

## **Movie Recommendation System**

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## Agenda



1	Why do we need data science?
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- What is Data science?
- 3 Life cycle of Data science
- Why Python is so popular?
- 5 Install python
- 6 Statistical visualization on Python user

- 7 What is recommendation system?
- 8 Types of recommendation system
- 9 Use case for UBCF
- 10 Use case for CBF
- Demo: Movie recommendation system

## Why do we need Data Science?





- In the past, we used to have data in a structured format but now as the volume of the data is increasing, so the number of structured data becomes very less, so to handle the massive amount of data we need data science techniques
- Those data can be used to get the proper business insights and the hidden trends from them.
- These insights helps the organization to predict the Future
- Using data science decision making can be faster and effective
- Helps to reduce the production cost
- Build model based on the data to give the ability to the machine to predicts on its own

#### What is Data Science?

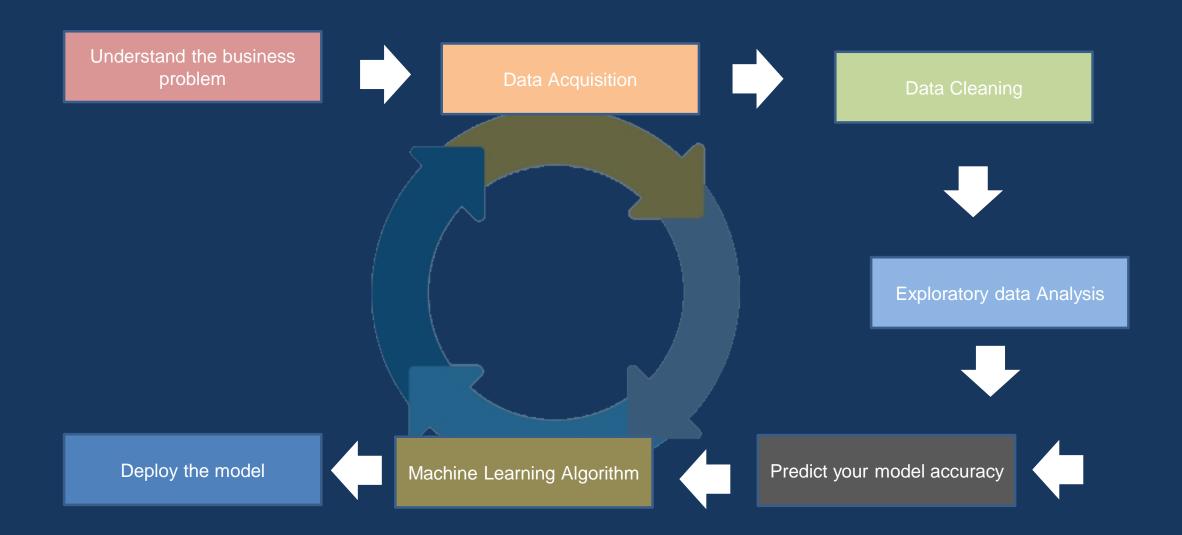




Data science is a process to get some meaningful information from the massive amount of data. In simple terms, read and study the data to get proper intuitive insights. Data Science is a mixture of various tools, algorithms, and machine learning and deep learning concepts to discover hidden patterns from the raw and unstructured data

