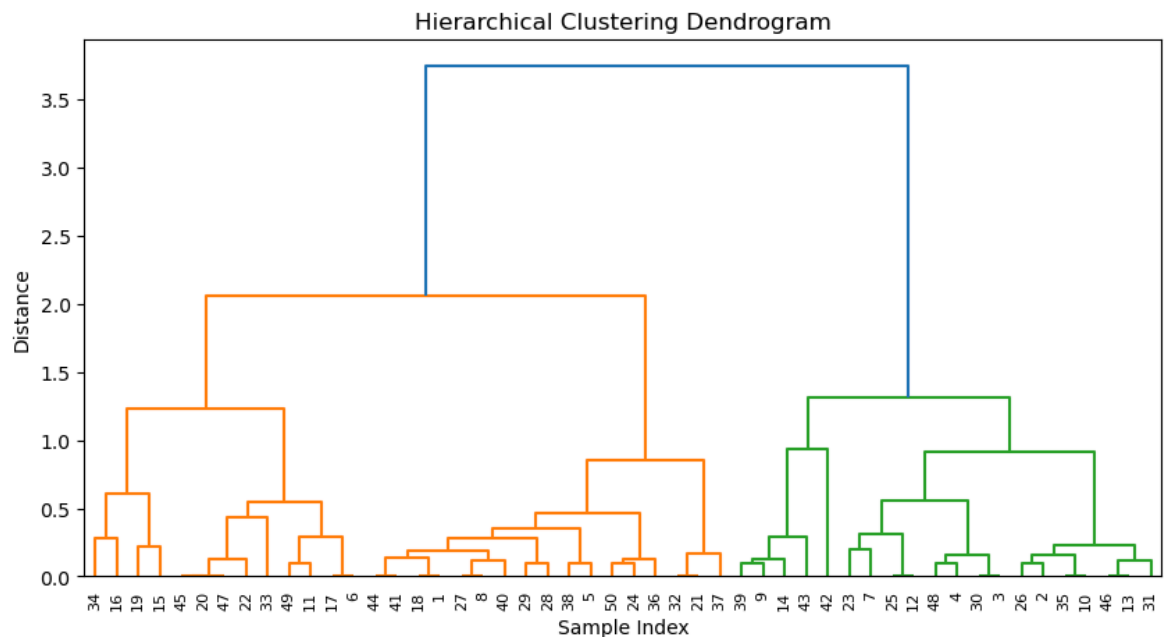


```
In [1]: 1 #now we disscused the heirarchical clustering.....
2 import numpy as np
3 import pandas as pd
4 import matplotlib.pyplot as plt
5 import math
```

```
In [2]: 1 import numpy as np
2 import pandas as pd
3 import matplotlib.pyplot as plt
4 import seaborn as sns
5 from scipy.cluster.hierarchy import dendrogram, linkage
6
7 dataset = pd.read_csv('iris')
8
9 # Select only sepal_length and sepal_width for clustering
10 data = dataset[['sepal_length', 'sepal_width']].values[:50]
11
12 # Perform hierarchical clustering
13 linked = linkage(data, method='ward')
14
15 # Create a dendrogram
16 plt.figure(figsize=(10, 5))
17 dendrogram(linked, orientation='top', labels=range(1, 51), distance_sort=
18 plt.title('Hierarchical Clustering Dendrogram')
19 plt.xlabel('Sample Index')
20 plt.ylabel('Distance')
21 plt.show()
22
```



```
In [4]: 1 import numpy as np
        2 import pandas as pd
        3 import matplotlib.pyplot as plt
        4 import math
        5 from scipy.cluster.hierarchy import dendrogram, linkage
```

```
In [5]: 1 data=pd.read_csv('iris')
        2 data
```

```
Out[5]:
```

	sepal_length	sepal_width	petal_length	petal_width	species
0	5.1	3.5	1.4	0.2	setosa
1	4.9	3.0	1.4	0.2	setosa
2	4.7	3.2	1.3	0.2	setosa
3	4.6	3.1	1.5	0.2	setosa
4	5.0	3.6	1.4	0.2	setosa
...
145	6.7	3.0	5.2	2.3	virginica
146	6.3	2.5	5.0	1.9	virginica
147	6.5	3.0	5.2	2.0	virginica
148	6.2	3.4	5.4	2.3	virginica
149	5.9	3.0	5.1	1.8	virginica

