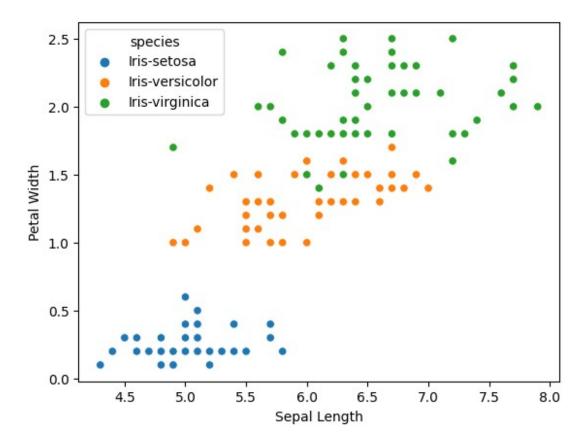
```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

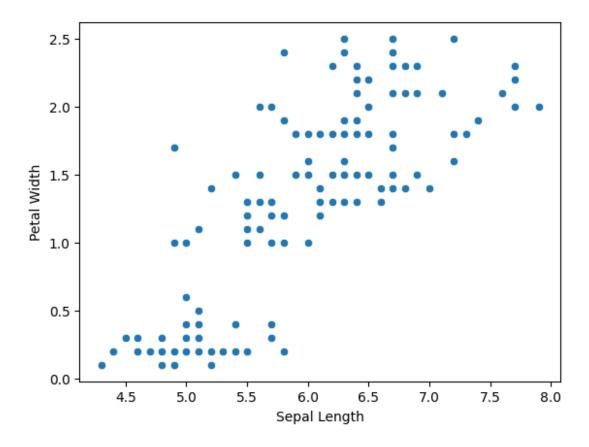
Preparing Data

```
data = pd.read_csv("datas/Iris.csv",index_col=0)
data.columns = ["Sepal Length", "Sepal Width", "Petal Length", "Petal
Width", "species"]
df = data
df.head(3)
    Sepal Length Sepal Width Petal Length Petal Width
                                                              species
Id
1
             5.1
                          3.5
                                        1.4
                                                     0.2 Iris-setosa
2
             4.9
                          3.0
                                        1.4
                                                     0.2 Iris-setosa
3
             4.7
                          3.2
                                        1.3
                                                     0.2 Iris-setosa
sns.scatterplot(data=df,x='Sepal Length',y='Petal
Width',hue='species')
<Axes: xlabel='Sepal Length', ylabel='Petal Width'>
```



X = df[['Sepal Length','Petal Width']]
sns.scatterplot(x=X['Sepal Length'],y=X['Petal Width'])

<Axes: xlabel='Sepal Length', ylabel='Petal Width'>



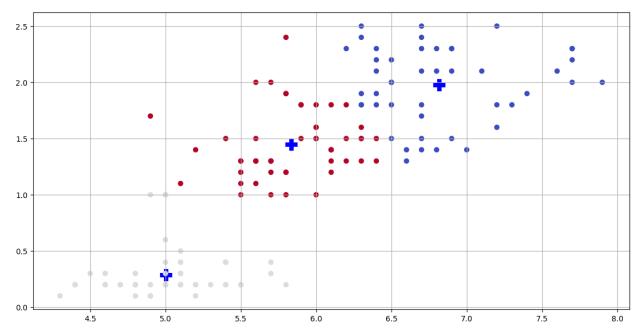
Model

```
class KMeans:
   def __init__(self,n_cluster=1):
        self.k = n cluster
   def find closest centroids(self,X,centroids):
        n = len(X)
        k = self.k
        index = np.zeros(n)
        for i in range(0,n):
            distances = []
            for j in range(0,k):
                dist = np.linalg.norm(X[i] - centroids[j])
                distances.append(dist)
            index[i] = np.argmin(distances)
        return index
   def _compute_centroid(self,X,index):
        centroids = np.zeros((self.k,X.shape[1]))
        for k in range(0, self.k):
            cluster = X[index == k]
            centroids[k] = np.mean(cluster,axis=0)
```

