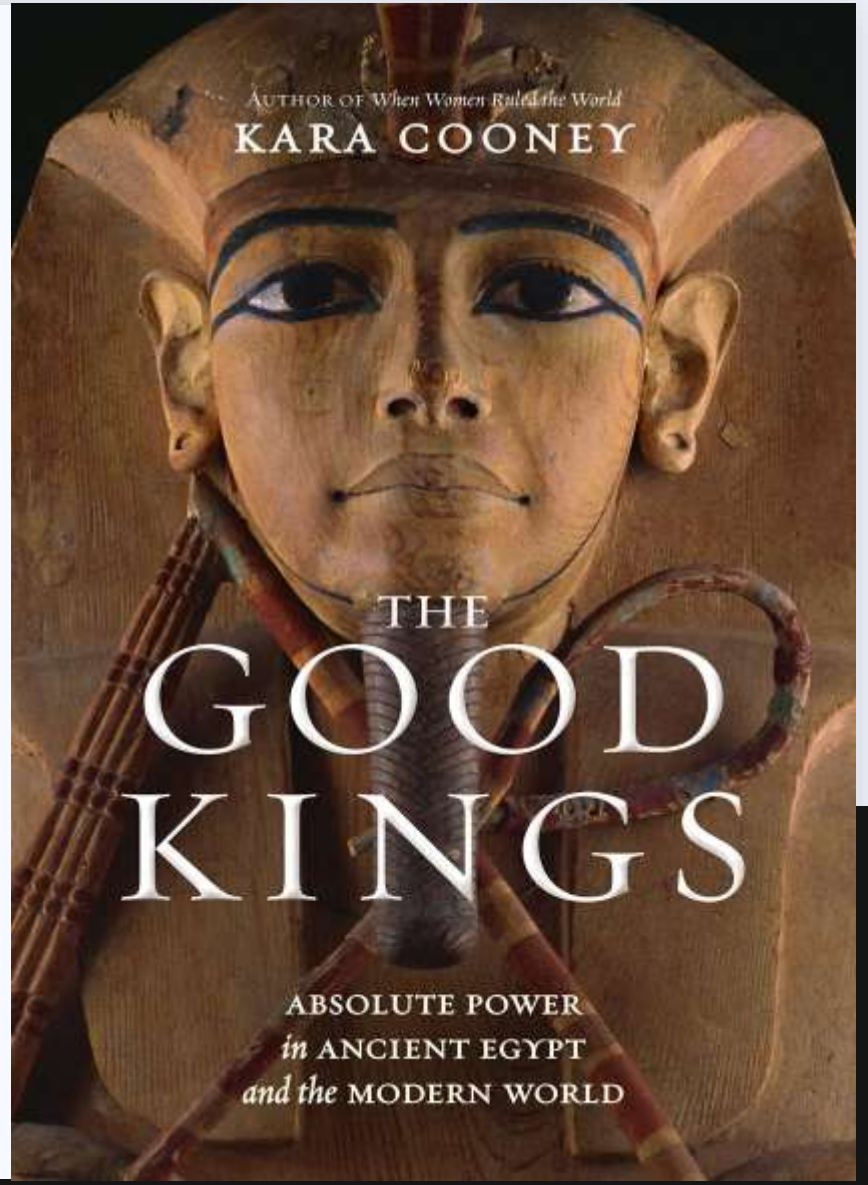
BOOK RECOMMENDER



INTRODUCTION

The project is primarily focused on building a **book recommendation system**, which typically falls under the domain of **collaborative filtering**. Collaborative filtering is an unsupervised learning approach used for recommendation tasks. It's neither regression nor classification, as the goal is to predict user preferences or item similarities rather than a specific numeric value or category.

The **k-nearest neighbors (KNN) algorithm** is commonly used for collaborative filtering in recommendation systems. In this case, it's used to find similar books based on user preferences or book characteristics. KNN is a type of unsupervised learning algorithm that belongs to the broader category of instance-based learning.

