

```
import pandas as pd
import numpy as np
from matplotlib import pyplot as plt
import seaborn as sns
```

```
df=pd.read_csv('/content/Employee.csv')
df.head()
```

	Education	JoiningYear	City	PaymentTier	Age	Gender	EverBenched	ExperienceInCurrentDomain	LeaveOrNot
0	Bachelors	2017	Bangalore	3	34	Male	No	0	0
1	Bachelors	2013	Pune	1	28	Female	No	3	1
2	Bachelors	2014	New Delhi	3	38	Female	No	2	0
3	Masters	2016	Bangalore	3	27	Male	No	5	1
4	Masters	2017	Pune	3	24	Male	Yes	2	1

Next steps:

[Generate code with df](#)[View recommended plots](#)[New interactive sheet](#)

df.tail()

	Education	JoiningYear	City	PaymentTier	Age	Gender	EverBenched	ExperienceInCurrentDomain	LeaveOrNot
4648	Bachelors	2013	Bangalore	3	26	Female	No	4	0
4649	Masters	2013	Pune	2	37	Male	No	2	1
4650	Masters	2018	New Delhi	3	27	Male	No	5	1
4651	Bachelors	2012	Bangalore	3	30	Male	Yes	2	0
4652	Bachelors	2015	Bangalore	3	33	Male	Yes	4	0

df.shape

(4653, 9)

df.describe()

	JoiningYear	PaymentTier	Age	ExperienceInCurrentDomain	LeaveOrNot
count	4653.000000	4653.000000	4653.000000	4653.000000	4653.000000
mean	2015.062970	2.698259	29.393295	2.905652	0.343864
std	1.863377	0.561435	4.826087	1.558240	0.475047
min	2012.000000	1.000000	22.000000	0.000000	0.000000
25%	2013.000000	3.000000	26.000000	2.000000	0.000000
50%	2015.000000	3.000000	28.000000	3.000000	0.000000
75%	2017.000000	3.000000	32.000000	4.000000	1.000000
max	2018.000000	3.000000	41.000000	7.000000	1.000000


```
df.replace({'Bachelors':1, 'Masters':2, 'PHD':3}, inplace=True)
df.head()
```



	Education	JoiningYear	City	PaymentTier	Age	Gender	EverBenched	ExperienceInCurrentDomain	LeaveOrNot
0	1	2017	Bangalore	3	34	Male	No	0	0
1	1	2013	Pune	1	28	Female	No	3	1
2	1	2014	New Delhi	3	38	Female	No	2	0
3	2	2016	Bangalore	3	27	Male	No	5	1
4	2	2017	Pune	3	24	Male	Yes	2	1

Next steps:

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```
df.replace({'Bangalore':1, 'Pune':2, 'New Delhi':3}, inplace=True)
df.head()
```


 <ipython-input-7-2c79d39905ab>:1: FutureWarning: Downcasting behavior in `replace` is deprecated and will be removed in a future ver
df.replace({'Bangalore':1, 'Pune':2, 'New Delhi':3}, inplace=True)



	Education	JoiningYear	City	PaymentTier	Age	Gender	EverBenched	ExperienceInCurrentDomain	LeaveOrNot	
0	1	2017	1	3	34	Male	No	0	0	
1	1	2013	2	1	28	Female	No	3	1	
2	1	2014	3	3	38	Female	No	2	0	
3	2	2016	1	3	27	Male	No	5	1	
4	2	2017	2	3	24	Male	Yes	2	1	

Next steps:

[Generate code with df](#)[View recommended plots](#)[New interactive sheet](#)

```
df.replace({'Male':1, 'Female':2}, inplace=True)
df.head()
```


 <ipython-input-8-cedbb1bc21a8>:1: FutureWarning: Downcasting behavior in `replace` is deprecated and will be removed in a future ver
df.replace({'Male':1, 'Female':2}, inplace=True)



	Education	JoiningYear	City	PaymentTier	Age	Gender	EverBenched	ExperienceInCurrentDomain	LeaveOrNot	
0	1	2017	1	3	34	1	No	0	0	
1	1	2013	2	1	28	2	No	3	1	
2	1	2014	3	3	38	2	No	2	0	
3	2	2016	1	3	27	1	No	5	1	
4	2	2017	2	3	24	1	Yes	2	1	

Next steps:

[Generate code with df](#)[View recommended plots](#)[New interactive sheet](#)

