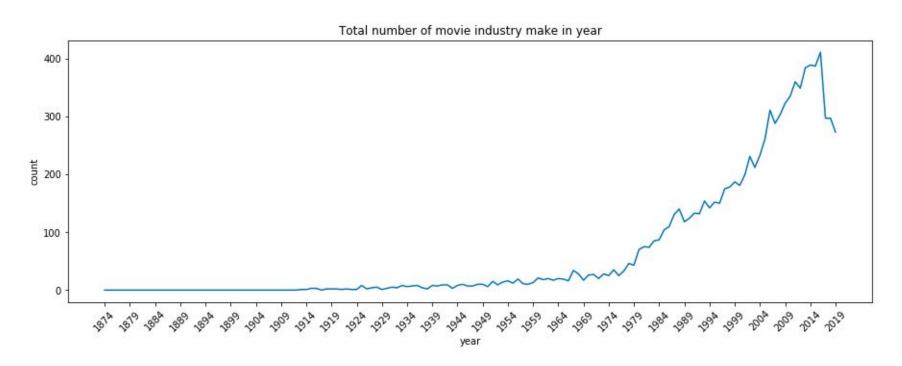
# MOVIE PREDICT AND RECOMENDATION

#### Word Cloud for title movie



#### Total number movie over year!



#### Total, Avg. revenue and number movie each year

	Average_revenue	Total_revenue	count	
year				
2016	7.627827e+07	3.135037e+10	411	
2015	7.414786e+07	2.869522e+10	387	
2014	7.113407e+07	2.767116e+10	389	
2013	7.039734e+07	2.703258e+10	384	
2012	7.569661e+07	2.641812e+10	349	
2011	6.939981e+07	2.498393e+10	360	
2010	7.205477e+07	2.413835e+10	335	
2009	7.440009e+07	2.403123e+10	323	
2008	6.991971e+07	2.118567e+10	303	
2007	7.076921e+07	2.038153e+10	288	

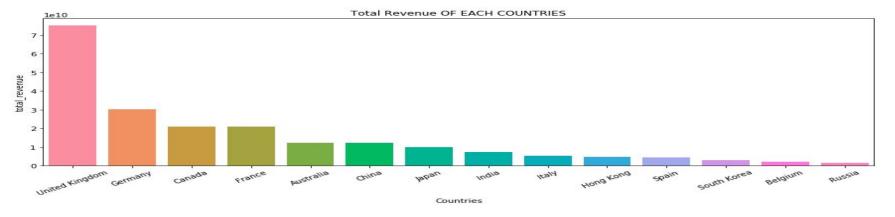
	Year year	r was first movie	runtime	
34693	1874	Passage of Venus	1.0	
34690	1878	Sallie Gardner at a Gallop	1.0	
70946	1881	Athlete Swinging a Pick	1.0	
41270	1883	Buffalo Running	1.0	
135279	1885	L'homme machine	1.0	

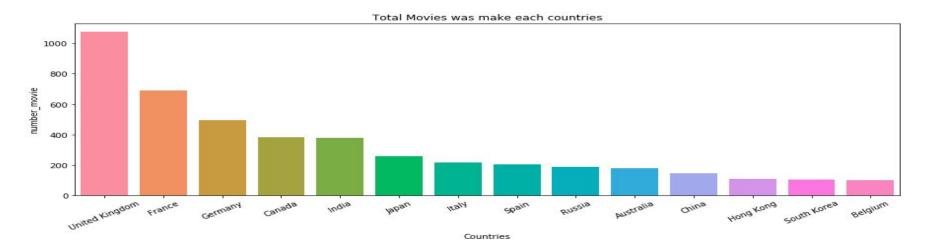
#### Year last movie in our data set title runtime year 2019 174446 Sunday 13.0 174441 2019 Queen + Béjart - Ballet For Life 58.0 The Ocean Washed Open Your Grave 3.0 174467 2019 Entropia 28.0 210551 2019 Jorge 20.0

# Total, Avg. revenue of each countries

	average_revenue	total_revenue	number_movie
Countries			
United States of America	7.714978e+07	4.976932e+11	6451
United Kingdom	6.990096e+07	7.514353e+10	1075
France	3.021624e+07	2.084921e+10	690
Germany	6.078097e+07	3.014736e+10	496
Canada	5.485054e+07	2.089806e+10	381
India	1.911960e+07	7.246328e+09	379
Japan	3.894279e+07	9.969354e+09	256
Italy	2.382187e+07	5.169346e+09	217
Spain	2.157035e+07	4.378782e+09	203
Russia	8.155989e+06	1.517014e+09	186

