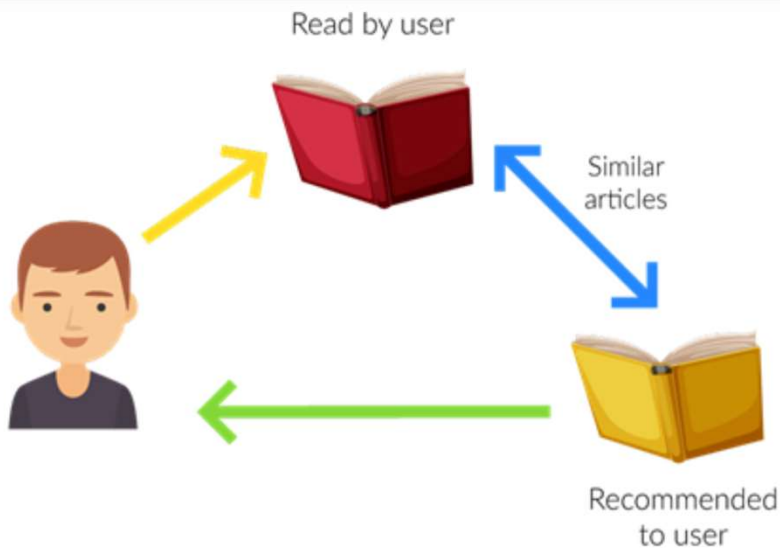




Recommendation System



Group 3

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Outline

- 01 Business Problem
- 02 Collaborative Filter Recommender System
- 03 Technology Used
- 04 System Instruction
- 05 Future Steps

Business Problem

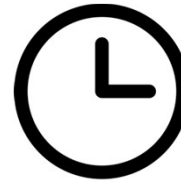
Value for Customer



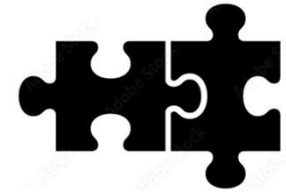
Improve Customer
Experience



Discover new things

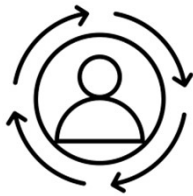


Reduced browsing
time



Complementary
items

Value for Business



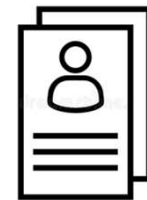
Improved Customer
Loyalty



Personalized Service



Increased sales &
conversion rate



More customer
knowledge

Item Based Recommender System

Users	Harry Potter	A Tale of Two Cities	The Silent Patient	Alice in Wonderland
	5	1	5	4
	4	4	3	3
	3	5	2	3
	1	2	4	

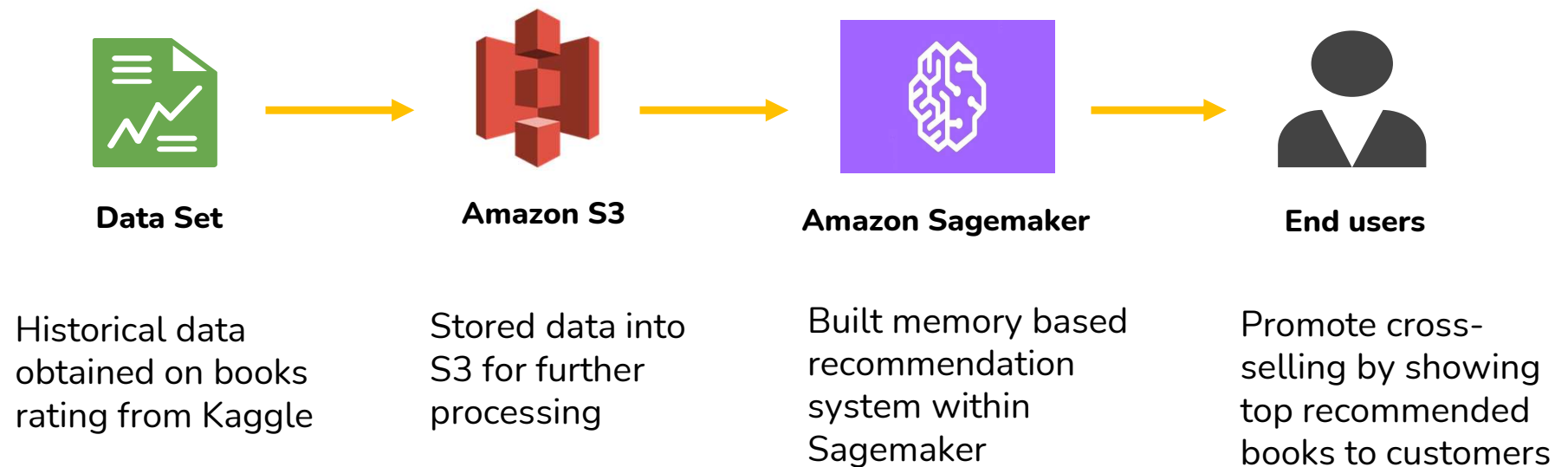
Item to Item Similarity

$$\text{Similarity}(\vec{A}, \vec{B}) = \frac{\vec{A} \cdot \vec{B}}{\|\vec{A}\| * \|\vec{B}\|}$$