## Life cycle of Data Science?





### **Most Popular Programming Languages For Data Science?**





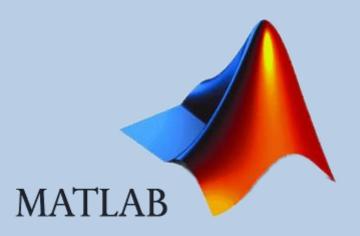








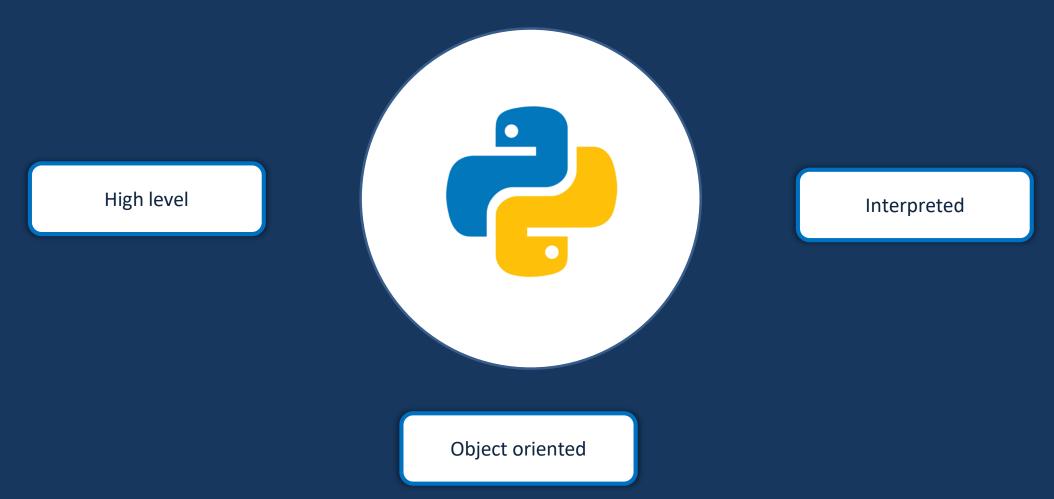




## **Introduction to Python**



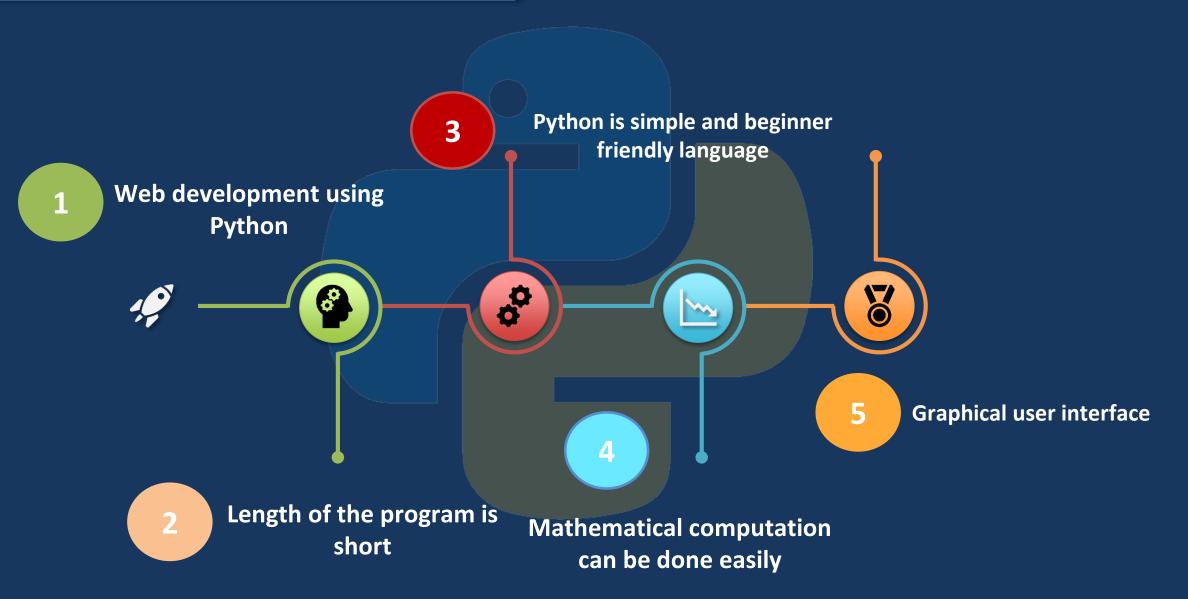
## Python is a popular high level, object oriented and interpreted language



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## Why should you learn Python?