## Total, Avg. revenue of each countries (continous)

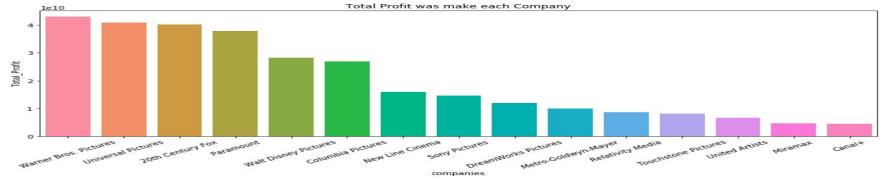


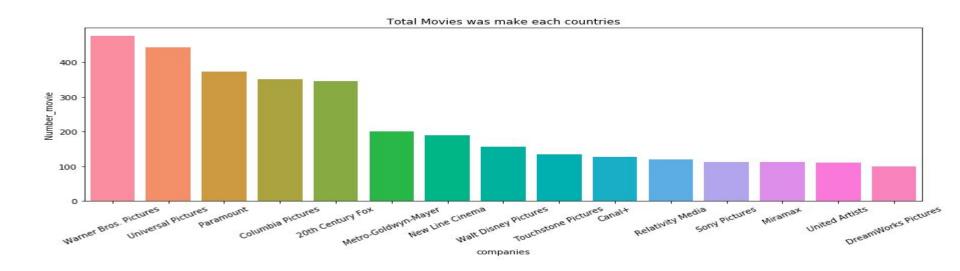


# Total, Avg profit of each company

	Avergae_profit	Total_Profit	Number_movie
companies			
Warner Bros. Pictures	9.032809e+07	4.299617e+10	476
Universal Pictures	9.229253e+07	4.088559e+10	443
Paramount	1.014296e+08	3.793466e+10	374
Columbia Pictures	7.684656e+07	2.697314e+10	351
20th Century Fox	1.164543e+08	4.017672e+10	345
Metro-Goldwyn-Mayer	5.015126e+07	1.003025e+10	200
New Line Cinema	8.415754e+07	1.598993e+10	190
Walt Disney Pictures	1.797291e+08	2.821747e+10	157
Touchstone Pictures	6.081461e+07	8.209972e+09	135
Canal+	3.521476e+07	4.437060e+09	126

### Total, Avg profit of each company (continous)

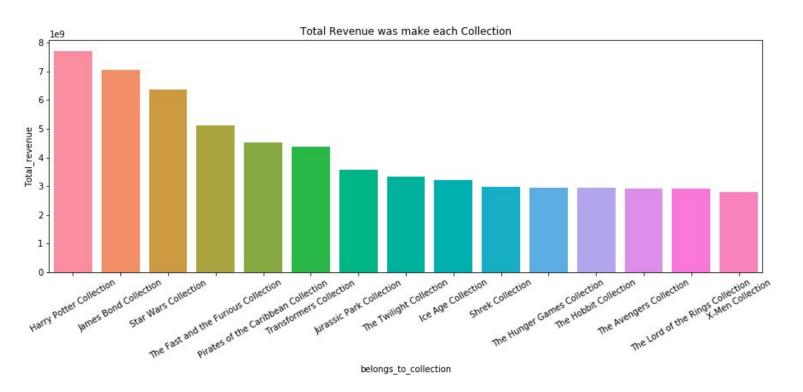




### Total, Avg. of each collection of movie

	Average_revenue	Total_revenue	Total_movie
belongs_to_collection			
Harry Potter Collection	9.633598e+08	7.706879e+09	8
James Bond Collection	2.827486e+08	7.068715e+09	25
Star Wars Collection	9.112054e+08	6.378438e+09	7
The Fast and the Furious Collection	6.406373e+08	5.125099e+09	8
Pirates of the Caribbean Collection	9.043154e+08	4.521577e+09	5
Transformers Collection	8.758574e+08	4.379287e+09	5
Jurassic Park Collection	8.948083e+08	3.579233e+09	4
The Twilight Collection	6.686215e+08	3.343107e+09	5
Ice Age Collection	6.433533e+08	3.216767e+09	5
Shrek Collection	7.411823e+08	2.964729e+09	4

