

Presented By:

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DIRECTOR

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## OBJECTIVE



To create a model that would be able to predict movies that have the greatest chance of winning Best Picture at the Academy Awards

## DATA SOURCES

The Oscar Awards have their own database with data from 1927-2023.

To optimize our time...we opted for a CSV file of these results from Kaggle.

To get more data, we ran each Best Picture nominee + winner through the OMDB API service. HOME / OSCARS / AWARDS DATABASES

### RESULTS



#### 1927/28 (1st)

#### **OUTSTANDING PICTURE**

The Caddo Company -- The Racket

Fox -- 7th Heaven

🖈 Paramount Famous Lasky -- Wings

#### 1928/29 (2nd)

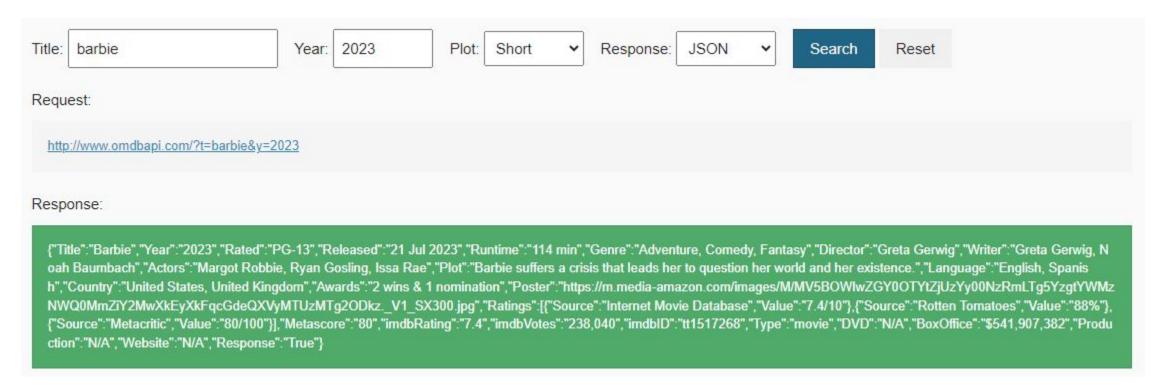
#### **OUTSTANDING PICTURE**

Feature Productions -- Alibi

[NOTE: THIS IS NOT AN OFFICIAL NOMINATION. There were no announcements of nominations, no certificates of nomination or honorable mention, and only the winners (\*) were revealed during the awards banquet on April 3, 1930. Though not official nominations, the additional names in each category, according to in-house records, were under consideration by the various boards of

# DATA SOURCES

### **OMDB** Search Result example:



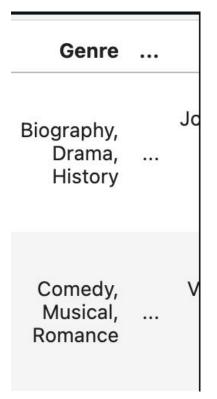
## FIRST CHALLENGE

- The title of "winner" at the Academy Awards has gone by different names over the years:
  - \* "Best Picture", "Outstanding Motion Picture", "Outstanding Picture", "Outstanding Production", "Best Motion Picture"
- Successfully able to filter the results to "Best Picture" and all previous aliases
- Compared our API Key with data frame



# DATA CLEANING

## GENRE COLUMN HAD MULTIPLE DATA PIECES



# USED THE SPLIT FUNCTION TO SEPARATE GENRES BUT...MORE CHALLENGES!

| Romance | Western | <br>Comedy | NaN | Fantasy | Sci-<br>Fi | Action | Drama | Short | Film-<br>Noir | Animation | Crime |
|---------|---------|------------|-----|---------|------------|--------|-------|-------|---------------|-----------|-------|
| 0       | 0       | <br>0      | 0   | 0       | 0          | 0      | 1     | 0     | 0             | 0         | 0     |
| 1       | 0       | <br>1      | 0   | 0       | 0          | 0      | 0     | 0     | 0             | 0         | 0     |
| 0       | 0       | <br>0      | 0   | 0       | 0          | 0      | 1     | 0     | 0             | 0         | 0     |
| 1       | 0       | <br>0      | 0   | 0       | 0          | 0      | 1     | 0     | 0             | 0         | 0     |
| 0       | 0       | <br>0      | 0   | 0       | 0          | 0      | 1     | 0     | 0             | 0         | 0     |

