44 4.2 5.0 1.8 4.2 4.2 5.0
## 836 0445a05e751e17de30ebdcbbcb575d59 5/6/2019 10:32 1.8 2.6 3.4 4.2 5.0 3.4
## 828 bfd1c69406d322d17312e965752813c2 5/2/2019 10:26 2.6 4.2 1.0 4.2 4.2 2.6
## 829 9cf05d7d516099c9533b98beb91993b9 5/2/2019 10:48 5.0 5.0 1.8 5.0 5.0 4.2
##      V07 V08 V09 V10 V11 V12 V13 V14 V15 V16 V17 V18 V19 V20 V21 V22 V23 V24 V25
## 838 4.2 2.6 3.4 4.2 4.2 3.4 3.4 4.2 5.0 3.4 5.0 3.4 1.8 2.6 2.6 2.6 4.2 3.4 5.0
## 837 5.0 3.4 4.2 3.4 2.6 3.4 3.4 4.2 4.2 4.2 4.2 4.2 3.4 2.6 3.4 4.2 4.2 4.2 2.6
## 835 3.4 3.4 4.2 3.4 2.6 2.6 4.2 5.0 3.4 4.2 5.0 4.2 2.6 2.6 1.8 3.4 5.0 3.4 1.8
## 836 2.6 2.6 5.0 3.4 2.6 4.2 2.6 3.4 4.2 3.4 4.2 4.2 3.4 1.8 2.6 3.4 4.2 5.0 1.8
## 828 3.4 1.8 4.2 4.2 1.8 2.6 3.4 5.0 4.2 4.2 5.0 4.2 3.4 1.8 1.0 3.4 4.2 3.4 1.8
## 829 1.0 5.0 5.0 5.0 1.0 5.0 5.0 5.0 5.0 5.0 5.0 3.4 3.4 3.4 4.2 5.0 5.0 3.4 1.8
##      V26 V27 V28 V29 V30 V31 V32 V33 V34 V35 V36 V37 V38 V39 V40 V41 V42 V43 V44
## 838 2.6 4.2 3.4 2.6 2.6 4.2 1.8 3.4 4.2 4.2 4.2 2.6 4.2 2.6 4.2 4.2 4.2 4.2 2.6
## 837 4.2 4.2 3.4 2.6 4.2 4.2 3.4 4.2 3.4 4.2 5.0 3.4 4.2 4.2 4.2 4.2 4.2 4.2 4.2
## 835 4.2 3.4 5.0 1.8 5.0 4.2 1.8 4.2 3.4 2.6 3.4 2.6 3.4 3.4 5.0 3.4 3.4 3.4 3.4
## 836 3.4 2.6 3.4 2.6 4.2 5.0 5.0 5.0 5.0 5.0 5.0 3.4 5.0 5.0 5.0 4.2 5.0 5.0 5.0
## 828 4.2 5.0 3.4 1.8 4.2 3.4 4.2 4.2 3.4 4.2 3.4 1.8 5.0 3.4 4.2 1.8 2.6 4.2 4.2
## 829 5.0 5.0 4.2 3.4 5.0 5.0 4.2 5.0 5.0 5.0 5.0 2.6 3.4 5.0 4.2 5.0 5.0 3.4 5.0
##      V45 V46 V47 V48 V49 V50 V51 V52 V53 V54 V55 V56 V57 V58 V59 V60 week year
## 838 4.2 4.2 2.6 3.4 2.6 4.2 1.8 4.2 2.6 3.4 4.2 4.2 1.8 4.2 2.6 4.2    15 2020
## 837 3.4 4.2 4.2 2.6 3.4 4.2 3.4 4.2 4.2 4.2 4.2 3.4 4.2 4.2 3.4 3.4    48 2019
## 835 4.2 5.0 3.4 4.2 3.4 4.2 2.6 3.4 5.0 5.0 3.4 3.4 3.4 3.4 1.8 3.4    19 2019
## 836 5.0 5.0 5.0 4.2 4.2 3.4 3.4 5.0 5.0 5.0 2.6 5.0 5.0 5.0 4.2 5.0    19 2019
## 828 4.2 4.2 4.2 2.6 3.4 1.8 2.6 2.6 5.0 4.2 3.4 2.6 2.6 4.2 4.2 4.2    18 2019
## 829 5.0 5.0 5.0 3.4 5.0 5.0 5.0 5.0 5.0 5.0 1.0 2.6 3.4 5.0 5.0 3.4    18 2019
```

4.2 Raw Dataset Dimensions