## Why Python is so popular?



1 Largest community for Learners and Collaborators

Open source

**3** Easy to learn and usable flexibility

Huge numbers of Python libraries and Frame work

Supports Big Data, Machine Learning and Cloud computing

Supports Automation

## **Installing Python**



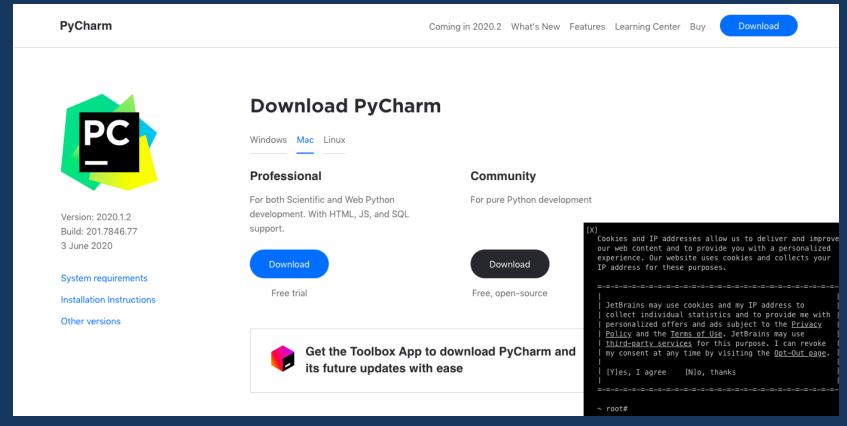
This is the site to install Python -> <a href="https://www.python.org/downloads/">https://www.python.org/downloads/</a>



### **Popular IDE for Python: Pycharm**



Site to install Python -> <a href="https://www.jetbrains.com/pycharm/download/#section=mac">https://www.jetbrains.com/pycharm/download/#section=mac</a>



## **Popular IDE for Python: Anaconda**



Anaconda installation site-> <a href="https://www.anaconda.com/products/individual">https://www.anaconda.com/products/individual</a>



**Individual Edition** 

# Your data science toolkit

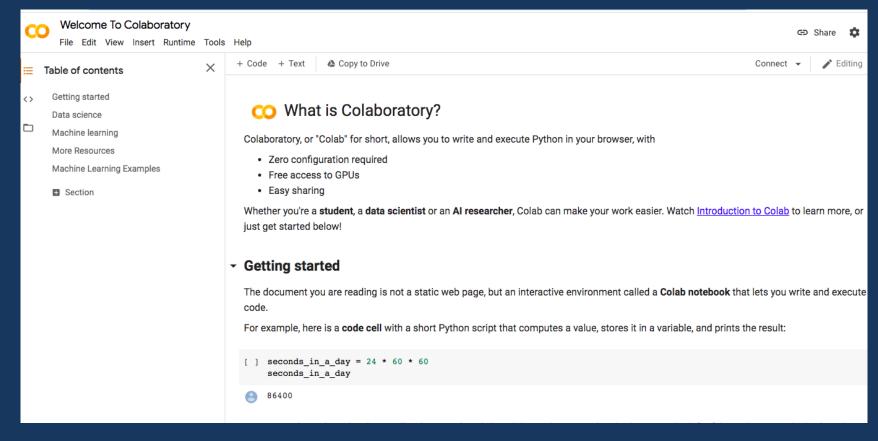
With over 20 million users worldwide, the open-source Individual Edition (Distribution) is the easiest way to perform Python/R data science and machine learning on a single machine. Developed for solo practitioners, it is the toolkit that equips you to work with thousands of open-source packages and libraries.

Download

### Popular IDE for Python: Google colab

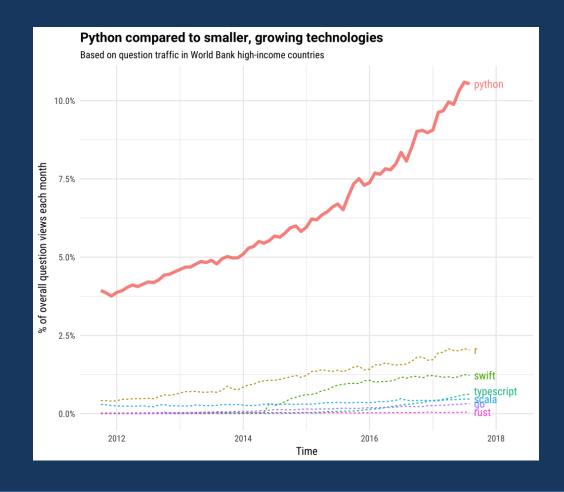


Google collaboratory link-> <a href="https://colab.research.google.com/notebooks/intro.ipynb">https://colab.research.google.com/notebooks/intro.ipynb</a>



