

# Capstone Project - III

## Team 3 : BOOK RECOMMENDATION SYSTEM

### Team Members

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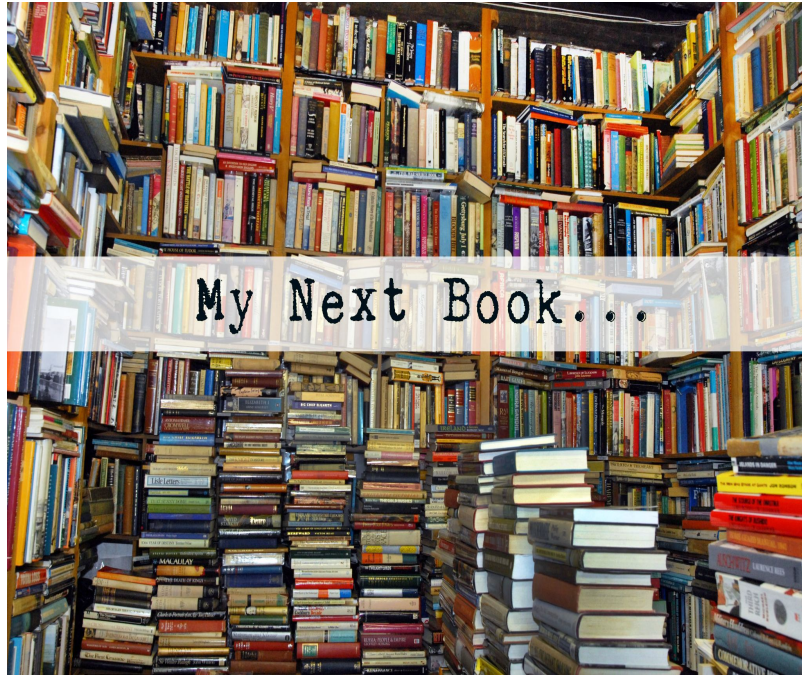
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- Analysis of different datasets
- Data Cleaning
- Outlier treatment
- Imputing missing values
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# Problem Statement



During the last few decades, with the rise of Youtube, Amazon, Netflix, and many other such web services, recommender systems have become much more important in our lives in terms of providing highly personalized and relevant content.

**The main objective is to create a recommendation system to recommend relevant books to users based on popularity and user interests.**

# Data Summary

The dataset is comprised of three csv files:: User\_df, Books\_df, Ratings\_df

Users\_dataset.

- User-ID (unique for each user)
  - Location (contains city, state and country separated by commas)
  - Age
- Shape of Dataset - (278858, 3)

Books\_dataset.

- ISBN (unique for each book)
  - Book-Title
  - Book-Author
  - Year-Of-Publication
  - Publisher
  - Image-URL-S
  - Image-URL-M
  - Image-URL-L
- Shape of Dataset - (271360, 8)

Ratings\_dataset.

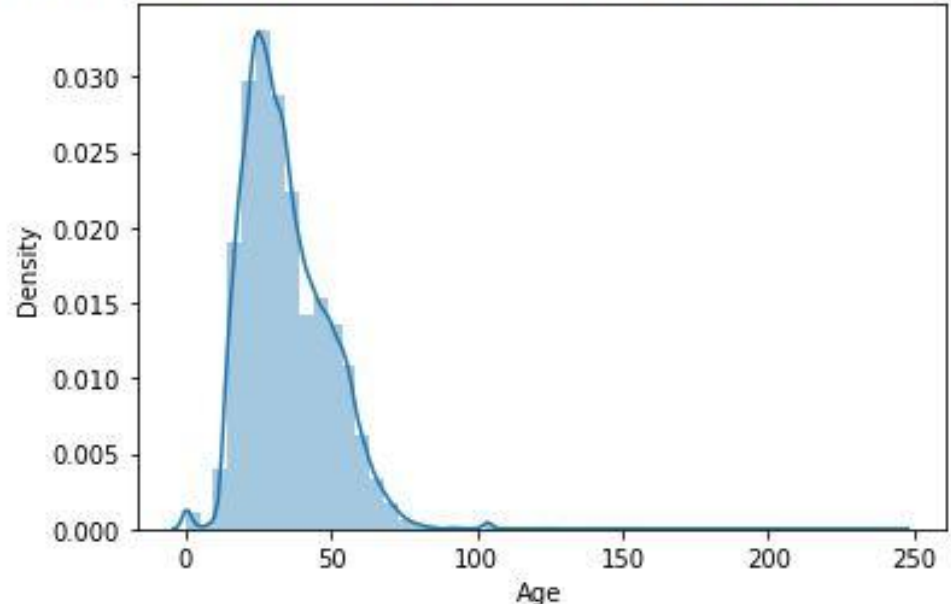
- User-ID
  - ISBN
  - Book-Rating
- Shape of Dataset - (1149780, 3)