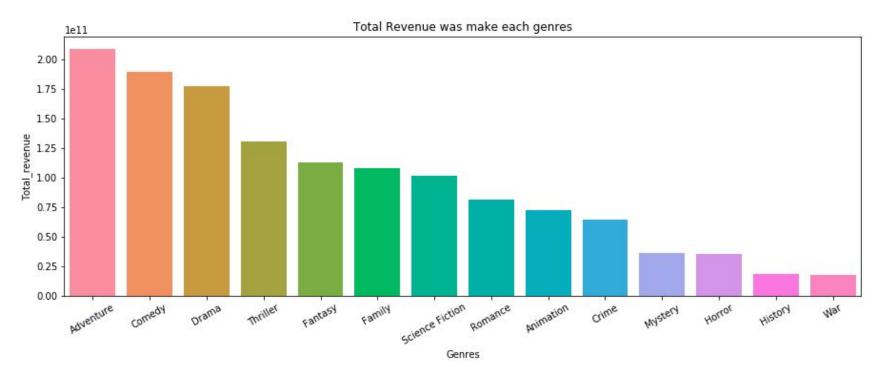
#### Total, Avg. of each collection of movie (continous)



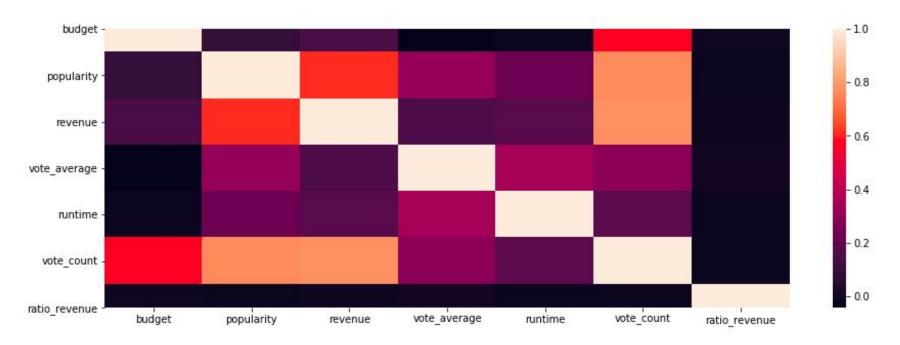
# Total, Avg. of Genres of movie

	Avergae_revenue	Total_revenue	Total_movie
Genres			
Action	1.025900e+08	2.134897e+11	2081
Adventure	1.640720e+08	2.086996e+11	1272
Comedy	5.798689e+07	1.889213e+11	3258
Drama	3.835340e+07	1.773078e+11	4623
Thriller	6.120269e+07	1.302393e+11	2128
Fantasy	1.463286e+08	1.122341e+11	767
Family	1.299991e+08	1.080293e+11	831
Science Fiction	1.181741e+08	1.016297e+11	860
Romance	4.568787e+07	8.073046e+10	1767
Animation	1.467283e+08	7.189686e+10	490

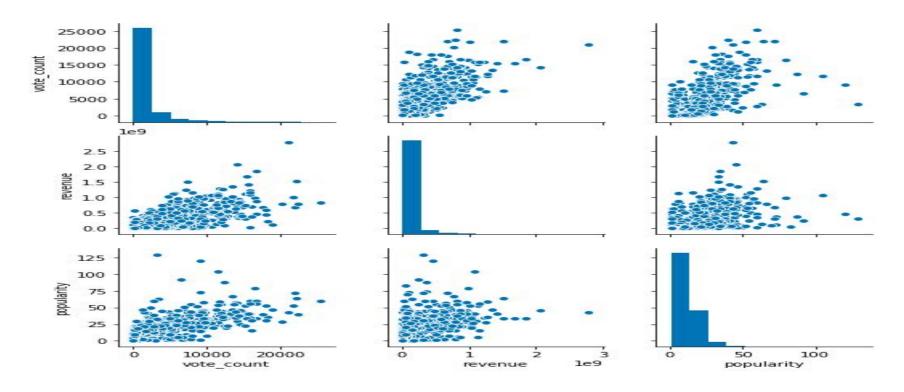
### Total, Avg. of Genres of movie (continous)



