

Objective

How well can we predict international box office based off of movie pre-release features and domestic box office information?

What features have a stronger correlation to box office success?

Data Collected



Box Office
Mojo
by IMDbPro



- 1,750 movies
- US wide released from 2000-2020
- Top grossing

Model Selection and Workflow

Cross Validation Tests	R^2
Linear Regression	0.535



Poly LASSO	0.626
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Runtime

Budget

Release Month

Metacritic Score

“Adventure” Genre

Director

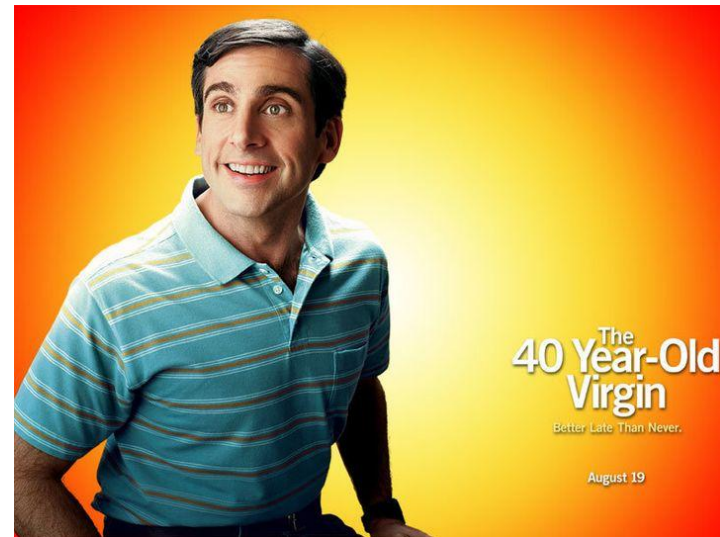
Dropping Outliers

Movie Predictions

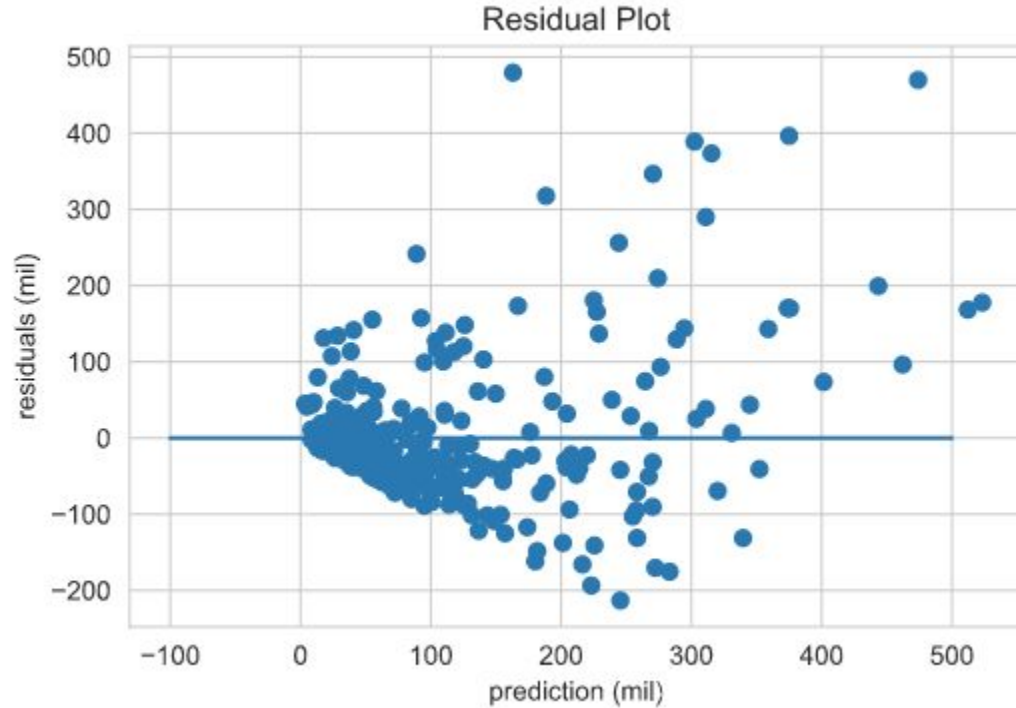
Film Title	Actual (mil)	Predicted (mil)	Error (mil)
The 40 Year Old Virgin	\$67.93	\$66.16	\$1.77
Rat Race	\$28.88	\$25.70	\$3.18
Transformers	\$691.28	\$302.40	\$388.88
Avengers	\$943.80	\$474.03	\$469.78