# Observations from Users\_df (Age)

```
1 sns.distplot(users.Age)
```

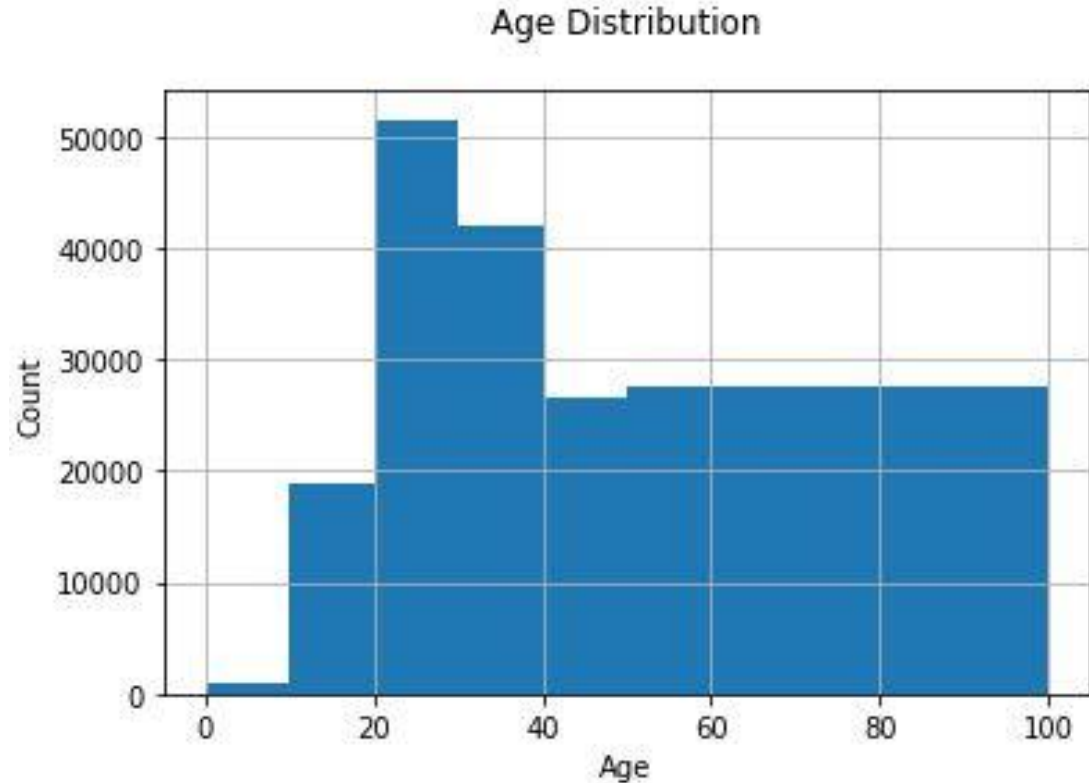
- The Age range given here is from 0 To 250.
- Outliers in the Age column.

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f5a11ac00d0>



# Observations from Users\_df (Age)

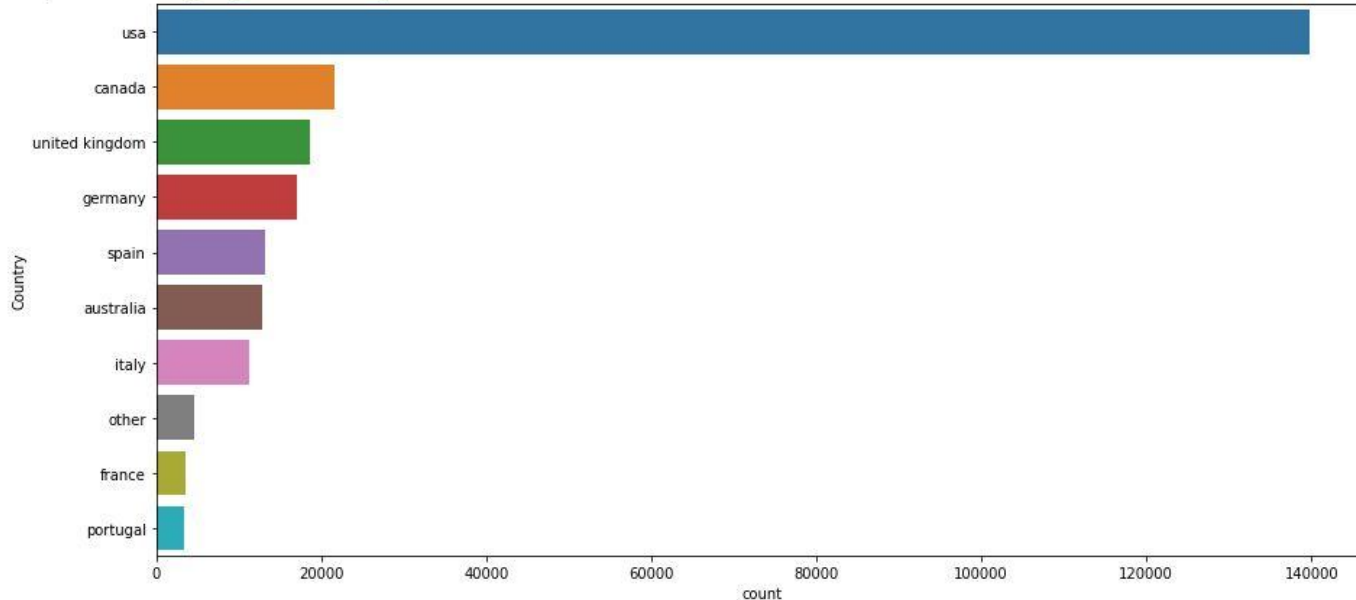
- The Age range distribution is right skewed
- Most active readers lie in age group 20- 40