```
dim(personality_raw)
```

```
## [1] 838 63
```

4.3 Clean Dataset Dimensions

```
dim(personality_clean)
```

```
## [1] 678 64
```

5 Custom Functions

Write functions for `doSummary` and `sampleVariance` and `doMode` ... test these functions in your homework on the “monte.shaffer@gmail.com” record from the clean dataset. Report your findings. For this “monte.shaffer@gmail.com” record, also create z-scores. `Plot(x,y)` where x is the raw scores for “monte.shaffer@gmail.com” and y is the z-scores from those raw scores. Include the plot in your assignment, and write 2 sentences describing what pattern you are seeing and why this pattern is present.

```
# Set up and get row for monte.shaffer@gmail.com
source_url(paste0(github.path,"master/functions/functions-custom.R"))
monte_row = personality_clean[which(personality_clean$md5_email ==
'b62c73cdaf59e0a13de495b84030734e'),]
```

5.1 doSummary

```
doSummary(monte_row)
```

```
##      SumSq      Sum  variance
## 1 771.04 208.8 0.7528136
```

```
##                                     md5_email mode Mean naNum Length
## 838 b62c73cdaf59e0a13de495b84030734e 4.2 3.48      0      64
##      TwoPassVariance.variance NaiveVariance.variance      Sd
## 838                      0.7528136                0.7528136 0.8676483
```

5.2 custom_mode

```
custom_mode(monte_row)
```

```
##                                     md5_email mode
## 838 b62c73cdaf59e0a13de495b84030734e 4.2
```

5.3 doSampleVariance Naive

```
doSampleVariance(monte_row, 'naive')
```

```
##      SumSq    Sum  variance  
## 1 771.04 208.8 0.7528136
```

5.4 doSampleVariance Two Pass

```
doSampleVariance(monte_row, 'Two Pass')
```

```
##      Sum1    Sum2  variance  
## 1 208.8 44.416 0.7528136
```

5.5 Z-scores

