

# 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

```