

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import scipy.cluster.hierarchy as shc
from scipy.spatial.distance import squareform, pdist
```

## Random Dataset

```
X = np.random.random_sample(size = 10)
Y = np.random.random_sample(size = 10)

point = ['P1', 'P2', 'P3', 'P4', 'P5', 'P6', 'P7', 'P8', 'P9', 'P10']

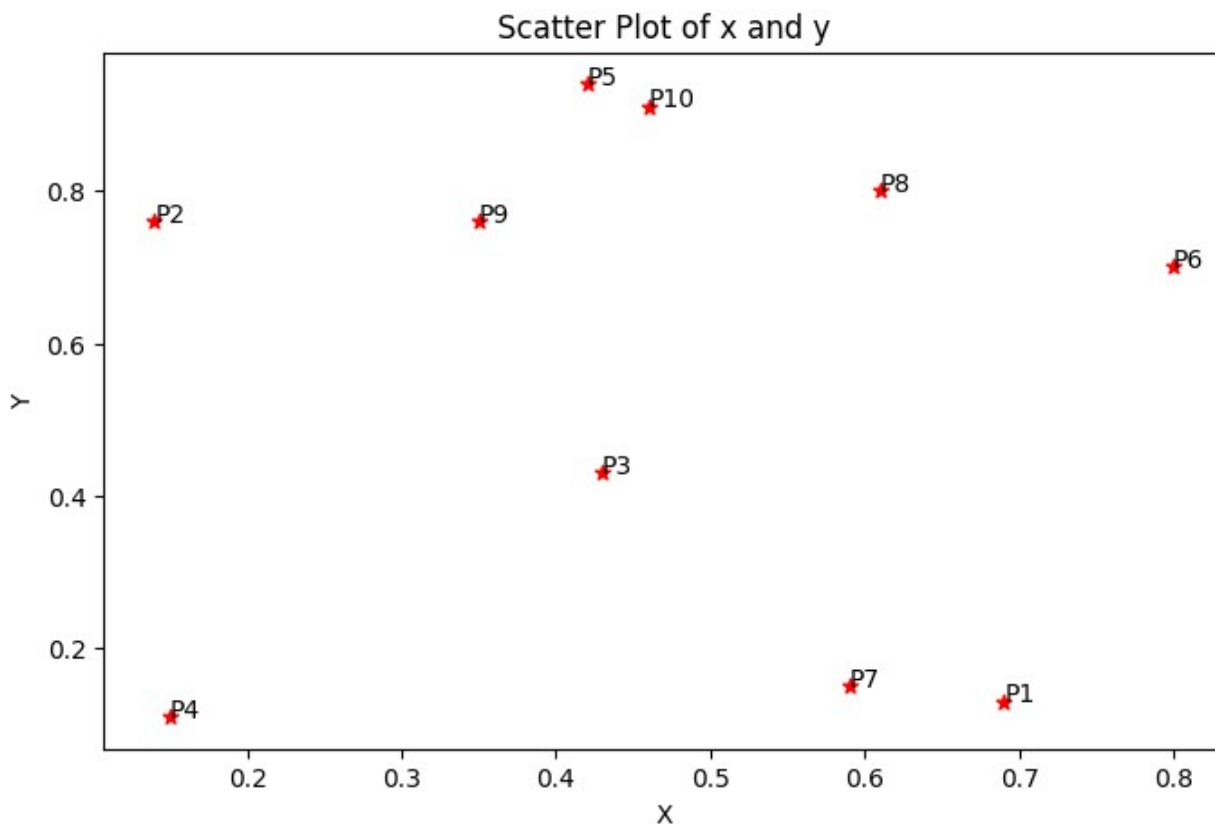
data = pd.DataFrame({'Point':point, 'X':np.round(X,2),
                    'Y':np.round(Y,2)})
data = data.set_index('Point')
```

data

	X	Y
Point		
P1	0.69	0.13
P2	0.14	0.76
P3	0.43	0.43
P4	0.15	0.11
P5	0.42	0.94
P6	0.80	0.70
P7	0.59	0.15
P8	0.61	0.80
P9	0.35	0.76
P10	0.46	0.91

## Visualize Dataset

```
plt.figure(figsize=(8,5))
plt.scatter(data['X'], data['Y'], c='r', marker='*')
plt.xlabel('X')
plt.ylabel('Y')
plt.title('Scatter Plot of x and y')
for i in data.irtuples():
    plt.annotate(i.Index, (i.X, i.Y))
```



## Distance Matrix

