

MOVIE RECOMMENDER SYSTEM

BRBSB BARUWATTA

16/ENG/016

Content

Problem

Background & Aim

Technology

Design & Implementation

Results

Challenges

Future improvements

Q & A

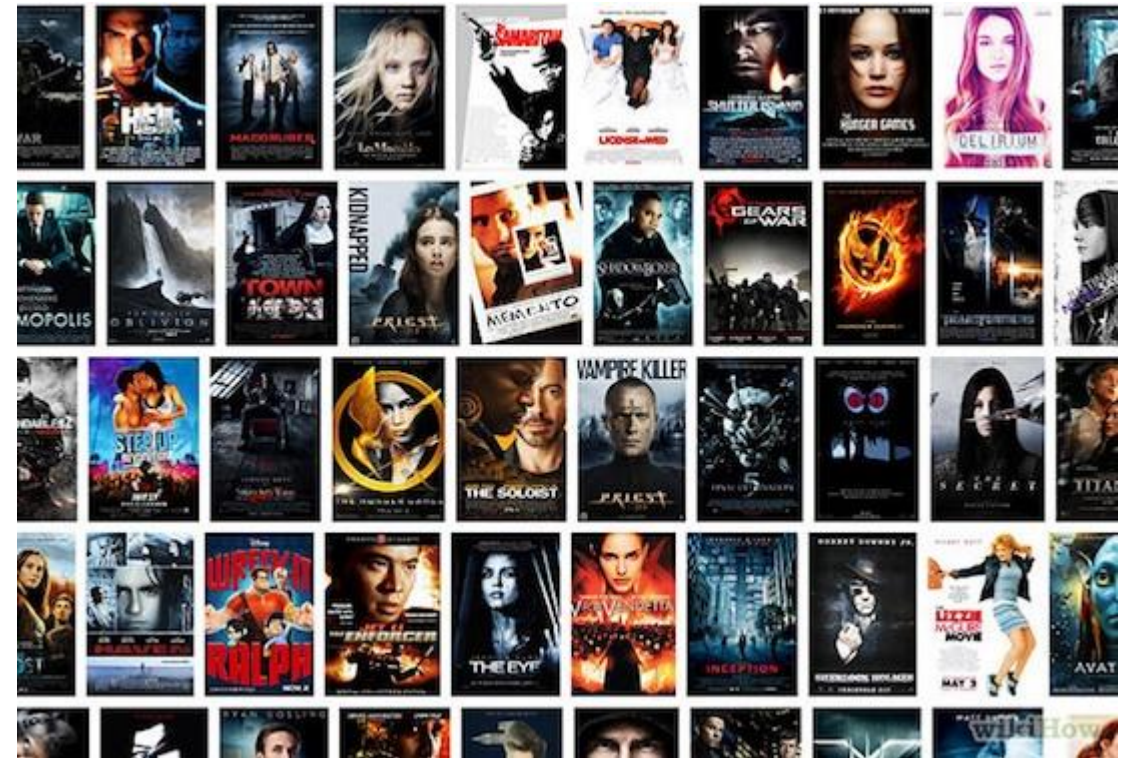
Problem

- ▶ Consumers Have many wants
- ▶ Choosing the most suitable item???
- ▶ Products are increasing rapidly
- ▶ Also problem in movie industry

Solution

Recommender System

Movie Recommender System

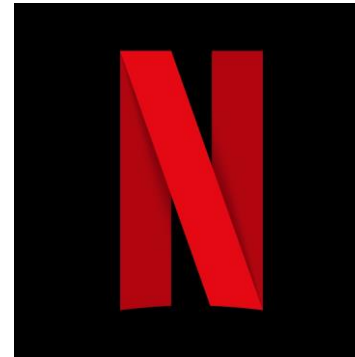


Background and Aim

Popular Videos Streaming and Movie Platforms
Increasing the Accuracy

Best recommender system perform 70% Netflix
“The BellKor Solution to the Netflix Grand Prize”