## DATA CLEANING CONT'D

- New challenge: NaN value was being assigned to a column
- Updated columns, successfully dropped NaN

```
df 1.columns
Index(['year_ceremony',
                                                                   'Runtime',
                                'winner',
                                                   'Rated',
                                               'imdbVotes',
           'Metascore',
                            'imdbRating',
                                                                 'BoxOffice',
            'Thriller',
                               'Family',
                                                  'Short',
                                                                   'Romance',
                               'Mystery',
             'Fantasy',
                                                 'Comedy',
                                                                   'Horror',
           'Animation',
                                'Music',
                                                                   'Action',
                                                 'Western',
                               'History',
                                                 'Musical',
                                                                    'Drama',
                   nan,
                             'Film-Noir',
                                                                       'War',
           'Biography',
                                                  'Crime',
              'Sci-Fi',
                                 'Sport',
                                               'Adventure'],
      dtype='object')
```

## MORE DATA CLEANING

 One aspect of data we needed to account for was some movies were duplicates because of remakes (i.e. "A Star Is Born")

|     | year_film | film           | Title/Year          |
|-----|-----------|----------------|---------------------|
| 71  | 1937      | A Star Is Born | A Star Is Born-1937 |
| 544 | 2018      | A Star Is Born | A Star Is Born-2018 |

## MORE DATA CLEANING

- Pieces of data that needed to be changed into integers include Runtime, IMDB Votes, and Box Office
- The dollar sign in Box Office needed to be removed
- We also needed to change much of our data into integers

```
# Remove "," and "$" from BoxOffice column so that we can change it to an integer
def BoxOffice_cleaning(dataframe):
    dataframe["BoxOffice"] = dataframe["BoxOffice"].str.replace(",","")
    dataframe["BoxOffice"] = dataframe["BoxOffice"].str.replace("$","")
    print(dataframe["BoxOffice"].unique())
BoxOffice_cleaning(combined_df)
```

| Runtime | imdbVotes | BoxOffice     |  |  |
|---------|-----------|---------------|--|--|
| 151 min | 98,933    | \$17,348,945  |  |  |
| 158 min | 80,669    | \$6,773,650   |  |  |
| 130 min | 627,105   | \$718,732,821 |  |  |
| 147 min | 150,623   | \$4,608,096   |  |  |
| 104 min | 33,492    | \$5,456,531   |  |  |
|         |           |               |  |  |

## FIRST MODEL ATTEMPT

Model Type: Logistic regression

Data Included: Runtime, Metascore, imdbRating, imdbVotes, BoxOffice, Genre, Rating, Year of Ceremony

#### Confusion Matrix

|               | Predicted loser | Predicted winner |  |
|---------------|-----------------|------------------|--|
| Actual loser  | 111             | 0                |  |
| Actual winner | 21              | 0                |  |

### Classification Report

|              | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| loser        | 0.84      | 1.00   | 0.91     | 111     |
| winner       | 0.00      | 0.00   | 0.00     | 21      |
| accuracy     |           |        | 0.84     | 132     |
| macro avg    | 0.42      | 0.50   | 0.46     | 132     |
| weighted avg | 0.71      | 0.84   | 0.77     | 132     |

## MODEL OPTIMIZATION

- Model types: Logistic Regression + Random Forest
- Data added: Golden Globes results, Director, Producer, Country
- Data transformed: Standard Scaling on numeric columns
- Data removed: Box Office
- Data limited: Years of Ceremony (past 50 years only, 1944 2020 only)

## BEST MODEL ATTEMPT

Method: logistic regression

Data Transformation: Standard Scaling on numeric columns.