```
zscores = zScores(monte_row)  
plot(zscores$raw.scores, zscores$z.score, main = 'Raw Scores Vs Z-Scores', xlab = 'raw scores', ylab = 'z-scores')
```

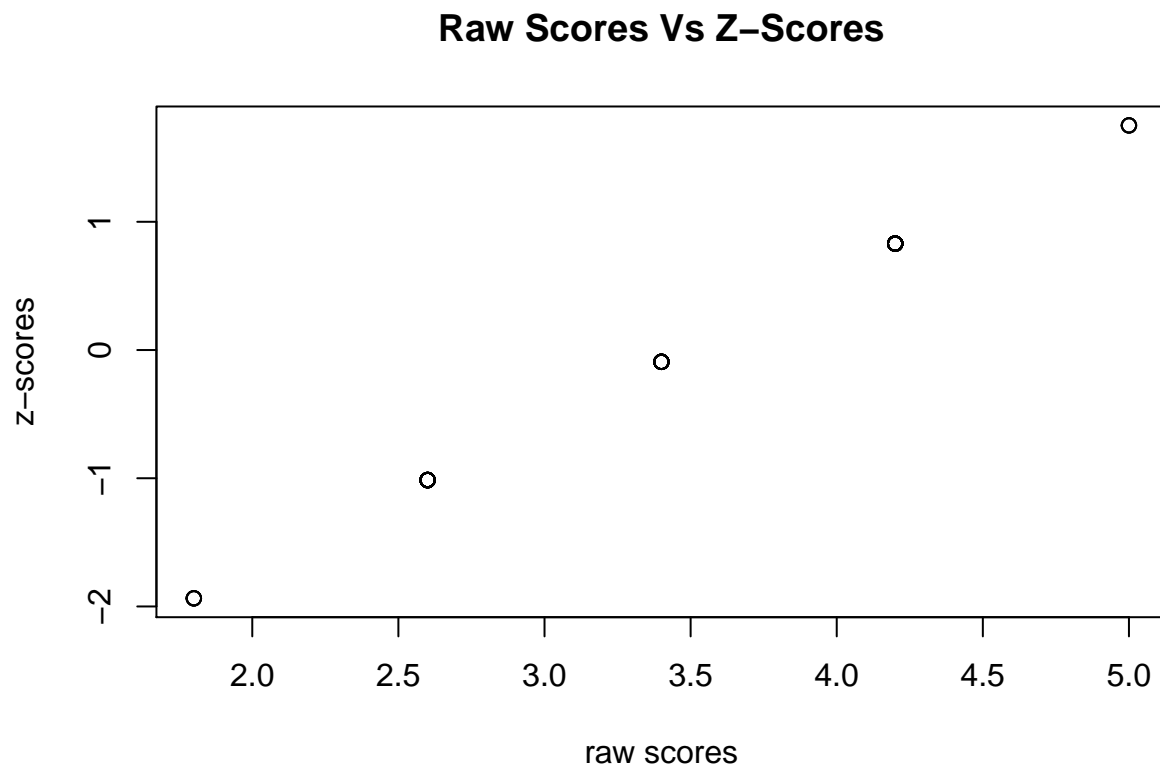


Figure 2: Plotting Raw Scores Against Z-scores

5.6 Response

After running the `doSummary()` on the record for `'monte.shaffer@gmail.com'`, it is apparent that the mode is 4.2, the mean is 3.48, there are no NAN values, the length is 64 columns, the two pass and naïve variances are both .7528136 and the standard deviation is .8676483. Considering the plot created from plotting raw scores against z-scores we can conclude from Figure 9 that there is a strong positive linear relationship between raw scores and z-scores. This is because the z-score is calculated from the raw values, meaning the two are highly correlated.

6 Will vs Denzel

Compare Will Smith and Denzel Washington. You will have to create a new variable `$millions.2000` that converts each movie's \$millions based on the \$year of the movie, so all dollars are in the same time frame. You will need inflation data from about 1980-2020 to make this work.

6.1 Get Data For Actors

```
source_url(paste0(github.path,"master/functions/functions-imdb.R"))

# Add mill.2000 column to Will's data frame with converted dollar values
nmid = "nm0000226"
will = grabFilmsForPerson(nmid)

nmid = "nm0000243"
denzel = grabFilmsForPerson(nmid)
```

6.2 Get Inflation Data

```
source_url(paste0(github.path,"master/functions/functions-inflation.R"))
inflation_data = grabInflation()
```

6.3 Adjust Denzel and Will Million Values

```
will = convertDollars(will,inflation_data)
denzel = convertDollars(denzel,inflation_data)

head(will$movies.50)
```

##	rank	title	ttid	year	rated	minutes			
## 1	1	I Am Legend	tt0480249	2007	PG-13	101			
## 2	2	Suicide Squad	tt1386697	2016	PG-13	123			
## 3	3	Independence Day	tt0116629	1996	PG-13	145			
## 4	4	Men in Black	tt0119654	1997	PG-13	98			
## 5	5	I, Robot	tt0343818	2004	PG-13	115			
## 6	6	The Pursuit of Happyness	tt0454921	2006	PG-13	117			
##		genre	ratings	metacritic	votes	millions	mill.2000		
## 1		Action, Adventure, Drama	7.2	65	675452	256.39	212.9349		
## 2		Action, Adventure, Fantasy	6.0	40	588391	325.10	233.2524		
## 3		Action, Adventure, Sci-Fi	7.0	59	520787	306.17	336.0260		
## 4		Action, Adventure, Comedy	7.3	71	507778	250.69	268.9646		
## 5		Action, Drama, Sci-Fi	7.1	59	491666	144.80	131.9987		
## 6		Biography, Drama	8.0	64	438369	163.57	139.7160		