# Observations from Users\_df (Location)

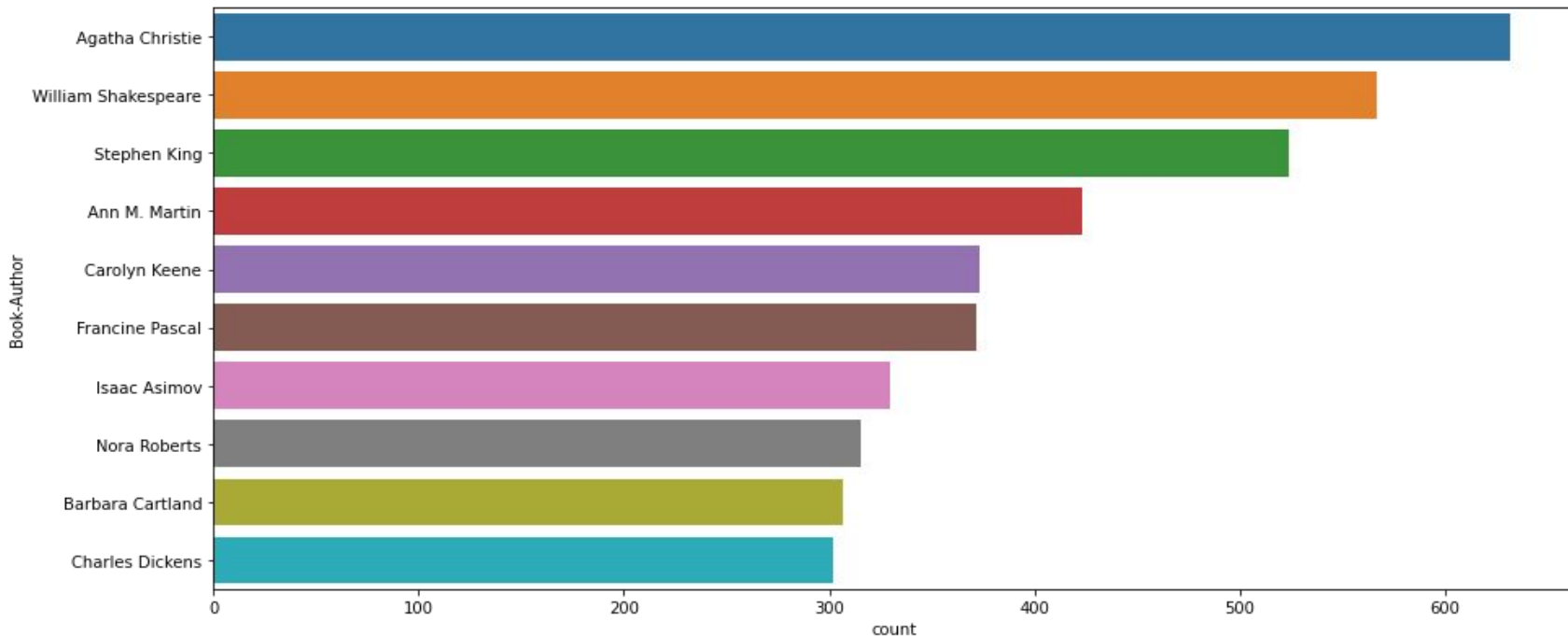
- Splitting Location column and analysing country.
- Most active readers are from USA.

