

Name: Likhith K Raj

LinkedIn: [LinkedIn Profile Link](#)

GitHub: [GitHub Profile Link](#)

Portfolio: [Portfolio Link](#)

Zeotap Data Science Intern Assignment

Task 2: Lookalike Model

Google Colab Link:

[https://colab.research.google.com/drive/1MHs2rP_Hkcs3IMrURgBBFk2PzUo_BKw?usp=drive link](https://colab.research.google.com/drive/1MHs2rP_Hkcs3IMrURgBBFk2PzUo_BKw?usp=drive_link)

Steps for solving the problem:

Step 0: Data Processing

Point: Merged the Customers.csv, Products.csv, and Transactions.csv files using common keys (CustomerID and ProductID). Cleaned the data by handling missing values, correcting data types, and removing redundant columns.

Step 1: Feature Selection

Point: Selected relevant features such as Quantity, TotalValue, and Price to capture customer purchasing behavior. Categorical data like Region was retained for one-hot encoding.

Step 2: Data Scaling

Point: Standardized the numerical features using StandardScaler to ensure all variables have equal importance when calculating similarity.

Step 3: Cosine Similarity Calculation

Point: Calculated the cosine similarity between customers based on their scaled numerical features and encoded categorical data.

Step 4: Identify Top Lookalikes

Point: For each customer, identified the **Top 3 Lookalikes** (excluding themselves) based on the highest similarity scores from the cosine similarity matrix.

Step 5: Save Lookalike Results

Point: Saved the lookalike results into a file Lookalike.csv with columns CustomerID, Lookalike_1, Lookalike_2, and Lookalike_3, including similarity scores for each lookalike.

Predications (Customized):

Here is the 2 case of predication:

- **Case 1: For customer id C0003.**

Enter CustomerID (e.g., C0001): C0003

Top 3 Lookalikes for Customer C0003:

Lookalike 1: C0190, Similarity Score: 0.9546214849405968

Lookalike 2: C0091, Similarity Score: 0.9086281452471642

Lookalike 3: C0174, Similarity Score: 0.9044670576598652

Mean Squared Error (MSE) between actual and predicted

lookalikes: 0.0006458004230247932

- **Case 2: For customer id C0199.**

Enter CustomerID (e.g., C0001): C0199

Top 3 Lookalikes for Customer C0199:

Lookalike 1: C0073, Similarity Score: 0.9850059276267502

Lookalike 2: C0132, Similarity Score: 0.974943257219227

Lookalike 3: C0019, Similarity Score: 0.9196573480364263

Mean Squared Error (MSE) between actual and predicted

lookalikes: 0.0009143822517061226

Top 3 lookalikes with their similarity scores for the first 20 customers:

CustomerID	Lookalike_1	Lookalike_2	Lookalike_3
C0001	C0076:0\ .9446	C0011:0\ .9432	C0137:0\ .9298
C0002	C0025:0\ .8750	C0157:0\ .8581	C0121:0\ .8524
C0003	C0190:0\ .9546	C0091:0\ .9086	C0174:0\ .9045
C0004	C0175:0\ .9434	C0109:0\ .9369	C0101:0\ .9319
C0005	C0186:0\ .9127	C0103:0\ .8282	C0131:0\ .8167
C0006	C0171:0\ .9533	C0107:0\ .9446	C0168:0\ .9300
C0007	C0078:0\ .9935	C0146:0\ .9875	C0092:0\ .9763
C0008	C0109:0\ .9596	C0034:0\ .9291	C0084:0\ .9177
C0009	C0061:0\ .9843	C0167:0\ .9547	C0128:0\ .9283
C0010	C0111:0\ .9570	C0121:0\ .9393	C0027:0\ .9352
C0011	C0001:0\ .9432	C0076:0\ .8773	C0184:0\ .8756
C0012	C0102:0\ .9505	C0099:0\ .9108	C0113:0\ .9108
C0013	C0148:0\ .9869	C0163:0\ .9490	C0096:0\ .8855
C0014	C0097:0\ .9675	C0119:0\ .9660	C0063:0\ .9580
C0015	C0058:0\ .9776	C0042:0\ .9404	C0092:0\ .9381
C0016	C0050:0\ .9721	C0079:0\ .9191	C0125:0\ .8953
C0017	C0041:0\ .9770	C0124:0\ .9729	C0075:0\ .9613
C0018	C0068:0\ .9630	C0065:0\ .8847	C0008:0\ .8441
C0019	C0073:0\ .9708	C0172:0\ .9479	C0199:0\ .9197

|C0020|C0015:0\ .9315|C0128:0\ .9302|C0058:0\ .9201|

Please consider my profile:

Strong Data Science Background:

Extensive experience in data analysis, machine learning, and NLP, demonstrated through internships, projects, and certifications.

Proven Impact on Business Outcomes:

Delivered a 15% revenue improvement at Leucine and developed predictive models with high accuracy for stock trading and sentiment analysis.

Research and Innovation Focus:

Published papers on machine learning applications and fine-tuned models like Gemma 2 for Kannada, blending technical skills with research.

Technical Proficiency and Continuous Learning:

Skilled in Python, SQL, Tableau, and TensorFlow, with 100+ LeetCode problems solved and active participation in Kaggle competitions.