Data added: Golden Globes results

Data limited: Years 1944-2020

#### Confusion Matrix

|               | Predicted loser | Predicted winner |
|---------------|-----------------|------------------|
| Actual loser  | 38              | 3                |
| Actual winner | 9               | 5                |

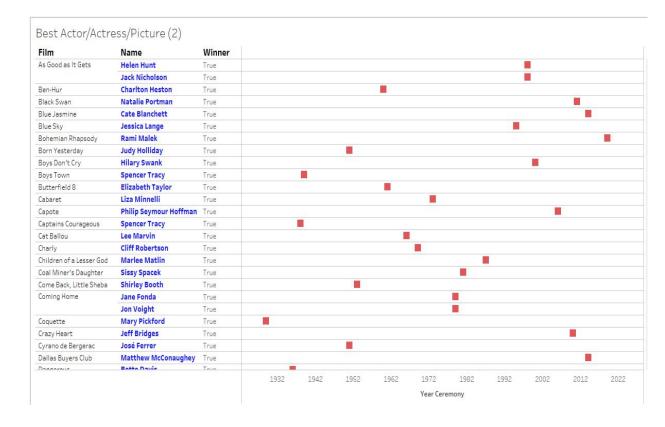
Accuracy Score : 0.7818181818181819

Balanced Accuracy Score : 0.64198606271777

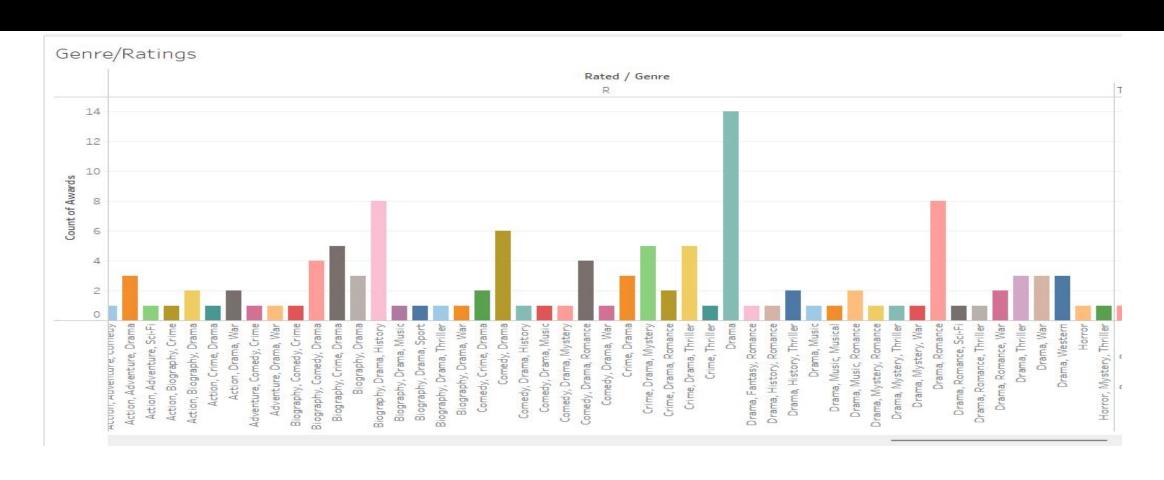
| Classification | n Report  |        |          |         |
|----------------|-----------|--------|----------|---------|
|                | precision | recall | f1-score | support |
| loser          | 0.81      | 0.93   | 0.86     | 41      |
| winner         | 0.62      | 0.36   | 0.45     | 14      |
| accuracy       |           |        | 0.78     | 55      |
| macro avg      | 0.72      | 0.64   | 0.66     | 55      |
| weighted avg   | 0.76      | 0.78   | 0.76     | 55      |

