# Predicting Opening Weekend Box Office Performance

From IMDB Search Frequency of Principal Cast Members

Christopher Giler



January 26, 2018

### Agenda

- 1. Project Objective
- 2. Data Overview
- 3. Key Features
- 4. Exploratory Data Analysis
- 5. Feature Engineering
- 6. Modeling Box Office Sales
- 7. Conclusion & Next Steps

### 1 Project Objective

Defining Project Goals



### The Problem Statement

How much does casting affect the hype surrounding a movie's release?

 Evaluate the impact of cast popularity on opening weekend box office ticket sales.



### 2 Data Overview

Scraping and Cleaning from Data Sources



### 2014 ~ 2017

4 years of box office data

### 994

Movies available w/ box office data

### 8

Total features for model (raw)

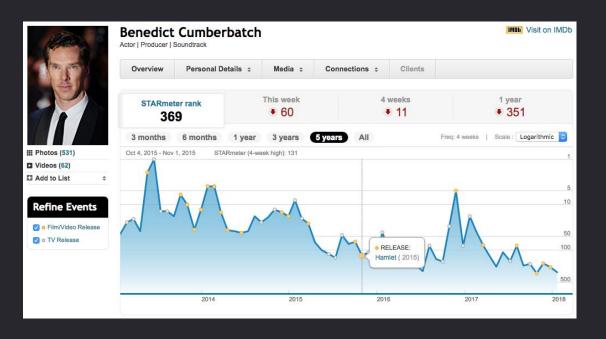
#### IMDB

- Release Date
- Opening Weekend Gross
- Metacritic / IMDB Reviews
- Genre
- MPAA Rating
- Principal Cast

### IMDB Pro

- STARmeter Data (next slide)
- Number Theaters

### **IMDB Pro's STARmeter**



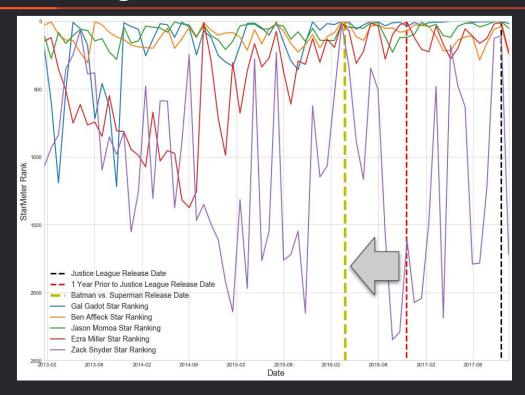
- Cast / Crew ranking based on IMDB search frequency
- Presented as time series data
- Scraped via Python / Selenium

### 3 Key Features

From IMDB Pro



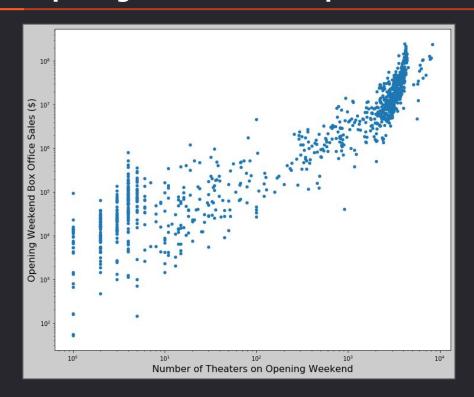
### Using IMDB Pro's StarMeter Data



 Feature defined as minimum star ranking prior to 12 months before release date.

A film's "Star Power" is based on average of top 5 star rankings

### **Opening Weekend Gross per Theater**



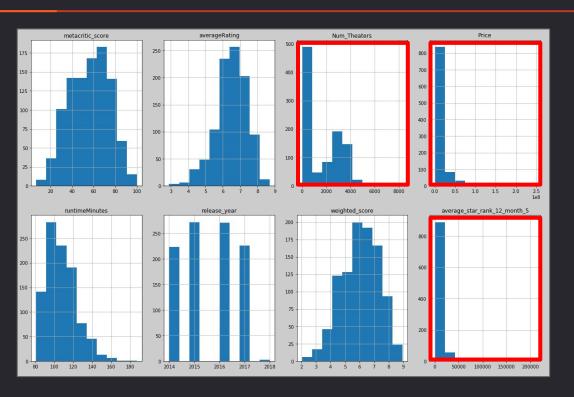
 Number of theaters is highly correlated with overall opening weekend gross