```
head(denzel$movies.50)
```

##	rank	title	ttid	year	rated	minutes	genre
## 1	1	American Gangster	tt0765429	2007	R	157	Biography, Crime, Drama
## 2	2	Training Day	tt0139654	2001	R	122	Crime, Drama, Thriller
## 3	3	Inside Man	tt0454848	2006	R	129	Crime, Drama, Mystery
## 4	4	The Equalizer	tt0455944	2014	R	132	Action, Crime, Thriller
## 5	5	Man on Fire	tt0328107	2004	R	146	Action, Crime, Drama
## 6	6	Flight	tt1907668	2012	R	138	Drama, Thriller

##	ratings	metacritic	votes	millions	mill.2000
## 1	7.8	76	384167	130.16	108.09943
## 2	7.7	69	382242	76.63	74.50980
## 3	7.6	76	332167	88.51	75.60229
## 4	7.2	57	326292	101.53	73.85216
## 5	7.7	47	324297	77.91	71.02224
## 6	7.3	76	320454	93.77	70.32934

6.4 Response

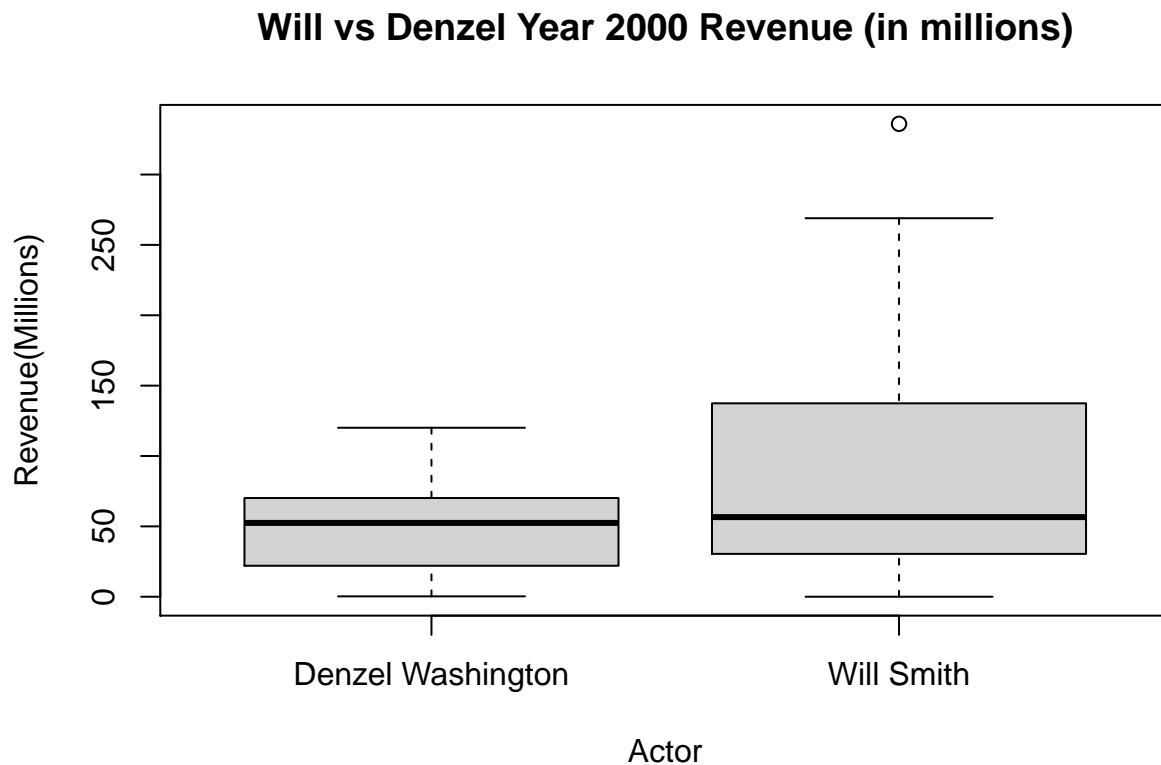
With converting all movie revenues to year 2000 dollars, we can tell that most of the movies Will Smith is in have higher revenue than the movies featuring Denzel Washington. We can also tell that Will Smith was in a movie that generated \$336.03 million, which is considered an outlier. In general, we can see that the median revenue is the almost same between Will Smith and Denzel Washington, and Will Smith has a higher maximum revenue value than Denzel Washington.

7 Will Vs Denzel

Build side-by-side box plots on several of the variables (including #6) to compare the two movie stars. After each box plot, write 2+ sentence describing what you are seeing, and what conclusions you can logically make. You will need to review what the box plot is showing with the box portion, the divider in the box, and the whiskers.

7.1 Comparing Will vs Denzel Year 2000 Millions

