# Stock Recommendation

# System

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#### **Contents**

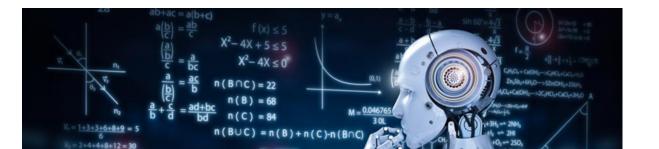


#### Introduction

- Information filtering technique
- Provides users with information which they may be interested in.
- Examples : Netflix, LinkedIn, Spotify
- Systems that are designed to recommend things to the user based on many different factors.
- These systems recommend products to the user they are most likely to purchase.







## Objectives

**01.** 

Conduct thorough research to build portfolios for people who have just started investing.

02.

Recommend a userdefined number of stocks according to the inputted stock.



## Types of Recommendation Systems



#### **Collaborative Filtering**

- Collaborative filtering uses similarities between users and items simultaneously to provide recommendations.
- Collaborative filtering models can recommend an item to user A based on the interests of a similar user B.
- For example, Netflix, Hotstar, Spotify and many more platforms use this system to recommend content to the users.



#### **Content-Based Filtering**

- Content-based filtering uses item features to recommend other items similar to what the user likes, based on their previous actions or explicit feedback.
- In this, items are ranked according to their relevancy and the most relevant ones are recommended to the user.
- For example, if a user listens to rock music every day, his Youtube recommendation feed will get full of rock music and music of related genres.

#### The Indian Stock Market

- Aggregation of buyers and sellers of stocks (also called shares),
- Represent ownership claims on businesses
- Secondary market where buying and selling between investors takes place.
   (Trading)
- Need an exchange platform for the trading to take place. The two exchanges in India are:
  - NSE (National Stock Exchange)
  - BSE (Bombay Stock Exchange)

#### BSE - Bombay Stock Exchange

- Dalal Street Mumbai
- Approximately 5439 stocks in the market,
- 8th largest stock exchange
- With an overall market capitalization in the world of more than ₹276.713 lakh crore
- Account for only around 4% of the Indian economy

## Recommendation System

#### Dataset

- 4186 Rows
- 19 Columns

Size (4186, 19) Ø df - DataFrame — □ ×

Index	S.No.	Name	CMP Rs.	P/E	Ind PE	Div Yld%	Debt/Ea	EPS 12M Rs.	. Profit a
0	1	R J Bio-Tech	7.5	nan	21.21	0	nan	-4.28	69.3
1	2	Forbes & Co	688.4	0.19	33.94	0	3.32	3392.22	8231
2	3	Nexus Surgical	14.4	18.33	32.87	0	0	0.79	530
3	4	SBEC Systems	9.43	6.99	32.87	0	nan	nan	542.
4	5	Sri Lak.Sar.Arni	51.8	11.2	7.73	0	nan	4.62	-48.
5	6	Baroda Rayon	266.05	1.68	9.61	0	11.81	163.37	420.
6	7	Standard Inds.	29.8	1.03	32.87	8.39	0.18	28.87	3245
7	8	EKI Energy	1699.15	10.28	24.98	0.29	0	165.31	1955
8	9	Dynavision	185	14.5	24.98	0	0	12.76	-3.3

#### Ratios used in the data set.

P/E = is the ratio for valuing a company that measures its current share price relative to its per-share earnings.

