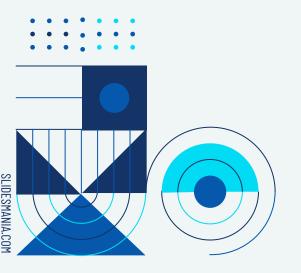
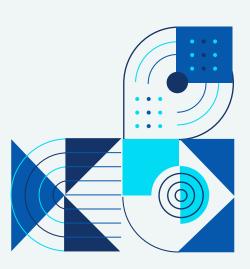
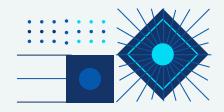


## COLLABORATIVE FILTERING AND CONTENT BASED RECOMMENDATION SYSTEM FOR SCALING BUSINESSES







## **Group Details**

#### **Supervisor:**

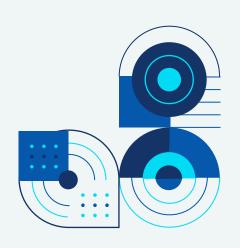
Prof Vrijendra Singh Sir

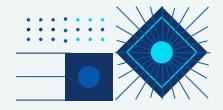
#### **Group Members:**

IIT2018031 - Raushan Raj

IIT2018069 - Suryasen Singh

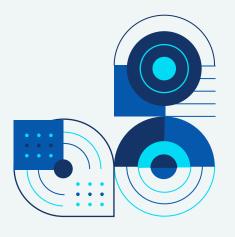
IIT2018071 - Rahul Kumar Yadav

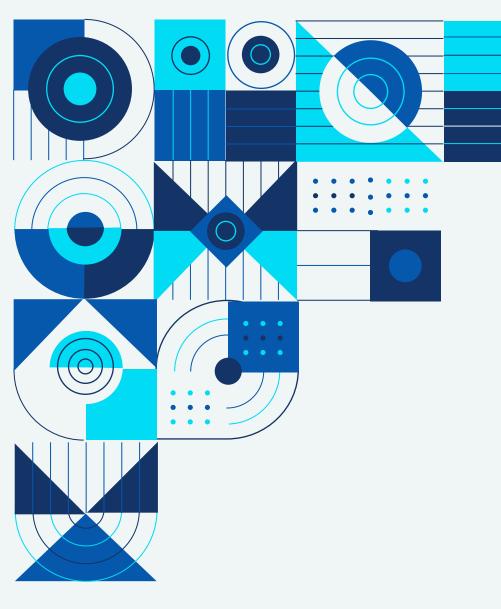




#### **Problem Statement**

- To develop a grocery and product recommendation system
- To develop a system which is majorly based on collaborative filtering technique





# Motivation

- Automation
- Human Effort
- Stable recommendation model

## LITERATURE SURVEY

| S.No. | Title  | Authors   | Year of Publication |
|-------|--|---|---------------------|
| 1.    | Movie Recommendation System     Using Item Based Collaborative     Filtering | Poonam Sharma , Lokesh Yadav                    | 2020                |
| 2.    | Deep Learning based product recommendation system.                           | Akshit Tayade1, Vidhi Sejpal2, Ankit Khivasara3 | 2021                |

## LITERATURE SURVEY

| S.No. | Title  | Authors   | Year of Publication |
|-------|--|---|---------------------|
| 3.    | Personalized Book<br>Recommendation System using<br>Machine Learning Algorithm | Dhiman Sarma, Tanni Mittra, and Mohammad<br>Shahadat Hossain            | 2021                |
| 4.    | Music Recommender System Based on Collaborative Filtering                      | AnuPrabha P S, HarsithaN, Vaishnavi K, Dr.P.Velvadivu and Dr.M.Sujithra | 2020                |
| 5.    | A Movie Recommender<br>System: MOVREC  | Manoj Kumar, D.K. Yadav, Ankur Singh, Vijay Kr. Gupta                   | 2015                |