## Statistical measurement on Python user





In recent time it is prominent that Python is one of the most popular language because of it's simplicity



Machine learning is a sub-set of artificial intelligence (AI) that allows the system to automatically learn and improve from experience without being explicitly programmed

	Time	V1	V2	V3	V4	V5
0	0.0	-1.359807	-0.072781	2.536347	1.378155	-0.338321
1	0.0	1.191857	0.266151	0.166480	0.448154	0.060018
2	1.0	-1.358354	-1.340163	1.773209	0.379780	-0.503198
3	1.0	-0.966272	-0.185226	1.792993	-0.863291	-0.010309
4	2.0	-1.158233	0.877737	1.548718	0.403034	-0.407193



	Time	V1	\ <sub>o</sub> V2	V3	V4
284802	172786.0	-11.881118	10.071785	-9.834783	-2.066656
284803	172787.0	-0.732789	-0.055080	2.035030	-0.738589
284804	172788.0	1.919565	-0.301254	-3.249640	-0.557828
284805	172788.0	-0.240440	0.530483	0.702510	0.689799
284806	172792.0	-0.533413	-0.189733	0.703337	-0.506271

**Training Data** 

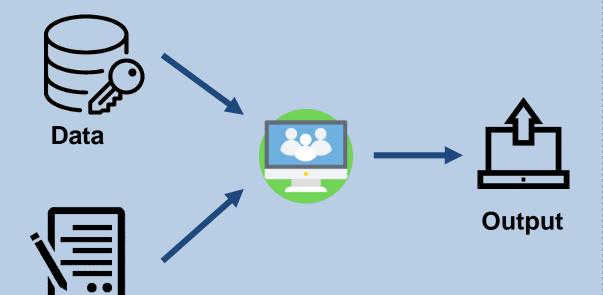
**Model Building** 

**Testing Data** 

## Traditional Vs Machine Learning

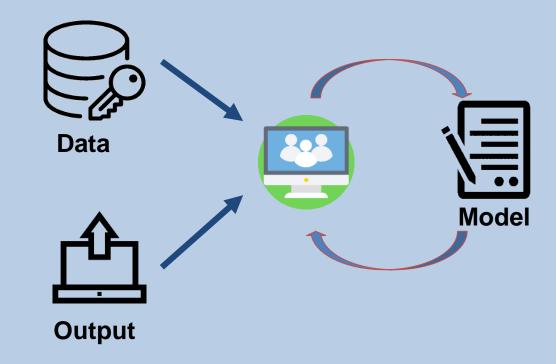


**Traditional Programming** 



**Program** 

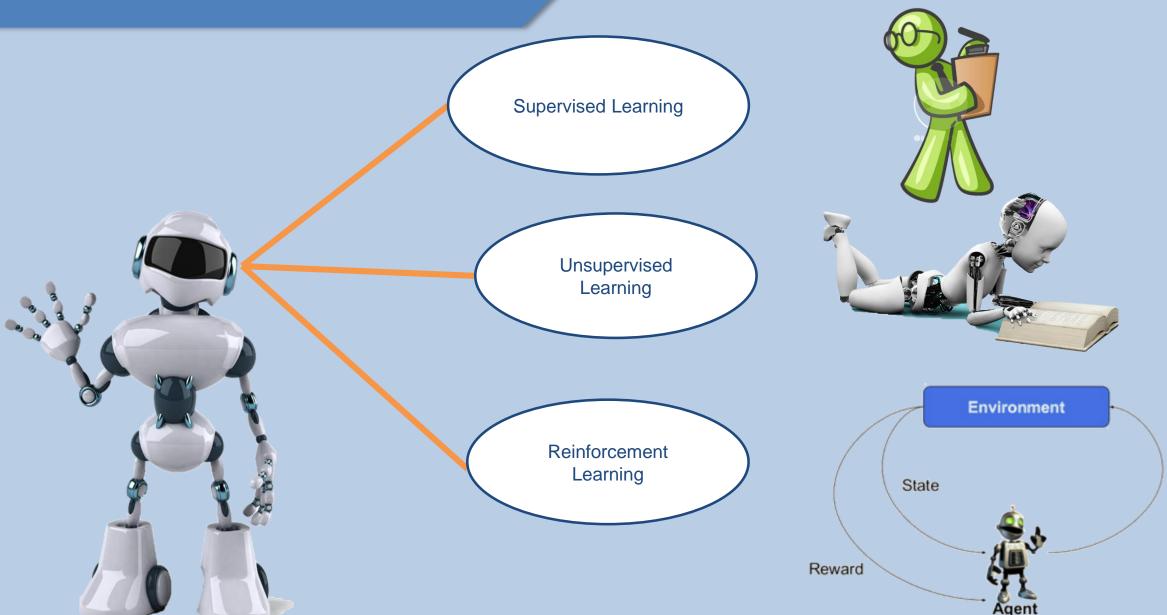