In the context of a book recommendation system, the primary learning task is generally considered unsupervised, as the goal is to find **patterns or similarities in user behavior or item features without explicit labels**.

THE DATASET

The dataset was obtained from Kaggle: https://www.kaggle.com/datasets/jealousleopard/goodreadsbooks?resource=download

- 12 columns
- 11126 unique entries

EXPLORING DATA

```
df.head()
```

	bookID	title	authors	average_rating	isbn	isbn13	language_code	num_pages	ratings_count	text_reviews_count	publication_date
0	1	Harry Potter and the Half-Blood Prince (Harry ...	J.K. Rowling/Mary GrandPré	4.57	0439785960	9780439785969	eng	652	2095690	27591	9/16/2006
1	2	Harry Potter and the Order of the Phoenix (Har...	J.K. Rowling/Mary GrandPré	4.49	0439358078	9780439358071	eng	870	2153167	29221	9/1/2004
2	4	Harry Potter and the Chamber of Secrets (Harry...	J.K. Rowling	4.42	0439554896	9780439554893	eng	352	6333	244	11/1/2003

EXPLORING DATA

```
: #Exploring the dataset:
  print(df.info())
  print(df.describe())
```

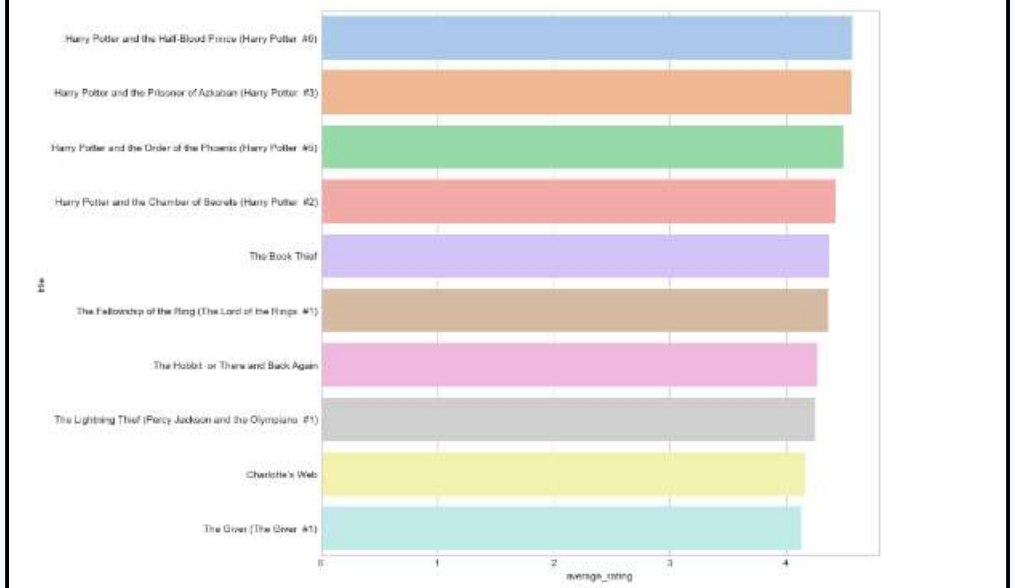
```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 11123 entries, 0 to 11122
Data columns (total 12 columns):
 #   Column              Non-Null Count  Dtype
---  ------              --------------  -----
 0   bookID              11123 non-null  int64
 1   title               11123 non-null  object
 2   authors             11123 non-null  object
 3   average_rating      11123 non-null  float64
 4   isbn                11123 non-null  object
 5   isbn13              11123 non-null  int64
 6   language_code       11123 non-null  object
 7     num_pages         11123 non-null  int64
 8   ratings_count       11123 non-null  int64
 9   text_reviews_count  11123 non-null  int64
 10  publication_date    11123 non-null  object
 11  publisher           11123 non-null  object
dtypes: float64(1), int64(5), object(6)
memory usage: 1.0+ MB
None
              bookID  average_rating        isbn13     num_pages  \
count   11123.000000    11123.000000  1.112300e+04  11123.000000
mean    21310.856963        3.934075  9.759880e+12    336.405556
std     13094.727252        0.350485  4.429758e+11    241.152626
min         1.000000        0.000000  8.987060e+09      0.000000
25%     10277.500000        3.770000  9.780345e+12    192.000000
```

