The Automated Script Reviewer

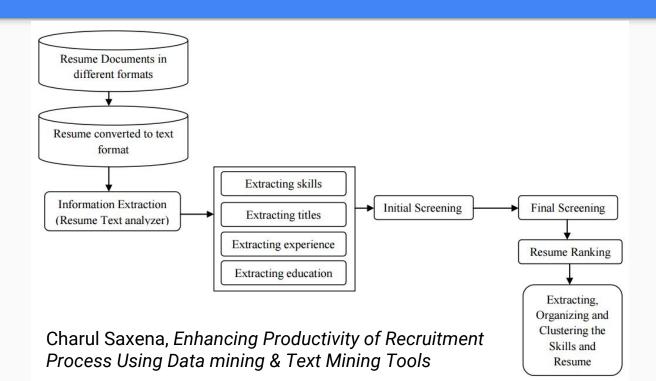
Exploring the relationship between scripts and ratings



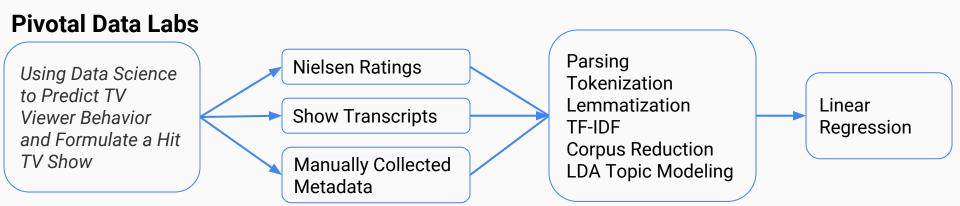


Katherine Schinkel & Marcus Rosti

The Resume Reviewer



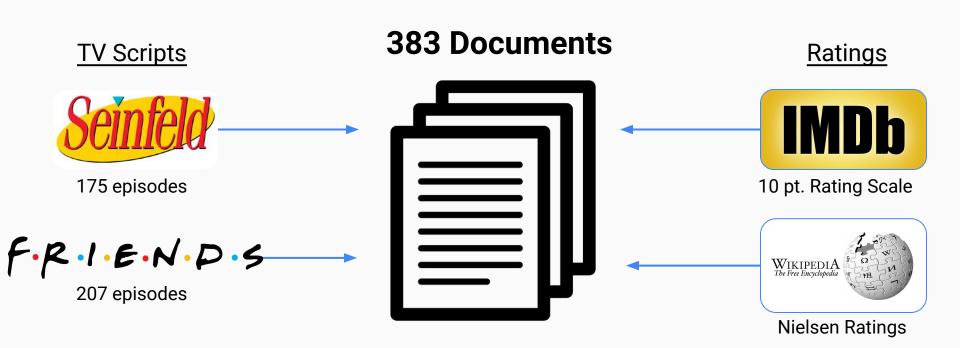
Can this method work for television scripts?



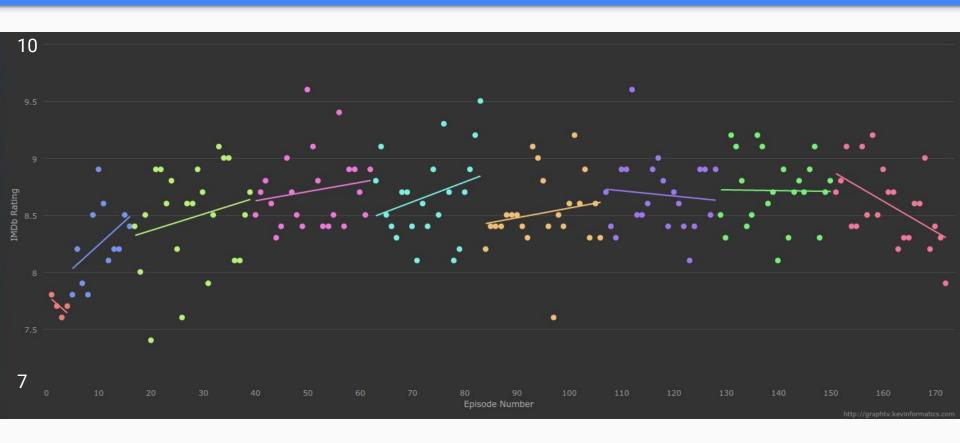
"the most relevant variables were derived from the transcript data"

https://blog.pivotal.io/data-science-pivotal/case-studies/using-data-science-to-predict-tv-viewer-behavior-and-formulate-a-hit-tv-show