Ind P/E = jst like above but of the overall industry to compare

Div Yld = percentage of the share price given as dividend

Debt/equity = parameter to measure companies debt

Profit grwth % = percentage of the growth of profit annually

#### Ratios used in the data set.

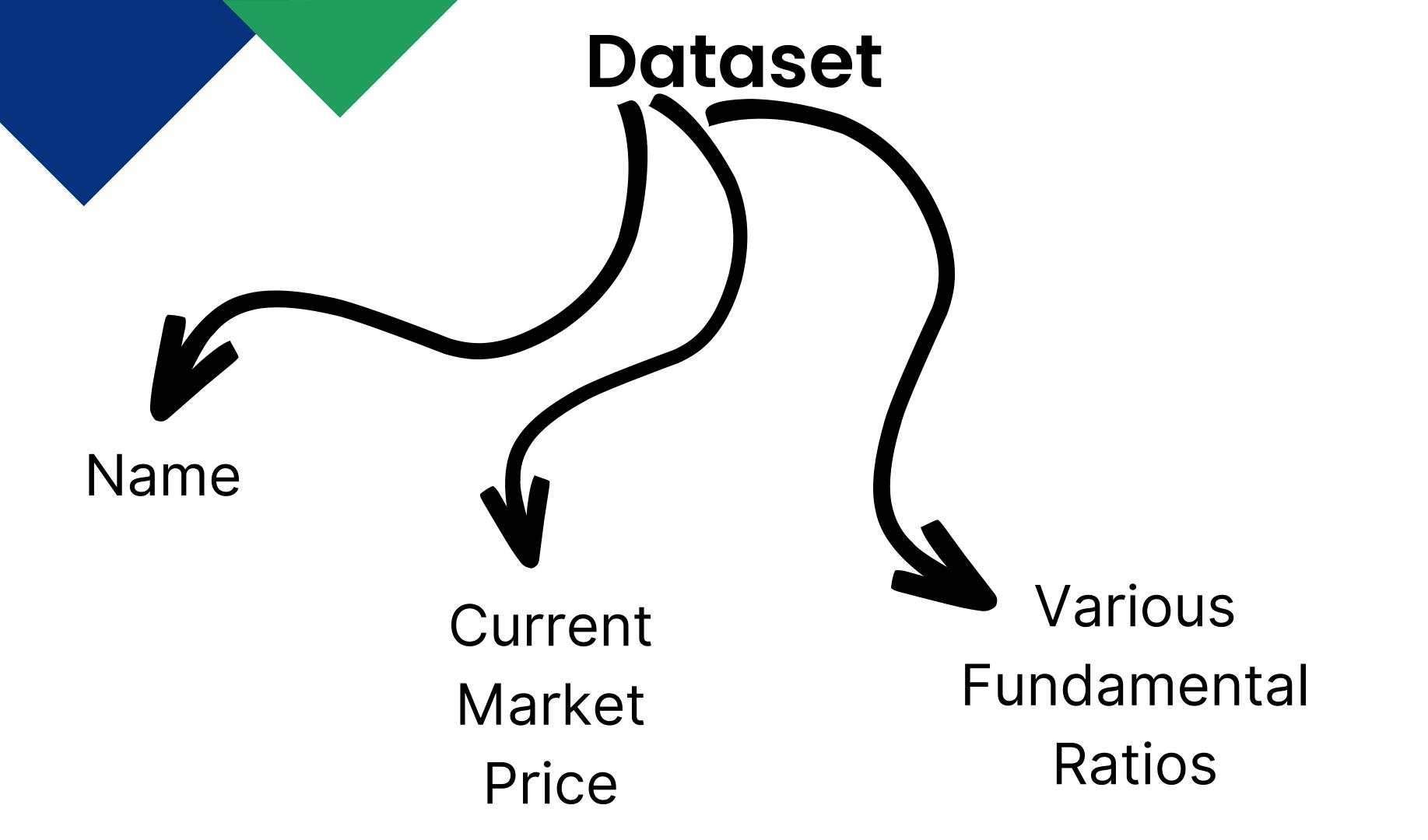
EV/EBITDA = valuation tool to compare the value of a company, debt included

sales var 10yrs= var in sales of past 10yrs

Profit var 10yrs = var in profit of past 10 yrs

ROE = return on equity

ROCE= return on capital employed.



#### Web Scraping

- Automatic method to obtain large amounts of data from websites
- Data is unstructured data in an HTML format
- Converted into structured data in a spreadsheet or a database
- Online services, particular API's or even creating your code for web scraping from scratch.
- Moneycontrol
- Financial Ratios
- Library used: Beautiful Soup

