# What are recommendation engines?

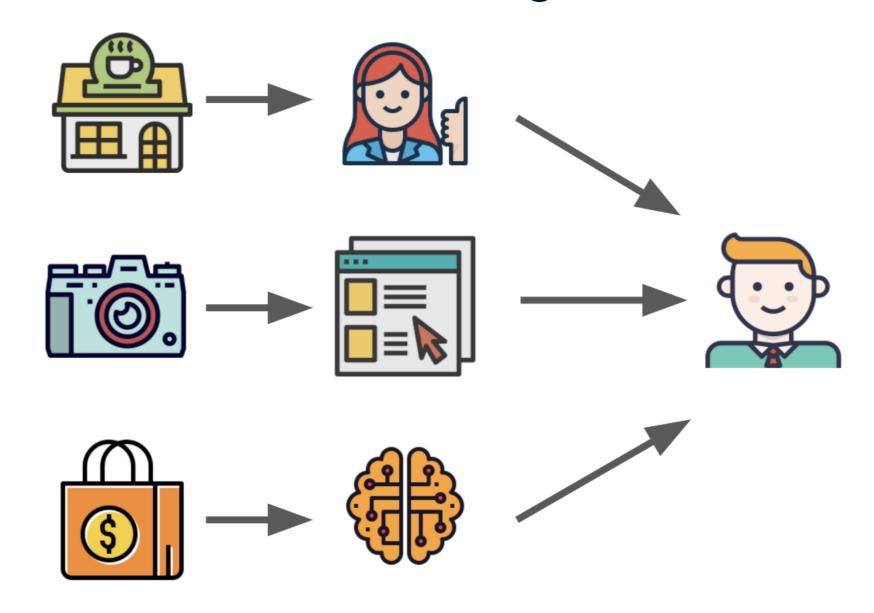
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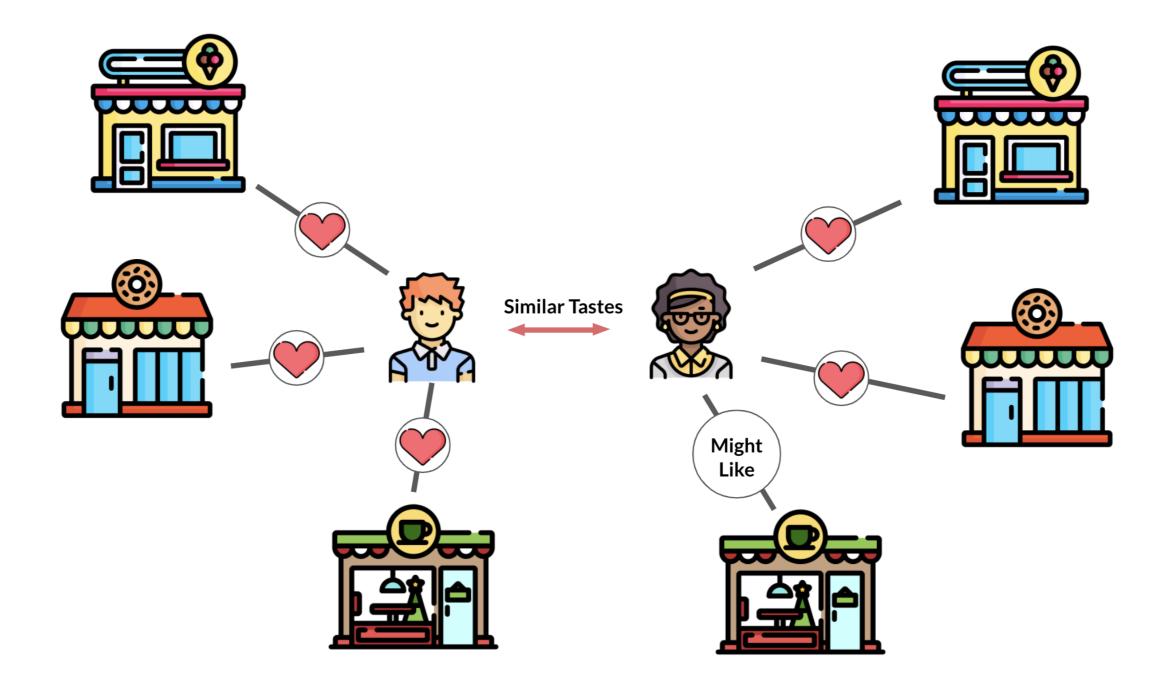
Rob O'Callaghan
Director of Data