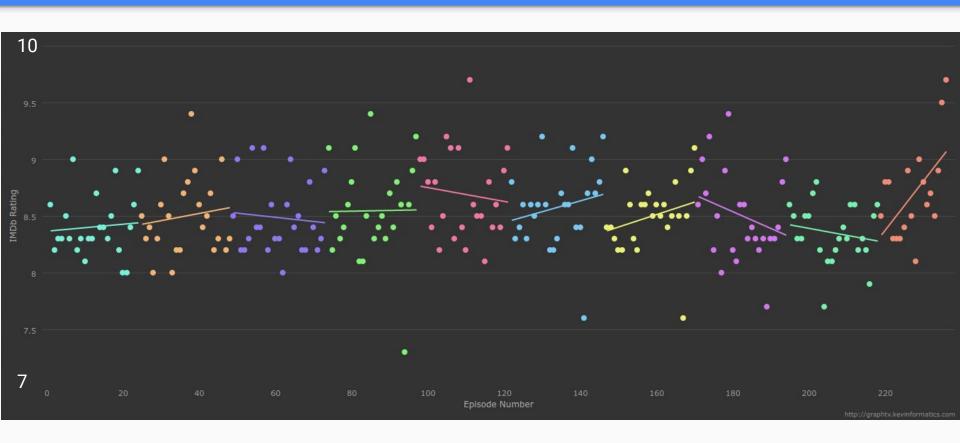
Data Collection



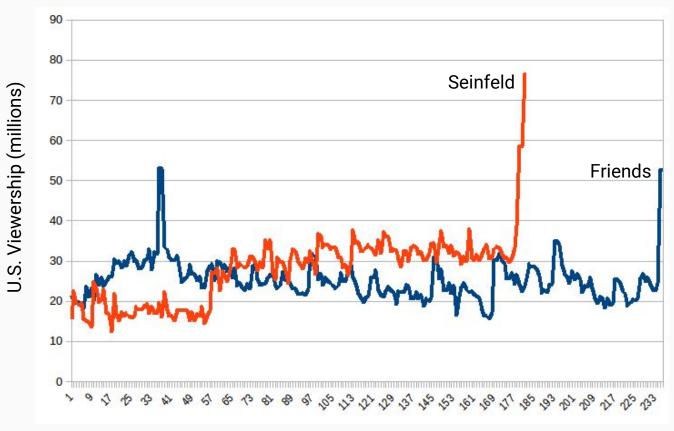
Seinfeld IMDb Ratings



Friends IMDb Ratings

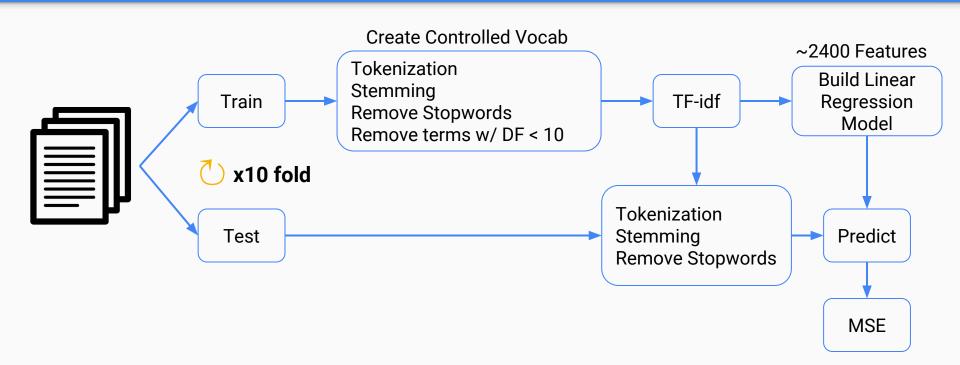


Nielsen Ratings



Episode Number - by Release Date

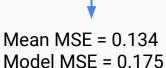
Base Modeling Approach



Base Modeling Evaluation



10 pt. Rating Scale



t = -1.713

Paired t-test

$$\frac{\bar{Y}_2 - \bar{Y}_1}{\sqrt{\sigma_1^2/n + \sigma_2^2/n}}$$

H₀: Model MSE = Mean MSE H_A: Model MSE < Mean MSE

We want t > 1.812

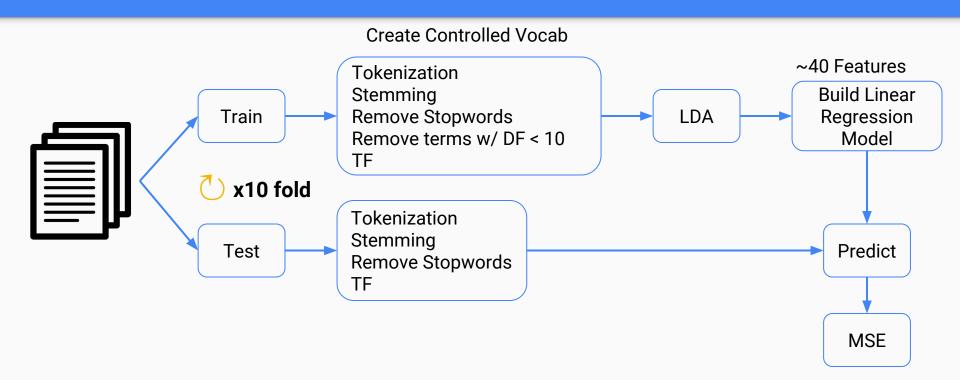


Nielsen Ratings



Mean MSE = 50.018 Model MSE = 70.269 t = -1.226

LDA Modeling Approach



LDA Modeling Evaluation



10 pt. Rating Scale



Mean MSE = 0.133 Model MSE = 0.132 t = 0.062

Paired t-test

$$\frac{\bar{Y}_2 - \bar{Y}_1}{\sqrt{\sigma_1^2/n + \sigma_2^2/n}}$$

H₀: Model MSE = Mean MSE H_A: Model MSE < Mean MSE

We want t > 1.812



Nielsen Ratings



Mean MSE = 45.858 Model MSE = 44.872 t = 0.096

Possible Improvements

Additional Feature Engineering (scene count, characters per scene count, characters per episode count)

Additional Modeling Techniques (Regression Trees, Lasso, ElasticNet Regression)

Consider multicollinearity of regression features

Use a classification approach with binning

Incorporate bigrams and/or trigrams

Questions?