# DATA INVESTIGATION WITH TABLEAU

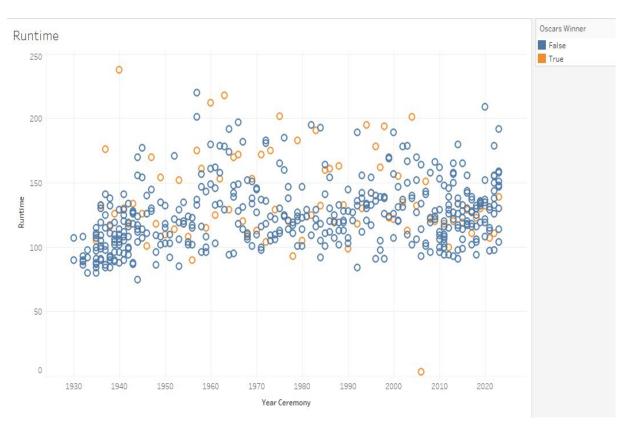
| Filters                  | Best Pic  | ture                           |                     |              |   |                     |                        |
|--------------------------|-----------|--------------------------------|---------------------|--------------|---|---------------------|------------------------|
| Category<br>Winner: True | Year Cere | Eilen                          | BEST MOTION PICTURE | BEST PICTURE | Winner / Category True OUTSTANDING MOTION PIC | OUTSTANDING DISTURE | OUTSTANDING DRODUSTION |
|                          |           | Wings                          | BEST MOTION PICTORE | DEST PICTORE | OUTSTANDING WOTTON PIC                        | OUTSTANDING PICTORE | OUTSTANDING PRODUCTION |
| Marks                    | 1/1/1929  | The Broadway Melody            |                     |              |   |                     |                        |
| ☐ Square ▼               | 1/1/1930  | All Quiet on the Western Front |                     |              |   | -                   |                        |
|                          | 1/1/1931  | Cimarron                       |                     |              |   |                     |                        |
| : 0 I                    | 1/1/1932  | Grand Hotel                    |                     |              |   |                     |                        |
| Color Size Label         | 1/1/1933  | Cavalcade                      |                     |              |   |                     |                        |
| · Q                      | 1/1/1935  | It Happened One Night          |                     |              |   |                     |                        |
| Detail Tool              | 1/1/1936  | Mutiny on the Bounty           |                     |              |   |                     | -                      |
| Catego                   | 1/1/1937  | The Great Ziegfeld             |                     |              |   |                     |                        |
| o Film                   | 1/1/1938  | The Life of Emile Zola         |                     |              |   |                     |                        |
|                          | 1/1/1939  | You Can't Take It with You     |                     |              |   |                     |                        |
| Name                     | 1/1/1940  | Gone with the Wind             |                     |              |   |                     |                        |
|                          | 1/1/1941  | Rebecca                        |                     |              |   |                     |                        |
|                          | 1/1/1942  | How Green Was My Valley        |                     |              |   |                     |                        |
|                          | 1/1/1943  | Mrs. Miniver                   |                     |              |   |                     |                        |
|                          | 1/1/1944  | Casablanca                     |                     |              |   |                     |                        |
|                          | 1/1/1945  | Going My Way                   | •                   |              |   |                     |                        |
|                          | 1/1/1946  | The Lost Weekend               |                     |              |   |                     |                        |
|                          | 1/1/1947  | The Best Years of Our Lives    |                     |              |   |                     |                        |
|                          | 1/1/1948  | Gentleman's Agreement          |                     |              |   |                     |                        |
|                          | 1/1/1949  | Hamlet                         |                     |              |   |                     |                        |
|                          | 1/1/1950  | All the King's Men             |                     |              |   |                     |                        |
|                          | 1/1/1951  | All about Eve                  |                     |              |   |                     |                        |
|                          | 1/1/1952  | An American in Paris           |                     |              |   |                     |                        |
|                          | 1/1/1953  | The Greatest Show on Earth     |                     |              |   |                     |                        |
|                          | 1/1/1954  | From Here to Eternity          |                     |              |   |                     |                        |