Prediction Computation

$$\text{rating}(U, I_i) = \frac{\sum_j \text{rating}(U, I_j) * s_{ij}}{\sum_j s_{ij}}$$

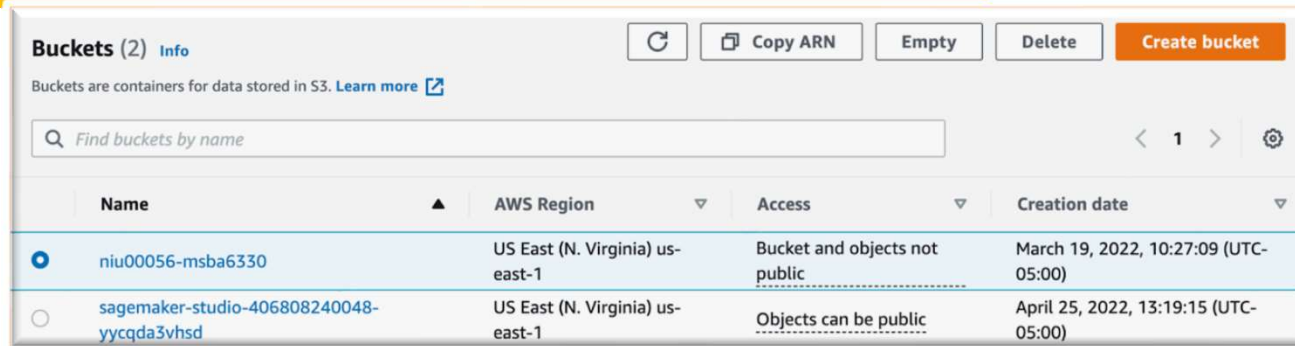
Technology Used



The implementation of recommendation system is quite straightforward and requires only a storage platform and Machine Learning tool

S3 Data Storage Process Flow

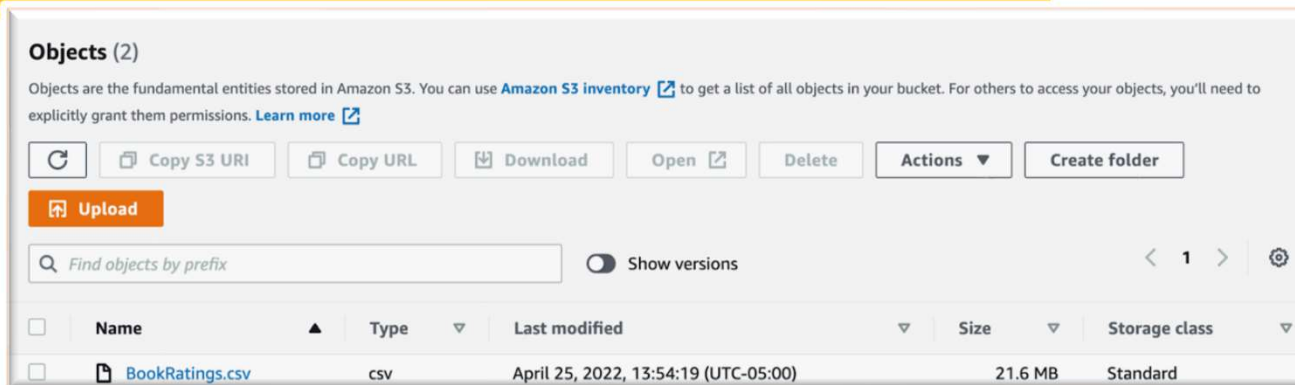
Step1. Create S3 Storage Bucket



The screenshot shows the AWS S3 Buckets console. At the top, there are buttons for 'Copy ARN', 'Empty', 'Delete', and 'Create bucket'. Below these is a search bar labeled 'Find buckets by name'. A table lists two buckets:

	Name	AWS Region	Access	Creation date
<input checked="" type="radio"/>	niu00056-msba6330	US East (N. Virginia) us-east-1	Bucket and objects not public	March 19, 2022, 10:27:09 (UTC-05:00)
<input type="radio"/>	sagemaker-studio-406808240048-yycqda3vhdsd	US East (N. Virginia) us-east-1	Objects can be public	April 25, 2022, 13:19:15 (UTC-05:00)