### 4 Data Analysis

Exploring the Data



### **Continuous Features**

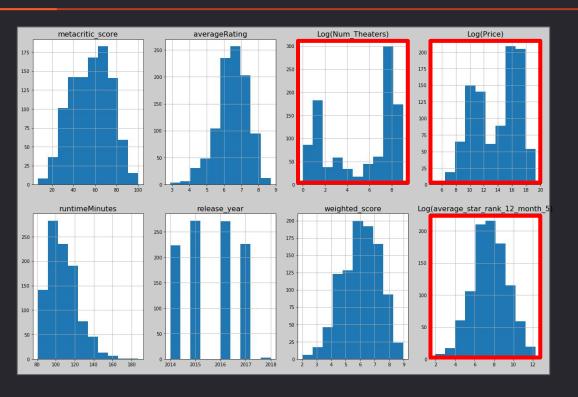


 All continuous features should have normal distribution.

### Issues:

- Opening WeekendGross
- Number of Theaters
- Star Power

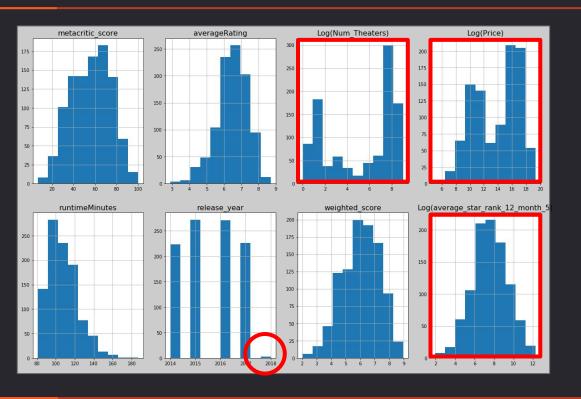
### **Continuous Features**



 All continuous features should have normal distribution.

 Logarithmic transform applied to skewed data.

### **Continuous Features**

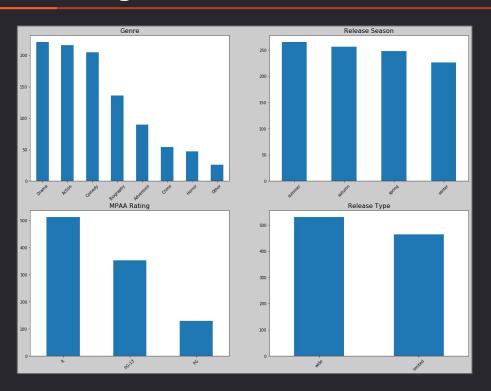


 All continuous features should have normal distribution.

Logarithmic transform applied to skewed data.

Removed all 2018 data points

### **Categorical Features**



- Movie Genres
   (Action, Drama, Comedy, etc.)
- Release Season (Spring, Summer, Autumn, Winter)
- MPAA Rating (PG, PG-13, R)
- Film Release Type (Wide vs. Limited)