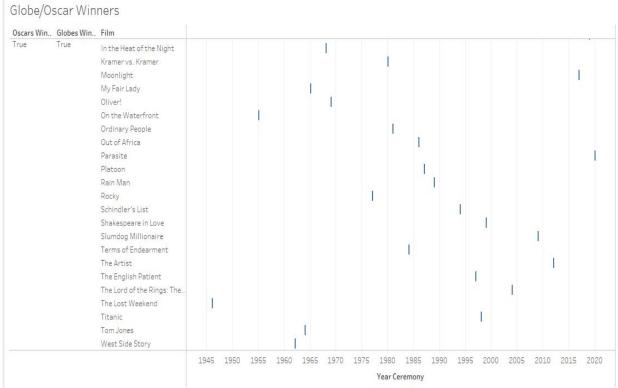


# DATA INVESTIGATION WITH TABLEAU



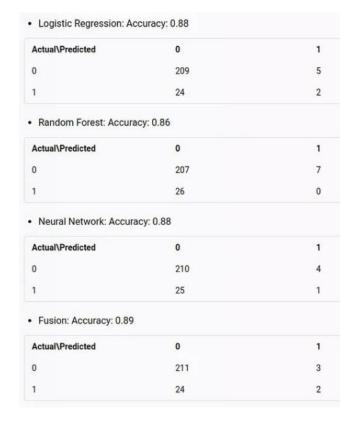
# DATA VISUALIZATION WITH TABLEAU





# COMPARING TO ANOTHER MODEL

- Difficult data to predict consider first challenge (names of categories changing)
- Importance of categories for predictions(our comparison model used 150+ categories)
- Sentiment Score





### 2024 Oscar Awards...what will win Best Picture?



## No hope for any of these?!

| Title                               | Prediction                    |
|-------------------------------------|-------------------------------|
| Barbie                              | It's an honor to be nominated |
| Oppenheimer                         | It's an honor to be nominated |
| Past Lives                          | It's an honor to be nominated |
| Spider-Man: Across the Spider-Verse | It's an honor to be nominated |
| Air                                 | It's an honor to be nominated |



But wait! Our model includes Golden Globes results...and the 2024 Golden Globes haven't happened yet either. Instead of telling our model that these movies didn't win at the Globes, what if we told the model that they did win?!

After all, it is possible for 3 of these 5 movies to ALL win at the Globes since the Globes splits Best Picture into several categories including Best Drama, Best Comedy/Musical, and Best Animated Film.

## And the winner is...

| Prediction                    |
|-------------------------------|
| It's an honor to be nominated |
| WINNER                        |
| It's an honor to be nominated |
| It's an honor to be nominated |
| It's an honor to be nominated |
|                               |

# Oppenheimer!

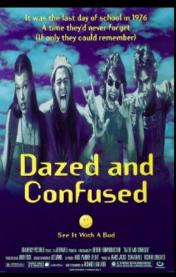


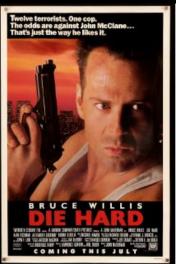
## But the Oscars don't always get it right either...

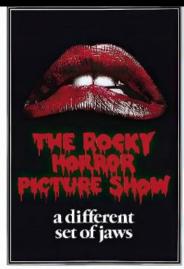
So we're going to run the model on some movies that we think should have been nominated. (Based on their representation of being classic "snubs")













## And the winner is...

| Prediction                    |
|-------------------------------|
| It's an honor to be nominated |
| WINNER                        |
|                               |







Singin' in the Rain!



# The Academy Awards