#### What are recommendation engines?

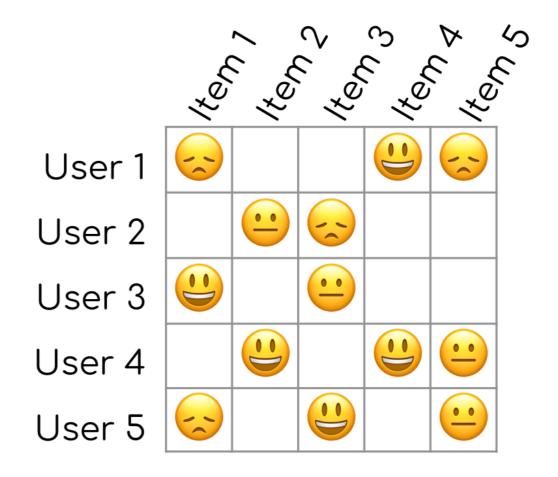


#### What kind of data do I need?



#### What kind of data do I need?

User ID	Item ID	Review
User 1	Item 1	
User 1	Item 4	
User 2	Item 2	•••
User 2	Item 4	



## What are recommendation engines useful for? Recommendation Engines Other Statistical Models

What movie should a viewer watch?

Will a diner enjoy a restaurant?

Will a movie sell a lot of tickets?

How much is a house worth?



#### Implicit vs. explicit data Implicit Feedback

A user often selecting the same genre of music.

An item is repeatedly purchased by the same consumer.

A viewer gives up halfway on a movie they were watching.

#### **Explicit Feedback**

Ratings given on movie sites like Rotten Tomatoes.

A user telling a social media platform that an ad is not relevant to their interests.

A user gives a thumbs up or thumbs down on YouTube.



### Let's practice!

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# Non-personalized recommendations

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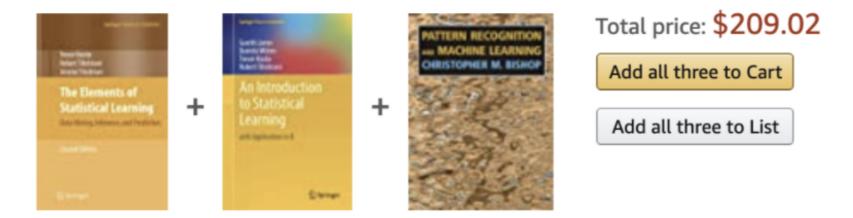


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#### Non-personalized ratings

#### Frequently bought together



- i These items are shipped from and sold by different sellers. Show details
- ▼ This item: The Elements of Statistical Learning: Data Mining, Inference, and Prediction, Second Edition... by Trevor Hastie Hardcover \$71.84
- An Introduction to Statistical Learning: with Applications in R (Springer Texts in Statistics) by Gareth James Hardcover \$65.39
- ✓ Pattern Recognition and Machine Learning (Information Science and Statistics) by Christopher M. Bishop Hardcover \$71.79

#### Finding the most popular items

book\_df DataFrame:

User	Book		
User_233	The Great Gatsby		
User_651	The Catcher in the Rye		
User_131	The Lord of the Rings		
User_965	Little Women		
User_651	Fifty Shades of Grey		
•••	•••		



#### Finding the most popular items

```
book_df['book'].value_counts()
```

```
40 Shades of Grey 524
Harry Potter and the Sorcerer's Stone 487
The Da Vinci Code 455
The Twilight Saga 401
Lord of the Rings 278
...
```



#### Finding the most popular items

```
print(book_df.value_counts().index)
```

user\_ratings DataFrame:

User	Book	Rating
User_233	The Great Gatsby	3.0
User_651	The Catcher in the Rye	5.0
User_131	The Lord of the Rings	3.0
User_965	Little Women	4.0
User_651	Fifty Shades of Grey	2.0
•••	•••	•••



```
avg_rating_df = user_ratings[["book", "rating"]].groupby(['book']).mean()
avg_rating_df.head()
```

```
title
Hamlet 4.1
The Da Vinci Code 2.1
Gone with the Wind 4.2
Fifty Shades of Grey 1.2
Wuthering Heights 3.9
...
```

```
sorted_avg_rating_df = avg_rating_df.sort_values(by="rating", ascending=False)
sorted_avg_rating_df.head()
```

```
title
The Girl in the Fog 5.0
Behind the Bell 5.0
Across the River and into the Trees 5.0
The Complete McGonagall 5.0
What Is to Be Done? 5.0
```



```
(user_ratings['title']=='The Girl in the Fog').sum()
```