```
dist = pd.DataFrame(squareform(pdist(data[['X', 'Y']]), 'euclidean'),
columns=data.index.values, index=data.index.values)
dist
```

	P1	P2	P3	P4	P5	P6
P7 \						
P1	0.000000	0.836301	0.396989	0.540370	0.853815	0.580517
P2	0.836301	0.000000	0.439318	0.650077	0.332866	0.662722
P3	0.396989	0.439318	0.000000	0.425206	0.510098	0.458039
P4	0.540370	0.650077	0.425206	0.000000	0.872812	0.877838
P5	0.853815	0.332866	0.510098	0.872812	0.000000	0.449444
P6	0.580517	0.662722	0.458039	0.877838	0.449444	0.000000
P7	0.101980	0.758024	0.322490	0.441814	0.808084	0.588727
P8	0.674759	0.471699	0.411461	0.829277	0.236008	0.214709

P9	0.715891	0.210000	0.339559	0.680074	0.193132	0.453982
	0.655515					
P10	0.813204	0.353412	0.480937	0.857963	0.050000	0.399625
	0.771038					

	P8	P9	P10
P1	0.674759	0.715891	0.813204
P2	0.471699	0.210000	0.353412
P3	0.411461	0.339559	0.480937
P4	0.829277	0.680074	0.857963
P5	0.236008	0.193132	0.050000
P6	0.214709	0.453982	0.399625
P7	0.650308	0.655515	0.771038
P8	0.000000	0.263059	0.186011
P9	0.263059	0.000000	0.186011
P10	0.186011	0.186011	0.000000

## Single-Link Clustering

```
def single_linkage(dist_matrix):
    n = len(dist_matrix)
    while n > 1:
        min_val = float('inf')
        min_index = None
        for i in range(n):
            for j in range(i+1, n):
                if dist_matrix.iloc[i, j] < min_val and
dist_matrix.index[i] != dist_matrix.columns[j]:
                    min_val = dist_matrix.iloc[i, j]
                    min_index = (i, j)

            if min_val == float('inf'):
                break

        i, j = min_index
        cluster1, cluster2 = dist_matrix.index[i],
dist_matrix.columns[j]
        print(f'Merging clusters {cluster1} and {cluster2} with
distance {min_val}')

        new_cluster = f'({cluster1},{cluster2})'

        dist_matrix[new_cluster] = dist_matrix[[cluster1,
cluster2]].min(axis=1)
        dist_matrix.loc[new_cluster] = dist_matrix.loc[[cluster1,
cluster2]].min(axis=0)
        dist_matrix = dist_matrix.drop([cluster1, cluster2], axis=0)
        dist_matrix = dist_matrix.drop([cluster1, cluster2], axis=1)
```

```

n -= 1

print(dist_matrix)
print("")