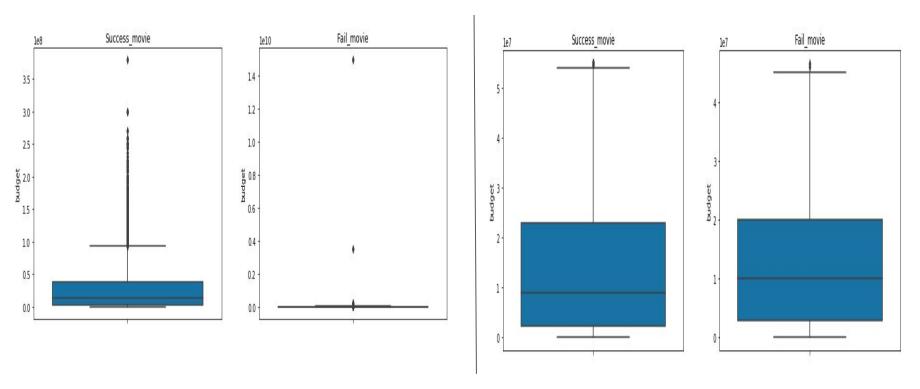
# Heatmap relation of each column



# Pairplot for vote\_count, revenue, popularity



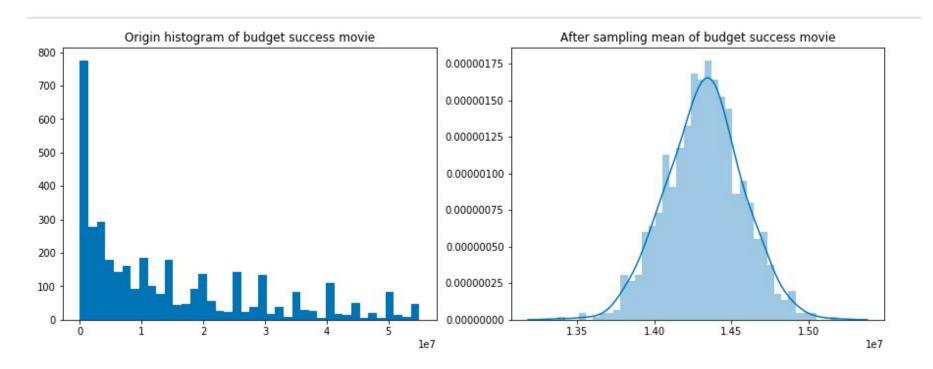
#### Boxplot detect outlier and remove



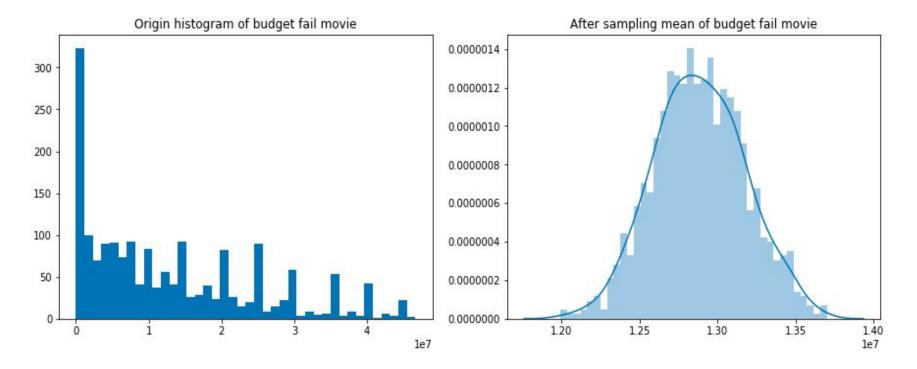
Before remove outlier

After remove outlier

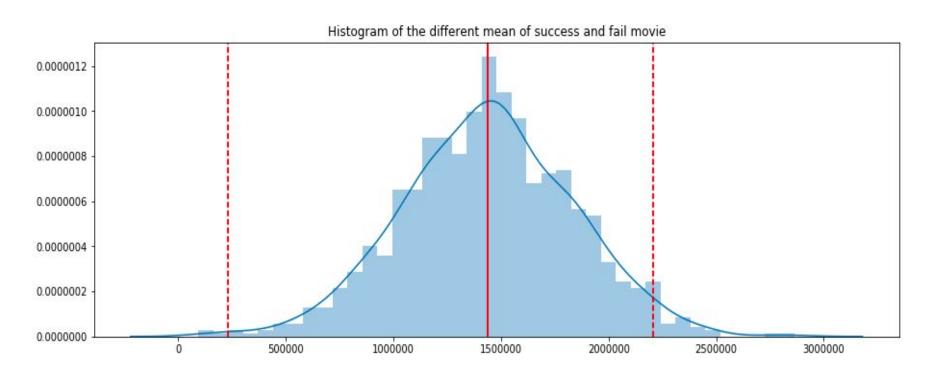
#### Resampling mean of success movie



#### Resampling mean of fail movie



#### Different mean of two resampling.



#### PREDICT MACHINE MODEL

1. Random forest classification

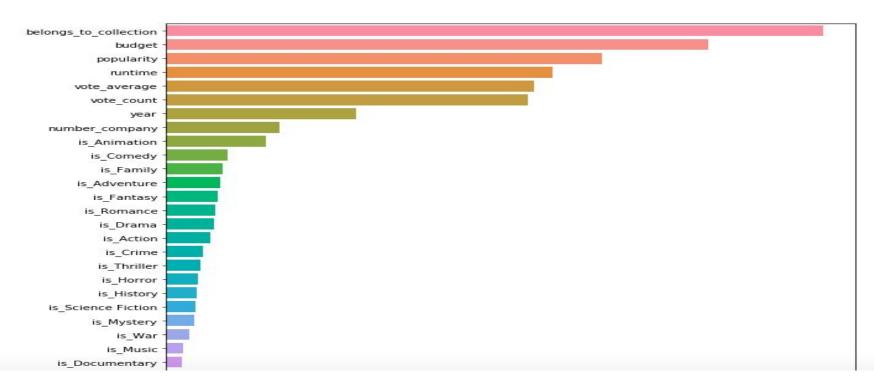
2. Gradient Boosting Classification

3. Logistic Regression model

#### Random Forest Classification

	precision	recall	f1-score	support
0	0.67	0.54	0.60	785
1	0.82	0.89	0.85	1860
accuracy			0.78	2645
macro avg	0.75	0.71	0.73	2645
weighted avg	0.78	0.78	0.78	2645

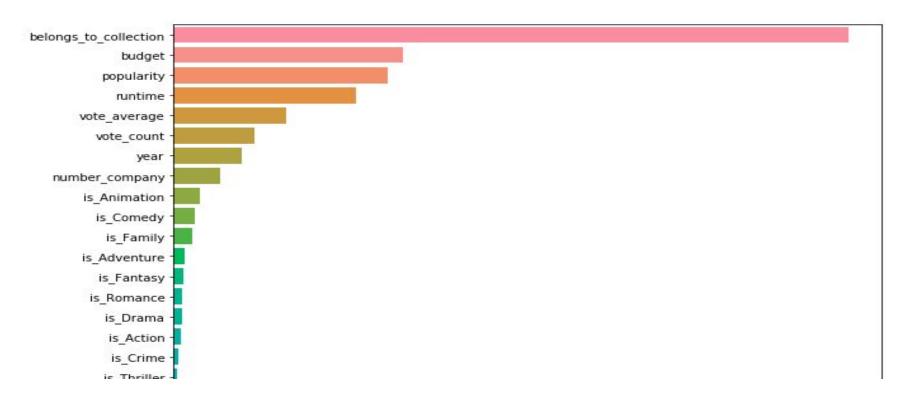
#### Feature importance



# **Gradient Boosting Classification**

	precision	recall	fl-score	support
0	0.67	0.54	0.60	785
1	0.82	0.88	0.85	1860
accuracy			0.78	2645
macro avg	0.74	0.71	0.73	2645
weighted avg	0.78	0.78	0.78	2645

# Feature Importance



# Logistic Regression model

precision	recall	f1-score	support
1.00	0.00	0.01	785
0.70	1.00	0.83	1860
		0.70	2645
0.85	0.50	0.42	2645
0.79	0.70	0.58	2645
	1.00 0.70	1.00 0.00 0.70 1.00 0.85 0.50	1.00 0.00 0.01 0.70 1.00 0.83 0.70 0.50 0.42

#### Conclusion

- Random forest:
  - Score: 0.78
  - Faster, and use more feature
- Gradient Boosting
  - Score: 0.78
  - Slower than random forest, and use less feature than random forest
- Logistic regression
  - Score: 0.70

#### RECOMMENDATION SYSTEM

- Basic recommendation
  - Top 20 movie
  - Top 15 movie for genre
- Correlation-base recommendation
- Model-based Collaborative filtering system recommendation

