

```
In [46]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import plotly.express as px
import seaborn as sns
```

```
In [47]: df = pd.read_csv('C:/Users/AjithKumar.Pola/Downloads/clustering_dataset.csv')
```

```
In [48]: df.head()
```

```
Out[48]:
```

	CUSTOMER_ID	BALANCE	BALANCE_FREQ	PURCHASES	ONE_OFF_PURCHASE	INSTALLMENT_PURCHASES
0	C10001	40.900749	0.818182	95.40	0.00	0.00
1	C10002	3202.467416	0.909091	0.00	0.00	0.00
2	C10003	2495.148862	1.000000	773.17	773.17	773.17
3	C10004	1666.670542	0.636364	1499.00	1499.00	1499.00
4	C10005	817.714335	1.000000	16.00	16.00	16.00

```
In [49]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 8950 entries, 0 to 8949
Data columns (total 18 columns):
#   Column                                     Non-Null Count  Dtype
---  -
0   CUSTOMER_ID                             8950 non-null   object
1   BALANCE                                  8950 non-null   float64
2   BALANCE_FREQ                             8950 non-null   float64
3   PURCHASES                               8950 non-null   float64
4   ONE_OFF_PURCHASE                         8950 non-null   float64
5   INSTALLMENT_PURCHASES                   8950 non-null   float64
6   CASH_IN_ADVANCE                         8950 non-null   float64
7   PURCHASE_FREQ                           8950 non-null   float64
8   ONE_OFF_PURCHASES_FREQUENCY             8950 non-null   float64
9   PURCHASES_INSTALLMENTS_FREQUENCY        8950 non-null   float64
10  CASH_ADVANCE_FREQUENCY                  8950 non-null   float64
11  CASH_ADVANCE_TRX                        8950 non-null   int64
12  PURCHASES_TRX                           8950 non-null   int64
13  CREDIT_LIMIT                            8949 non-null   float64
14  PAYMENTS                                8950 non-null   float64
15  MINIMUM_PAYMENTS                        8637 non-null   float64
16  PRC_FULL_PAYMENT                        8950 non-null   float64
17  TENURE                                  8950 non-null   int64
dtypes: float64(14), int64(3), object(1)
memory usage: 1.2+ MB
```

```
In [50]: df.describe()
```

```
Out[50]:
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	BALANCE	BALANCE_FREQ	PURCHASES	ONE_OFF_PURCHASE	INSTALLMENT_PURCHASES
count	8950.000000	8950.000000	8950.000000	8950.000000	8950.000000
mean	1564.474828	0.877271	1003.204834	592.437371	411.061736
std	2081.531879	0.236904	2136.634782	1659.887917	904.338126
min	0.000000	0.000000	0.000000	0.000000	0.000000
25%	128.281915	0.888889	39.635000	0.000000	0.000000
50%	873.385231	1.000000	361.280000	38.000000	89.000000
75%	2054.140036	1.000000	1110.130000	577.405000	468.630000
max	19043.138560	1.000000	49039.570000	40761.250000	22500.000000

```
In [51]: df.drop(['CUSTOMER_ID'], axis = 1, inplace= True)
# checking the modified dataset:
df.head()
```

```
Out[51]:
```

	BALANCE	BALANCE_FREQ	PURCHASES	ONE_OFF_PURCHASE	INSTALLMENT_PURCHASES
0	40.900749	0.818182	95.40	0.00	95.40
1	3202.467416	0.909091	0.00	0.00	0.00
2	2495.148862	1.000000	773.17	773.17	0.00
3	1666.670542	0.636364	1499.00	1499.00	0.00
4	817.714335	1.000000	16.00	16.00	0.00

```
In [52]: df.isnull().sum()
```

```
Out[52]:
```

BALANCE	0
BALANCE_FREQ	0
PURCHASES	0
ONE_OFF_PURCHASE	0
INSTALLMENT_PURCHASES	0
CASH_IN_ADVANCE	0
PURCHASE_FREQ	0
ONE_OFF_PURCHASES_FREQUENCY	0
PURCHASES_INSTALLMENTS_FREQUENCY	0
CASH_ADVANCE_FREQUENCY	0
CASH_ADVANCE_TRX	0
PURCHASES_TRX	0
CREDIT_LIMIT	1
PAYMENTS	0
MINIMUM_PAYMENTS	313
PRC_FULL_PAYMENT	0
TENURE	0

dtype: int64

```
In [53]: df ['MINIMUM_PAYMENTS'] = df ['MINIMUM_PAYMENTS'].fillna (0)
df ['CREDIT_LIMIT'] = df ['CREDIT_LIMIT'].fillna (0)
```

```
In [54]: df.isnull().sum()
```

```
Out[54]: BALANCE                                0
BALANCE_FREQ                                0
PURCHASES                                0
ONE_OFF_PURCHASE                                0
INSTALLMENT_PURCHASES                        0
CASH_IN_ADVANCE                            0
PURCHASE_FREQ                                0
ONE_OFF_PURCHASES_FREQUENCY                0
PURCHASES_INSTALLMENTS_FREQUENCY          0
CASH_ADVANCE_FREQUENCY                    0
CASH_ADVANCE_TRX                          0
PURCHASES_TRX                             0
CREDIT_LIMIT                              0
PAYMENTS                                  0
MINIMUM_PAYMENTS                          0
PRC_FULL_PAYMENT                          0
TENURE                                    0
dtype: int64
```