#### **Tools and Libraries**







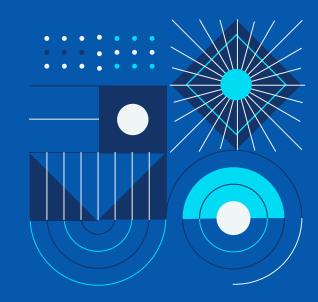








# METHODOLOGY



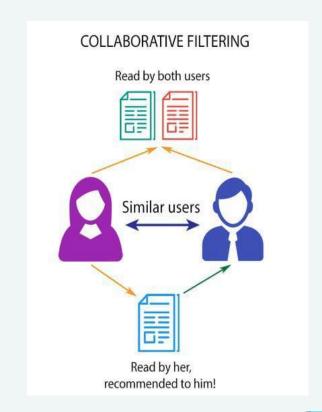


### Techniques Used

#### 1. Collaborative Filtering

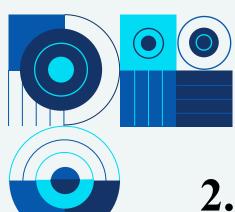
- Identify relationships between pieces of data.
- Frequently used in recommendation system.

| Users  | Item 1 | Item 2 | Item 3 | Item 4 |
|--------|--------|--------|--------|--------|
| User 1 | 5      | 4      |        | 5      |
| User 2 | 4      |        | 3      |        |
| User 3 |        | 1      |        | 2      |
| User 4 | 1      | 2      |        |        |



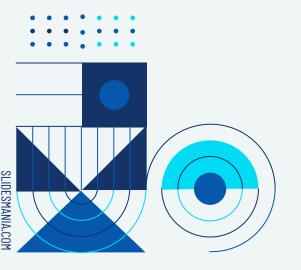
User 1 and 2 have similar interest likewise User 3 and 4 are also showing similar kind of interest.

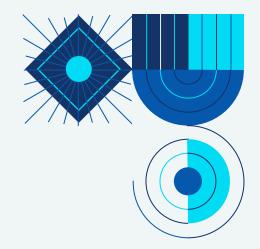


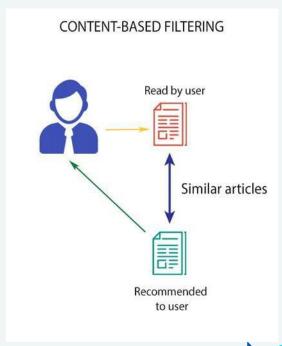


#### 2. Content Based Filtering

- Based on user preferences for product features
- It doesn't need any data about other users, since the recommendations are specific to this user











#### **Data Set Overview:**



- ☐ Amazon product ratings dataset.
- ☐ Flipkart product dataset.

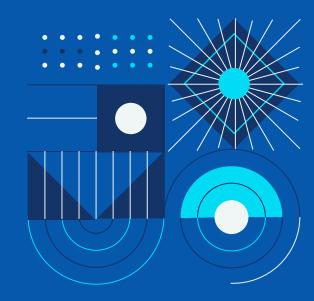




## Implementation:

Our model has been implemented using three different techniques:-

- Popularity based recommendation system
- Content based recommendation system
- Item to item based collaborative filtering



## Procedural Steps:



**Importing Libraries** 



Loading of Dataset



Data preprocessing



Generation of similarity score



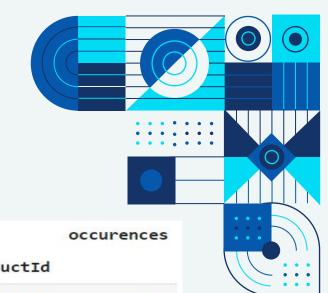


#### Popularity based recommendation

- 1. Grouping of product based on rating
- 2. Reverse sorting

3. Recommendation





|            | occurences |
|------------|------------|
| ProductId  |            |
| B001MA0QY2 | 7533       |
| B0009V1YR8 | 2869       |
| B0043OYFKU | 2477       |
| B0000YUXI0 | 2143       |
| B003V265QW | 2088       |
| B000ZMBSPE | 2041       |
| B003BQ6QXK | 1918       |
| B004OHQR1Q | 1885       |
| B00121UVU0 | 1838       |
| B000FS05VG | 1589       |