$R^2 = 0.626$

RMSE: 95.75



Final Model Residual Plot



$R^2 = 0.626$
RMSE: 95.75

Future Work

- Features:
 - Franchise
 - Popular int'l actors or directors
 - Popular int'l genres
- Log Regression
 - Too skewed?



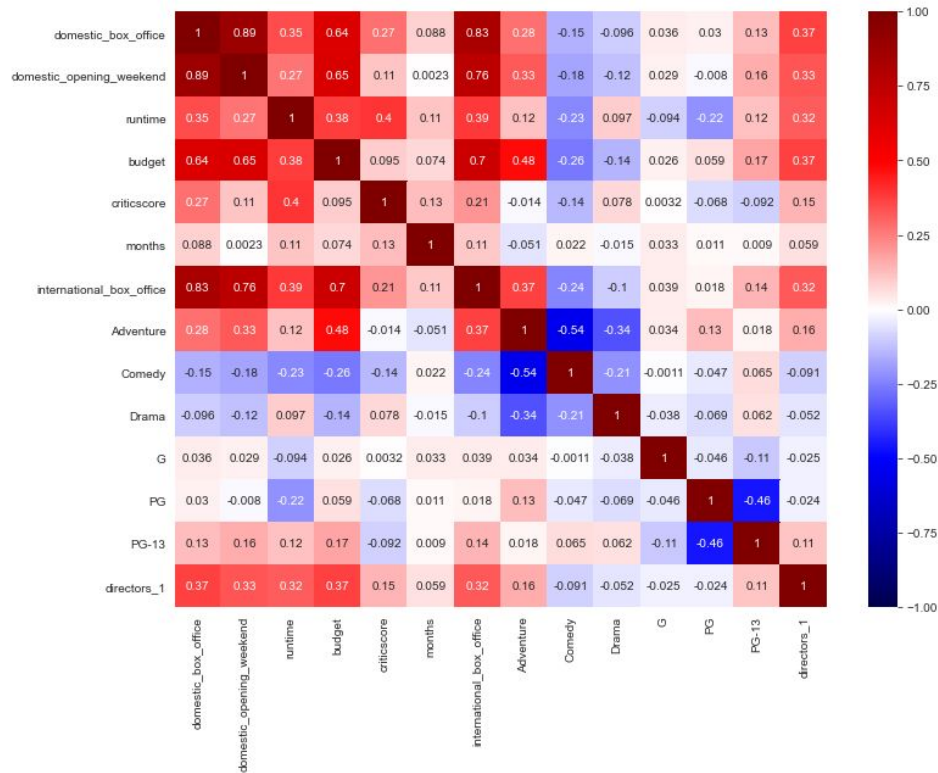
Appendix 1

OLS Regression Results

Dep. Variable:	international_box_office	R-squared:	0.750
Model:	OLS	Adj. R-squared:	0.748
Method:	Least Squares	F-statistic:	376.7
Date:	Wed, 14 Apr 2021	Prob (F-statistic):	0.00
Time:	21:33:56	Log-Likelihood:	-8824.8
No. Observations:	1519	AIC:	1.768e+04
Df Residuals:	1506	BIC:	1.774e+04
Df Model:	12		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
Intercept	-106.8134	16.594	-6.437	0.000	-139.362	-74.264
domestic_box_office	1.1888	0.059	20.126	0.000	1.073	1.305
domestic_opening_weekend	0.1487	0.178	0.833	0.405	-0.201	0.499
runtime	0.6602	0.146	4.527	0.000	0.374	0.946
budget	0.6313	0.054	11.614	0.000	0.525	0.738
criticscore	-0.2225	0.149	-1.496	0.135	-0.514	0.069
months	1.6078	0.607	2.647	0.008	0.416	2.799
Adventure	17.1199	6.523	2.625	0.009	4.325	29.914
Comedy	-14.2784	6.699	-2.131	0.033	-27.419	-1.138
Drama	-0.6663	7.942	-0.084	0.933	-16.244	14.912
G	18.3626	20.650	0.889	0.374	-22.143	58.868
PG	-4.3948	5.951	-0.739	0.460	-16.068	7.278
directors_1	-14.2485	5.070	-2.810	0.005	-24.193	-4.304

