

hierarchical

April 3, 2022

1 Hierarical Clustring (Agglomerative-single Linkage)

1.0.1 Importing packages

```
[ ]: import numpy as np
import pandas as pd
from matplotlib import pyplot as plt
from scipy.cluster.hierarchy import dendrogram
from sklearn.cluster import AgglomerativeClustering
```

1.1 Reading the dateset & Renaming columns lables

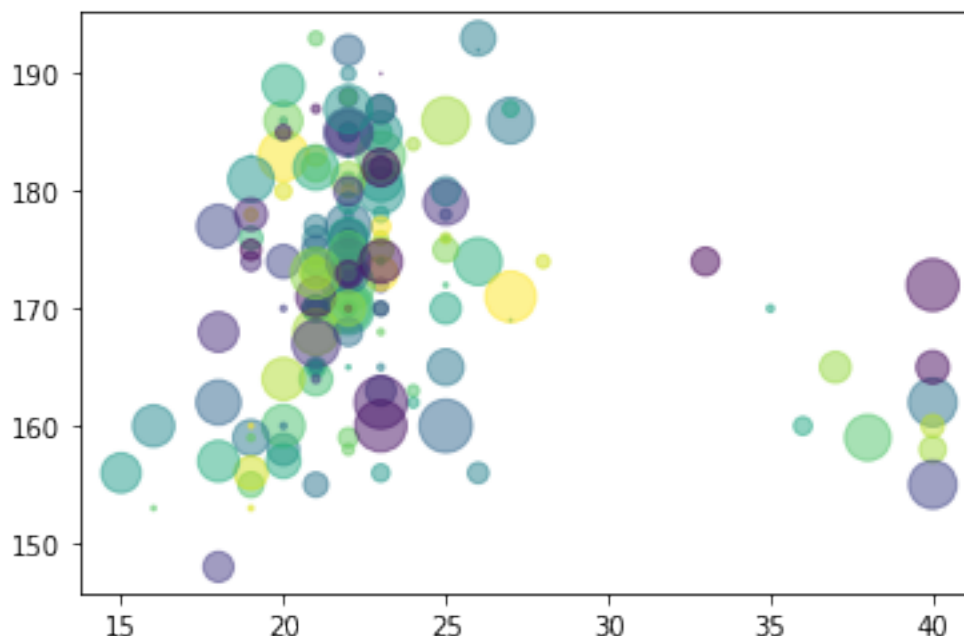
```
[ ]: col_names = ['timeStamp', 'Gender', 'Grade', 'Age', 'Length', 'Weight',
↳ 'ShoesSize']
dataframe = pd.read_csv("../human_features.csv", names = col_names,
↳ skiprows=(0, ))
```

1.2 Visualizing the data

```
[ ]: x = dataframe.loc[:, 'Age']
y = dataframe.loc[:, 'Length']
```

```
[ ]: n = x.shape[0]
colors = np.random.rand(n)
size = pow(20 * np.random.rand(n), 2)
plt.scatter(x, y, s=size, c=colors, alpha=0.5)
```

```
[ ]: <matplotlib.collections.PathCollection at 0x7f422a9afdc0>
```



1.3 Creating Agglomerative Clustering model

```
[ ]: cluster_model = AgglomerativeClustering(distance_threshold=0, n_clusters=None)
```