```
boxplot(denzel$movies.50$mill.2000,will$movies.50$mill.2000,main = 'Will vs Denzel Year 2000 Revenue (i
```

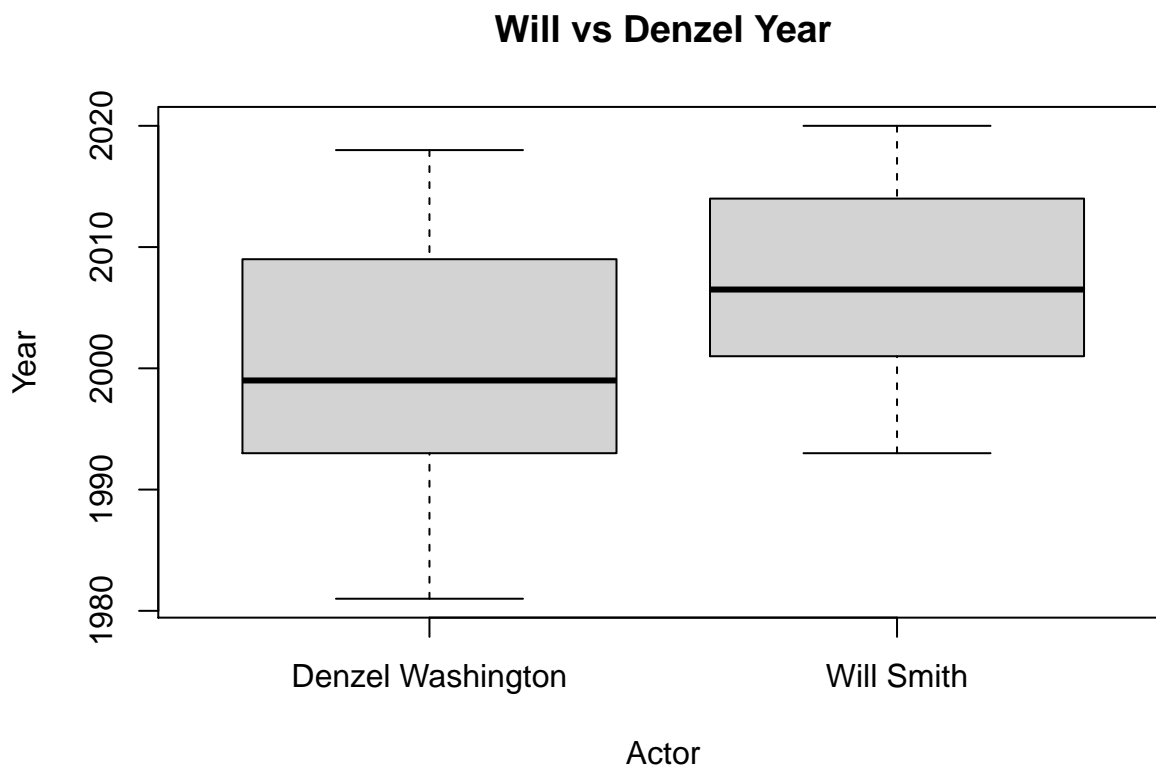



7.1.1 Response

Building a boxplot comparing the revenues (scaled to year 2000) between Will Smith and Denzel Washington (Figure 10), we can tell that Will Smith has a slightly higher median revenue than Denzel Washington. We can also conclude that fifty percent of Will Smith's revenue values are between \$30.45 million and \$137.66 million and fifty percent of Denzel Washington's revenue values are between \$22.39 million and \$70.06 million. This shows us that Will Smith's revenue values are spread out more when compared to Denzel Washington's values.

7.2 Comparing Will vs Denzel Years