- Yehuda Koren



Aim

- ▶ Build a hybrid recommendation system which perform better



Technologies

- ▶ Movielens data set 100k data set
- ▶ Anaconda for data analysis
- ▶ Surprise Library

Methodology

► TOP N Architecture

**(one) anatomy of a
top-N recommender**



Methodology

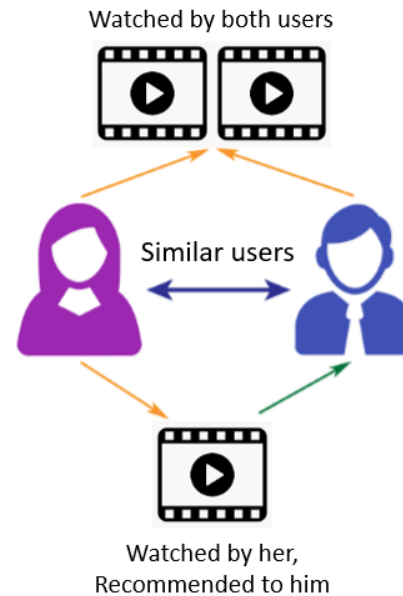
Theory

Find relationship between consumer – item based on action or attributes

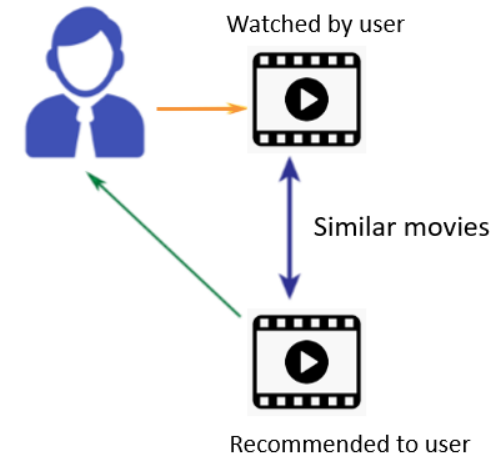
Hybrid

Assign Weight Factor

Collaborative Filtering



Content-Based Filtering



Methodology

BENCHMARKS

Accuracy

MSE

RMSE

Hit Rate = Hits/Users

**mean absolute
error (MAE)**

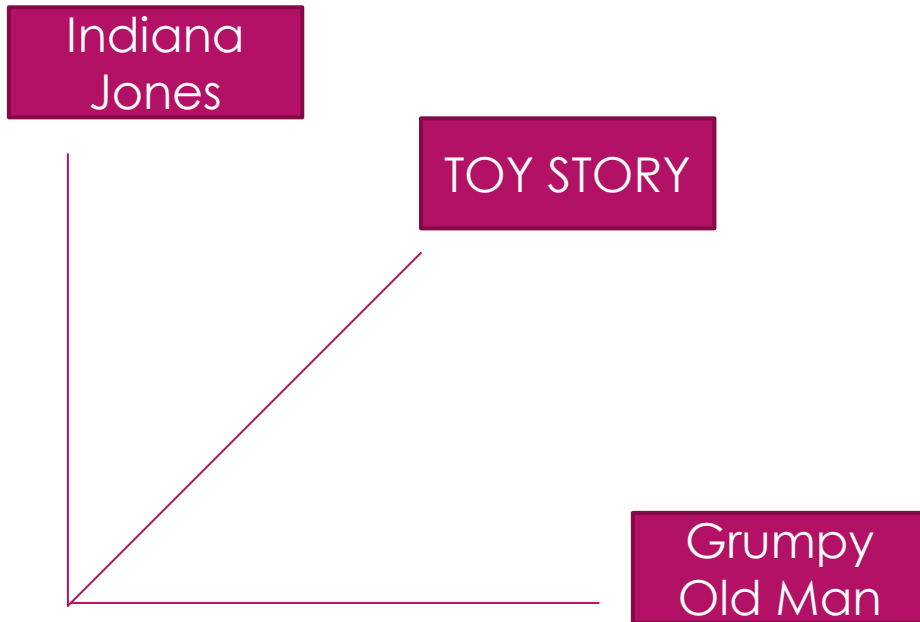
$$\frac{\sum_{i=1}^n |y_i - x_i|}{n}$$

**root mean square
error (RMSE)**

$$\sqrt{\frac{\sum_{i=1}^n (y_i - x_i)^2}{n}}$$

Methodology

► Content Based Filtering KNN



	Co me dy	Adve nture	Horro r	Roma nce	actio n	scifi
Matrix	0	0	0	0		1	1
Doo little	1	1	0	0		0	0