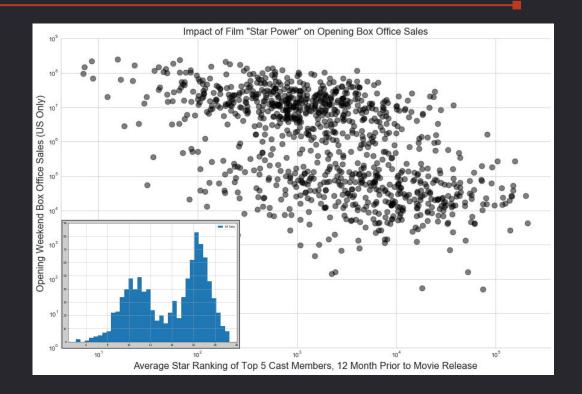
### 5 Feature Engineering

Modifying Model Features



### **Star Power**

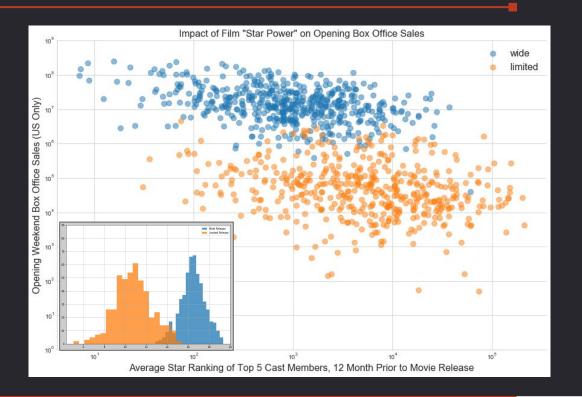
Opening Weekend
 Gross vs. Star
 Ranking alone does
 not show a clear
 linear relationship.



### **Star Power**

Opening Weekend
 Gross vs. Star
 Ranking alone does
 not show a clear
 linear relationship.

"Limited Release":# Theaters < 600</li>

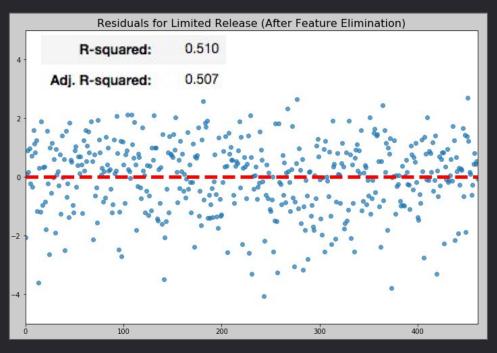


## 6 **Modeling Box Office Sales**

Final Model Results

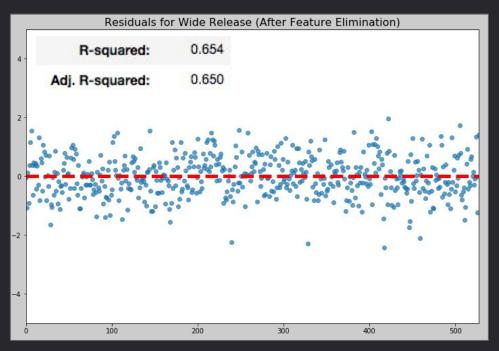


### **Limited Release - Feature Selection**



- Model trained on all data.
- Features selected based on individual p-values.
- Remaining features:
  - Num\_Theaters
  - Runtime\_Minutes
  - Star\_Power

### Wide Release - Feature Selection



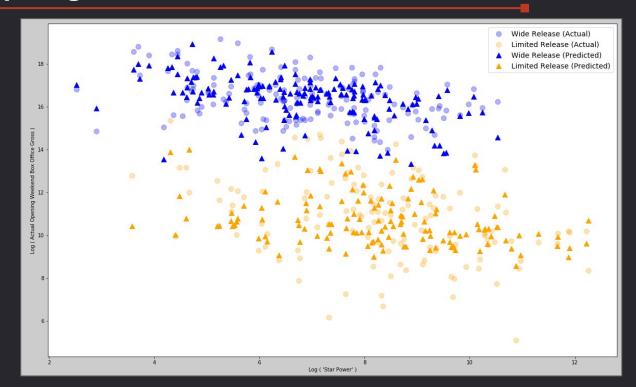
- Model trained on all data.
- Features selected based on individual p-values.
- Remaining features:
  - Num\_Theaters
  - Runtime\_Minutes
  - Star\_Power
  - Release\_Year
  - Rated\_R
  - Release\_in\_Spring

### Predicted vs. Actual Opening Weekend Gross

70-30 random train-test split

Test data shown

Predicted results within expected range



### Residuals for Train/Test Split

- Model trained 70% of data, and validated on remaining 30%.
- Cross-Validation score aligns with observed results.