```
df.replace({'Yes':1, 'No':2 }, inplace=True)
df.head()
```



 <ipython-input-9-689c9e7f3934>:1: FutureWarning: Downcasting behavior in `replace` is deprecated and will be removed in a future ver
df.replace({'Yes':1, 'No':2 }, inplace=True)

	Education	JoiningYear	City	PaymentTier	Age	Gender	EverBenched	ExperienceInCurrentDomain	LeaveOrNot	
0	1	2017	1	3	34	1	2	0	0	
1	1	2013	2	1	28	2	2	3	1	
2	1	2014	3	3	38	2	2	2	0	
3	2	2016	1	3	27	1	2	5	1	
4	2	2017	2	3	24	1	1	2	1	

Next steps:

[Generate code with df](#)[View recommended plots](#)[New interactive sheet](#)

```
df.to_csv("/content/Employee-new.csv",index=False)
new_df=pd.read_csv("/content/Employee-new.csv")
new_df.head()
```

	Education	JoiningYear	City	PaymentTier	Age	Gender	EverBenched	ExperienceInCurrentDomain	LeaveOrNot	
0	1	2017	1	3	34	1	2	0	0	
1	1	2013	2	1	28	2	2	3	1	
2	1	2014	3	3	38	2	2	2	0	
3	2	2016	1	3	27	1	2	5	1	
4	2	2017	2	3	24	1	1	2	1	

Next steps:

[Generate code with new_df](#)[View recommended plots](#)[New interactive sheet](#)

Step 1: Choosing the Right Number of Clusters Using the Elbow Method

```
from sklearn.cluster import KMeans
import matplotlib.pyplot as plt
```

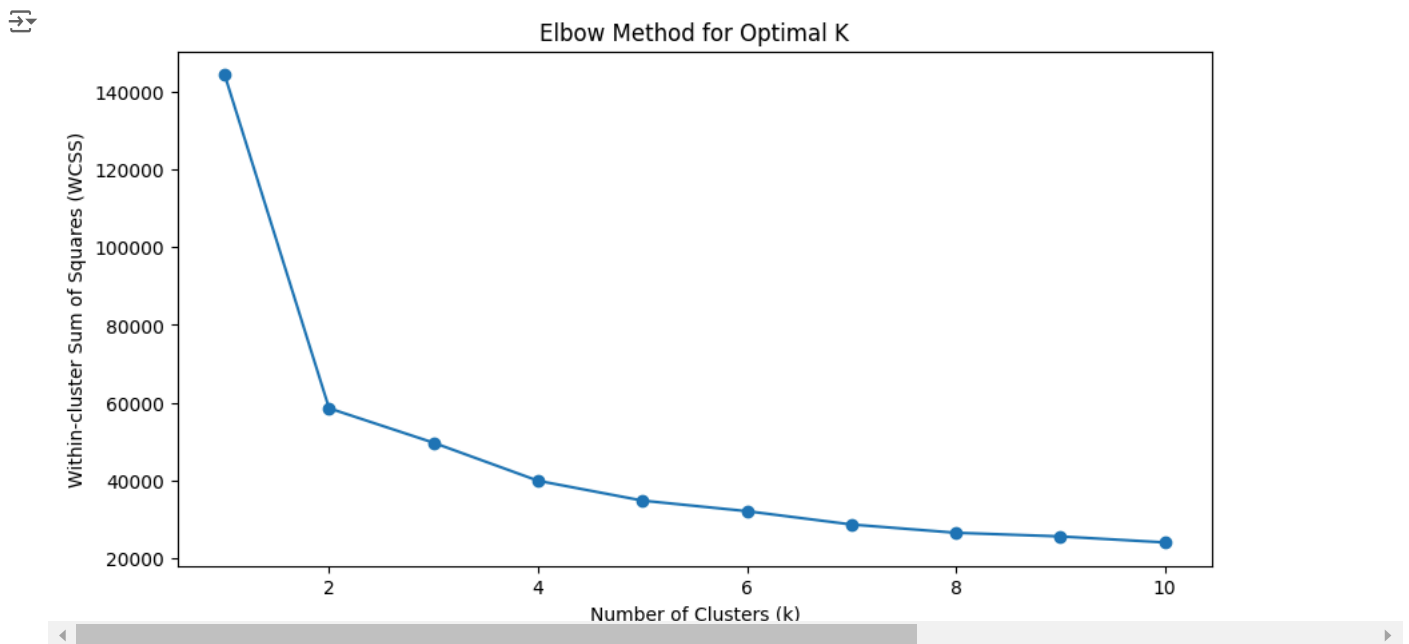
```
# Sample DataFrame (replace `df` with your actual DataFrame)
```

```
df=pd.read_csv('/content/Employee-new.csv')