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f5a118b2750>



# Observations from Book\_df (Authors)

Agatha Christie wrote highest number of books in our given dataset

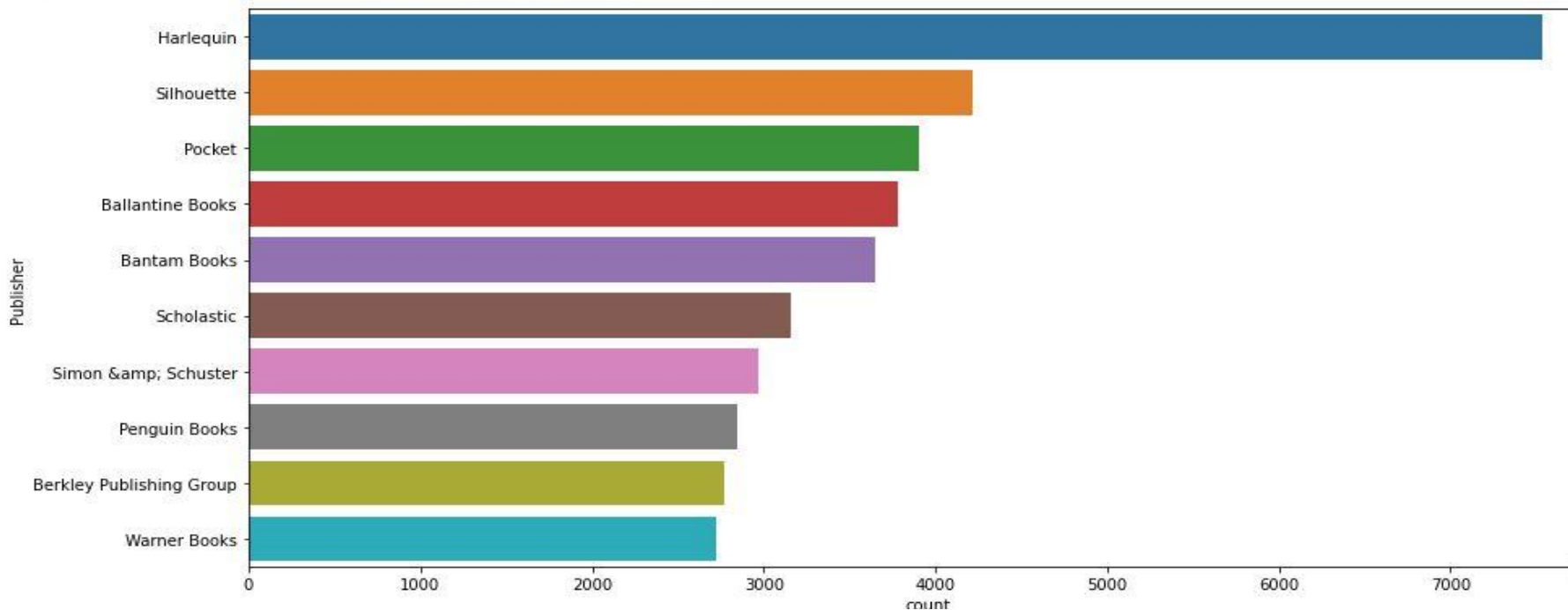




# Observations from Book\_df (Publishers)

Harlequin published highest number of books in our given dataset

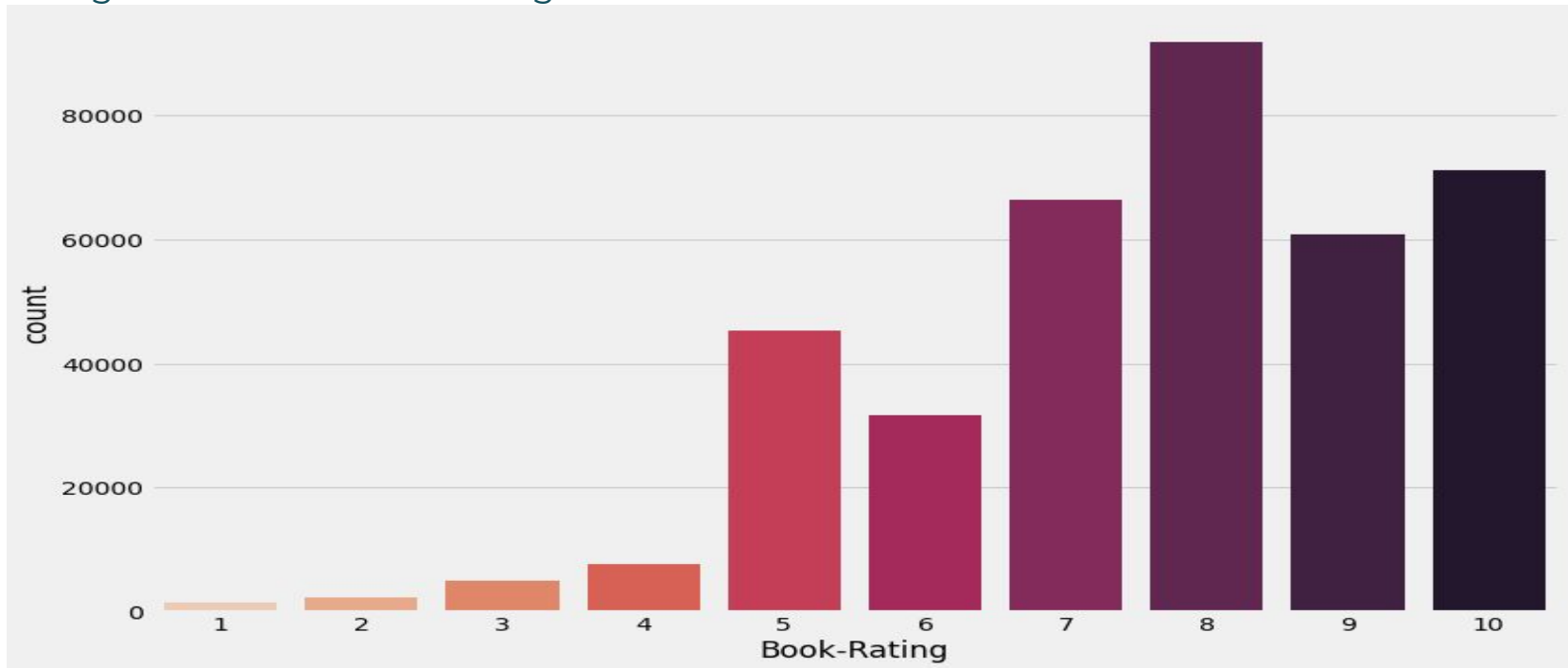
```
<matplotlib.axes._subplots.AxesSubplot at 0x7f5a1194a3d0>
```



# Observations from Ratings\_df (Book\_Rating)



- Higher ratings are more common amongst users
- Rating 8 has been rated the highest number of times



# Data Cleaning

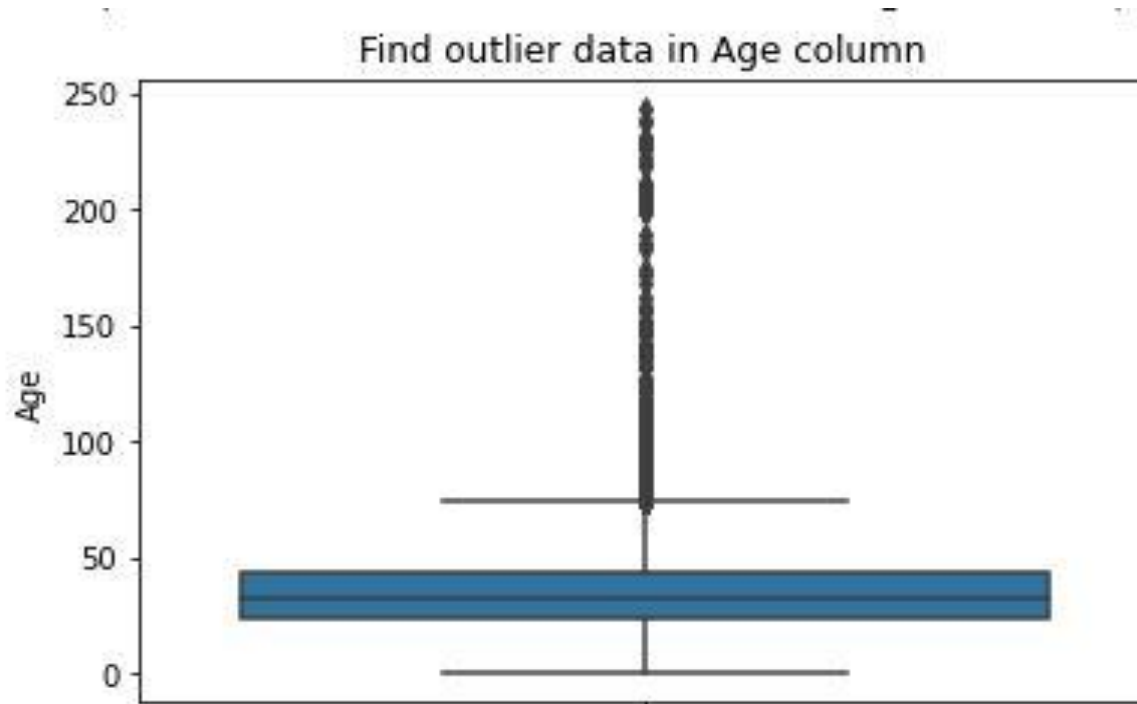
## 1. Null Value Imputation:

Age column has 40% missing values

|   | index    | Missing Values | % of Total Values | Data_type |
|---|----------|----------------|-------------------|-----------|
| 0 | Age      | 110762         | 39.72             | float64   |
| 1 | User-ID  | 0              | 0.00              | int64     |
| 2 | Location | 0              | 0.00              | object    |

# Imputing missing values

- Outliers in Age column
- Age has positive Skewness (right tail) so we can use median to fill Nan values,



# Data Cleaning

## 1. Null Value Imputation:

```
books_df.isnull().sum()
```

|                     |       |
|---------------------|-------|
| ISBN                | 0     |
| Book-Title          | 0     |
| Book-Author         | 1     |
| Year-Of-Publication | 0     |
| Publisher           | 2     |
| Image-URL-S         | 0     |
| Image-URL-M         | 0     |
| Image-URL-L         | 3     |
| dtype:              | int64 |