150 rows × 5 columns

```
In [6]: 1 data.isnull().sum()
```

```
Out[6]: sepal_length    0
        sepal_width    0
        petal_length    0
        petal_width    0
        species        0
        dtype: int64
```

```
In [7]: 1 data.duplicated().sum()  
2 dataset=data.drop_duplicates()  
3 dataset
```

```
Out[7]:
```

	sepal_length	sepal_width	petal_length	petal_width	species
0	5.1	3.5	1.4	0.2	setosa
1	4.9	3.0	1.4	0.2	setosa
2	4.7	3.2	1.3	0.2	setosa
3	4.6	3.1	1.5	0.2	setosa
4	5.0	3.6	1.4	0.2	setosa
...
145	6.7	3.0	5.2	2.3	virginica
146	6.3	2.5	5.0	1.9	virginica
147	6.5	3.0	5.2	2.0	virginica
148	6.2	3.4	5.4	2.3	virginica
149	5.9	3.0	5.1	1.8	virginica

149 rows × 5 columns

```
In [8]: 1 input=dataset.iloc[:,[2,3]].values[:50]
        2 input
```

```
Out[8]: array([[1.4, 0.2],
               [1.4, 0.2],
               [1.3, 0.2],
               [1.5, 0.2],
               [1.4, 0.2],
               [1.7, 0.4],
               [1.4, 0.3],
               [1.5, 0.2],
               [1.4, 0.2],
               [1.5, 0.1],
               [1.5, 0.2],
               [1.6, 0.2],
               [1.4, 0.1],
               [1.1, 0.1],
               [1.2, 0.2],
               [1.5, 0.4],
               [1.3, 0.4],
               [1.4, 0.3],
               [1.7, 0.3],
               [1.5, 0.3],
               [1.7, 0.2],
               [1.5, 0.4],
               [1. , 0.2],
               [1.7, 0.5],
               [1.9, 0.2],
               [1.6, 0.2],
               [1.6, 0.4],
               [1.5, 0.2],
               [1.4, 0.2],
               [1.6, 0.2],
               [1.6, 0.2],
               [1.5, 0.4],
               [1.5, 0.1],
               [1.4, 0.2],
               [1.5, 0.2],
               [1.2, 0.2],
               [1.3, 0.2],
               [1.4, 0.1],
               [1.3, 0.2],
               [1.5, 0.2],
               [1.3, 0.3],
               [1.3, 0.3],
               [1.3, 0.2],
               [1.6, 0.6],
               [1.9, 0.4],
               [1.4, 0.3],
               [1.6, 0.2],
               [1.4, 0.2],
               [1.5, 0.2],
               [1.4, 0.2]])
```

```
In [9]: 1 linked=linkage(input,method='ward')
```