# Elbow Method function
def elbow_method(df, max_clusters=10):
    wcss = [] # List to store within-cluster sum of squares
    for i in range(1, max_clusters + 1):
        kmeans = KMeans(n_clusters=i, init='k-means++', random_state=42)
        kmeans.fit(df)
        wcss.append(kmeans.inertia_) # Append WCSS for each k

    # Plotting the Elbow Curve
    plt.figure(figsize=(10, 5))
    plt.plot(range(1, max_clusters + 1), wcss, marker='o')
    plt.title('Elbow Method for Optimal K')
    plt.xlabel('Number of Clusters (k)')
    plt.ylabel('Within-cluster Sum of Squares (WCSS)')
    plt.show()

# Call the elbow method function to find the optimal number of clusters
elbow_method(df)
```



Step 2: Apply K-means Clustering

```
from sklearn.cluster import KMeans

# Define the optimal number of clusters based on the Elbow Method plot
optimal_clusters = 4 # Replace this with the number you choose from the elbow plot
kmeans = KMeans(n_clusters=optimal_clusters, init='k-means++', random_state=42)
kmeans_labels = kmeans.fit_predict(df)

# kmeans_labels contains the cluster labels for each data point
```

Step 3: Apply K-medoids Clustering

```
!pip install scikit-learn-extra
```

```
Collecting scikit-learn-extra
  Downloading scikit_learn_extra-0.3.0-cp310-cp310-manylinux_2_17_x86_64.manylinux2014_x86_64.whl.metadata (3.6 kB)
Requirement already satisfied: numpy>=1.13.3 in /usr/local/lib/python3.10/dist-packages (from scikit-learn-extra) (1.26.4)
Requirement already satisfied: scipy>=0.19.1 in /usr/local/lib/python3.10/dist-packages (from scikit-learn-extra) (1.13.1)
Requirement already satisfied: scikit-learn>=0.23.0 in /usr/local/lib/python3.10/dist-packages (from scikit-learn-extra) (1.5.2)
Requirement already satisfied: joblib>=1.2.0 in /usr/local/lib/python3.10/dist-packages (from scikit-learn>=0.23.0->scikit-learn-ext
Requirement already satisfied: threadpoolctl>=3.1.0 in /usr/local/lib/python3.10/dist-packages (from scikit-learn>=0.23.0->scikit-l
  Downloading scikit_learn_extra-0.3.0-cp310-cp310-manylinux_2_17_x86_64.manylinux2014_x86_64.whl (2.0 MB)
    2.0/2.0 MB 21.0 MB/s eta 0:00:00
Installing collected packages: scikit-learn-extra
Successfully installed scikit-learn-extra-0.3.0
```

```
from sklearn_extra.cluster import KMedoids
```

```
# Apply K-medoids Clustering with the chosen number of clusters
kmedoids = KMedoids(n_clusters=optimal_clusters, random_state=42)
kmedoids_labels = kmedoids.fit_predict(df)
```

```
# kmedoids_labels contains the cluster labels for each data point
```

```
↳ /usr/local/lib/python3.10/dist-packages/sklearn_extra/cluster/_k_medoids.py:329: UserWarning: Cluster 1 is empty! self.labels_[self
warnings.warn(
/usr/local/lib/python3.10/dist-packages/sklearn_extra/cluster/_k_medoids.py:329: UserWarning: Cluster 2 is empty! self.labels_[self
warnings.warn(
/usr/local/lib/python3.10/dist-packages/sklearn_extra/cluster/_k_medoids.py:329: UserWarning: Cluster 3 is empty! self.labels_[self
warnings.warn(
```

Step 4: Visualizing the Clusters for Both K-means and K-medoids

```
import matplotlib.pyplot as plt
```

```
# Plotting K-means Clusters
plt.figure(figsize=(12, 5))
plt.subplot(1, 2, 1)
plt.scatter(df.iloc[:, 0], df.iloc[:, 1], c=kmeans_labels, cmap='viridis')
plt.scatter(kmeans.cluster_centers_[0], kmeans.cluster_centers_[1], s=300, c='red', marker='X', label='Centroids')
plt.title('K-means Clustering')
plt.xlabel('Feature 1')
plt.ylabel('Feature 2')
plt.legend()
```

```
# Plotting K-medoids Clusters
plt.subplot(1, 2, 2)
plt.scatter(df.iloc[:, 0], df.iloc[:, 1], c=kmedoids_labels, cmap='viridis')
plt.scatter(df.iloc[kmedoids.medoid_indices_, 0], df.iloc[kmedoids.medoid_indices_, 1], s=300, c='blue', marker='X', label='Medoids')
plt.title('K-medoids Clustering')
plt.xlabel('Feature 1')
plt.ylabel('Feature 2')
plt.legend()
```

```
plt.show()
```