1.3.1 Training the model on the data

```
[ ]: clustring = cluster_model.fit(dataframe.loc[:, ['Age', 'Length']])
```

```
[ ]: clustring.labels_
```

```
[ ]: array([120, 125, 157, 156, 155, 118, 153, 117, 152, 154, 116, 151, 150,
          115, 134, 148,  58,  86, 114, 128, 143, 142, 111, 112, 144, 147,
          110, 107, 133, 106,  76, 149, 131,  57,  91, 138, 113,  78, 105,
          132, 104, 108,  99, 139,  89,  85, 103,  71, 123,  55, 141,  70,
           80,  77, 137, 122,  97,  87,  84,  56,  68,  92, 126,  51, 109,
           53, 140,  45, 146, 101,  27,  52, 145,  83, 121,  28,  82, 136,
          119,  93,  90,  41,  95,  81,  62,  25,  54, 135,  75,  79, 102,
           42,  67, 124,  94,  98,  26, 127,  37,  61,  63,  96,  46, 129,
          100,  39, 130,  40,  60,  20,  30,  73,  88,  69,  59,  47,  72,
           74,  49,  66,  36,  44,  29,  50,  24,  13,  64,  65,  48,  43,
           38,  21,  34,  23,  14,  33,  31,  22,  32,  19,  15,  11,  16,
           35,  12,  10,   6,  18,   9,   7,   5,   4,   2,  17,   8,   3,
           1,   0])
```

```
[ ]: clustring.labels_.shape
```

```
[ ]: (158,)
```

```
[ ]: clustering.children_
```

```
[ ]: array([[ 2, 37],
           [ 3, 71],
           [ 4, 39],
           [ 9, 91],
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           [ 8, 57],
           [11, 30],
           [12, 14],
           [31, 165],
           [15, 105],
           [111, 112],
           [68, 73],
           [72, 154],
           [24, 162],
           [20, 107],
           [21, 115],
           [50, 164],
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```

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[266, 283],
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```

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[301, 303],
[296, 300],
[298, 304],
[289, 293],
[292, 302],
[299, 306],
[307, 308],
[305, 309],
[311, 312],
[310, 313]])

```

```
[ ]: clustring.children_.shape
```

```
[ ]: (157, 2)
```

```

[ ]: def plot_dendrogram(model, **kwargs):
    # Create linkage matrix and then plot the dendrogram

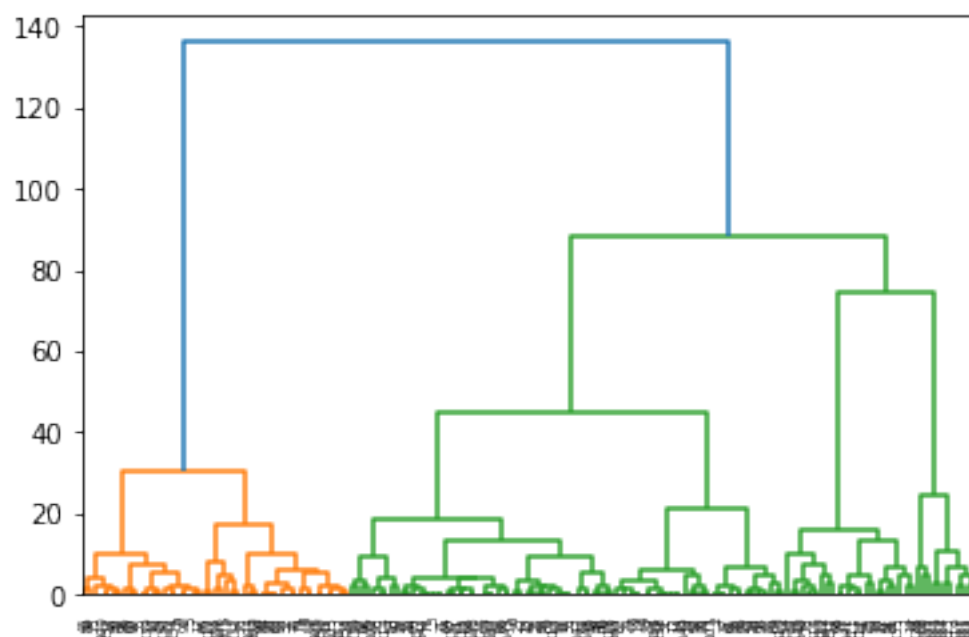
    # create the counts of samples under each node
    counts = np.zeros(model.children_.shape[0])
    n_samples = len(model.labels_)
    for i, merge in enumerate(model.children_):
        current_count = 0
        for child_idx in merge:
            if child_idx < n_samples:
                current_count += 1 # leaf node
            else:
                current_count += counts[child_idx - n_samples]
        counts[i] = current_count

    linkage_matrix = np.column_stack(
        [model.children_, model.distances_, counts]
    ).astype(float)

    # Plot the corresponding dendrogram
    dendrogram(linkage_matrix, **kwargs)

```

```
[ ]: plot_dendrogram(clustering)
```



```
[ ]: plot_dendrogram(clustering, truncate_mode="level", p=3 )
```