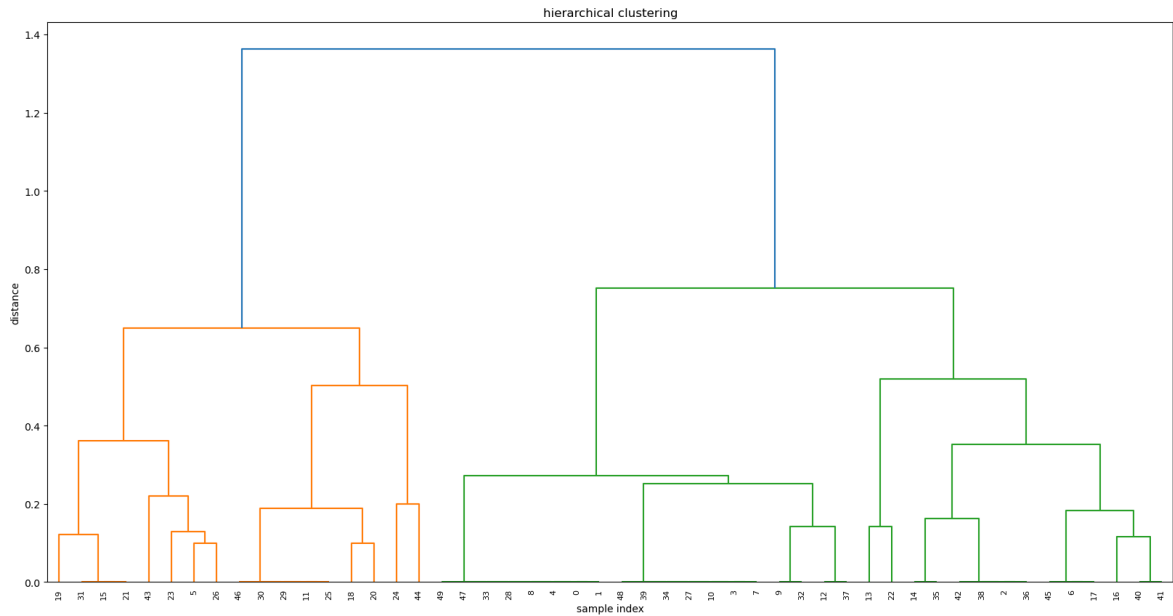
```
In [12]: 1 linked[:50]
```

```
Out[12]: array([[ 0.          ,  1.          ,  0.          ,  2.          ],
 [ 4.          , 50.          ,  0.          ,  3.          ],
 [ 2.          , 36.          ,  0.          ,  2.          ],
 [ 3.          ,  7.          ,  0.          ,  2.          ],
 [ 8.          , 51.          ,  0.          ,  4.          ],
 [ 6.          , 17.          ,  0.          ,  2.          ],
 [10.          , 53.          ,  0.          ,  3.          ],
 [28.          , 54.          ,  0.          ,  5.          ],
 [ 9.          , 32.          ,  0.          ,  2.          ],
 [27.          , 56.          ,  0.          ,  4.          ],
 [11.          , 25.          ,  0.          ,  2.          ],
 [12.          , 37.          ,  0.          ,  2.          ],
 [14.          , 35.          ,  0.          ,  2.          ],
 [15.          , 21.          ,  0.          ,  2.          ],
 [40.          , 41.          ,  0.          ,  2.          ],
 [45.          , 55.          ,  0.          ,  3.          ],
 [38.          , 52.          ,  0.          ,  3.          ],
 [42.          , 66.          ,  0.          ,  4.          ],
 [29.          , 60.          ,  0.          ,  3.          ],
 [30.          , 68.          ,  0.          ,  4.          ],
 [46.          , 69.          ,  0.          ,  5.          ],
 [34.          , 59.          ,  0.          ,  5.          ],
 [39.          , 71.          ,  0.          ,  6.          ],
 [48.          , 72.          ,  0.          ,  7.          ],
 [31.          , 63.          ,  0.          ,  3.          ],
 [33.          , 57.          ,  0.          ,  6.          ],
 [47.          , 75.          ,  0.          ,  7.          ],
 [49.          , 76.          ,  0.          ,  8.          ],
 [ 5.          , 26.          ,  0.1          ,  2.          ],
 [18.          , 20.          ,  0.1          ,  2.          ],
