





TITLE OF PROJECT REPORT " MOVIE WATCH "

A PROJECT REPORT Submitted by:

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in partial fulfillment for the award of the degree of

BACHELOR OF TECHNOLOGY DEGREE

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in CSE(AI)

Introduction:

In the digital era, **understanding user behavior** is critical for platforms like Netflix, Amazon Prime, or YouTube, where content is vast and user preferences vary widely. One of the most powerful tools in this space is **clustering**, which helps group users based on similar behavior patterns.

This project focuses on clustering users based on their **movie-watching patterns**, using three main factors:

- Time of watching (hour of the day)
- Genre preference
- Average rating given

This helps identify similar user groups for personalization, targeting, or recommendations.

Methodology:

1. Data Preprocessing:

- Dataset has 3 columns: watch_time_hour, genre_preference, and avg_rating_given.
- All 100 rows are clean with no missing values.

2. Encoding:

- o genre_preference is a text column (like "comedy", "thriller").
- We convert this to numbers using Label Encoding (e.g., "comedy" → 0, "action" → 1, etc.).

3. Normalization:

 Features are scaled using **StandardScaler** so all variables contribute equally to clustering.

4. Clustering with KMeans:

- We apply **KMeans with 3 clusters** (you can change this).
- Each user is assigned to one of these clusters based on similarity in the 3 features.

5. Visualization:

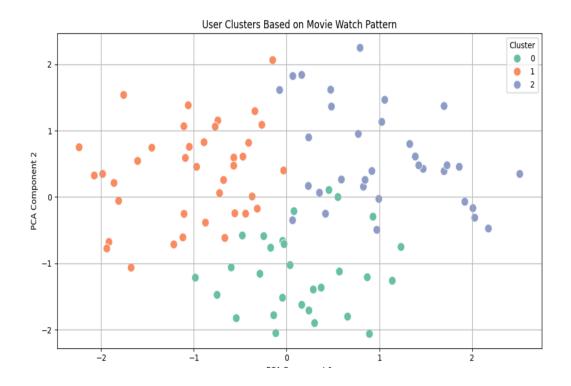
- We reduce the 3D feature space to 2D using **PCA** for visualization.
- o A scatter plot shows users in colored clusters.

Code Summary:

```
import pandas as pd
from sklearn.preprocessing import StandardScaler, LabelEncoder
from sklearn.cluster import KMeans
from sklearn.decomposition import PCA
import matplotlib.pyplot as plt
import seaborn as sns
# Load the dataset
df = pd.read csv("movie watch.csv")
# Encode the categorical genre column
le = LabelEncoder()
df['genre encoded'] = le.fit transform(df['genre preference'])
# Select and scale the features
features = df[['watch_time_hour', 'genre_encoded', 'avg_rating_given']]
scaler = StandardScaler()
scaled features = scaler.fit transform(features)
```

```
# Apply KMeans clustering
kmeans = KMeans(n clusters=3, random state=42)
df['cluster'] = kmeans.fit_predict(scaled_features)
# Reduce dimensions with PCA for visualization
pca = PCA(n components=2)
pca_components = pca.fit_transform(scaled_features)
df['pca1'] = pca components[:, 0]
df['pca2'] = pca_components[:, 1]
# Plot the clusters
plt.figure(figsize=(10, 6))
sns.scatterplot(data=df, x='pca1', y='pca2', hue='cluster', palette='Set2',
s=100)
plt.title('User Clusters Based on Movie Watch Pattern')
plt.xlabel('PCA Component 1')
plt.ylabel('PCA Component 2')
plt.legend(title='Cluster')
plt.grid(True)
plt.tight_layout()
plt.show()
```

OUTPUT/RESULTS:



References / Credits:

This clustering project was made possible using publicly available tools and libraries from the Python ecosystem, along with structured user movie interaction data. Below are the key resources and acknowledgments:

Libraries and Tools:

- Pandas: For data handling and manipulation
- https://pandas.pydata.org/
- Scikit-learn: For machine learning algorithms like KMeans and PCA https://scikit-learn.org/
- Matplotlib & Seaborn: For data visualization and plotting

https://matplotlib.org/

https://seaborn.pydata.org/

FPDF for Python (optional if using

PDF): For report generation

https://pyfpdf.github.io/

Conceptual References:

- KMeans Clustering Algorithm –
 Used to group users based on feature similarity.
- Principal Component Analysis (PCA) – Used to visualize multidimensional data in 2D.
- Data Preprocessing Techniques –
 Label Encoding and Standard Scaling.

Author / Contributor:

- Analysis and implementation by: Prince
- Date: 22nd April 2025

Dataset:

 Simulated or collected dataset of user movie-watching patterns
 (For real-world use, ensure compliance with data privacy laws like GDPR/CCPA)