## Data Pre-processing

- Importing Dataset
- Checking for Null Values
- Imputing Null Values

```
In [18]: print("% of NULL : ",df.isnull().sum()/len(df)*100)
% of NULL : S.No.
                                      0.000000
                         0.000000
Name
CMP Rs.
                         0.000000
P/E
                         0.000000
Ind PE
                         0.000000
Div Yld%
                         0.430005
Debt/Eq
                         9.268992
EPS 12M Rs.
                         0.477783
Profit growth%
                         1.505017
                         0.000000
PEG
EV/EBITDA
                         0.453894
Sales Var 10Yrs%
                         0.000000
EBIDT Ann Rs.Cr.
                         0.023889
Profit Var 10Yrs%
                         0.000000
Avg PAT 10Yrs Rs.Cr.
                        22.240803
ROE%
                        10.033445
Prom. Hold.%
                         0.477783
Sales Var 3Yrs%
                        11.657907
ROCE%
                         3.702819
dtype: float64
```

```
In [20]: print("% of NULL : ",df.isnull().sum()/len(df)*100)
% of NULL: S.No.
                                     0.0
Name
                        0.0
CMP Rs.
                        0.0
P/E
                        0.0
Ind PE
                        0.0
Div Yld%
                        0.0
Debt/Eq
                        0.0
EPS 12M Rs.
                        0.0
Profit growth%
                        0.0
                        0.0
EV/EBITDA
                        0.0
Sales Var 10Yrs%
                        0.0
EBIDT Ann Rs.Cr.
                        0.0
Profit Var 10Yrs%
                        0.0
Avg PAT 10Yrs Rs.Cr.
                        0.0
ROE%
                        0.0
Prom. Hold.%
                        0.0
Sales Var 3Yrs%
                        0.0
ROCE%
                        0.0
dtype: float64
```

Before

After

## **Cosine Similarity**

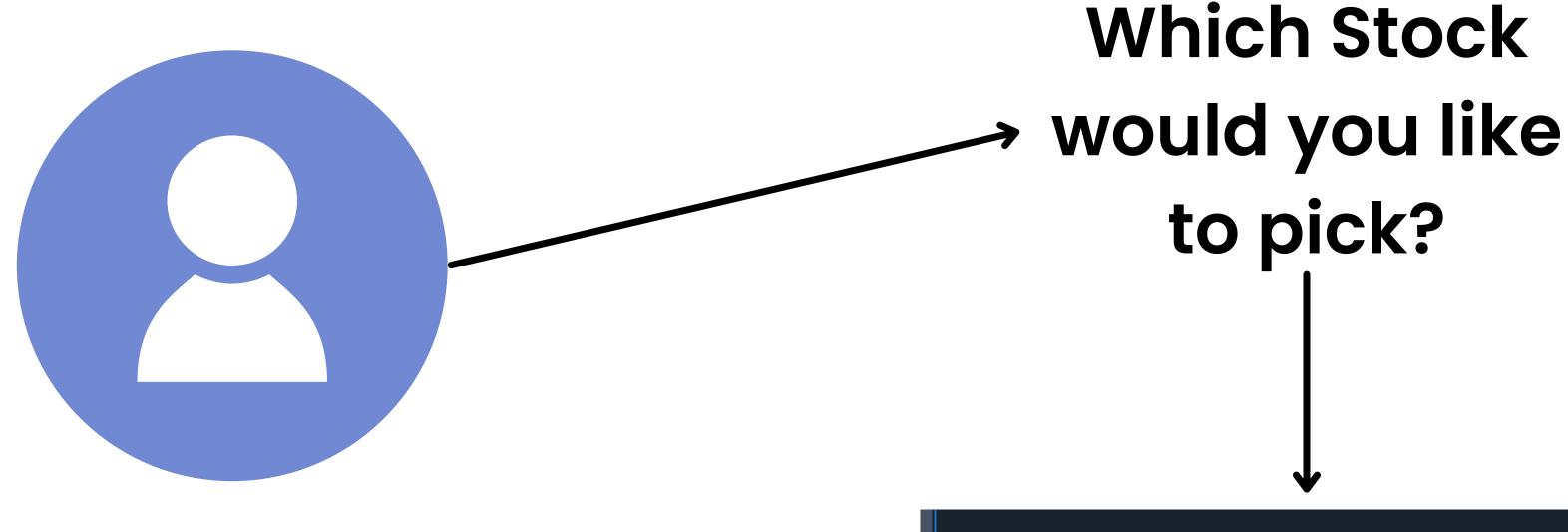
- It is a metric, helpful in determining, how similar the data objects are irrespective of their size
- Data objects in a dataset are treated as a vector.
- It's formula is:
   Cos(x, y) = x . y / ||x|| \* ||y||
   where,
- x . y = product (dot) of the vectors 'x' and 'y'.
- ||x|| and ||y|| = length of the two vectors 'x' and 'y'.
- ||x|| \* ||y|| = cross product of the two vectors 'x' and 'y'.

