

A decorative background featuring a network diagram. It consists of numerous nodes, represented by circles of varying sizes and colors (gray, blue, and white with blue outlines), connected by thin gray lines. The nodes are distributed across the slide, with a higher concentration in the top-left and bottom-right corners, leaving the center area clear for the text.

Project Luther

Too many movies?

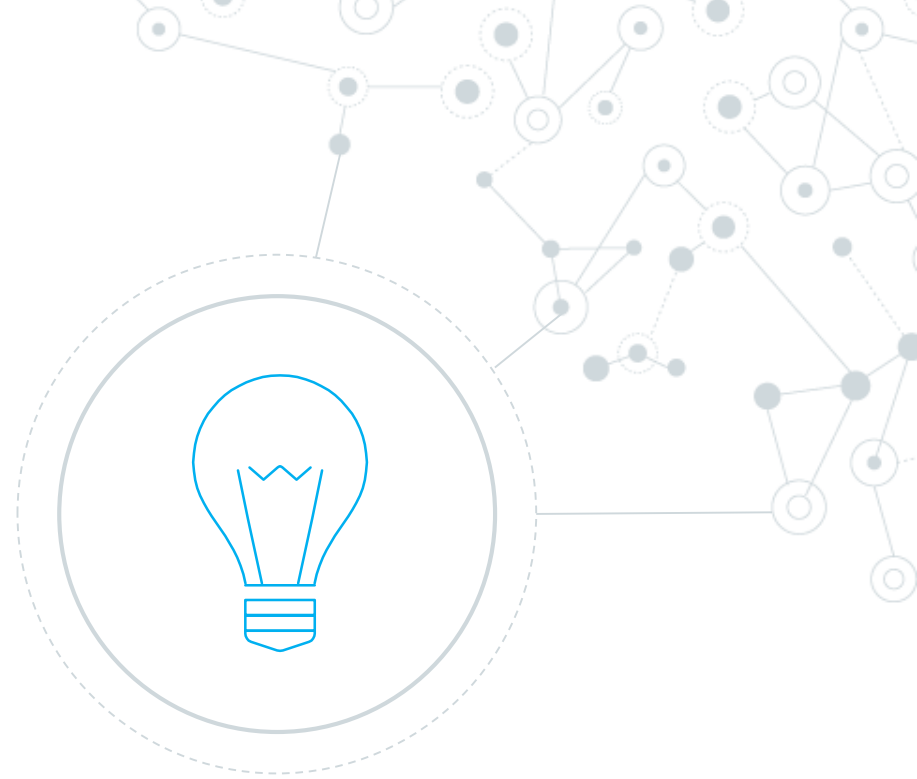
Situation

- © Luther Films plans to release a film in July 2016 but they found out that there will be a number of similar (same-genre) movies that will be released within +/- 30 days ("density")
- © Luther funded the production with high-yield debt and is worried that the competition might affect its ability to service the debt. On the other hand, interest will accrue longer if the release is delayed
- © The managers at Luther Films reached out to Metis to help them on this issue and that Metis come back with a recommendation the next day

The idea

Competition decreases
revenue, or does it?
The average American
watches a movie every
2 months

L.E.K Box Office Research



Approach

- ◎ Given the time and budget constraints...
- ◎ Data collection through web scraping
- ◎ Modeling through linear regression

Data Munging

- ◎ Data: Movies from 2009 to 2015
- ◎ Data source: BoxOfficeMojo
- ◎ Webpages scraped: 4,400
- ◎ Missing data treatment: Complete cases only
- ◎ Filter out: removed blockbusters, foreign movies, any movie with box office gross and budget below US\$1 million
- ◎ Centered “# of theaters” and “# of days in release” variables to address severe multicollinearity
- ◎ Collapsed “genre” and “distributor” categorical variables
- ◎ Final # of observations: 730

Beware of the Outliers

- ◎ Distributed by Kenn (Viselman)
- ◎ Domestic gross of US\$1 million
- ◎ Budget of US\$20 million
- ◎ While not statistically significant, OLS shows that getting Kenn as the distributor increases domestic box office gross by US\$48 million relative to “Other Distributors”

Source: BoxOfficeMojo



Regression Model

- ⦿ Evaluated OLS and Elastic Net Regression
- ⦿ Used 100% of data since focus in on coefficient
- ⦿ Model equation:

adj. dom. gross = 1 + adj. budget + runtime + # days in release + max # of theaters (C) + (max # of theaters (C))² + density + genre + distributor + month

	R ²	Adj. R ²	Density Coefficient p-value
Ordinary Least Squares	73%	73%	6%
Elastic Net Regression	72%	72%	21%

US\$ 2,200,000

less revenue for every same-genre movie that
will be released within 30 days Luther Film's
movie release date ($p < 0.07$, HC3)



“

Luther Films should consider releasing the movie in June, a month earlier, to catch the spring/summer rush and potentially increase revenue by
US\$15 million ($p < 0.09$, HC3)

Other Ideas

- ◎ Density measured by # of high budget movies and (just the) # of movies released within +/- 30 days
- ◎ Scraped OMDb data for movie ratings from Rotten Tomatoes, IMDB, Metacritic
- ◎ Scraped RGB pixel data of movie posters

Thanks!

Any questions?