print("Single-Link Clustering:")
single_linkage(dist.copy())

```

Single-Link Clustering:

Merging clusters P5 and P10 with distance 0.04999999999999998

	P1	P2	P3	P4	P6	
P7 \						
P1	0.000000	0.836301	0.396989	0.540370	0.580517	0.101980
P2	0.836301	0.000000	0.439318	0.650077	0.662722	0.758024
P3	0.396989	0.439318	0.000000	0.425206	0.458039	0.322490
P4	0.540370	0.650077	0.425206	0.000000	0.877838	0.441814
P6	0.580517	0.662722	0.458039	0.877838	0.000000	0.588727
P7	0.101980	0.758024	0.322490	0.441814	0.588727	0.000000
P8	0.674759	0.471699	0.411461	0.829277	0.214709	0.650308
P9	0.715891	0.210000	0.339559	0.680074	0.453982	0.655515
(P5,P10)	0.813204	0.332866	0.480937	0.857963	0.399625	0.771038

	P8	P9	(P5,P10)
P1	0.674759	0.715891	0.813204
P2	0.471699	0.210000	0.332866
P3	0.411461	0.339559	0.480937
P4	0.829277	0.680074	0.857963
P6	0.214709	0.453982	0.399625
P7	0.650308	0.655515	0.771038
P8	0.000000	0.263059	0.186011
P9	0.263059	0.000000	0.186011
(P5,P10)	0.186011	0.186011	0.000000

Merging clusters P1 and P7 with distance 0.10198039027185567

	P2	P3	P4	P6	P8	
P9 \						
P2	0.000000	0.439318	0.650077	0.662722	0.471699	0.210000
P3	0.439318	0.000000	0.425206	0.458039	0.411461	0.339559
P4	0.650077	0.425206	0.000000	0.877838	0.829277	0.680074

P6	0.662722	0.458039	0.877838	0.000000	0.214709	0.453982
P8	0.471699	0.411461	0.829277	0.214709	0.000000	0.263059
P9	0.210000	0.339559	0.680074	0.453982	0.263059	0.000000
(P5,P10)	0.332866	0.480937	0.857963	0.399625	0.186011	0.186011
(P1,P7)	0.758024	0.322490	0.441814	0.580517	0.650308	0.655515

	(P5,P10)	(P1,P7)
P2	0.332866	0.758024
P3	0.480937	0.322490
P4	0.857963	0.441814
P6	0.399625	0.580517
P8	0.186011	0.650308
P9	0.186011	0.655515
(P5,P10)	0.000000	0.771038
(P1,P7)	0.771038	0.000000

Merging clusters P8 and (P5,P10) with distance 0.1860107523773827

	P2	P3	P4	P6	P9
(P1,P7) \					
P2	0.000000	0.439318	0.650077	0.662722	0.210000
0.758024					
P3	0.439318	0.000000	0.425206	0.458039	0.339559
0.322490					
P4	0.650077	0.425206	0.000000	0.877838	0.680074
0.441814					
P6	0.662722	0.458039	0.877838	0.000000	0.453982
0.580517					
P9	0.210000	0.339559	0.680074	0.453982	0.000000
0.655515					
(P1,P7)	0.758024	0.322490	0.441814	0.580517	0.655515
0.000000					
(P8,(P5,P10))	0.332866	0.411461	0.829277	0.214709	0.186011
0.650308					

	(P8,(P5,P10))
P2	0.332866
P3	0.411461
P4	0.829277
P6	0.214709
P9	0.186011
(P1,P7)	0.650308
(P8,(P5,P10))	0.000000

Merging clusters P9 and (P8,(P5,P10)) with distance 0.1860107523773828

P2 P3 P4 P6

(P1,P7) \					
P2	0.000000	0.439318	0.650077	0.662722	0.758024
P3	0.439318	0.000000	0.425206	0.458039	0.322490
P4	0.650077	0.425206	0.000000	0.877838	0.441814
P6	0.662722	0.458039	0.877838	0.000000	0.580517
(P1,P7)	0.758024	0.322490	0.441814	0.580517	0.000000
(P9,(P8,(P5,P10)))	0.210000	0.339559	0.680074	0.214709	0.650308

	(P9,(P8,(P5,P10)))
P2	0.210000
P3	0.339559
P4	0.680074
P6	0.214709
(P1,P7)	0.650308
(P9,(P8,(P5,P10)))	0.000000

Merging clusters P2 and (P9,(P8,(P5,P10))) with distance  
0.20999999999999996

	P3	P4	P6	(P1,P7) \
P3	0.000000	0.425206	0.458039	0.322490
P4	0.425206	0.000000	0.877838	0.441814
P6	0.458039	0.877838	0.000000	0.580517
(P1,P7)	0.322490	0.441814	0.580517	0.000000
(P2,(P9,(P8,(P5,P10))))	0.339559	0.650077	0.214709	0.650308

	(P2,(P9,(P8,(P5,P10))))
P3	0.339559
P4	0.650077
P6	0.214709
(P1,P7)	0.650308
(P2,(P9,(P8,(P5,P10))))	0.000000

Merging clusters P6 and (P2,(P9,(P8,(P5,P10)))) with distance  
0.21470910553583897

	P3	P4	(P1,P7) \
P3	0.000000	0.425206	0.322490
P4	0.425206	0.000000	0.441814
(P1,P7)	0.322490	0.441814	0.000000
(P6,(P2,(P9,(P8,(P5,P10))))	0.339559	0.650077	0.580517

	(P6,(P2,(P9,(P8,(P5,P10))))
P3	0.339559
P4	0.650077
(P1,P7)	0.580517

```

(P6,(P2,(P9,(P8,(P5,P10)))))) 0.000000
Merging clusters P3 and (P1,P7) with distance 0.322490309931942
P4 (P6,(P2,(P9,(P8,
(P5,P10)))))) \
P4 0.000000 0.650077
(P6,(P2,(P9,(P8,(P5,P10)))))) 0.650077 0.000000
(P3,(P1,P7)) 0.425206 0.339559

(P3,(P1,P7))
P4 0.425206
(P6,(P2,(P9,(P8,(P5,P10)))))) 0.339559
(P3,(P1,P7)) 0.000000
Merging clusters (P6,(P2,(P9,(P8,(P5,P10)))))) and (P3,(P1,P7)) with
distance 0.3395585369269929
P4 \
P4 0.000000
((P6,(P2,(P9,(P8,(P5,P10))))), (P3,(P1,P7))) 0.425206
((P6,(P2,(P9,(P8,
(P5,P10))))), (P3,(P1,P7)))
P4
0.425206
((P6,(P2,(P9,(P8,(P5,P10))))), (P3,(P1,P7)))
0.000000
Merging clusters P4 and ((P6,(P2,(P9,(P8,(P5,P10))))), (P3,(P1,P7)))
with distance 0.425205832509386
(P4,((P6,(P2,(P9,
(P8,(P5,P10))))), (P3,(P1,P7))))
(P4,((P6,(P2,(P9,(P8,(P5,P10))))), (P3,(P1,P7))))
0.0