**Machine Learning** 



## **Types Of Machine Learning**



Action



### **Recommendation Engine**

greatlearning

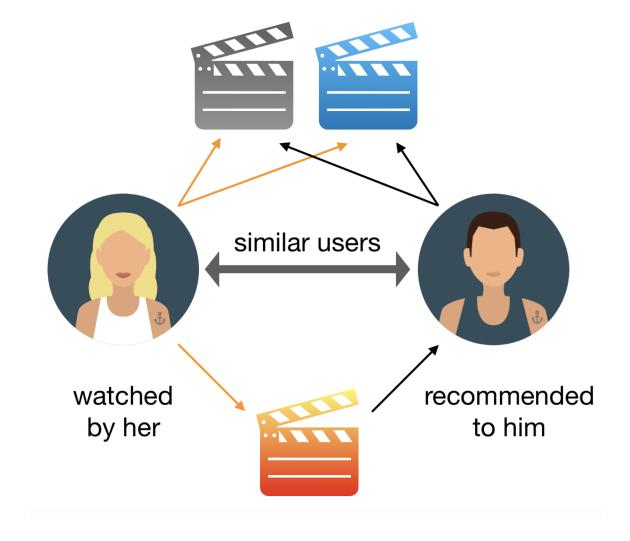
Learning for Life

- Recommendation engine is like a recommender system
- Predicts the choice of the user
- Helps the users to discover new products or content according to their past that they may not have come across otherwise



Product Recommendation





## **Recommendation System**

### **Applications of Recommendation System**





LinkedIn job matching algorithms has improved the performance by 50%

#### Netflix values recommendations at half a billion dollars to the company



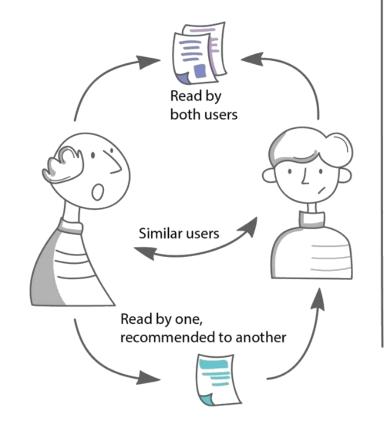


Instagram switches to use algorithmic feed

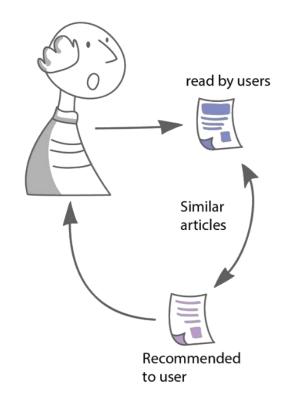
## **Types of Recommendation System**



## User-based filtering



## Content-based filtering

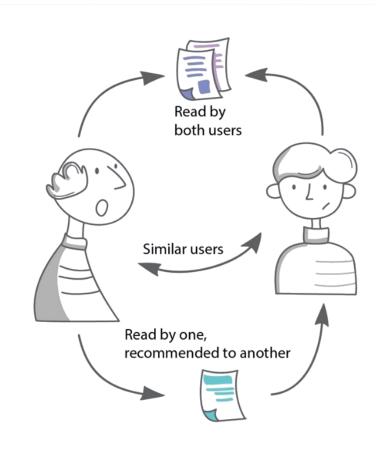


### **Types of Recommendation System: User-based**



User-based filtering

Building a model from a user's past behavior as well as similar decisions made by other users. This model is then used to predict items that the user may have an interest in

