

Netflix Content Clustering and Recommendation

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import pandas as pd
import numpy as np

from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.decomposition import TruncatedSVD
from sklearn.cluster import KMeans
from sklearn.metrics import silhouette_score
from sklearn.metrics.pairwise import cosine_similarity

df = pd.read_csv("/content/NetflixSimple.csv") # Load Netflix CSV

text_cols = ["description", "listed_in", "cast", "director"] # Select
text columns and fill missing values

for col in text_cols:
    if col not in df.columns:
        df[col] = ""
    df[col] = df[col].fillna("")

df["bag_of_content"] = (
    df["description"] + " " +
    df["listed_in"] + " " +
    df["cast"] + " " +
    df["director"]
)

# Combine text columns into one
df = df[df["bag_of_content"].str.strip() != ""]

tfidf = TfidfVectorizer(
    stop_words="english", # Convert text to TF-IDF numbers
    max_features=10000 # control dimensionality
)
tfidf_matrix = tfidf.fit_transform(df["bag_of_content"])

svd = TruncatedSVD(
    n_components=100, # reduce data size
    random_state=42
)
reduced_features = svd.fit_transform(tfidf_matrix)
```

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k = 20 # Divide content into similar clusters
kmeans = KMeans(
    n_clusters=k,
    random_state=42,
    n_init="auto"
)
cluster_labels = kmeans.fit_predict(reduced_features)
df["cluster"] = cluster_labels

score = silhouette_score(reduced_features, cluster_labels) # Check clustering score
print("Silhouette score:", score)

Silhouette score: 0.07009307629448445

# Basic analysis of content types and countries
print(df["type"].value_counts())
print(df["country"].value_counts().head(10))

type
Movie      5377
TV Show    2410
Name: count, dtype: int64
country
United States    2555
India            923
United Kingdom   397
Japan            226
South Korea      183
Canada           177
Spain            134
France           115
Egypt            101
Turkey           100
Name: count, dtype: int64

similarity_matrix = cosine_similarity(reduced_features)

title_col = "title" # Check how similar titles are

# Map title to index
title_to_index = {t.lower(): i for i, t in
    enumerate(df[title_col].astype(str))}

def recommend_similar(title, top_n=10):
    t = title.lower()
    if t not in title_to_index:
        print("Title not found.")
        return []

    idx = title_to_index[t]

```

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sim_scores = similarity_matrix[idx]

# Get similar results but remove same item
similar_indices = np.argsort(sim_scores)[::-1] # descending
similar_indices = [i for i in similar_indices if i != idx][:top_n]

return df.iloc[similar_indices][[title_col, "cluster"]]

# Example usage:
print(recommend_similar("3 Idiots", top_n=10))

```

| | title | cluster |
|------|----------------------|---------|
| 1758 | Dil Dhadakne Do | 1 |
| 4872 | PK | 1 |
| 1757 | Dil Chahta Hai | 1 |
| 4485 | No Entry | 1 |
| 7371 | Upstarts | 1 |
| 5097 | Rang De Basanti | 1 |
| 3362 | Khushi | 1 |
| 4961 | Prem Ratan Dhan Payo | 1 |
| 3128 | Jatt James Bond | 1 |
| 3287 | Kai Po Che! | 13 |

```

print(recommend_similar("PK", top_n=10))
print(recommend_similar("Dangal", top_n=10))

```

| | title | cluster |
|------|-------------------------------------|---------|
| 271 | Aarakshan | 1 |
| 4276 | Mumbai Cha Raja | 1 |
| 7371 | Upstarts | 1 |
| 4721 | Paan Singh Tomar | 1 |
| 4634 | Once Upon a Time in Mumbai Dobaara! | 1 |
| 663 | Baazaar | 1 |
| 2006 | English Babu Desi Mem | 1 |
| 100 | 3 Idiots | 1 |
| 5579 | Shorgul | 13 |
| 1758 | Dil Dhadakne Do | 1 |

| | title | cluster |
|------|-----------------------------|---------|
| 7157 | Torbaaz | 1 |
| 1739 | Dhobi Ghat (Mumbai Diaries) | 1 |
| 4721 | Paan Singh Tomar | 1 |
| 3812 | Lucky: No Time for Love | 1 |
| 4276 | Mumbai Cha Raja | 1 |
| 2859 | Hum Aapke Hain Koun | 1 |
| 1896 | Duplicate | 1 |
| 5722 | Soorma | 13 |
| 3477 | Kya Kehna | 1 |
| 1257 | Chaahat | 1 |