## Cosine Similarity

Stock Name	R J Bio-Tech	Forbes & Co	Nexus Surgical	SBEC Systems	Sri Lak.Sar.Arni	
R J Bio-Tech	1	0.143105	0.758555	0.902039	0.99084	
Forbes & Co	0.143105	1	0.379515	0.472768	0.076129	
Nexus Surgical	0.758555	0.379515	1	0.822166	0.718931	

**Cosine Similarity** HDFC Bank Ltd. Sector: Banks - Private Sector Tata Consultancy Services Ltd. Sector: Computers - Software Cipla Ltd. Sector: Pharmaceuticals Mahindra and Mahindra Ltd. USER A Sector: Auto - Cars & Jeeps

#### **Cosine Similarity**



USER A

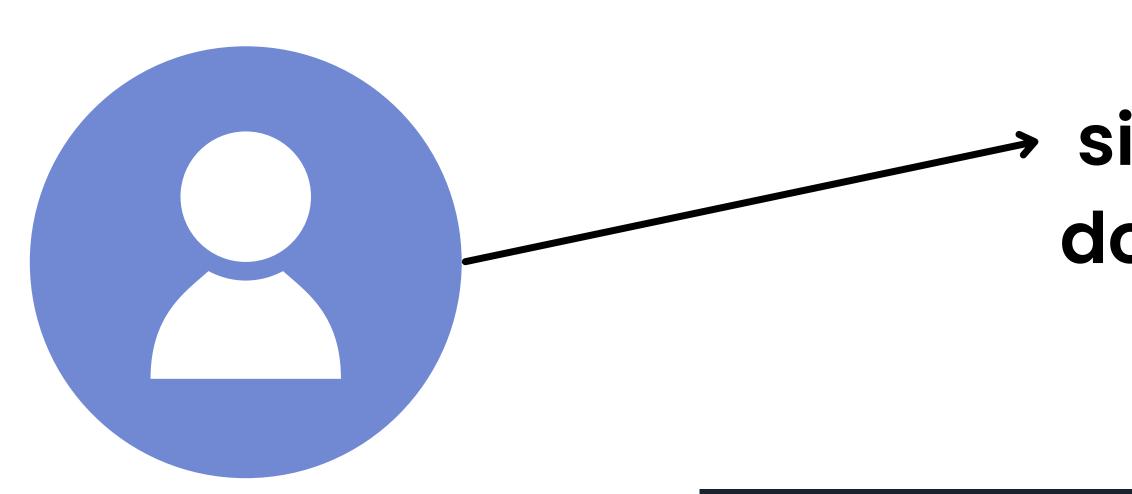
Which stock would you like to pick?
HDFC Bank

HDFC Bank

→ Stored as variable "name"

## Cosine Similarity

USER A



How many number of similar stocks do you want to find?

How many number of similar stocks do you want to find?

Stored as variable "n"

## Output

Cosine Values for 1st 10 stocks in the dataset

Stock Name	HDFC Bank
R J Bio-Tech	0.051516
Forbes & Co	0.112999
Nexus Surgical	0.076263
SBEC Systems	0.184029
Sri Lak.Sar.Arni	0.082247
Baroda Rayon	0.207064
Standard Inds.	0.083489
EKI Energy	0.259046
Dynavision	0.327087
Ghosts	0.685

Index	HDFC Bank		Index	HDFC Bank
0	0.051516		2356	1
1	0.112999		1990	0.98352
2	0.076263		1782	0.983427
3	0.184029	Rearranging this in descending order	1762	0.98184
4	0.082247		2323	0.979711
5	0.207064		760	0.97884

## Output

```
for i in n_similar:
    print(df._get_value(i, 'Name'))
```

**Stock Name** 

Amber Enterp.