Regression Model	5-Fold Cross-Validation R^2 Score
All Data (Base)	0.16
Limited Release	0.49
Wide Release	0.64



### **Applying Predictive Model**

2018 Opening Weekend Box Office Predictions

Movie Title	Release Date	Predicted Gross	Actual Gross	Release Type	% Error
The Post	2018-01-12	\$19,783,540	\$19,887,979	wide	-0.53 %
Paddington 2	2018-01-12	\$18,109,110	\$11,001,961	wide	64.60 %

### 7 Conclusion

And Next Steps



#### **Conclusions**

### **Results**

- "Star Power" is a factor in predicting opening weekend box office performance.
- Splitting data into two models proved effective.
- Genre removed as feature

### **Next Steps**

- New features to consider:
  - Box office competition
  - Is Sequel? (Title/Brand recognition)
  - Other metrics for cast / crew popularity (Google Trends, Twitter)

### THANKS!

ANY QUESTIONS?





### Scraping IMDB Pro's StarMeter

- Visualization uses JavaScript SVG Element
- Scraping using Selenium

### All Data - Feature Elimination

	coe	f std err	t	P> t	[0.025	0.975]
Inte	rcept 18.0114	63.439	0.284	0.777	-106.482	142.504
Q('Num_Thea	iters') -0.2987	0.028	-10.613	0.000	-0.354	-0.243
Q('runtimeMin	utes') 2.0799	0.268	7.751	0.000	1.553	2.607
Q('release_	year') -0.0079	0.031	-0.250	0.802	-0.070	0.054
Q('dum_is_Ac	ction') 0.1864	0.222	0.838	0.402	-0.250	0.623
Q('dum_is_Adven	ture') 0.3005	0.235	1.280	0.201	-0.160	0.761
Q('dum_is_Biogra	aphy') 0.0634	0.225	0.281	0.778	-0.379	0.506
Q('dum_is_Con	nedy') 0.1231	0.218	0.564	0.573	-0.305	0.551
Q('dum_is_C	rime') -0.051	0.254	-0.202	0.840	-0.549	0.447
Q('dum_is_Dr	ama') -0.1605	0.217	-0.739	0.460	-0.587	0.266
Q('dum_is_Ho	orror') 0.0166	0.260	0.064	0.949	-0.493	0.527
Q('dum_rated_Po	G-13') -0.2297	0.122	-1.879	0.061	-0.470	0.010
Q('dum_rate	ed_R') -0.5150	0.122	-4.206	0.000	-0.756	-0.275
Q('dum_release_in_sp	ring') 0.3025	0.093	3.250	0.001	0.120	0.485
Q('dum_release_in_sum	mer') 0.2754	0.092	3.000	0.003	0.095	0.456
Q('dum_release_in_w	inter') 0.1084	0.096	1.123	0.262	-0.081	0.298
Q('dum_release_lim	nited') -1.2143	0.167	-7.289	0.000	-1.541	-0.887
Q('average_star_rank_12_mon	th_5') -0.1476	0.023	-6.429	0.000	-0.193	-0.103
Omnibus: 34.397 D	urbin-Watson:	1.912				
Prob(Omnibus): 0.000 Jaro	que-Bera (JB):	40.181				
Skew: -0.402	Prob(JB):	1.88e-09				
Kurtosis: 3.574	Cond. No.	3.91e+06				

Dep. Variable	: open	ing_per_thea	ter	R-squa	red:	0.193		
Mode	l:	0	LS A	dj. R-squa	red:	0.187		
Method	ı:	Least Squa		F-stati		33.56		
Date		ri. 26 Jan 20		b (F-statis		i.02e-42		
Time	20 BA	00:02		g-Likelih		-1438.0		
No. Observations			90		AIC:	2892.		
Df Residuals			82		BIC:	2931.		
Df Mode			7		DIC.	2001.		
		nonrob						
Covariance Type	1:	nonrob	usı					
			coef	std err		t P> t	[0.025	0.975]
		Intercept	2.7809	1.248	2.22	9 0.026	0.333	5.229
-	Q('Num	Theaters')	-0.2861	0.028	-10.33	0.000	-0.340	-0.232
G	('runtim	eMinutes')	1.9218	0.251	7.65	0.000	1.429	2.415
	Q('dum	_rated_R')	-0.3633	0.071	-5.13	9 0.000	-0.502	-0.225
Q('dum_	release_	in_spring')	0.2571	0.082	3.13	3 0.002	0.096	0.418
Q('dum_re	lease_in	_summer')	0.2296	0.080	2.85	0.004	0.072	0.388
Q('dur	n_releas	e_limited')	-1.2593	0.166	-7.60	0.000	-1.584	-0.934
Q('average_star_	rank_12	_month_5')	-0.1442	0.022	-6.61	4 0.000	-0.187	-0.101
Omnibus:	28.847	Dbir. N		1,904				
		Durbin-V						
Prob(Omnibus):	0.000	Jarque-Be		33.255				
Skew:	-0.360		ob(JB):	6.01e-08				
Kurtosis:	3.535	Co	nd. No.	406.				