 [16.          , 64.          ,  0.11547005,  3.          ],
 [19.          , 74.          ,  0.12247449,  4.          ],
 [23.          , 78.          ,  0.12909944,  3.          ],
 [13.          , 22.          ,  0.14142136,  2.          ],
 [58.          , 61.          ,  0.14142136,  4.          ],
 [62.          , 67.          ,  0.16329932,  6.          ],
 [65.          , 80.          ,  0.18257419,  6.          ],
 [70.          , 79.          ,  0.18898224,  7.          ],
 [24.          , 44.          ,  0.2          ,  2.          ],
 [43.          , 82.          ,  0.21984843,  4.          ],
 [73.          , 84.          ,  0.25226249, 11.          ],
 [77.          , 90.          ,  0.2725039 , 19.          ],
 [85.          , 86.          ,  0.35118846, 12.          ],
 [81.          , 89.          ,  0.36055513,  8.          ],
 [87.          , 88.          ,  0.50205925,  9.          ],
 [83.          , 92.          ,  0.518698 , 14.          ],
 [93.          , 94.          ,  0.64888038, 17.          ],
 [91.          , 95.          ,  0.75158222, 33.          ],
 [96.          , 97.          ,  1.36260472, 50.          ]])
```

```

In [20]: 1 # dendrogram(linked, orientation='top', labels=range(1, 51), distance_sort
2 plt.figure(figsize=(20,10))
3 #this method is divisive hierarchical algorithm
4 dendrogram(linked,orientation='top',distance_sort='accending', show_leaf_
5 plt.title('hierarchical clustering')
6 plt.xlabel('sample index')
7 plt.ylabel('distance')
8
9 plt.show()

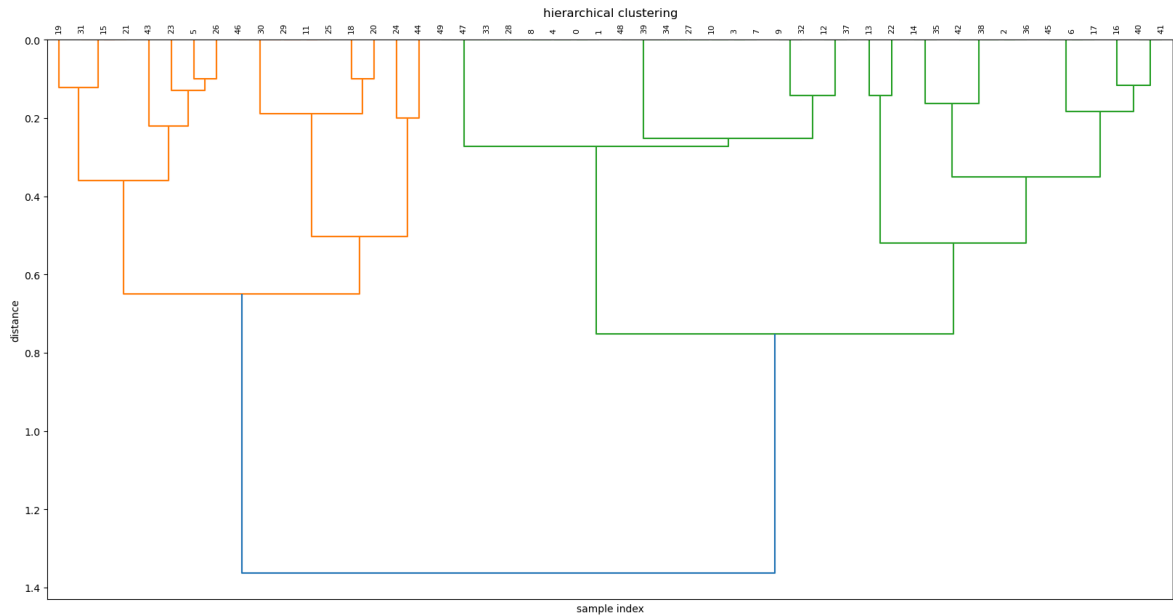
```



```

In [21]: 1 # dendrogram(linked, orientation='top', labels=range(1, 51), distance_sort
2 plt.figure(figsize=(20,10))
3 # this is agglomerative hierarchical clustering ...
4 dendrogram(linked,orientation='bottom',distance_sort='decending', show_le
5 plt.title('hierarchical clustering')
6 plt.xlabel('sample index')
7 plt.ylabel('distance')
8
9 plt.show()

```



```

In [ ]: 1

```