```
return centroids
    def _random_initialize_centroids(self,X,K):
        indices = np.random.choice(X.shape[0], size=K, replace=False)
        centroids = X[indices]
        return centroids
    def fit(self,X,max iter=10):
        initial centroids =
self._random_initialize_centroids(X,self.k)
        index = np.zeros(X.shape[0])
        centroids = initial centroids
        for i in range(max iter):
            index=self. find closest centroids(X,centroids)
            centroids=self. compute centroid(X,index)
        self.index=index
        self.centroids = centroids
def plot cluster(X,index,centroids):
    plt.figure(figsize=(14, 7))
    plt.scatter(x=centroids[:, 0], y=centroids[:, 1], marker="+",
c='b', linewidth=16)
    plt.scatter(X[:, 0], X[:, 1], c=index, cmap='coolwarm')
    plt.grid()
    plt.savefig("K-Means iris")
    plt.show()
```

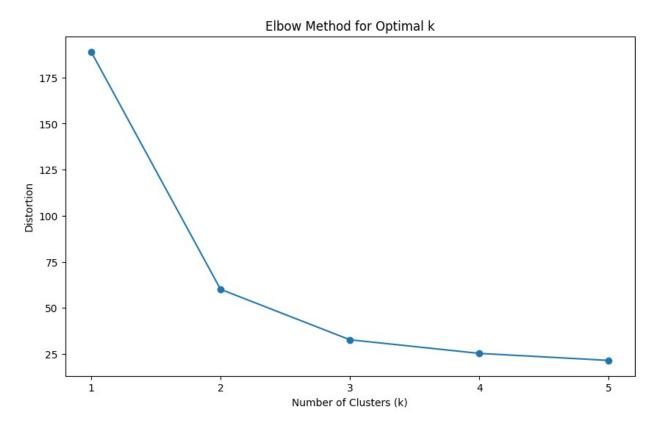
Model Evaluation

```
# initial_centroids = np.array([[5.0,0.3],[6.5,1.5],[7.8,2.4]])
model = KMeans(3)
model.fit(X.values,10)
plot_cluster(X.values,model.index,model.centroids)
```



```
from sklearn.metrics import silhouette score
def calculate distortion(X, centroids, cluster indices):
    distortion = 0
    for i in range(0,centroids.shape[0]):
        cluster points = X[cluster indices == i]
        centroid = centroids[i]
        distortion += np.sum(np.linalq.norm(cluster points - centroid,
axis=1)**2)
    return distortion
def elbow method(X, max k, max iter):
    distortions = []
    silhouette scores =[]
    for k in range(1, \max k + 1):
        model kmeans = KMeans(k)
        model kmeans.fit(X, max iter)
        cluster indices, centroids =
model_kmeans.index,model kmeans.centroids
        distortion = calculate distortion(X, centroids,
cluster indices)
        distortions.append(distortion)
        if k > 1: # Silhouette Score requires at least 2 clusters
            silhouette = silhouette score(X, cluster indices)
            silhouette scores.append(silhouette)
            print(f"At k = \{k\} distortion is \{distortion\} and silhoutee
score is {silhouette}")
    # Plot the elbow graph
```

```
plt.figure(figsize=(10, 6))
    plt.plot(range(1, max k + 1), distortions, marker='o')
    plt.title("Elbow Method for Optimal k")
    plt.xlabel("Number of Clusters (k)")
    plt.ylabel("Distortion")
    plt.xticks(range(1, max k + 1))
    plt.show()
    return distortions, silhouette scores
distortions, silhouette = elbow_method(X.values, 5, 10)
At k = 2 distortion is 60.13669920795727 and silhoutee score is
0.5705593986178453
At k = 3 distortion is 32.76801587301587and silhoutee score is
0.5051309106461155
At k = 4 distortion is 25.380192158385093and silhoutee score is
0.4457401981463308
At k = 5 distortion is 21.57256938490981and silhoutee score is
0.40381998058843466
```



Using SK-Learn

```
from sklearn.cluster import KMeans as sklKMeans

k = 3
model_1 =
sklKMeans(n_clusters=k,init='random',n_init='auto',max_iter=10).fit(X.
values)

print("Distortion Value
",calculate_distortion(X,model_1.cluster_centers_,model_1.labels_))
print("Silhouette ",silhouette_score(X.values,model_1.labels_))

Distortion Value 32.757081714581716
Silhouette 0.5038084350159749
```