### **Limited Release - Feature Elimination**

	coe	f std err	t	P> t	[0.025	0.975]
Intercept	-248.3468	112.604	-2.205	0.028	-469.649	-27.044
Q('Num_Theaters')	0.6578	0.035	19.028	0.000	0.590	0.726
Q('runtimeMinutes')	2.4225	0.467	5.192	0.000	1.505	3.340
Q('release_year')	0.1226	0.056	2.196	0.029	0.013	0.232
Q('dum_is_Action')	-0.3678	0.361	-1.018	0.309	-1.078	0.342
Q('dum_is_Adventure')	0.3924	0.386	1.016	0.310	-0.367	1.151
Q('dum_is_Biography')	0.4816	0.319	1.508	0.132	-0.146	1.109
Q('dum_is_Comedy')	0.4056	0.308	1.317	0.188	-0.199	1.011
Q('dum_is_Crime')	0.1261	0.370	0.341	0.733	-0.600	0.852
Q('dum_is_Drama')	0.0119	0.305	0.039	0.969	-0.587	0.611
Q('dum_is_Horror')	-0.7249	0.452	-1.602	0.110	-1.614	0.164
Q('dum_rated_PG-13')	-0.0586	0.243	-0.241	0.809	-0.536	0.418
Q('dum_rated_R')	-0.2042	0.236	-0.866	0.387	-0.668	0.259
Q('dum_release_in_spring')	0.1934	0.152	1.276	0.203	-0.104	0.491
Q('dum_release_in_summer')	0.3278	0.155	2.118	0.035	0.024	0.632
Q('dum_release_in_winter')	0.1579	0.180	0.879	0.380	-0.195	0.511
Q('average_star_rank_12_month_5')	-0.1151	0.035	-3.267	0.001	-0.184	-0.046
Omnibus: 7.749 Durbin-V	Watson:	2.001				
Prob(Omnibus): 0.021 Jarque-Be	era (JB):	7.721				
Skew: -0.314 Pr	rob(JB):	0.0211				
Kurtosis: 3.076 Co	nd. No. 4	.09e+06				

Dep. Variable:		Price	R	-squared:	0.51	0		
Model:		OLS	Adj. R	-squared:	0.50	07		
Method	Lea	st Squares	F	F-statistic:		.7		
Date	Fri, 20	6 Jan 2018	Prob (F-	statistic):	1.57e-7	0		
Time:		00:02:10	Log-Li	kelihood:	-749.8	86		
No. Observations:		462		AIC:	150	8.		
Df Residuals:		458		BIC:	152	4.		
Df Model:		3						
Covariance Type:		nonrobust						
			coef	std err	t	P> t	[0.025	0.975]
		Intercept	-1.1327	2.109	-0.537	0.591	-5.277	3.011
(	('Num_	Theaters')	0.6559	0.033	19.590	0.000	0.590	0.722
Q	('runtim	eMinutes')	2.4769	0.442	5.605	0.000	1.609	3.345
Q('average_star_r	ank_12	_month_5')	-0.1320	0.034	-3.855	0.000	-0.199	-0.065
Omnibus:	19.133	Durbin-V	Vatson:	2.010				
Prob(Omnibus):	0.000	Jarque-Be	ra (JB):	20.372				
Skew:	-0.503	Pr	ob(JB):	3.77e-05				
Kurtosis:	3.212	Co	nd. No.	380.				