#### **Basic Recommendation**

recommend\_top\_20\_movie()

genres	title	
Drama Film-Noir Mystery Romance	In a Lonely Place (1950)	0
Animation Comedy Romance	Paperman (2012)	1
Horror Mystery Thriller	Diabolique (Les diaboliques) (1955)	2
Crime Drama Thriller War	Paradise Now (2005)	3
Drama War	Best Years of Our Lives, The (1946)	4
Action Comedy	Drunken Master (Jui kuen) (1978)	5
Drama	Inherit the Wind (1960)	6
Crime Drama Mystery Thriller	Tell No One (Ne le dis à personne) (2006)	7
Animation Children Comedy	For the Birds (2000)	8
Animation Drama Romance	Wind Rises, The (Kaze tachinu) (2013)	9
Adventure Comedy Musical	Court Jester, The (1956)	10
Crime Drama	Godfather, The (1972)	11
Crime Drama	Shawshank Redemption, The (1994)	12
Adventure Comedy Romance	Tom Jones (1963)	13
Action Drama	Gladiator (1992)	14
Crime Drama	On the Waterfront (1954)	15
Comedy Drama	Kid, The (1921)	16
Documentary	When We Were Kings (1996)	17
Comedy Drama	Carnal Knowledge (1971)	18

#### Basic recommendation

get\_recommend\_genre("Drama")

genres	title	
Drama War	Best Years of Our Lives, The (1946)	0
Drama	Inherit the Wind (1960)	1
Crime Drama	Godfather, The (1972)	2
Crime Drama	Shawshank Redemption, The (1994)	3
Action Drama	Gladiator (1992)	4
Crime Drama	On the Waterfront (1954)	5
Drama	All About Eve (1950)	6
Drama War	Ran (1985)	7
Comedy Drama War	Mister Roberts (1955)	8
Crime Drama	Godfather: Part II, The (1974)	9
Drama War	Paths of Glory (1957)	10
Drama War	Lifeboat (1944)	11
Action Drama	Rush (2013)	12
Comedy Drama Romance	Modern Times (1936)	13
Comedy Drama Romance	Philadelphia Story, The (1940)	14

#### Correlation base recommendation

- Use Pearson's r correlation
  - to recommend a movie that is most similar to the movie that user have early watch
- Users can search movies by name, or
- the system will recommend the next movie based on the user have been watched.

#### Correlation recommendation

```
get_recommendation_movie_corr('Dangerous Minds (1995)')
/Users/hungnguyen/miniconda3/lib/python3.7/site-packages/numpy/lifreedom <= 0 for slice
    c = cov(x, y, rowvar)
/Users/hungnguyen/miniconda3/lib/python3.7/site-packages/numpy/lifrero encountered in true_divide
    c *= np.true_divide(1, fact)</pre>
```

genres	title	
Drama Thriller	And Justice for All (1979)	0
Adventure Drama Thriller	127 Hours (2010)	1
Crime Film-Noir	2 Days in the Valley (1996)	2
Action Crime Thriller	2 Fast 2 Furious (Fast and the Furious 2, The)	3
Crime Drama Romance Thriller	21 (2008)	4
Action Fantasy War IMAX	300 (2007)	5
Comedy Romance	40 Days and 40 Nights (2002)	6
Comedy Drama Romance	About Last Night (1986)	7
Mystery Thriller	Absolute Power (1997)	8
Children Comedy Fantasy	Addams Family, The (1991)	9
Comedy Drama	Adventureland (2009)	10
Adventure Comedy	Adventures in Babysitting (1987)	11
Adventure Comedy Fantasy	Adventures of Baron Munchausen, The (1988)	12
Action Sci-Fi	Aeon Flux (2005)	13

#### Model-base collaborative filter system

- Use Singular Value Decomposition (SVD)
- Recommend base on model
  - Faster

# Model-base collaborative filter system

model\_base\_recommendation('Dangerous Minds (1995)')

genres	title	
Drama	Dangerous Minds (1995)	0
Action Drama Sci-Fi Thriller	Outbreak (1995)	1
Action Adventure Sci-Fi	Waterworld (1995)	2
Drama Romance War Western	Legends of the Fall (1994)	3
Action Drama Western	Tombstone (1993)	4
Adventure Drama Western	Dances with Wolves (1990)	5
Adventure Drama IMAX	Apollo 13 (1995)	6
Action Adventure Thriller	Cliffhanger (1993)	7
Comedy	Ace Ventura: Pet Detective (1994)	8
Drama Thriller	Firm, The (1993)	9
Action Romance Thriller	Speed (1994)	10
Action Adventure Comedy Romance Thriller	True Lies (1994)	11
Action Crime Thriller	Die Hard: With a Vengeance (1995)	12
Drama Thriller War	Crimson Tide (1995)	13
Action Crime Thriller	Net, The (1995)	14