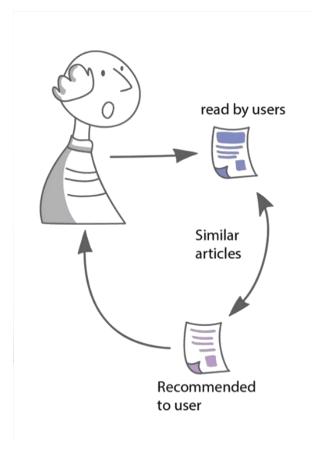


## **Types of Recommendation System: User-based**



Content-based filtering

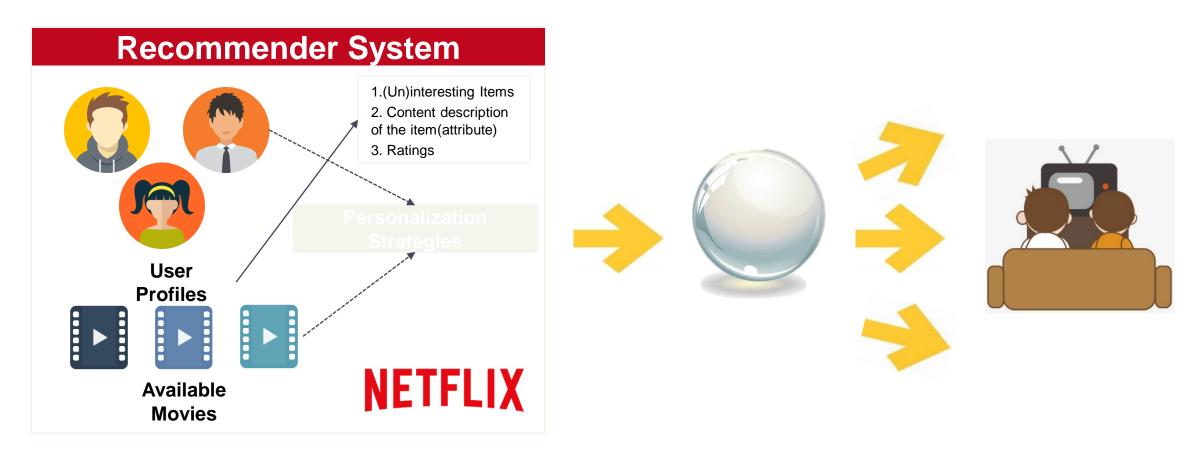
Utilizes a series of discrete characteristics of an item in order to recommend additional items with similar properties to the user



## Netflix: Hybrid Recommender System



## A Hybrid recommender system is based on both the concept UBF and CBF



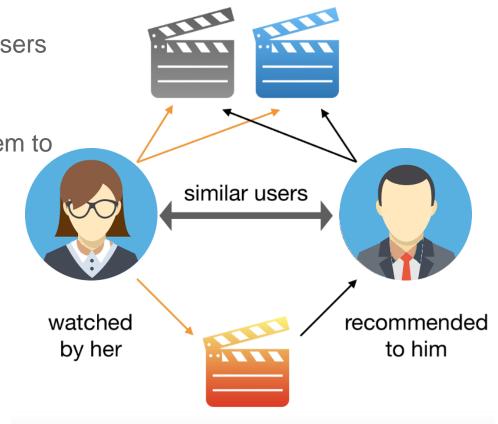


 Algorithm finds the large group of users and also searches users with similar tastes

 Algorithm looks at different things they like and combines them to create a ranked list of suggestions

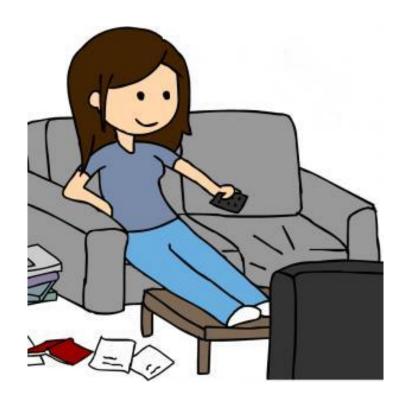
Algorithms used for measuring user or item similarity:

- ❖ K nearest neighbor (k-NN)
- Pearson Correlation



watched by both users





"Liza just watched the movie Jocker."