Step2. Upload Data into Bucket



The screenshot shows the AWS S3 Objects console for a specific bucket. At the top, there are buttons for 'Copy S3 URI', 'Copy URL', 'Download', 'Open', 'Delete', 'Actions', and 'Create folder'. Below these is an 'Upload' button and a search bar labeled 'Find objects by prefix'. A toggle switch for 'Show versions' is also present. A table lists the objects in the bucket:

	Name	Type	Last modified	Size	Storage class
<input type="checkbox"/>	BookRatings.csv	csv	April 25, 2022, 13:54:19 (UTC-05:00)	21.6 MB	Standard

Model Development Using Sagemaker

Step1. Calculate Book Similarity

```
#Recommender System
# store the original dataset in 'df', and create the copy of df, df1 = df.copy().
def book_recommender(user, num_neighbors, num_recommendation):
    number_neighbors = num_neighbors

    knn = NearestNeighbors(metric='cosine', algorithm='brute')
    knn.fit(df.values)
    distances, indices = knn.kneighbors(df.values, n_neighbors=number_neighbors)

    user_index = df.columns.tolist().index(user)

    for m,t in list(enumerate(df.index)):
        if df.iloc[m, user_index] == 0:
            sim_books = indices[m].tolist()
            book_distances = distances[m].tolist()

            if m in sim_books:
                id_book = sim_books.index(m)
                sim_books.remove(m)
                book_distances.pop(id_book)

            else:
                sim_books = sim_books[:num_neighbors-1]
                book_distances = book_distances[:num_neighbors-1]

            book_similarity = [1-x for x in book_distances]
            book_similarity_copy = book_similarity.copy()
            nominator = 0
```

Step2. Make Predictions

```
for s in range(0, len(book_similarity)):
    if df.iloc[sim_books[s], user_index] == 0:
        if len(book_similarity_copy) == (number_neighbors - 1):
            book_similarity_copy.pop(s)

        else:
            book_similarity_copy.pop(s-(len(book_similarity)-len(book_similarity_copy)))

    else:
        nominator = nominator + book_similarity[s]*df.iloc[sim_books[s],user_index]

if len(book_similarity_copy) > 0:
    if sum(book_similarity_copy) > 0:
        predicted_r = nominator/sum(book_similarity_copy)

    else:
        predicted_r = 0

else:
    predicted_r = 0

df1.iloc[m,user_index] = predicted_r
recommend books(user, num_recommendation)
```

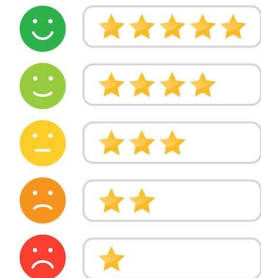
System Instruction - Result



User ID



Items



Ratings



Personalized Recommendations



The list of the Recommended Books

- 1: book_038097438X - Paradise Fever: Growing Up in the Shadow of the New Age - predicted rating:10.0
- 2: book_0452281784 - Bad Heir Day - predicted rating:10.0
- 3: book_0670889202 - Penny Dreadful - predicted rating:10.0
- 4: book_006092411X - The Living : A Novel - predicted rating:9.0
- 5: book_1573227374 - Missing Women and Others - predicted rating:9.0

Books Already Read

The list of the Books user_243 Has Read

- book_0060915544 - The Bean Trees
- book_0060977493 - The God of Small Things
- book_0140272100 - Vanished
- book_0316601950 - The Pilot's Wife : A Novel
- book_0316776963 - Me Talk Pretty One Day
- book_0316899984 - River, Cross My Heart
- book_0375400117 - Memoirs of a Geisha
- book_0385316895 - Legacy of Silence
- book_0385720106 - A Map of the World
- book_0425163407 - Unnatural Exposure
- book_044023722X - A Painted House
- book_0446364800 - The General's Daughter
- book_0446606383 - The Midnight Club
- book_0449006522 - Manhattan Hunt Club
- book_0553580388 - The Patient
- book_0786863986 - A Monk Swimming
- book_0803251718 - Crazy Horse

Limitations

1

Limited
computational
resources

2

Number of
recommendations

3

Measurement of
recommendation
system

4

Finding niche
items

Conclusion

- Business value
- How recommendation system works
- Our book recommendation system
 - Technology used
 - Prediction process
 - Recommendation output
- Limitations of recommender system
- Benefits
 - Improved customer experience
 - Personalized service
 - Increased sales