# 60. Silhouette Score (Practical)

```
In [1]: import pandas as pd
         import matplotlib.pyplot as plt
         import seaborn as sns
In [2]: dataset = pd.read_csv(r'Data/iris_raw.csv')
         dataset.head(3)
Out[2]:
            sepal_length sepal_width petal_length petal_width
         0
                     5.1
                                  3.5
                                               1.4
                                                            0.2
         1
                     4.9
                                  3.0
                                                            0.2
                                               1.4
         2
                     4.7
                                  3.2
                                               1.3
                                                            0.2
```

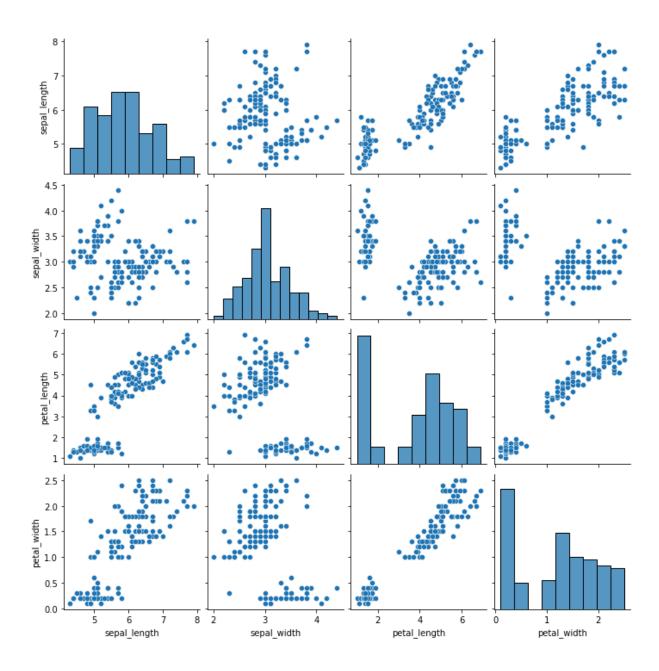
## 54.1 Making Clusters of Data

• Use K-mean clustering when your data is linearly separable

## Check the data if it is linearly separable

```
In [3]: sns.pairplot(data=dataset)
   plt.show()

C:\Users\rashi\AppData\Local\Programs\Python\Python39\lib\site-packages\seaborn\axis
   grid.py:123: UserWarning: The figure layout has changed to tight
   self._figure.tight_layout(*args, **kwargs)
```



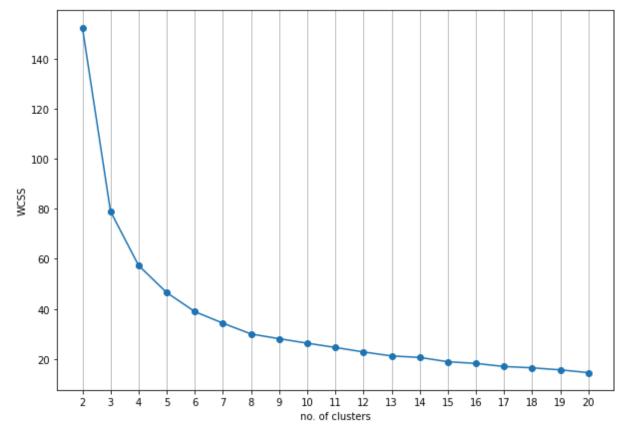
- In supervised learning, the data is split into training and testing data
- In unsupervised learning, data is not split into training and testing data b/c the data is unlabelled

### 54.1.1 Find Number of clusters

```
In [4]: from sklearn.cluster import KMeans
In [5]: # Use a loop to find best number of clusters from 2 to 20
wcss = []

for i in range(2,21):
    km = KMeans(n_clusters=i, init='k-means++')
    km.fit(dataset)
    wcss.append(km.inertia_) # it assings value of wcss {Elbow graph}
```

```
In [6]: plt.figure(figsize=(10,7))
   plt.plot([i for i in range(2,21)], wcss, marker='o')
   plt.xlabel('no. of clusters')
   plt.xticks([i for i in range(2,21)])
   plt.ylabel('WCSS')
   plt.grid(axis='x')
   plt.show()
```



#### Elbow point = 3

#### It means that will have 3 number of clusters

Out[9]:		sepal_length	sepal_width	petal_length	petal_width	Predict
	0	5.1	3.5	1.4	0.2	0
	1	4.9	3.0	1.4	0.2	0
	2	4.7	3.2	1.3	0.2	0
	3	4.6	3.1	1.5	0.2	0
	4	5.0	3.6	1.4	0.2	0
	•••					•••
	145	6.7	3.0	5.2	2.3	2
	146	6.3	2.5	5.0	1.9	1
	147	6.5	3.0	5.2	2.0	2
	148	6.2	3.4	5.4	2.3	2
	149	5.9	3.0	5.1	1.8	1

150 rows × 5 columns

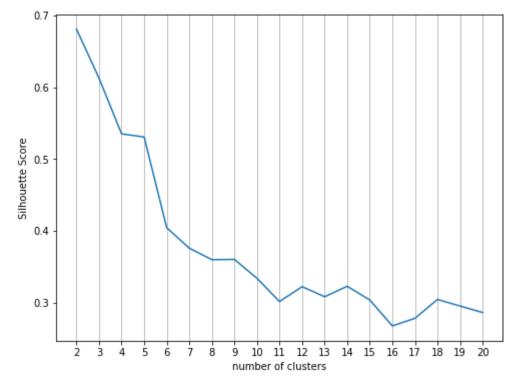
# 54.2 Apply Silhouette Score to validate above results

We are not sure about the results. So will apply loop to determine which silhouette\_score is best and determine what is acutal number of clusters.

```
kmn1.fit(dataset)
ss.append(silhouette_score(dataset, labels=kmn1.labels_))
```

• We are going to make graph b/w ss vs #clusters

```
In [39]: plt.figure(figsize=(8,6))
   plt.plot(n_clusters, ss)
   plt.xlabel('number of clusters')
   plt.ylabel('Silhouette Score')
   plt.xticks(n_clusters)
   plt.grid(axis='x')
   plt.show()
```



#### **Best Silhouette Score = 2**

In [ ]: