

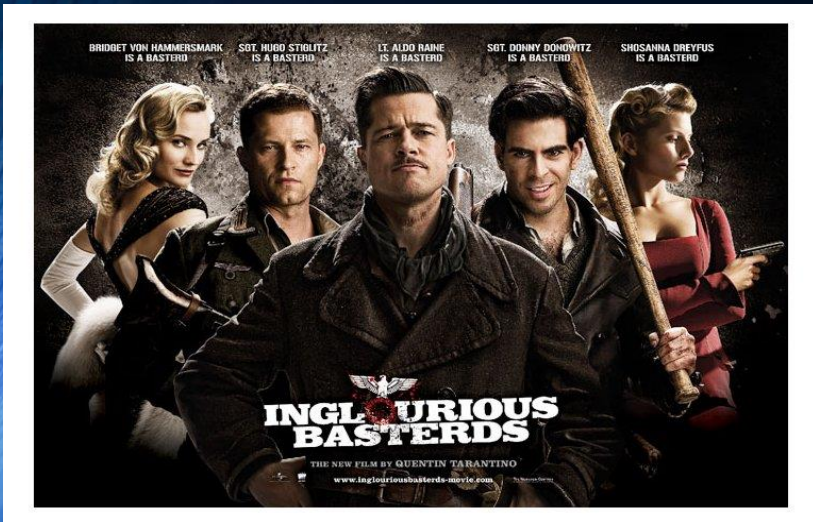
Movie Fun

SEAN WILLIAMS
RANDY TREAT
MARK WILLIAMS



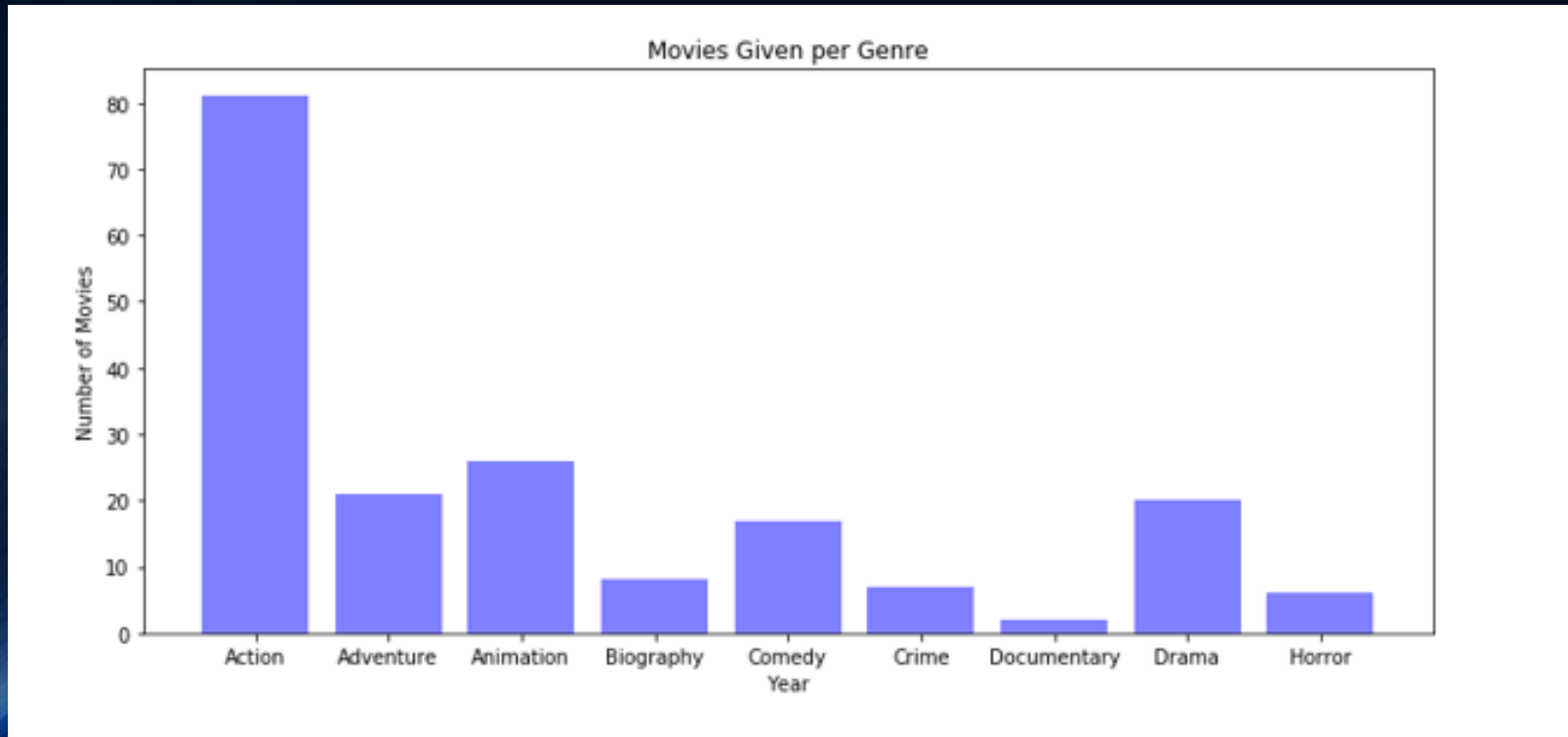
Summary

- (Enter finalized hypothesis here)
- Why we asked questions we did?
- Able to answer questions? Satisfied with results?
- Conclusion



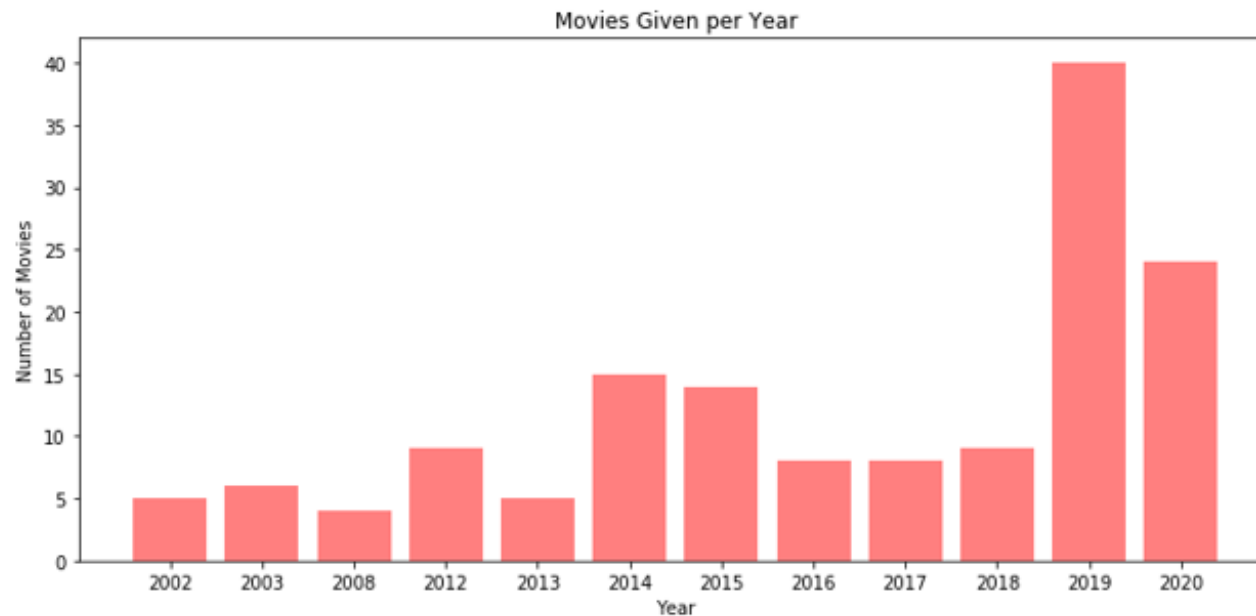
What genre of movies are generated when using Random Lists?

- Use of Random Lists data
- Narrowed scope
- Only included genres with one or more films



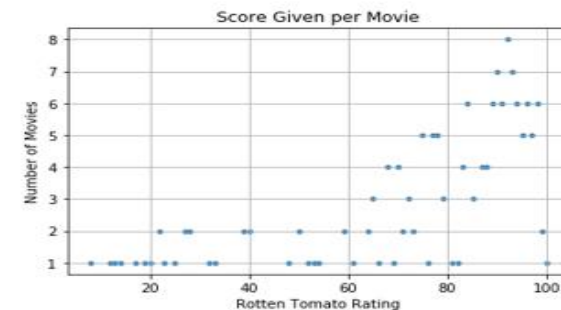
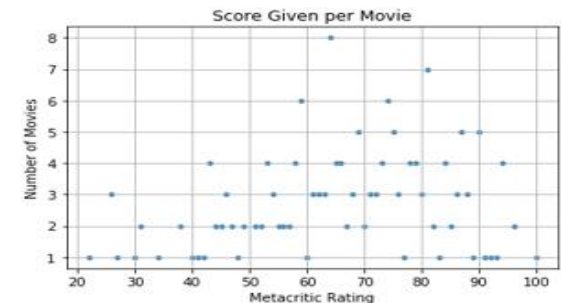
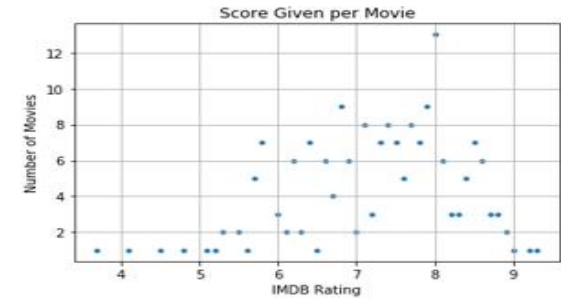
Are the movies from a specific time period? Are they newer films, older films, or is there a mix?

- Aimed to isolate the years with the most movies
- Excluded years that didn't include more than four titles
- Findings



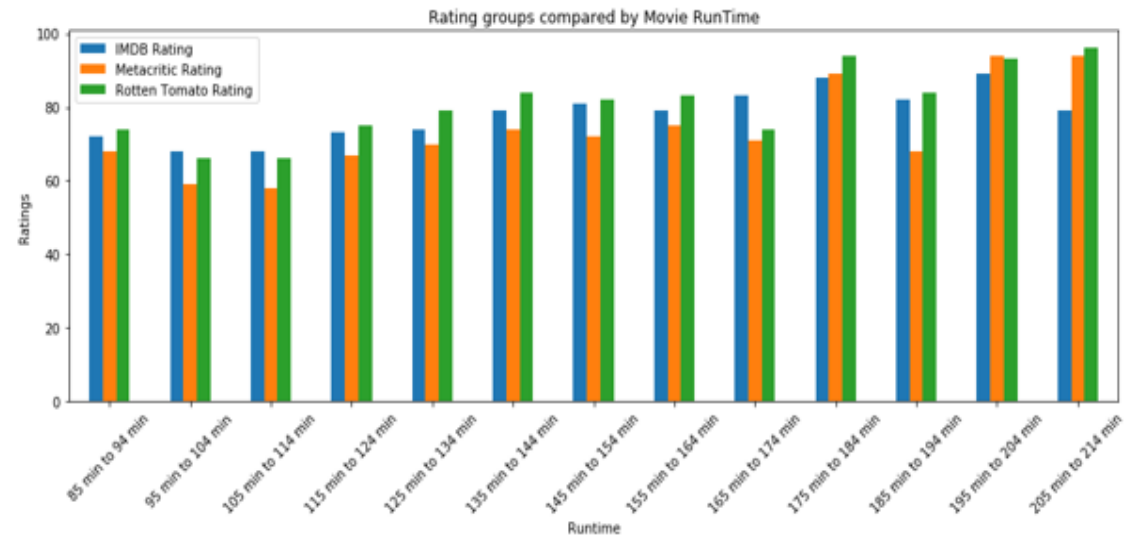
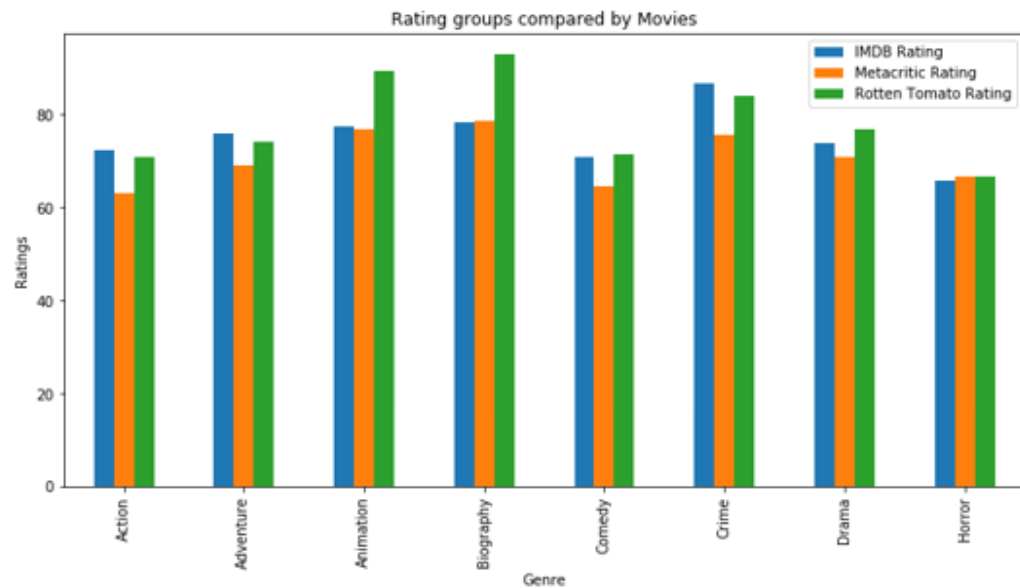
What type of movies are included in the generator? Are there any major differences in our ratings sources?