NULL VALUES

```
6]: #finding null values:
    df.isnull().sum()

6]: bookID                0
    title                 0
    authors               0
    average_rating        0
    isbn                  0
    isbn13                0
    language_code         0
      num_pages           0
    ratings_count         0
    text_reviews_count    0
    publication_date      0
    publisher             0
    dtype: int64
```

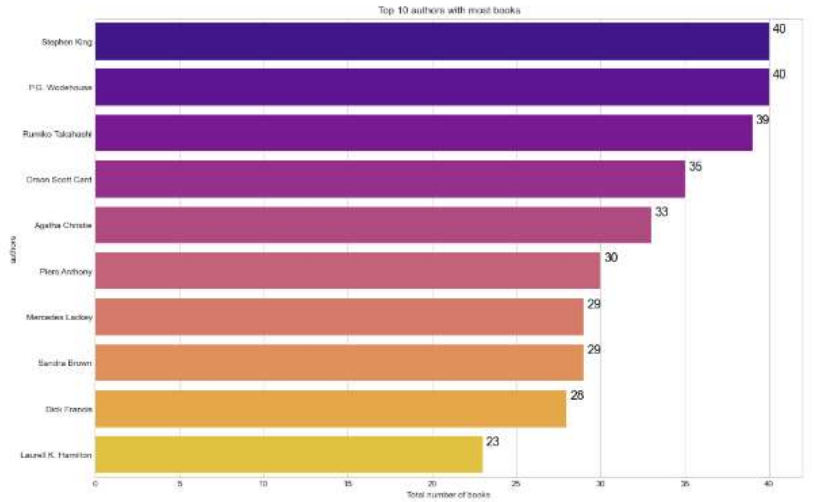
DATA VISUALIZATION

```
# Select rows from the DataFrame where 'ratings_count' is greater than 1000000
top_ten = df[df['ratings_count'] > 1000000]

# Sort the selected rows based on 'average_rating' in descending order
top_ten_sorted = top_ten.sort_values(by='average_rating', ascending=False)

# Set the style of the plot to 'seaborn-whitegrid'
plt.style.use('seaborn-whitegrid')

# Create a new figure with a size of 10x10 inches
plt.figure(figsize=(10, 10))

# Select the top 10 rows from the sorted DataFrame based on 'average_rating'
data = top_ten_sorted.head(10)

sns.barplot(x="average_rating", y="title", data=data, palette='pastel')
plt.show()
```



DATA VISUALIZATION

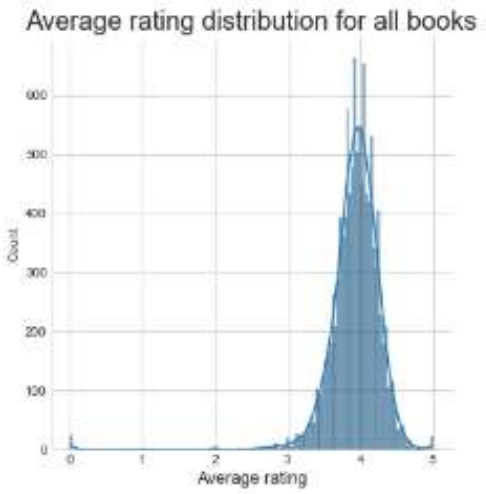
```
26]: #Top 10 authors

     # Group the DataFrame by 'authors' and count the number of books for each author
     most_books = df.groupby('authors')['title'].count().reset_index()

     # Sort the grouped data by the count of books in descending order and select the top 10
     most_books_sorted = most_books.sort_values('title', ascending=False).head(10)

     # Set the 'authors' column as the index for better readability in the plot
     most_books_sorted = most_books_sorted.set_index('authors')

     plt.figure(figsize=(15, 10))
     ax = sns.barplot(x=most_books_sorted['title'], y=most_books_sorted.index, palette='plasma')
     ax.set_title("Top 10 authors with most books")
     ax.set_xlabel("Total number of books")

     # Add annotations (total number of books) to each bar in the plot
     totals = []
     for i in ax.patches:
         totals.append(i.get_width())

     total = sum(totals)

     for i in ax.patches:
         # Display the count of books next to each bar, with a slight offset for better visibility
         ax.text(i.get_width() + 0.2, i.get_y() + 0.2, str(round(i.get_width())), fontsize=15, color='black')

     plt.show()
```