Top 10 most popular products

#### Content based recommendation:

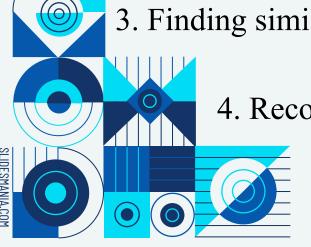
- 1. Preparation of documents
- 2. Text vectorization

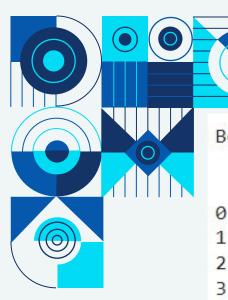
```
from sklearn.feature_extraction.text import TfidfVectorizer
documents = df2['description']
tf = TfidfVectorizer(stop_words='english')
tfidf_matrix = tf.fit_transform(documents)
```



4. Recommendation

from sklearn.metrics.pairwise import cosine\_similarity
cosine\_sim = cosine\_similarity(tfidf\_matrix, tfidf\_matrix)



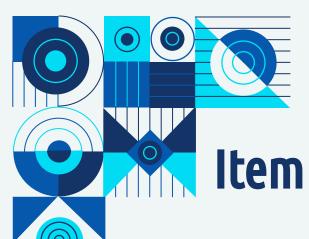


Below are the recommendations for the product - HRS ULTIMATE MEN Chest Pads

|    | Product name                                    | Similarity score |
|----|---|------------------|
| 0  | HRS ULTIMATE MEN Chest Pads                     | 1.000000         |
| 1  | HRS ULTIMATE BOY Chest Pads                     | 0.891930         |
| 2  | HRS ULTIMATE BOY Elbow Pads                     | 0.740572         |
| 3  | HRS CLUB BOY Thigh Pads                         | 0.721257         |
| 4  | Parth Collection English Y-pad                  | 0.171612         |
| 5  | India Inc Women's Solid Casual Shirt            | 0.159057         |
| 6  | TeeMoods Casual Full Sleeve Striped Women's Top | 0.154496         |
| 7  | TeeMoods Casual Full Sleeve Striped Women's Top | 0.154496         |
| 8  | TeeMoods Casual Full Sleeve Striped Women's Top | 0.154496         |
| 9  | TeeMoods Casual Full Sleeve Striped Women's Top | 0.154496         |
| 10 | TeeMoods Casual Full Sleeve Striped Women's Top | 0.154496         |

## Top 10 similar products





## Item - Item based collaborative filtering

- 1. Preparation of Utility matrix.
- 2. Removal of insignificant values and handling null values

```
utility_matrix = df3.pivot_table(values='Rating', index='UserId', columns='ProductId')
utility_matrix = utilty_matrix.dropna(thresh=10, axis=1).fillna(0)
```

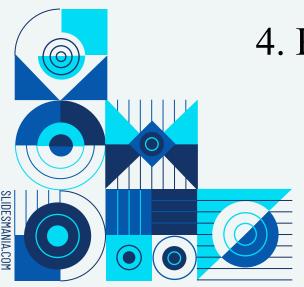






3. Finding relationship between products

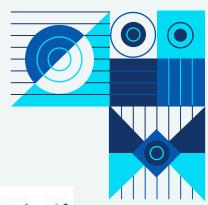
```
df_item_similarity = utility_matrix.corr(method='pearson')
df_item_similarity.head(10)
```



4. Preparation of user history and recommendation



## Top 10 recommendations



```
Below are the recommendation based on the User history - [('1403790965', 4), ('B0002VHBTU', 5), ('535795531X', 2)]
```

```
2.499806
B0002VHBTU
1403790965
              1.499566
              0.075251
B000052YKM
B000280SI0
              0.041307
B00028MLG6
              0.034251
              0.031249
B00014DMQE
B0000C0XL8
              0.030753
B000142P12
              0.023551
B0000DD8VH
             -0.000462
B0000589YB
             -0.000480
dtype: float64
```

