

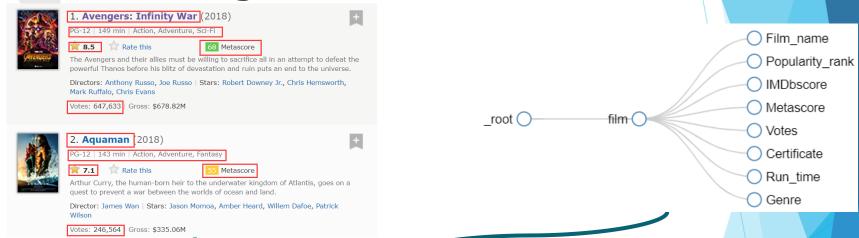
## Predict the IMDb score of films released in 2018

Stat 418 project

Zhou, Yichen



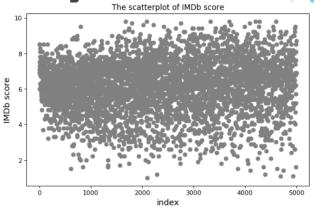
## Attaining Data - Web Scraper

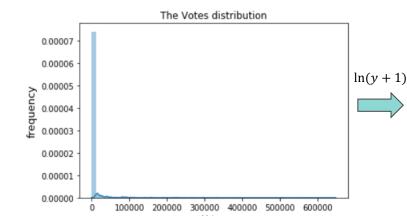


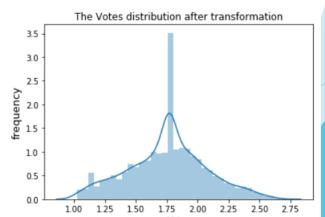
web- scraper- order	web-scraper-start-url	Film_name	Popularity_rank	IMDbscore	Metascore	Votes	Certificate	Run_time	Genre
o 1557261010- 11301	https://www.imdb.com/search/title? title_type=f	Avengers: Infinity War	1	8.5	68.0	647469.0	PG-12	149 min	Action, Adventure, Sci-Fi
1 1557261010- 11302	https://www.imdb.com/search/title? title_type=f	Aquaman	2	7.1	55.0	246474.0	PG-12	143 min	Action, Adventure, Fantasy
2 1557261010- 11303	https://www.imdb.com/search/title? title_type=f	Arctic	3	6.9	71.0	13690.0	PG-13	98 min	Adventure, Drama

## Exploratory data analysis

<class 'pandas.core.frame.DataFrame'> RangeIndex: 5000 entries, 0 to 4999 Data columns (total 10 columns): web-scraper-order 5000 non-null object web-scraper-start-url 5000 non-null object Film name 5000 non-null object Popularity rank 5000 non-null int64 IMDbscore 4459 non-null float64 Metascore 563 non-null float64 Votes 4459 non-null float64 Certificate 1367 non-null object 4284 non-null object Run time 4989 non-null object Genre dtypes: float64(3), int64(1), object(6) memory usage: 390.7+ KB







## **Further Research**

- Input: film information data(Popularity rank, Meta score, Votes, Certificate, Run time, Genre)
- Output: estimation of the IMDb score of the film.
- Complete the EDA: Genre variable, plots to show relationships between some variables
- Model Fitting: KRR, XGBoost.
- Model Evaluating: Loss function = MSE, R-Squared score