DATA VISUALIZATION

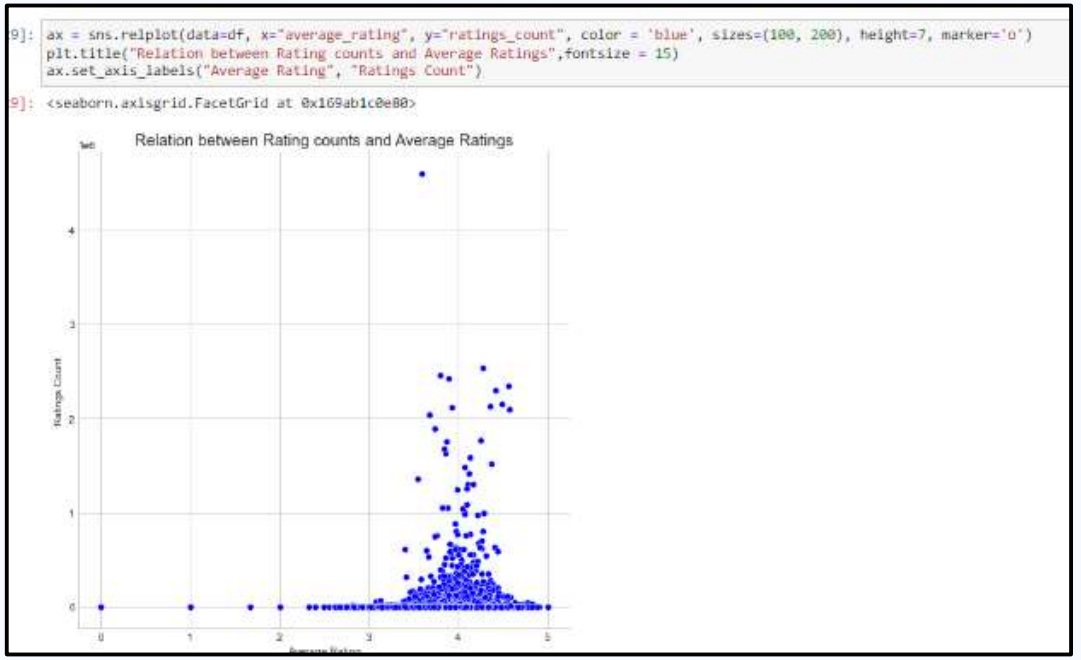
[23]:
```
#Average rating distribution:

# Convert 'average_rating' to float
df['average_rating'] = df['average_rating'].astype(float)

# Create a figure-level distribution plot
fig = plt.figure(figsize=[15, 10])
sns.displot(df['average_rating'], kde=True)
plt.title('Average rating distribution for all books', fontsize=20)
plt.xlabel('Average rating', fontsize=13)
plt.show()
```

```
<Figure size 1080x720 with 0 Axes>
```