**Peeti Securities** 

Skipper

Novartis India

**Smruthi Organic** 

MAS FINANC SER

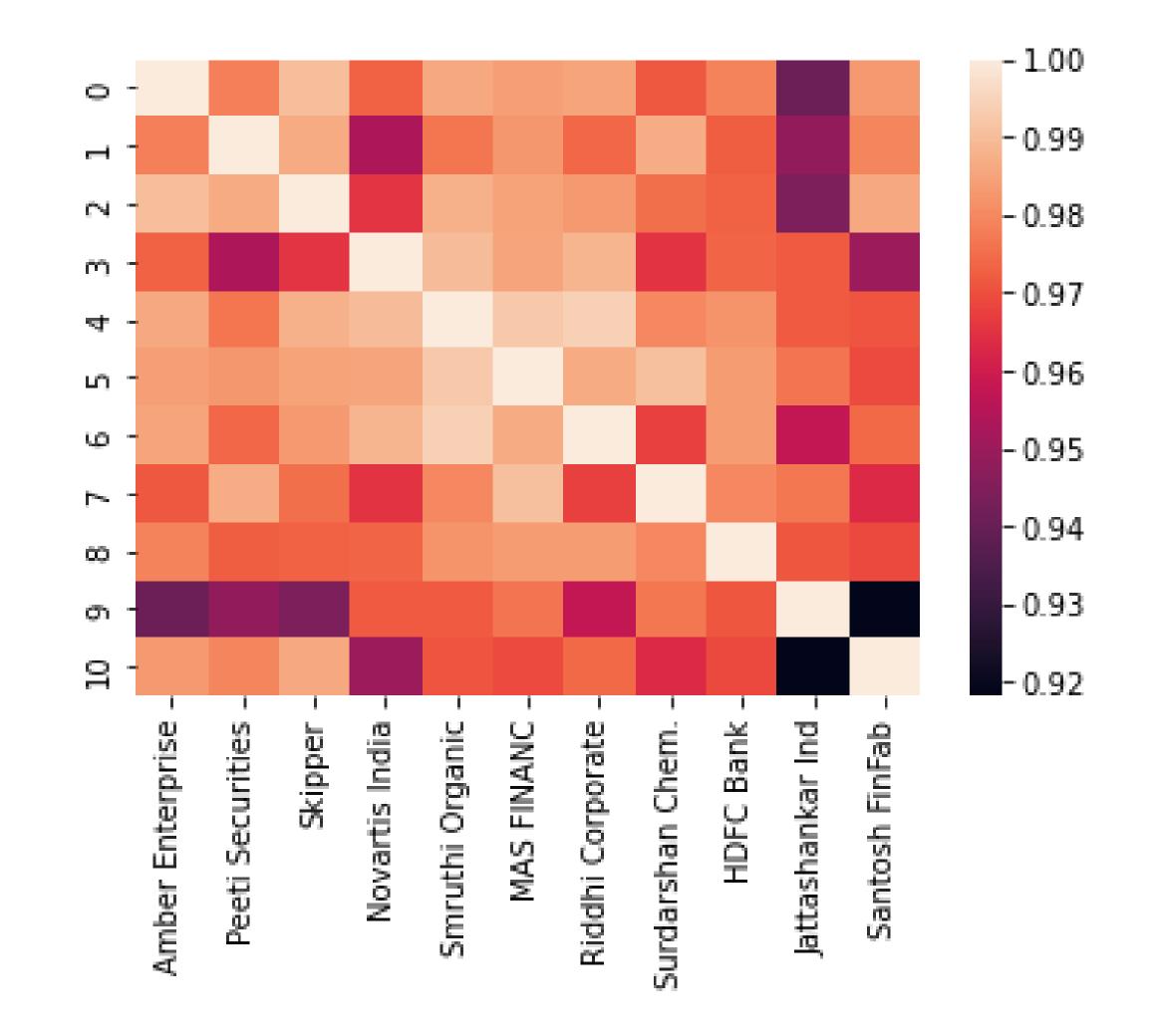
Riddhi Corporate

Sudarshan Chem.

Jattashankar Ind

Santosh Fin Fab

## Output



#### Conclusion

- Successfully recommended stocks using cosine similarity
- Web Scraping to obtain the dataset
- 'n' number of similar stocks can be invested in based on fundamental analysis

#### Limitations

- Industry Wise segregation
- Lack of Accuracy

## Future Scope

- Creation of a Bot to filter stocks industry wise and recommend them
- Stock Price prediction

# Taga You.