### Wide Release - Feature Elimination

			coef	std err	t	P> t	[0.025	0.975]
	Interd	ept 1	89.1722	57.438	3.294	0.001	76.329	302.015
	Q('Num_Theate	rs')	1.6340	0.077	21.183	0.000	1.482	1.786
(	Q('runtimeMinut	es¹)	1.4929	0.253	5.912	0.000	0.997	1.989
	Q('release_ye	ar')	-0.0951	0.028	-3.350	0.001	-0.151	-0.039
	Q('dum_is_Action	on')	-0.2478	0.256	-0.968	0.334	-0.751	0.255
Q('d	dum_is_Adventu	ire')	-0.2825	0.259	-1.090	0.276	-0.791	0.227
Q('d	dum_is_Biograp	hy')	-0.4186	0.272	-1.537	0.125	-0.954	0.116
Q	('dum_is_Come	dy')	-0.3339	0.261	-1.279	0.202	-0.847	0.179
	Q('dum_is_Crin	ne')	-0.2966	0.293	-1.014	0.311	-0.872	0.278
	Q('dum_is_Dran	na')	-0.5223	0.264	-1.975	0.049	-1.042	-0.003
	Q('dum_is_Horr	or')	-0.1273	0.280	-0.455	0.649	-0.677	0.423
Q('	dum_rated_PG-	13')	-0.1804	0.105	-1.722	0.086	-0.386	0.025
	Q('dum_rated	_R')	-0.3989	0.107	-3.737	0.000	-0.609	-0.189
Q('dum	release_in_spri	ng')	0.3277	0.090	3.636	0.000	0.151	0.505
Q('dum_re	elease_in_summ	er')	0.1569	0.086	1.830	0.068	-0.012	0.325
Q('dum	release_in_wint	er')	0.0947	0.085	1.119	0.264	-0.072	0.261
Q('average_star_	rank_12_month	_5')	-0.0911	0.025	-3.606	0.000	-0.141	-0.041
Omnibus:	2.339 <b>Durt</b>	oin-Wa	tson:	1.767				
Prob(Omnibus):	0.310 Jarque	e-Bera	(JB):	2.238				
Skew:	-0.085	Prob	(JB):	0.327				
Kurtosis:	3.270	Cond	l. No. 3	.86e+06				

Dep. Variable:		Price	R	-sc	quared:	0.65	64		
Model		OLS	Adj. R-squared:		quared:	0.65	60		
Method	: Lea	ast Squares	F-st		tatistic:	163	.8		
Date	: Fri, 2	6 Jan 2018	Prob (F	-st	atistic):	1.73e-11	6		
Time	:	00:02:11	Log-L	ike	elihood:	-553.7	5		
No. Observations:	:	528			AIC:	112	2.		
Df Residuals	:	521			BIC:	115	1.		
Df Model:	:	6							
Covariance Type:	:	nonrobust							
			co	ef	std err	t	P>ltl	[0.025	0.975]
		Intercept	168.74	12	57.124	2.954	0.003	56.520	280.962
(	Q('Num	_Theaters')	1.726	50	0.071	24.449	0.000	1.587	1.865
Q	('runtim	eMinutes')	1.199	96	0.233	5.140	0.000	0.741	1.658
50.00	Q('rele	ease_year')	-0.084	18	0.028	-3.004	0.003	-0.140	-0.029
3	Q('dum	n_rated_R')	-0.24	14	0.065	-3.785	0.000	-0.371	-0.118
Q('dum_r	elease_	in_spring')	0.248	31	0.074	3.354	0.001	0.103	0.393
Q('average_star_r	ank_12	_month_5')	-0.08	18	0.024	-3.400	0.001	-0.129	-0.035
Omnibus:	1.843	Durbin-W	/atson:		1.780				
Prob(Omnibus):	0.398	Jarque-Be	ra (JB):		1.670				
Skew:	-0.081	Pr	ob(JB):		0.434				
Kurtosis:	3.223	Co	nd. No.	3.	81e+06				