Similarity Matrixes (Cosine Sim & Exponential Sim)

$$\text{similarity} = \cos(\theta) = \frac{\mathbf{A} \cdot \mathbf{B}}{\|\mathbf{A}\| \|\mathbf{B}\|} = \frac{\sum_{i=1}^n A_i B_i}{\sqrt{\sum_{i=1}^n A_i^2} \sqrt{\sum_{i=1}^n B_i^2}}$$

Methodology

SVDpp

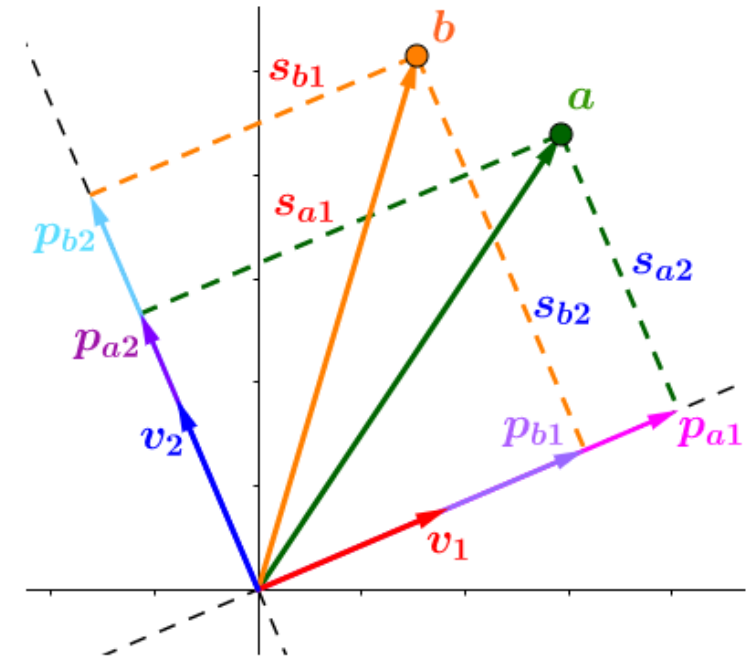
$$A = U \Sigma V^T$$

SVDpp = GV+A

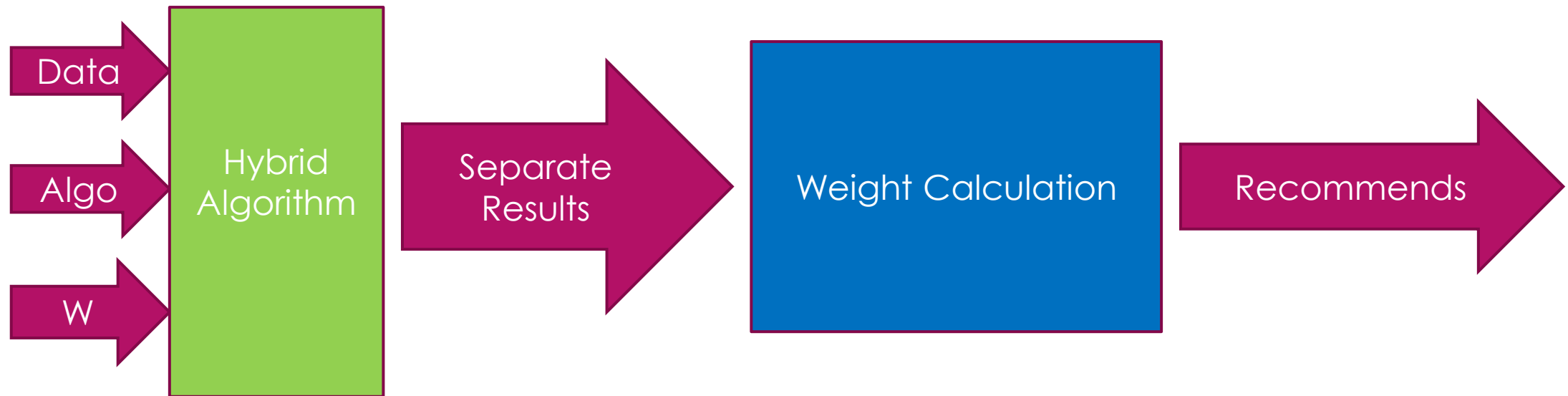
	Bob	Ted	Ann
Indiana Jones	4	3	4
Star Wars	5	3	5
Empire Strikes Back	5	3	5
Incredibles	4	5	5
Casablanca	4	4	2