# Replacing strings by int values

|        | ISBN       | Book-<br>Title  | Book-<br>Author | Year-Of-<br>Publication |   |
|--------|------------|---|-----------------|-------------------------|---|
| 209538 | 078946697X | DK<br>Readers:<br>Creating<br>the X-<br>Men,<br>How It<br>All Beg...    | 2000            | DK<br>Publishing<br>Inc | h |
| 221678 | 0789466953 | DK<br>Readers:<br>Creating<br>the X-<br>Men,<br>How<br>Comic<br>Book... | 2000            | DK<br>Publishing<br>Inc | h |

# Different Models

## 1.)Popularity Based Recommendation

Book weighted average formula:

$$\text{Weighted Rating(WR)}=[vR/(v+m)]+[mC/(v+m)]$$

Where,

v is the number of votes for the books;

m is the minimum votes required to be listed in the chart;

R is the average rating of the book; and

C is the mean vote across the whole report.

# Different Models

| Book-Title   | Total_No_Of_Users_Rated | Avg_Rating | Score    |
|--|-------------------------|------------|----------|
| 0 Harry Potter and the Goblet of Fire (Book 4)                               | 137                     | 9.262774   | 8.741835 |
| 1 Harry Potter and the Sorcerer's Stone (Harry Potter (Paperback))           | 313                     | 8.939297   | 8.716469 |
| 2 Harry Potter and the Order of the Phoenix (Book 5)                         | 206                     | 9.033981   | 8.700403 |
| 3 To Kill a Mockingbird  | 214                     | 8.943925   | 8.640679 |
| 4 Harry Potter and the Prisoner of Azkaban (Book 3)                          | 133                     | 9.082707   | 8.609690 |
| 5 The Return of the King (The Lord of the Rings, Part 3)                     | 77                      | 9.402597   | 8.596517 |
| 6 Harry Potter and the Prisoner of Azkaban (Book 3)                          | 141                     | 9.035461   | 8.595653 |
| 7 Harry Potter and the Sorcerer's Stone (Book 1)                             | 119                     | 8.983193   | 8.508791 |
| 8 Harry Potter and the Chamber of Secrets (Book 2)                           | 189                     | 8.783069   | 8.490549 |
| 9 Harry Potter and the Chamber of Secrets (Book 2)                           | 126                     | 8.920635   | 8.484783 |
| 10 The Two Towers (The Lord of the Rings, Part 2)                            | 83                      | 9.120482   | 8.470128 |
| 11 Harry Potter and the Goblet of Fire (Book 4)                              | 110                     | 8.954545   | 8.466143 |
| 12 The Fellowship of the Ring (The Lord of the Rings, Part 1)                | 131                     | 8.839695   | 8.441584 |
| 13 The Hobbit : The Enchanting Prelude to The Lord of the Rings              | 161                     | 8.739130   | 8.422706 |
| 14 Ender's Game (Ender Wiggins Saga (Paperback))                             | 117                     | 8.837607   | 8.409441 |
| 15 Tuesdays with Morrie: An Old Man, a Young Man, and Life's Greatest Lesson | 200                     | 8.615000   | 8.375412 |
| 16 Charlotte's Web (Trophy Newbery)  | 68                      | 9.073529   | 8.372037 |
| 17 Dune (Remembering Tomorrow)   | 75                      | 8.973333   | 8.353301 |
| 18 A Prayer for Owen Meany   | 181                     | 8.607735   | 8.351465 |
| 19 Fahrenheit 451  | 164                     | 8.628049   | 8.346969 |



# Different Models

## 2.)Model based collaborative filtering

### SVD

```
test_rmse    1.602152
test_mae     1.239638
fit_time     5.437686
test_time    0.472132
dtype: float64
```

### NMF

```
test_rmse    2.626532
test_mae     2.242070
fit_time     8.057059
test_time    0.546524
dtype: float64
```

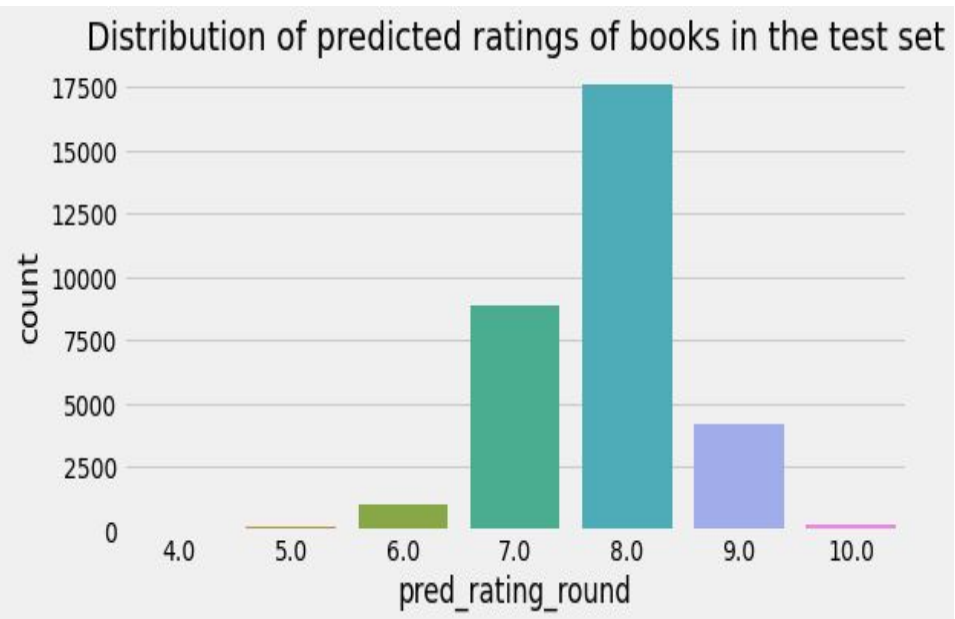
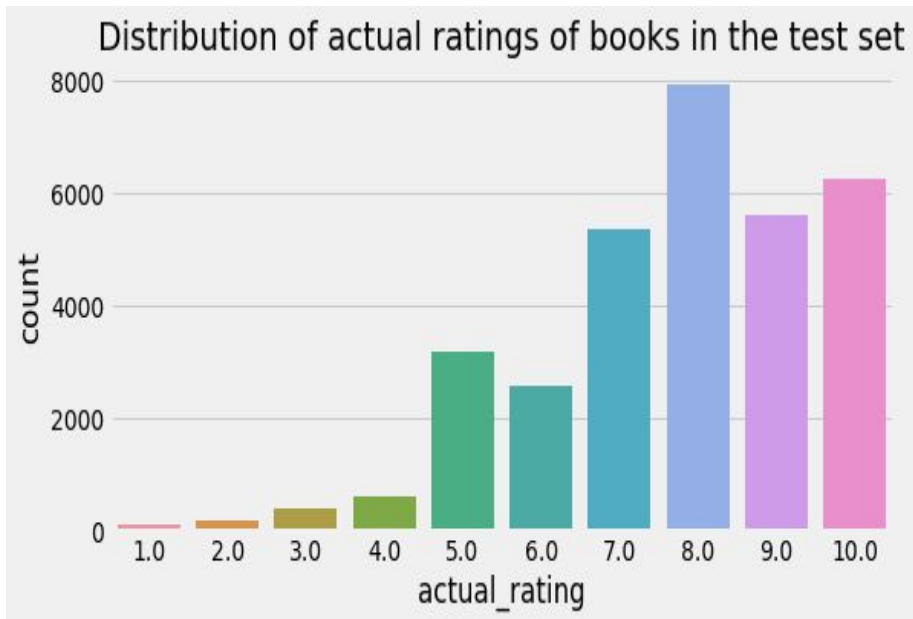
# Different Models