- Describe goal of models
- Pulled ratings from various sources; Why we chose those sources
- Good movie threshold; How we determined
- Findings



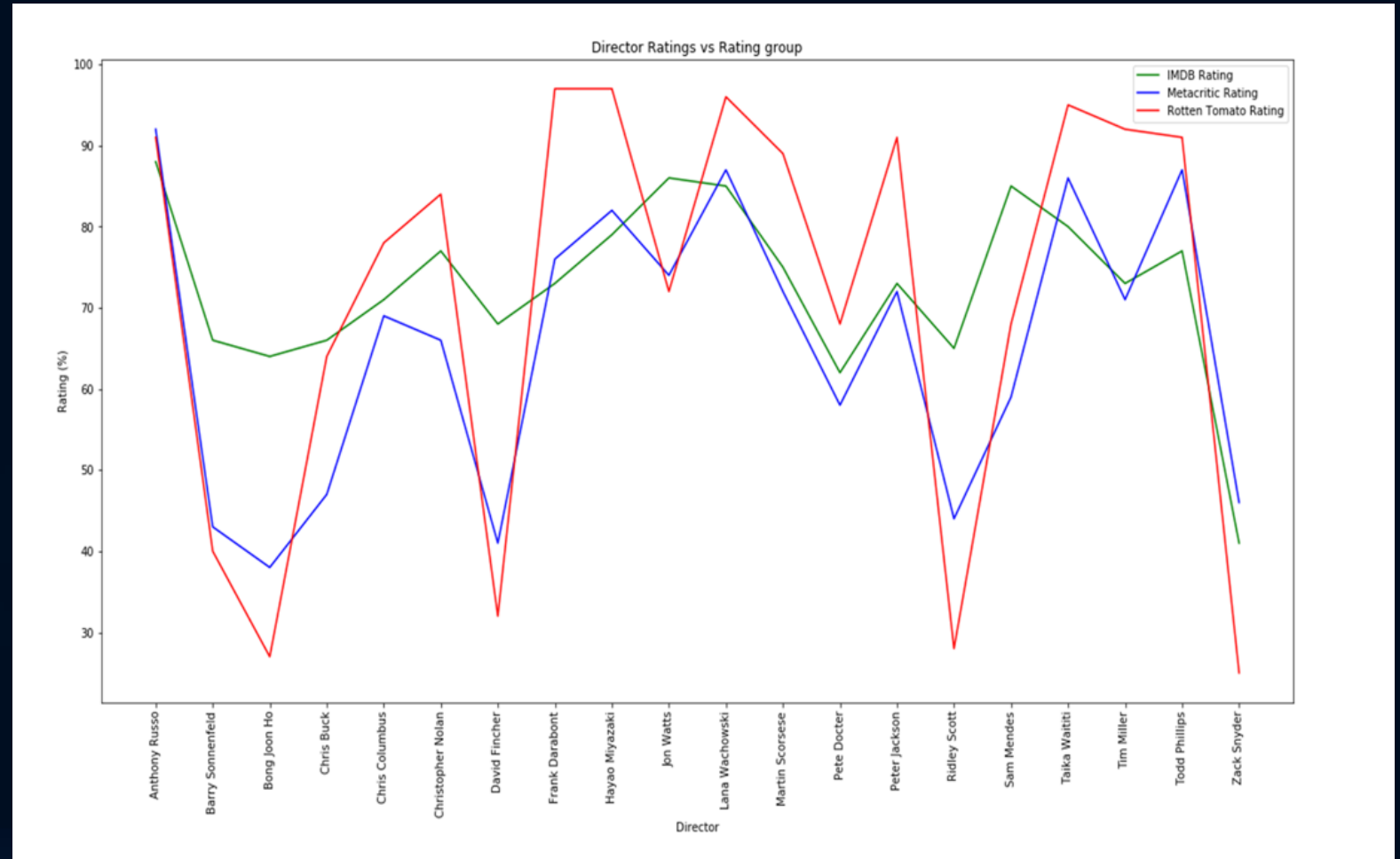
Is there any correlation between movie ratings based off directors, genre, and run time