	"Action"	"Sci-Fi"	"Classic"
Indiana Jones	0.6	0.3	0.1
Star Wars	0.4	0.6	0
Empire Strikes Back	0.4	0.6	0
Incredibles	0.8	0.2	0
Casablanca	0.2	0	0.8



Methodology



Results

```
Evaluating Random ...  
Evaluating accuracy...  
Analysis complete.
```

Algorithm	RMSE	MAE
ContentKNN	1.0460	0.8198
Random	1.4385	1.1478

Legend:

RMSE: Root Mean Squared Error. Lower values mean better accuracy.
MAE: Mean Absolute Error. Lower values mean better accuracy.

```
...done.  
Analysis complete.
```

Algorithm	RMSE	MAE
SVDpp	0.9002	0.6958
ContentKNN	0.9375	0.7263
Hybrid	0.8998	0.6961

Legend:

Results

```
Computing recommendations...
```

```
We recommend:
```

```
Inception (2010) 3.9710086765133914  
Touch of Evil (1958) 3.9349401871607217  
Fight Club (1999) 3.885915398141114  
French Connection, The (1971) 3.8558946140756425  
City of God (Cidade de Deus) (2002) 3.8515050078395734  
Star Trek: First Contact (1996) 3.850014218802074  
Independence Day (a.k.a. ID4) (1996) 3.833685147457443  
Indiana Jones and the Last Crusade (1989) 3.824705292208889  
Heat (1995) 3.8064215431021307  
True Grit (2010) 3.796431973892063
```

Hybrid Algorithm

```
...done.  
Computing recommendations...
```

```
We recommend:
```

```
Presidio, The (1988) 3.841314676872932  
Femme Nikita, La (Nikita) (1990) 3.839613347087336  
Wyatt Earp (1994) 3.8125061475551796  
Shooter, The (1997) 3.8125061475551796  
Bad Girls (1994) 3.8125061475551796  
The Hateful Eight (2015) 3.812506147555179  
True Grit (2010) 3.812506147555179  
Open Range (2003) 3.812506147555179  
Big Easy, The (1987) 3.7835412549266985  
Point Break (1991) 3.764158410102279
```

```
Using recommender Hybrid
```

KNN Algorithm

Results

Algorithm	MAE	RMSE	Bench(RMSE)
KNNBasic Cont	0.8198	1.046	1.154
KNN tuned Cont	0.7263	0.9375	0.952
SVDpp	0.6958	0.9002	0.926
Random	1.1478	1.4385	1.501
Hybrid	0.6961	0.8998	-

Challenges


- ▶ Scaling Up the System

100k to 20 million

- ▶ Predict Rating problem
- ▶ Success is based on real test
- ▶ Resources
- ▶ SVDpp Need to modify

Future Works

- ▶ Test Hybrid Method with different Algorithms
- ▶ Increase the number of algorithms for hybrid system



Q & A