HW7

We load and split the USArrests dataset.

```
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.decomposition import PCA

data = pd.read_csv("usarrests.csv", index_col=0)

# Features for use in clustering.
X = data[["Murder", "Assault", "UrbanPop", "Rape"]]

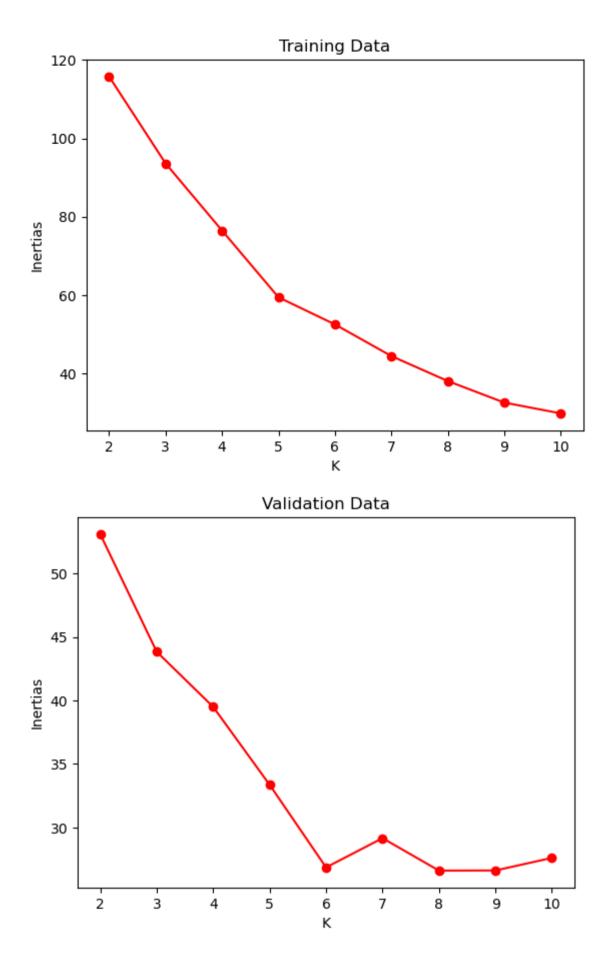
# For whitening datasets for KMeans clustering.
whitener = PCA(whiten=True)

# Split into training and validation sets.
X_train, X_val = train_test_split(X, random_state=0)

# Whiten.
X_train = whitener.fit_transform(X_train)
X_val = whitener.transform(X_val)
```

Fit KMeans clusters to the training set for K in [2, 10] using a random_state of 0 . Plot the training inertias as a function of K. In a separate figure, plot the validation inertias as a function of K.

```
In [123... import warnings
         warnings.filterwarnings('ignore', message='Kmeans')
         from sklearn.cluster import KMeans
         import numpy as np
         import matplotlib.pyplot as plt
         klist = []
         ilist = []
         vpred = []
         cen=[]
         for k in range(2,11):
             kmeans = KMeans(n_clusters=k, random_state=0)
             km=kmeans.fit(X train)
             v=km.predict(X_val)
             vpred.append(v)
             klist.append(k)
             ilist.append(km.inertia_)
             cen.append(km.cluster_centers_)
         plt.plot(klist,ilist,'o-',color="red")
         plt.title("Training Data")
         plt.xlabel("K")
         plt.ylabel("Inertias")
         plt.show()
         def inertia(X_val, vlist):
             lst=[]
             for i in range(X_val.shape[0]):
                 lst.append(np.sum((X_val[i] - vlist[i])**2))
             return sum(lst)
         res=[]
         for i in range(len(vpred)):
             vlist=[]
             for j in vpred[i]:
                 vlist.append(cen[i][j])
             res.append(inertia(X_val,vlist))
         plt.plot(klist,res,'o-',color="red")
         plt.title("Validation Data")
         plt.xlabel("K")
         plt.ylabel("Inertias")
         plt.show()
```



Based on the training plot, what value of K would the elbow method recommend selecting? Explain.

Answer: The optimal K is 5. This is because after K=5, the inertia, which is the sum of squared distances of samples to their closest cluster center, starts decreasing in a linear fashion, and according to the elbow method, K=5 is the optimal point.

Justify the elbow method based on the validation plot.

Answer: From the validation plot, we choose the optimal K as 5. Because when K=6, the intertia is way too low that might indicates an overfitting, so we need to pick the K before overfitting which is 5. K=5 is the same as the optimal K we chose in the training plot, thus, justified the elbow method.