```
boxplot(denzel$movies.50$year,will$movies.50$year,main = 'Will vs Denzel Year', names = c('Denzel Washi
```

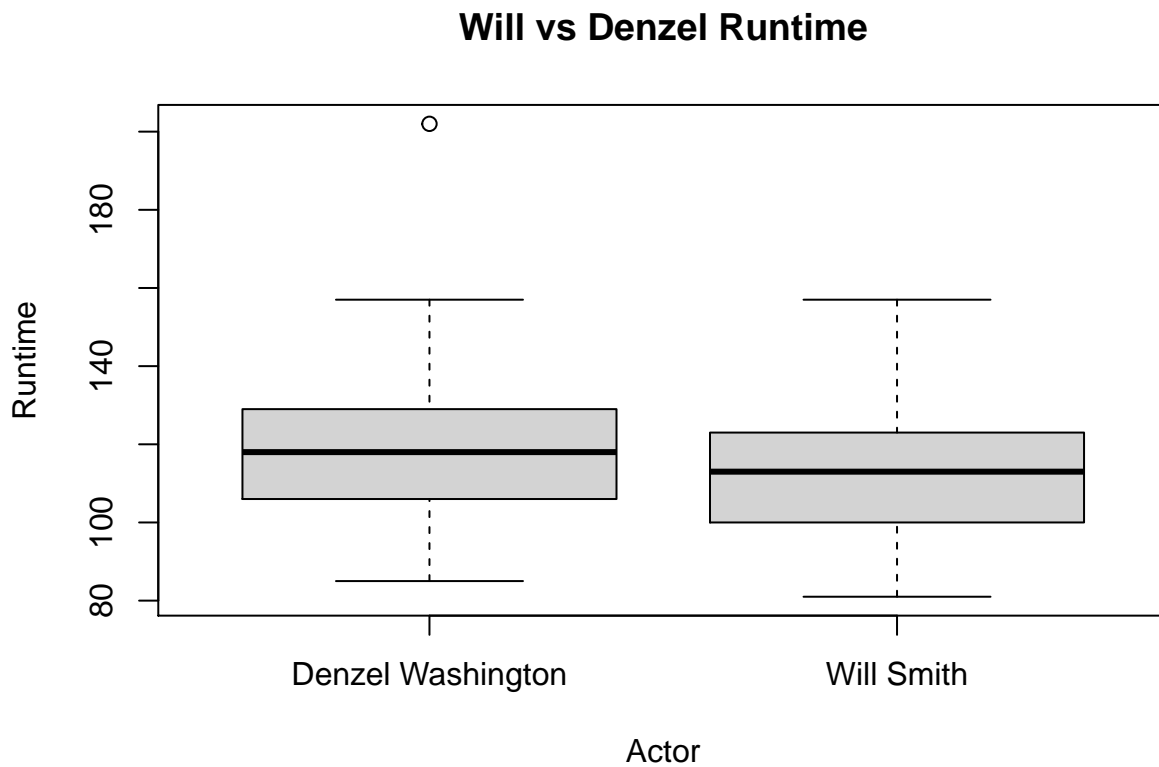


7.2.1 Response

Comparing Will Smith and Denzel Washington's years using a boxplot (Figure 11), there are several key findings. We can conclude that, for this data set, Denzel Washington appears in older movies when compared to Will Smith. We can also determine that Denzel Washington's most recent film was before 2020, while Will Smith's most recent film was in 2020. We can also conclude that half of the movies Will Smith has been featured in were after 2006, and the other half before 2006. Likewise, we can determine that half of the movies Denzel Washington has appeared in were after 1999, and the other half before 1999.

7.3 Comparing Will vs Denzel Runtime

```
boxplot(denzel$movies.50$minutes,will$movies.50$minutes,main = 'Will vs Denzel Runtime', names = c('Denzel', 'Will'))
```



7.3.1 Response

Using a boxplot to compare Will Smith and Denzel Washington's movie runtime(Figure 12), we can conclude that Denzel Washington has appeared in the longest movie in the data set. We can also conclude that Will Smith appeared in the shortest movie in the data set. Furthermore, we can conclude that the upper half of Denzel Washington's runtime values are higher than Will Smith's runtime values.