Omnibus:	809.317	Durbin-Watson:	2.218
Prob(Omnibus):	0.000	Jarque-Bera (JB):	34756.383
Skew:	1.789	Prob(JB):	0.00
Kurtosis:	26.159	Cond. No.	1.89e+03

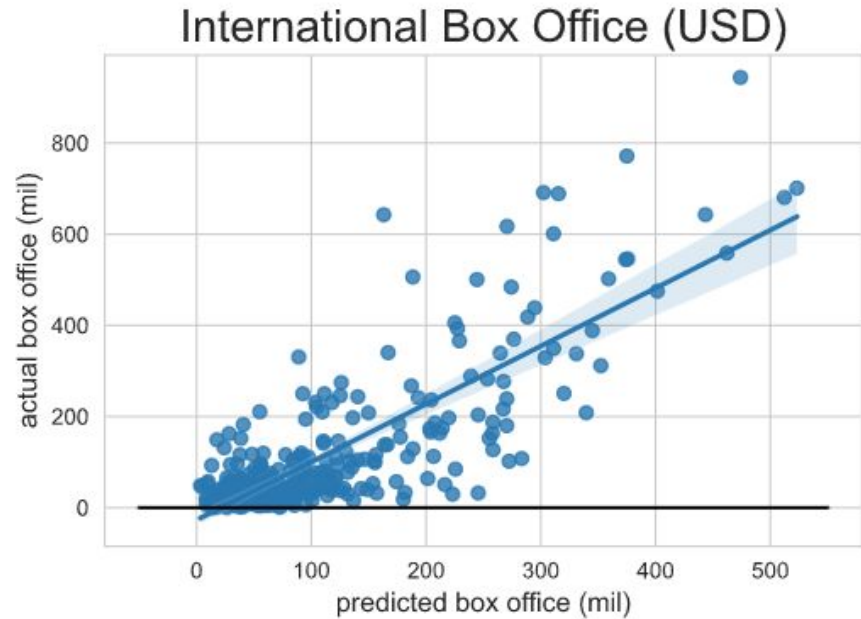


Appendix 2

final model regression formula

	feature	coef
0	intercept	1.174032e+02
1	1	2.487151e-11
2	runtime	-2.947956e+00
3	budget	-1.415960e-01
4	criticscore	7.754058e-01
5	months	-7.347438e+00
6	Adventure	-1.698209e+01
7	directors_1	-3.572943e+01
8	runtime^2	1.550250e-02
9	runtime budget	7.583133e-03
10	runtime criticscore	-2.119288e-02
11	runtime months	3.271938e-02
12	runtime Adventure	6.925195e-02
13	runtime directors_1	5.808105e-01
14	budget^2	-2.438303e-03
15	budget criticscore	1.482026e-02
16	budget months	6.016295e-03
17	budget Adventure	5.934067e-01
18	budget directors_1	-2.604903e-02
19	criticscore^2	1.638269e-02
20	criticscore months	7.981884e-03
21	criticscore Adventure	-7.546504e-02
22	criticscore directors_1	-3.495226e-01
23	months^2	1.601039e-01
24	months Adventure	1.573396e+00
25	months directors_1	4.269101e+00
26	Adventure^2	-1.698209e+01
27	Adventure directors_1	1.148067e+01
28	directors_1^2	-3.572943e+01

Appendix 3



final model regplot