## SVD Model Results

|       | user_id | isbn       | actual_rating | pred_rating | impossible | pred_rating_round | abs_err  |
|-------|---------|------------|---------------|-------------|------------|-------------------|----------|
| 15594 | 62862   | 0385335482 | 8.0           | 7.978811    | False      | 8.0               | 0.021189 |
| 30626 | 193938  | 0385497288 | 8.0           | 7.882566    | False      | 8.0               | 0.117434 |
| 27451 | 234401  | 0812540026 | 8.0           | 7.316338    | False      | 7.0               | 0.683662 |
| 14130 | 89602   | 0060987529 | 8.0           | 6.649098    | False      | 7.0               | 1.350902 |
| 18074 | 86189   | 0312186886 | 10.0          | 7.303280    | False      | 7.0               | 2.696720 |

# Different Models

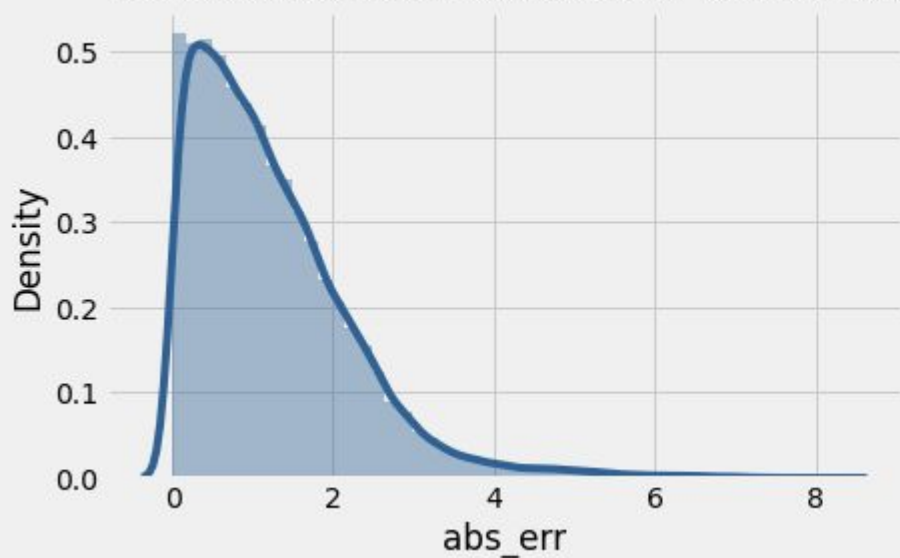
## SVD Model Results



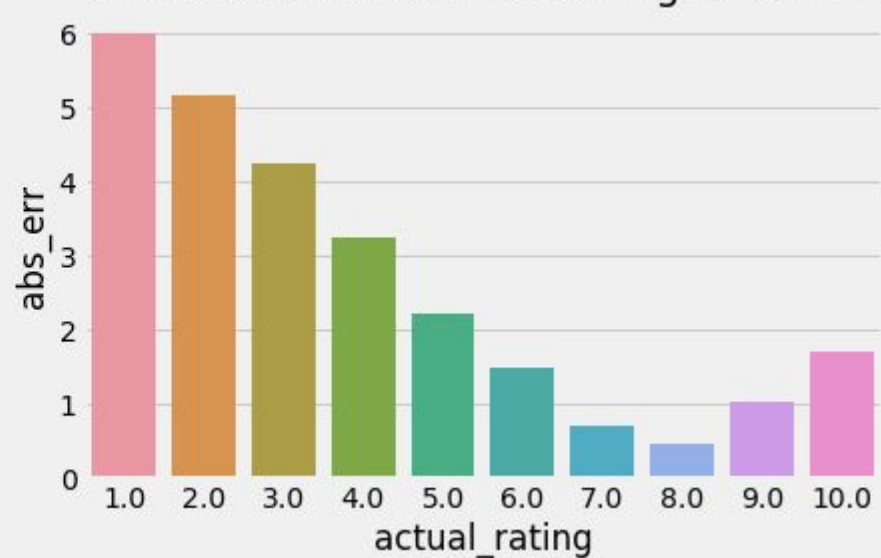
# Different Models

## SVD Model Results

Distribution of absolute error in test set



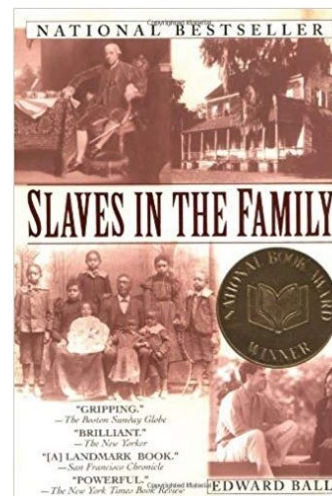
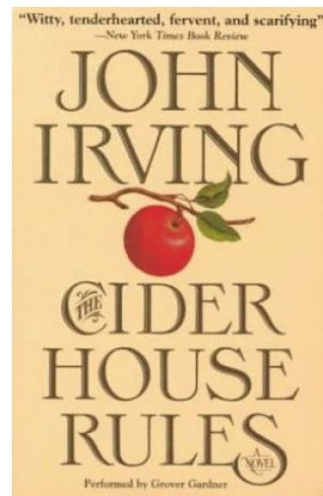
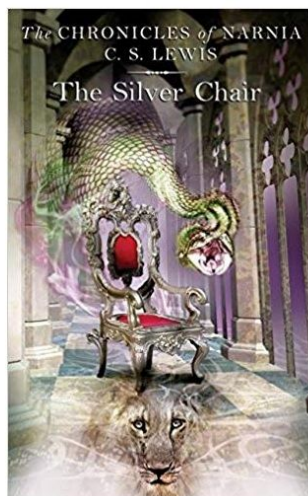
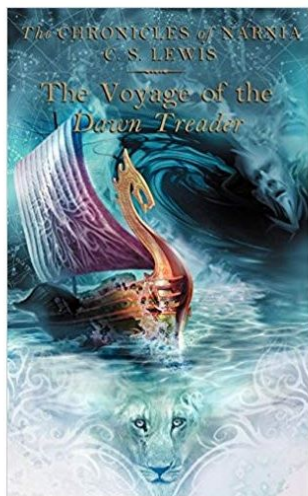
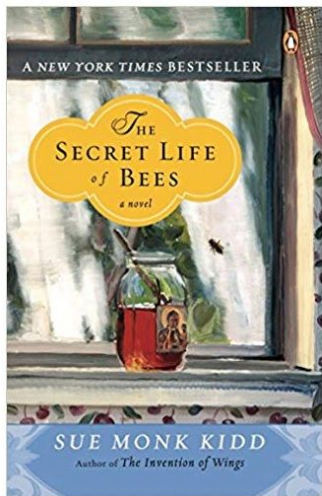
Mean absolute error for rating in test set



# Different Models

User-ID - 193458

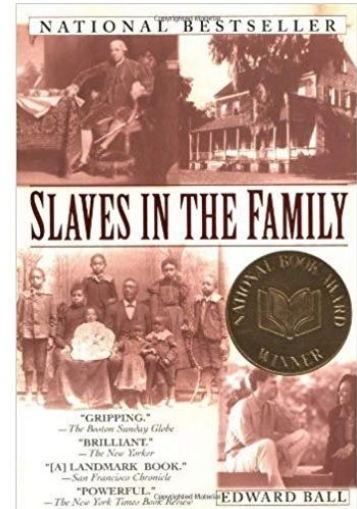
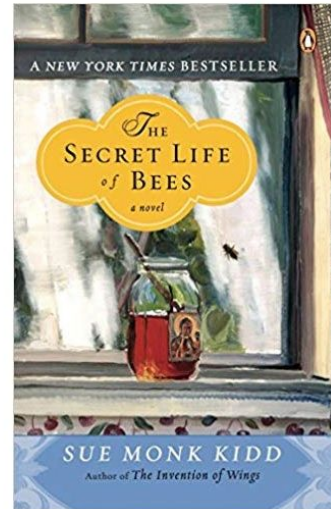
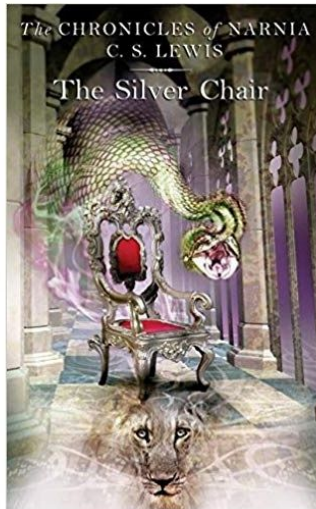
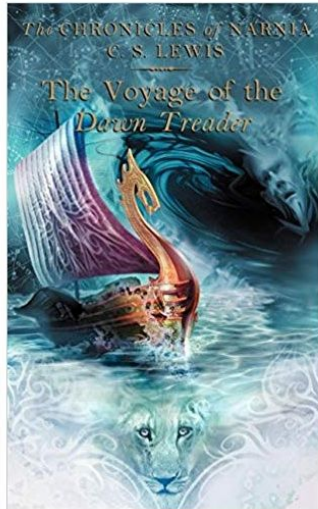
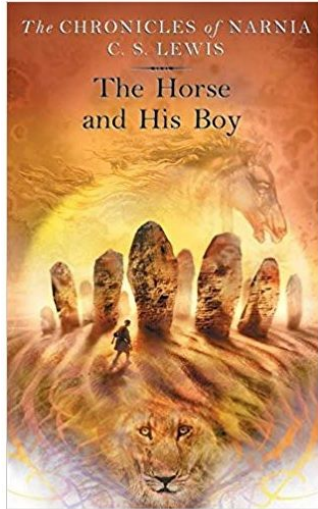
Test set: predicted top rated books





# Different Models

Test set: actual top rated books



# Collaborative Filtering-(Item-Item based)

## 3.) Collaborative Filtering-(Item-Item based)

- Cosine Similarity
- Nearest Neighbour

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Recommendations for Angels & Demons:

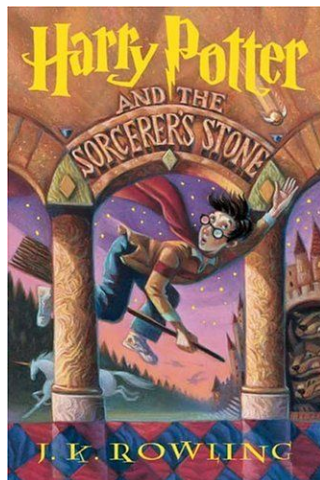
- 1: The Da Vinci Code, with distance of 0.8275555141289059:
- 2: Digital Fortress : A Thriller, with distance of 0.83781217691282:
- 3: Deception Point, with distance of 0.8422605379839627:
- 4: Prey: A Novel, with distance of 0.9216969275206289:
- 5: The Cat Who Knew a Cardinal, with distance of 0.9280814355076102:

# Different Models

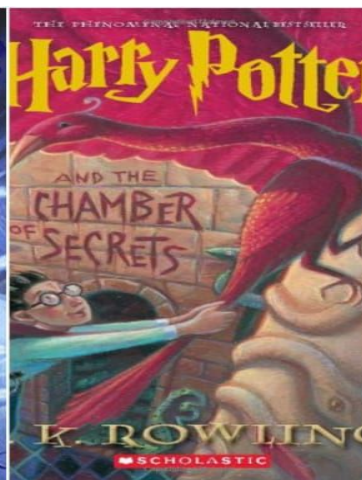
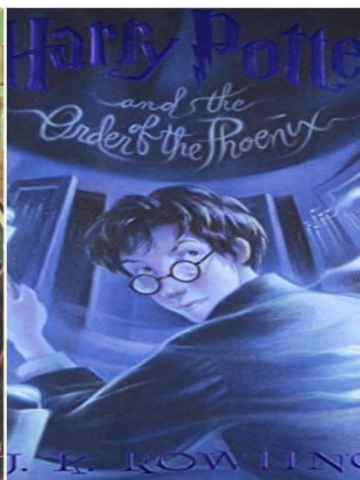
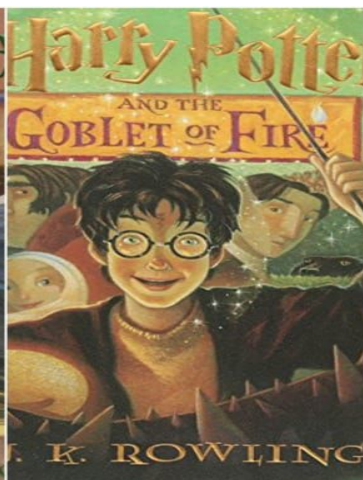
## SVD and Correlation

Recommendations for Harry Potter and the Sorcerer's Stone (Book 1)

Input



Output





# Different Models

## 4.) Collaborative Filtering-(User-Item based)

Enter User ID from above list for book recommendation 69078

Recommendation for User-ID = 69078

|   | ISBN       | Book-Title                                      | recStrength |
|---|------------|---|-------------|
| 0 | 0446310786 | To Kill a Mockingbird                           | 0.842       |
| 1 | 0345370775 | Jurassic Park                                   | 0.802       |
| 2 | 0312966970 | Four To Score (A Stephanie Plum Novel)          | 0.675       |
| 3 | 0316769487 | The Catcher in the Rye                          | 0.673       |
| 4 | 0345361792 | A Prayer for Owen Meany                         | 0.646       |
| 5 | 0440214041 | The Pelican Brief                               | 0.621       |
| 6 | 044021145X | The Firm  | 0.617       |
| 7 | 0440211727 | A Time to Kill                                  | 0.617       |
| 8 | 0060928336 | Divine Secrets of the Ya-Ya Sisterhood: A Novel | 0.606       |
| 9 | 0312924585 | Silence of the Lambs                            | 0.600       |

# Different Models

## Model Results

Global metrics:

```
{'modelName': 'Collaborative Filtering', 'recall@5': 0.2357298474945534, 'recall@10': 0.3057371096586783}
```

|    | hits@5_count | hits@10_count | interacted_count | recall@5 | recall@10 | User-ID |
|----|--------------|---------------|------------------|----------|-----------|---------|
| 10 | 252          | 343           | 1389             | 0.181    | 0.247     | 11676   |
| 31 | 189          | 245           | 1138             | 0.166    | 0.215     | 98391   |
| 45 | 17           | 30            | 380              | 0.045    | 0.079     | 189835  |
| 30 | 83           | 104           | 369              | 0.225    | 0.282     | 153662  |
| 70 | 29           | 33            | 236              | 0.123    | 0.140     | 23902   |
| 7  | 30           | 49            | 204              | 0.147    | 0.240     | 235105  |
| 47 | 22           | 32            | 203              | 0.108    | 0.158     | 76499   |
| 50 | 23           | 35            | 193              | 0.119    | 0.181     | 171118  |
| 42 | 55           | 68            | 192              | 0.286    | 0.354     | 16795   |
| 43 | 23           | 31            | 188              | 0.122    | 0.165     | 248718  |

# Conclusion

- In EDA, the Top-10 most rated books were essentially novels. Books like *The Lovely Bone* and *The Secret Life of Bees* were very well perceived.
- Majority of the readers were of the age bracket 20-35 and most of them came from North American and European countries namely USA, Canada, UK, Germany and Spain.
- If we look at the ratings distribution, most of the books have high ratings with maximum books being rated 8. Ratings below 5 are few in number.
- Author with the most books was Agatha Christie, William Shakespeare and Stephen King.
- For modelling, it was observed that for model based collaborative filtering SVD technique worked way better than NMF with lower Mean Absolute Error (MAE) .

# Challenges

- Handling of sparsity was a major challenge as well since the user interactions were not present for the majority of the books.
- Understanding the metric for evaluation was a challenge as well.
- Since the data consisted of text data, data cleaning was a major challenge in features like Location etc..
- Decision making on missing value imputations and outlier treatment was quite challenging as well.

# Future Scope

- Given more information regarding the books dataset, namely features like Genre, Description etc, we could implement a content-filtering based recommendation system and compare the results with the existing collaborative-filtering based system.
- We would like to explore various clustering approaches for clustering the users based on Age, Location etc., and then implement voting algorithms to recommend items to the user depending on the cluster into which it belongs.

**Thank You**  
**Q & A**