```
In [26]: from sklearn import preprocessing
from sklearn.preprocessing import StandardScaler, normalize
from sklearn.manifold import TSNE
from sklearn.model_selection import train_test_split
from sklearn.cluster import KMeans # Class to develop kmeans model
from sklearn import metrics
from sklearn.metrics import silhouette_score # base for clustering
from yellowbrick.cluster import SilhouetteVisualizer
from sklearn.mixture import GaussianMixture
```

```
In [10]: !pip install yellowbrick
```

```
Collecting yellowbrick
  Downloading yellowbrick-1.5-py3-none-any.whl (282 kB)
    ----- 282.6/282.6 kB 1.9 MB/s eta 0:00:00
Requirement already satisfied: scikit-learn>=1.0.0 in c:\users\ajithkumar.pola\to
shiba\jupyter\lib\site-packages (from yellowbrick) (1.0.2)
Requirement already satisfied: numpy>=1.16.0 in c:\users\ajithkumar.pola\to
shiba\jupyter\lib\site-packages (from yellowbrick) (1.21.5)
Requirement already satisfied: cycler>=0.10.0 in c:\users\ajithkumar.pola\to
shiba\jupyter\lib\site-packages (from yellowbrick) (0.11.0)
Requirement already satisfied: scipy>=1.0.0 in c:\users\ajithkumar.pola\to
shiba\jupyter\lib\site-packages (from yellowbrick) (1.9.1)
Requirement already satisfied: matplotlib!=3.0.0,>=2.0.2 in c:\users\ajithkuma
r.pola\to
shiba\jupyter\lib\site-packages (from yellowbrick) (3.5.2)
Requirement already satisfied: pillow>=6.2.0 in c:\users\ajithkumar.pola\to
shiba\jupyter\lib\site-packages (from matplotlib!=3.0.0,>=2.0.2->yellowbrick) (9.2.
0)
Requirement already satisfied: pyparsing>=2.2.1 in c:\users\ajithkumar.pola\to
shiba\jupyter\lib\site-packages (from matplotlib!=3.0.0,>=2.0.2->yellowbrick) (3.0.9)
Requirement already satisfied: fonttools>=4.22.0 in c:\users\ajithkumar.pola\to
shiba\jupyter\lib\site-packages (from matplotlib!=3.0.0,>=2.0.2->yellowbrick) (4.25.0)
Requirement already satisfied: python-dateutil>=2.7 in c:\users\ajithkumar.pola\to
shiba\jupyter\lib\site-packages (from matplotlib!=3.0.0,>=2.0.2->yellowbric
k) (2.8.2)
Requirement already satisfied: packaging>=20.0 in c:\users\ajithkumar.pola\to
shiba\jupyter\lib\site-packages (from matplotlib!=3.0.0,>=2.0.2->yellowbrick) (2
1.3)
Requirement already satisfied: kiwisolver>=1.0.1 in c:\users\ajithkumar.pola\to
shiba\jupyter\lib\site-packages (from matplotlib!=3.0.0,>=2.0.2->yellowbrick) (1.4.2)
Requirement already satisfied: threadpoolctl>=2.0.0 in c:\users\ajithkumar.pola\to
shiba\jupyter\lib\site-packages (from scikit-learn>=1.0.0->yellowbrick) (2.
2.0)
Requirement already satisfied: joblib>=0.11 in c:\users\ajithkumar.pola\to
shiba\jupyter\lib\site-packages (from scikit-learn>=1.0.0->yellowbrick) (1.1.0)
Requirement already satisfied: six>=1.5 in c:\users\ajithkumar.pola\to
shiba\jupyter\lib\site-packages (from python-dateutil>=2.7->matplotlib!=3.0.0,>=2.0.2->y
ellowbrick) (1.16.0)
Installing collected packages: yellowbrick
Successfully installed yellowbrick-1.5
```

```
In [27]: from sklearn import preprocessing
from sklearn.preprocessing import StandardScaler, normalize
from sklearn.manifold import TSNE
from sklearn.model_selection import train_test_split
from sklearn.cluster import KMeans # Class to develop kmeans model
from sklearn import metrics
from sklearn.metrics import silhouette_score # base for clustering
from yellowbrick.cluster import SilhouetteVisualizer
from sklearn.mixture import GaussianMixture
```

```
In [28]: import warnings
warnings.filterwarnings('ignore')
import os
```

```
In [30]: from sklearn.decomposition import PCA
```

```
In [31]: #Scaling the data
scaler = StandardScaler()
scaled_df = scaler.fit_transform(df)

# Normalizing the Data
normalized_df = normalize(scaled_df)

# Converting the numpy array into a pandas DataFrame
normalized_df = pd.DataFrame(normalized_df)

# Reducing the dimensions of the data
pca = PCA(n_components = 2)
X_principal = pca.fit_transform(normalized_df)
X_principal = pd.DataFrame(X_principal)
X_principal.columns = ['P1', 'P2']

X_principal.head(2)
```

```
Out[31]:
```

	P1	P2
0	-0.490758	-0.679041
1	-0.518463	0.545358

```
In [32]: gmm = GaussianMixture(n_components = 3)
gmm.fit(X_principal)
```

```
Out[32]: GaussianMixture(n_components=3)
```

```
In [45]: plt.scatter(X_principal['P1'], X_principal['P2'], c = GaussianMixture(n_components=3).fit(X_principal).predict(X_principal), s=50, edgecolor='k')
plt.show()
```