- Explain process for determining how we'd compare all three scores
- Formatting of all three ratings similarly
- Searched for patters in models
- Findings



Did Random Lists Provide Good Movies Based on Director?

- The movie directors had to be trimmed down to a more manageable number
- Directors with only 1 movie were filtered out
- Directors had to be above a 50 rating to be considered good
- 4 of the 20 directors had very low average scores with Metacritic and Rotten Tomatoes. This may be due to their genre, but we were unable to correlate this
- Overall, if going by ratings from all 3 rating types 80 directors are considered good



Data Cleanup

- Exploration and cleanup
- Describe insights not anticipated:
- Problems that arose and how they were resolved
- ~~• Discuss interesting figures we developed. Can use help of Jupyter notebook~~

Exploration and cleanup

- Exploration: Movie year, genre, director, ratings (Rotten Tomatoes, Metacritic, IMDB), and runtime were reviewed for correlations to see if the movies were considered good
- Clean up
 - Movie year, genre, and director had to be reduced in number due the number of values that were returned with low values, no values, N/A and none. This makes for cleaner and easier to read graphs
 - Director and genre returned multiple values in a single column of a database and had to be split and the first row from the split was kept for simplicity
 - Numeric values had to be converted to integers and floats because they were represented by strings in the data drawn from the OMDB website
 - When comparing director/ratings and genre/ratings, the mean of each rating type was calculated to keep the graphs a manageable size

Unanticipated Insight

According to Rotten Tomatoes and Metacritic, viewers do not like the horror genre

- Slide 7 show a major dip in ratings for Bong Joon Ho, David Fincher, Ridley Scott and Zack Snyder
- Due to the way the genres were split in clean up, this correlation would have been missed
- A quick internet search shows that all these directors are best known for horror movies