1

```
(user_ratings['title']=='Valley of the Dolls').sum()
```

1

```
(user_ratings['title']=='Across the River and into the Trees').sum()
```

1



#### Finding the most liked popular items

```
book_frequency = user_ratings["book"].value_counts()
print(book_frequency)
```

```
40 Shades of Grey 524
Harry Potter and the Sorcerer's Stone 487
...
```

```
frequently_reviewed_books = book_frequency[book_frequency > 100].index
print(frequently_reviewed_books)
```

```
Index([u'The Lord of the Rings', u'To Kill a Mockingbird', u'Of Mice and Men', u'1984', u'Hamlet'])
```



#### Finding the most liked popular items

```
frequent_books_df = user_ratings_df[user_ratings_df["book"].isin(frequently_reviewed_books)]
```

```
frequent_books_avgs = frequently_reviewed_books[["title", "rating"]].groupby('title').mean()
print(frequent_books_avgs.sort_values(by="rating", ascending=False).head())
```

```
rating
title
To Kill a Mockingbird 4.7
1984. 4.7
Harry Potter and the Sorcerer's Stone 4.6
...
```



### Let's practice!

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# Non-personalized suggestions

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#### Identifying pairs

```
O User_223 The Great Gatsby <---| Read by the same user

1 User_223 The Catcher in the Rye <---|

2 User_131 The Lord of the Rings

3 User_965 Little Women <---| Read by the same user

4 User_965 Fifty Shades of Grey <---|

.....
```

#### Permutations versus combinations

User	book_title	
User_233	The Great Gatsby	
User_233	The Catcher in the Rye	

#### To:

	Book A	Book B
0	The Great Gatsby	The Catcher in the Rye
1	The Catcher in the Rye	The Great Gatsby

Books seen with The Great Gatsby -> The Catcher in the Rye

Books seen with The Catcher in the Rye -> The Great Gatsby



```
from itertools import permutations

def create_pairs(x):
    return pairs
```



• permutations(list, length\_of\_permutations)) Generates iterable object containing all permutations

```
from itertools import permutations

def create_pairs(x):
   pairs = list(permutations(x.values, 2))

return pairs
```

- permutations(list, length\_of\_permutations)) Generates iterable object containing all permutations
- list() Converts this object to a usable list

- permutations(list, length\_of\_permutations)) Generates iterable object containing all permutations
- list() Converts this object to a usable list
- pd.DataFrame() Converts the list to a DataFrame containing the columns book\_a and book\_b

#### Applying the function to the data

```
book_pairs = book_df.groupby('userId')['book_title'].apply(perm_function)
print(book_pairs.head())
```

		book_a	book_b
userId			
User_223	0	The Great Gatsby	The Catcher in the Rye
	1	The Catcher in the Rye	The Great Gatsby
User_965	0	Little Women	40 Shades of Grey
	1	40 Shades of Grey	Little Women
User_773	0	The Twilight Saga	Harry Potter and the Sorcerer's Stone
			•••



#### Cleaning up the results

```
book_pairs = book_pairs.reset_index(drop=True)
print(book_pairs.head())
```

```
book_b
                     book_a
0
           The Great Gatsby
                                               The Catcher in the Rye
     The Catcher in the Rye
1
                                                     The Great Gatsby
3
               Little Women
                                                    40 Shades of Grey
                                                         Little Women
          40 Shades of Grey
5
          The Twilight Saga
                               Harry Potter and the Sorcerer's Stone
```



#### Counting the pairings

```
pair_counts = book_pairs.groupby(['book_a', 'book_b']).size()
```

```
book_a book_b
The Twilight Saga Fifty Shades of Grey 16
Pride and Prejudice 12
...
```

```
pair_counts_df = pair_counts.to_frame(name = 'size').reset_index()
print(pair_counts_df.head())
```

```
book_abook_bsize1 The Twilight SagaFifty Shades of Grey162 The Twilight SagaPride and Prejudice12...
```



#### Looking up recommendations

```
pair_counts_sorted = pair_counts_df.sort_values('size', ascending=False)
```

```
pair_counts_sorted[pair_counts_sorted['book_a'] == 'Lord of the Rings']
```

```
book_a book_b size

137 Lord of the Rings The Hobbit 12

147 Lord of the Rings Harry Potter and the Sorcerer's Stone 10

143 Lord of the Rings The Colour of Magic 9

...
```



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