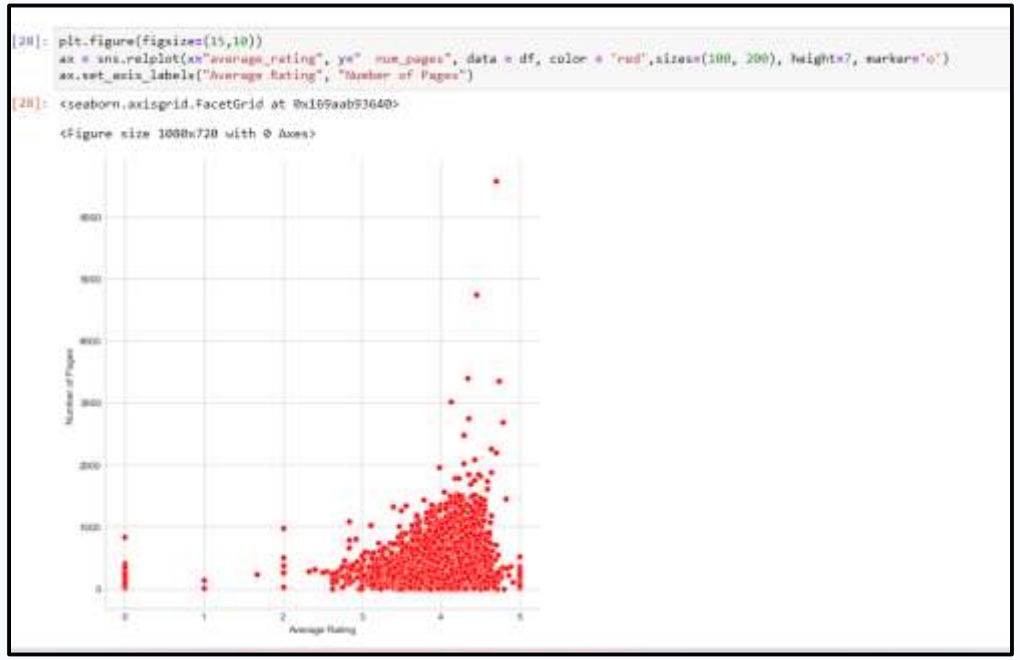


DATA VISUALIZATION

```
9]: ax = sns.relplot(data=df, x="average_rating", y="ratings_count", color = 'blue', sizes=(100, 200), height=7, marker='o')
    plt.title("Relation between Rating counts and Average Ratings",fontsize = 15)
    ax.set_axis_labels("Average Rating", "Ratings Count")

9]: <seaborn.axisgrid.FacetGrid at 0x169ab1c0e80>
```



DATA VISUALIZATION



THE MODEL

11123 rows × 34 columns

```
[2]: #Scaling the values
     from sklearn.preprocessing import MinMaxScaler
     min_max_scaler = MinMaxScaler()
     features = min_max_scaler.fit_transform(features)

     #Building the model
     model = neighbors.NearestNeighbors(n_neighbors=6, algorithm='ball_tree')
     model.fit(features)
     dist, idlist = model.kneighbors(features)
```

```
2]: def BookRecommender(book_name):
```

EXAMPLE OUTPUT

```
Recommender(book_name):
itialize an empty list to store recommended book names
_list_name = []

nd the index of the input book_name in the DataFrame
_id = df2[df2['title'] == book_name].index

tract the first (and only) element from the index array
_id = book_id[0]

erate through the indices of the k-nearest neighbors for the input book_
newid in idlist[book_id]:
# Append the title of each recommended book to the book_list_name
book_list_name.append(df2.loc[newid].title)

turn the list of recommended book names
rn book_list_name

he BookRecommender function with a specific book name
s = BookRecommender('Harry Potter and the Half-Blood Prince (Harry Potte

y the list of recommended book names
s[1:]
```

```
Potter and the Order of the Phoenix (Harry Potter  #5)',
llowship of the Ring (The Lord of the Rings  #1)',
Potter and the Chamber of Secrets (Harry Potter  #2)',
Potter and the Prisoner of Azkaban (Harry Potter  #3)',
ghtning Thief (Percy Jackson and the Olympians  #1)']
```

OUTPUT WITH BOOK COVER