

Hands-on: K-means Algorithm



Demo- K-means Algorithm

- We will be using the famous Iris Dataset, collected in the 1930's by Edgar Anderson.
- In this example, we are going to train a random forest classification algorithm to predict the class in the test data

	A	B	C	D	E	F
1	class	petal_len	petal_wid	sepal_len	sepal_width	
2	Iris-virgin	5.5	1.8	6.4	3.1	
3	Iris-virgin	5.9	2.3	6.8	3.2	
4	Iris-virgin	5.4	2.3	6.2	3.4	
5	Iris-virgin	4.8	1.8	6	3	
6	Iris-virgin	5.1	2.3	6.9	3.1	
7	Iris-virgin	5.6	2.4	6.3	3.4	
8	Iris-virgin	5.2	2.3	6.7	3	
9	Iris-virgin	6.7	2	7.7	2.8	
10	Iris-virgin	5.8	2.2	6.5	3	
11	Iris-virgin	5.3	1.9	6.4	2.7	
12	Iris-virgin	5	2	5.7	2.5	
13	Iris-virgin	5.1	1.9	5.8	2.7	

K-means Algorithm

Loading the famous Iris dataset:

1

```
In [5]: from sklearn.datasets import load_iris
```

Having a glance at the shape:

2

```
In [3]: iris_data = load_iris()
```

eano



K-means Algorithm

Building the model:

1

```
In [12]: from sklearn.cluster import KMeans
```

Set up the number of clusters:

2

```
In [25]: # set the number of clusters  
kmeans = KMeans(n_clusters = 3)
```

1

2

```
array([1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,  
       1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,  
       1, 1, 1, 1, 1, 1, 0, 0, 2, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,  
       0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 2, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,  
       0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2,  
       2, 2, 2, 0, 0, 2, 2, 2, 2, 0, 2, 0, 2, 2, 0, 2, 2, 2, 2, 2, 2,  
       2, 0, 2, 2, 2, 2, 0, 2, 2, 2, 0, 2, 2, 2, 0, 2, 2, 0])
```

3

```
array([[5.006      , 3.428      , 1.462      , 0.246      ],
       [5.9016129 , 2.7483871 , 4.39354839, 1.43387097],
       [6.85       , 3.07368421, 5.74210526, 2.07105263]])
```

K-means Algorithm

Evaluating the Algorithm:

```
In [22]: import pandas as pd  
pd.crosstab(iris_data.target, kmodel.labels_)
```



col_0	0	1	2
row_0			
0	50	0	0
1	0	48	2
2	0	14	36

- 0's -> All 50 of them labeled as 0,
- 1's -> 48 of them labeled correctly, 2 of them labeled incorrectly
- 2's -> 36 of them labeled correctly, 14 of them labeled incorrectly