# Classification of Film Reviews using Machine Learning

MINI PROJECT

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#### What

Given a movie review, analyze sentiment to predict the star rating.

- Without using NLP techniques, find best machine learning approach
- Evaluate performance of learning models with various parameters
- Create webapp to perform prediction

# Why

- Quantify reviews that don't have ratings
- Analyze comments about movie trailers/previews to gauge public opinion

- NLP approaches often don't generalize while machine learning might
- Solution may be domain specific, hence needs parameter tuning

#### How

- ▶ Scikit-learn
  - ► MultinomialNB
  - ▶ LinearSVC
  - ▶ LinearSVR
  - LogisticRegression
  - ▶ LinearRegression
- IPython (Jupyter) for experiments
- ► Flask for webapp
- ► ACL IMDB Dataset

## ACL IMDB dataset

- ▶ 50,000 reviews classified as pos/neg and split into test/train
- Directory structure is perfect for scikit-learn
- We reorganize into star categories (Filenames are <id>\_<stars>.txt)
- Form aggregate dataset for custom splits

- ▶ 5 & 6 star reviews not available
- ▶ 1 & 10 star reviews 2x as many as other classes

### A few terms

- Classification: From a set of distinct target classes, assign one to the input
- Regression: Given the input, predict the output as a continuous variable
- Accuracy: How many inputs are correctly classified
- ▶ Confusion matrix: Shows for each class, how many of its inputs were classified as which class. Can be more useful than accuracy.

## Searching for the best models

- First as positive/negative sentiment, then as stars
- Experiment workflow
  - 1. Load training and testing sets
  - 2. Convert reviews to tf-idf vectors
  - Grid-search chosen algorithm on all combinations of relevant params
  - 4. Test the optimal model and record metrics

Notebooks

# Command line app

- ▶ Framework that uses scikit-learn to
  - ▶ Train classifiers and dump to disk (python pickle)
  - Classify a review using a dumped classifier
  - ▶ Various other utilities to deal with the dataset and dumps
  - ▶ Can be used as a library or standalone

Code walkthrough

## Dumps

- Python allows serialization of objects
- Instead of training classifier before every prediction, load from disk
- Dumping process is very slow
- Dumps need large amounts of storage space
- ▶ Loading is slow but faster than training

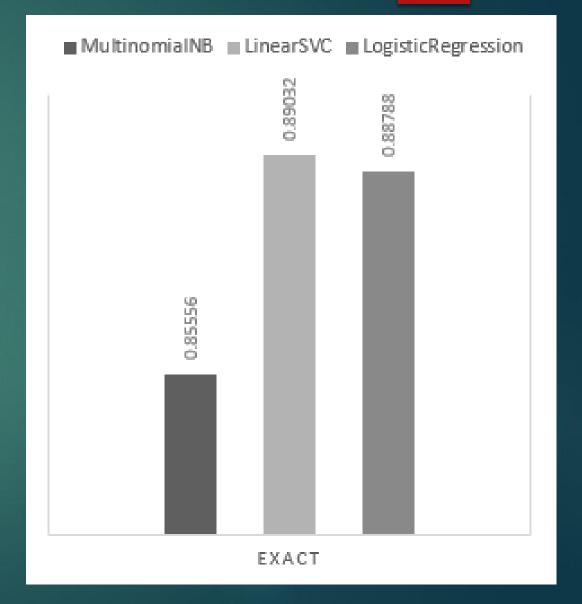
# Webapp

- ▶ Loading dumped classifiers causes delay
- Solution: load once and provide an interface
- Models used
  - ► LinearRegression for star ratings
  - ▶ LinearSVC for positive/negative
- ▶ Flask for backend
- Jinja template for frontend

Webapp demo

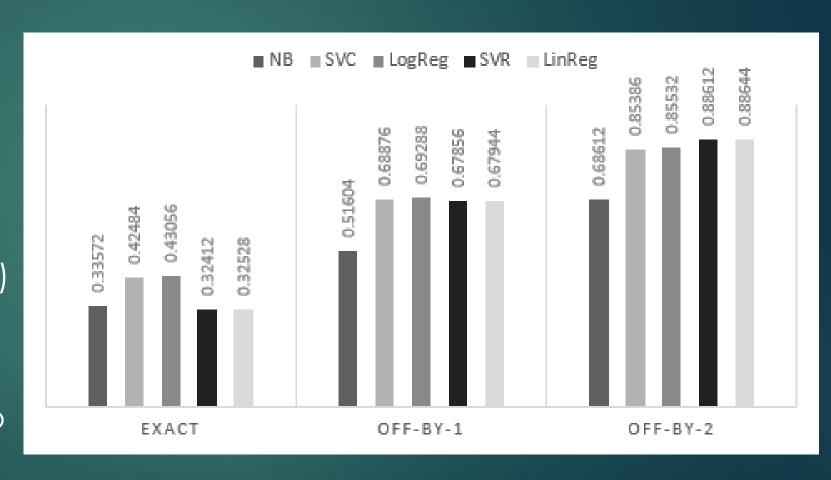
## Results – Binary classification

- All 3 classifiers (MultinomialNB, LinearSVC, LogisticRegression) perform well
- ▶ LinearSVC best at 89% accuracy
- Satisfactory solution reached



#### Results – Star classification

- Regressors perform poorly with exact accuracy in mind
- Regressors perform acceptably if off-byone errors are allowed (LinearRegression: 68%)
- Positive/negative distinction maintained
- But accuracy needs to be improved



Thank you