Let's see how the recommendation engine recommends her the movies?



Generate a list by the machine of users who have seen the following movies:





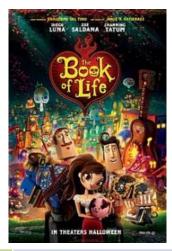




Sam	Yes	Yes	Yes	Yes
Jay	No	Yes	No	No
Ratan	No	Yes	Yes	No
Dev	No	No	No	Yes



- List of Liza's watched movies
- Find the same taste user







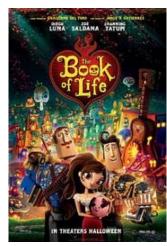


Sam	Yes	Yes	Yes	Yes
Jay	No	Yes	No	No
Ratan	No	Yes	Yes	No
Dev	No	No	No	Yes
Liza	?	Yes	?	?



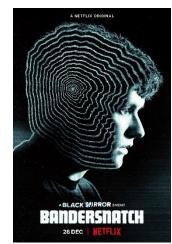
- List of Liza's watched movies
- Find the same taste user
- From the user's similarity, it's found the probable movie for Liza "Avengers" which gets more votes,

so it gets recommended to Liza









Sam	Yes	Yes	Yes	Yes
Jay	No	Yes	No	No
Ratan	No	Yes	Yes	No
Dev	No	No	No	Yes
Liza	?	Yes	?	?

1 vote

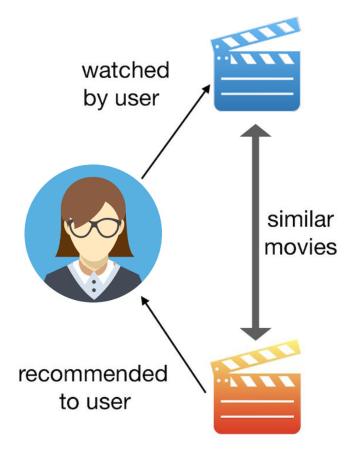
2 votes



1 vote

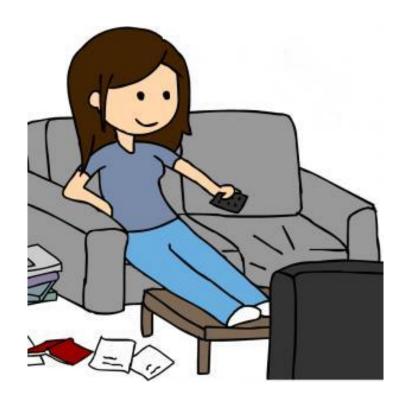


- Contents are the central entities
- Works with data that the user provides, either explicitly (rating)
   or implicitly (clicking on a link, purchase history)
- Based on that data, the suggestions are given to the user
- Engine's accuracy increases with more input given to it



## Content-Based Filtering (CBF): Example



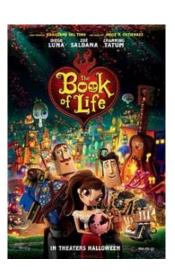


"Liza just watched the movie Jocker."

Let's see how the recommendation engine recommends her the movies?



- Generate a list of features about the
  - movies like:
    - Actors
    - Directors
    - Themes
    - Story
    - Characters





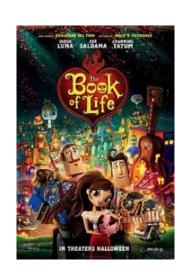




:



Compare columns of each movies with column of the movie *Jocker*









Hero	Yes	Yes	No	Yes
Horror	No	No	No	Yes
Theme	Yes	Yes	Yes	Yes
IMDB rating 8+	Yes	No	Yes	No
Comedy	Yes	No	No	No





## Demo: Movie Recommendation system



## Thank You