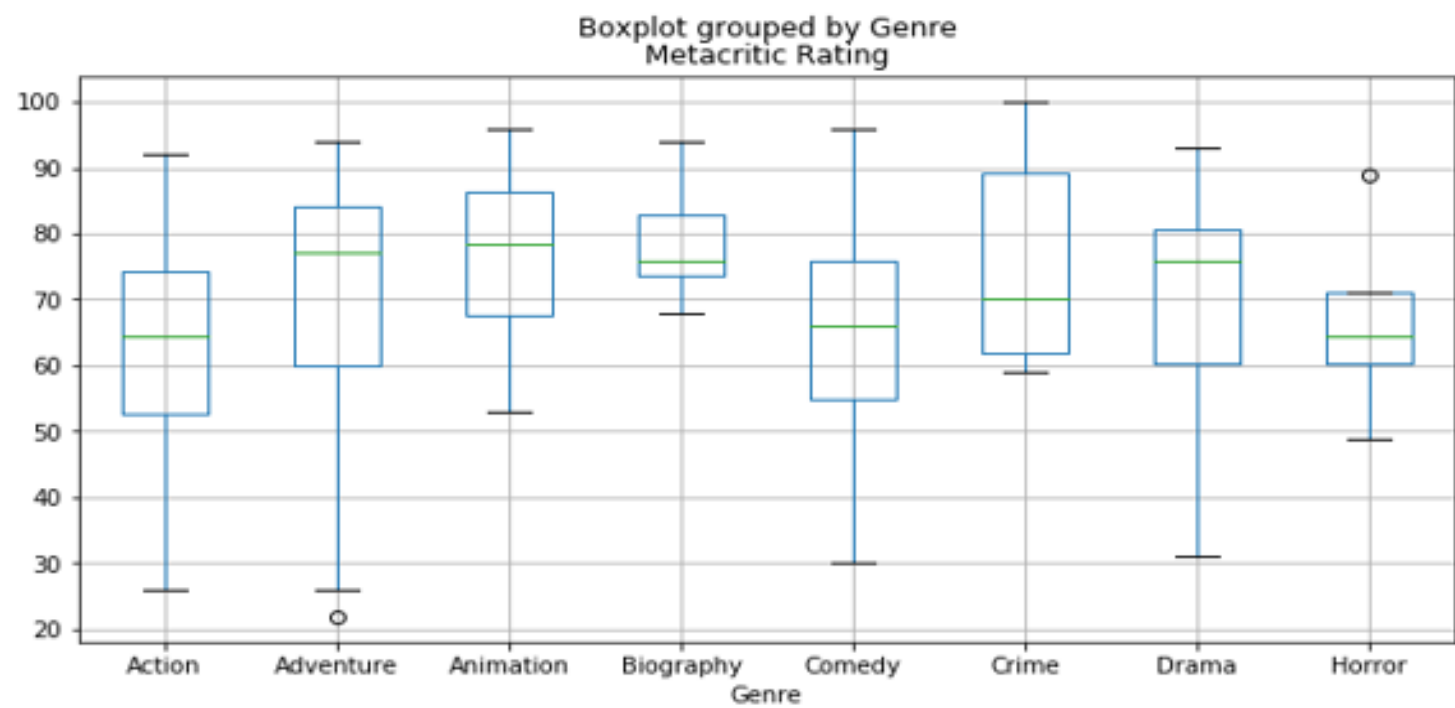
Problems

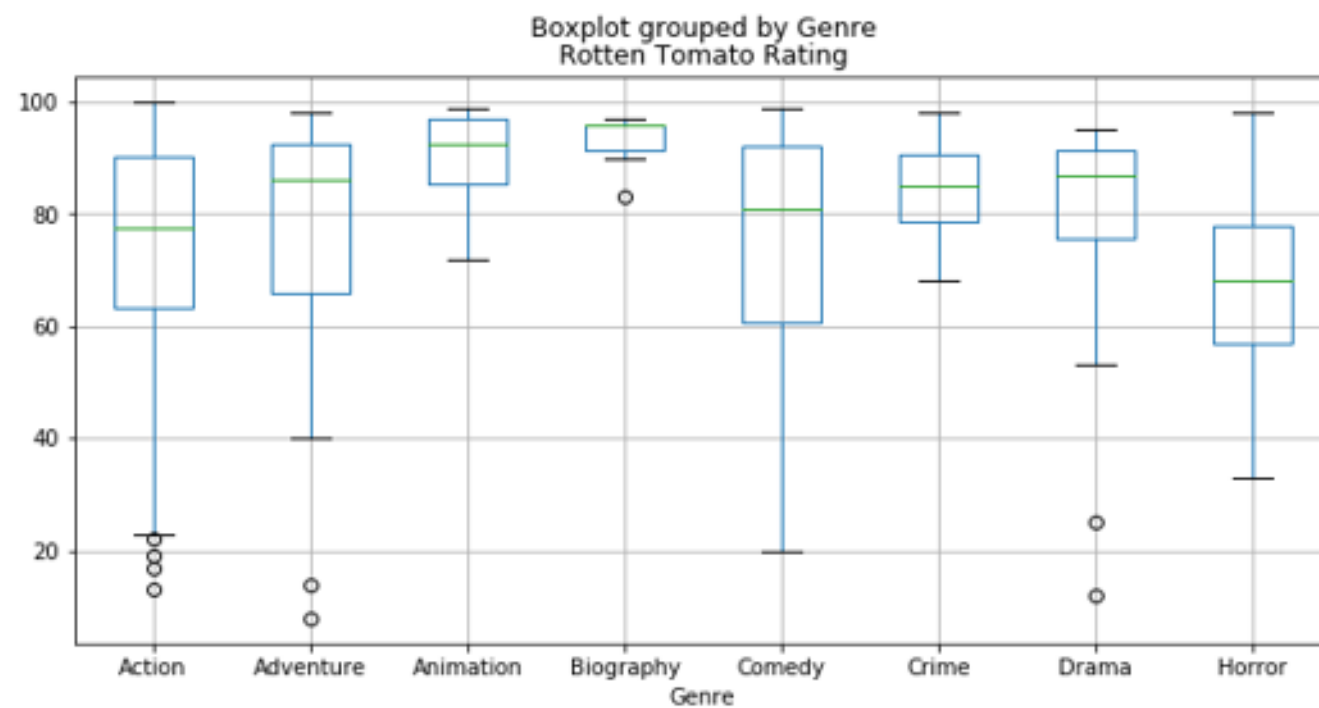
Most of the issue that were run into were coding issues:

- Making sure that data values were properly converted to numbers from strings
- Using code in the correct format (ex: splitting columns)
- Making sure variables created were used in sequence and adjusted for in upstream and downstream code

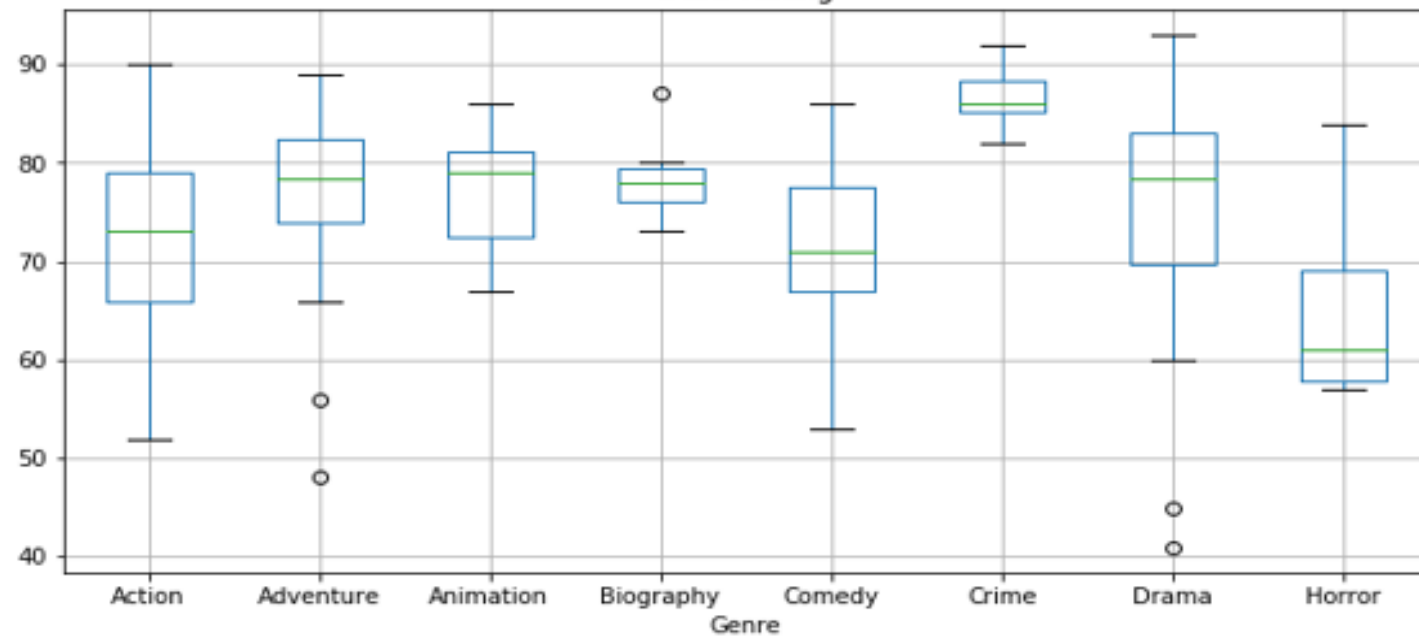
Data Analysis

- Discuss the steps we took to analyze our data and answer each question
- Also use charts/figures here to help explain. Can put visualizations on add'l slides if needed





Boxplot grouped by Genre
IMDB Rating



Conclusion/Q&A

-Discuss our findings. Did we find what we expected? If not, why?
What conclusions can we draw from analysis?

-Describe any difficulties, and how we resolved.
What other questions would we ask, but couldn't with data?
Also, what would we ask if we had more time?
(Ex: 2 add'l weeks)

-Q/A