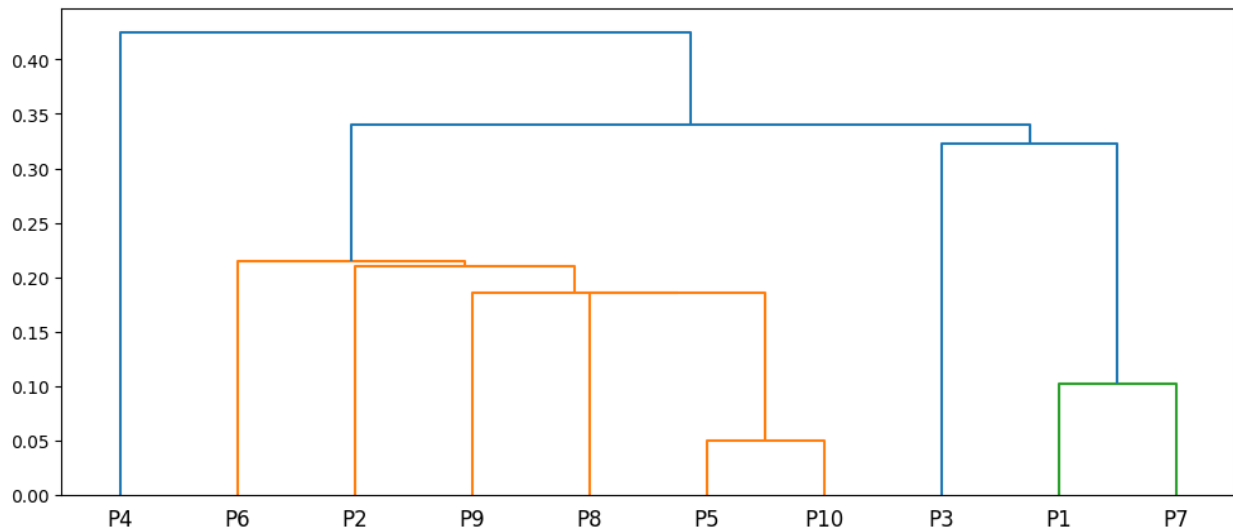
```

## Visualize Dendrogram for Single-Link Cluster

```

plt.figure(figsize=(12,5))
dend = shc.dendrogram(shc.linkage(data[['X', 'Y']], method='single'),
labels=data.index)

```



## Complete-Link Clustering

```
def complete_linkage(dist_matrix):
    n = len(dist_matrix)
    while n > 1:
        min_val = float('inf')
        min_index = None
        for i in range(n):
            for j in range(i+1, n):
                if dist_matrix.iloc[i, j] < min_val and
dist_matrix.index[i] != dist_matrix.columns[j]:
                    min_val = dist_matrix.iloc[i, j]
                    min_index = (i, j)

        if min_val == float('inf'):
            break

        i, j = min_index
        cluster1, cluster2 = dist_matrix.index[i],
dist_matrix.columns[j]

        print(f'Merging clusters {cluster1} and {cluster2} with
distance {min_val}')
        new_cluster = f'({cluster1},{cluster2})'

        dist_matrix[new_cluster] = dist_matrix[[cluster1,
cluster2]].max(axis=1)
        dist_matrix.loc[new_cluster] = dist_matrix.loc[[cluster1,
cluster2]].max(axis=0)
        dist_matrix = dist_matrix.drop([cluster1, cluster2], axis=0)
        dist_matrix = dist_matrix.drop([cluster1, cluster2], axis=1)

        n -= 1
```



```

print(dist_matrix)
print("")

print("Complete-Link Clustering:")
complete_linkage(dist.copy())

```

Complete-Link Clustering:

Merging clusters P5 and P10 with distance 0.04999999999999998

	P1	P2	P3	P4	P6	
P7 \						
P1	0.000000	0.836301	0.396989	0.540370	0.580517	0.101980
P2	0.836301	0.000000	0.439318	0.650077	0.662722	0.758024
P3	0.396989	0.439318	0.000000	0.425206	0.458039	0.322490
P4	0.540370	0.650077	0.425206	0.000000	0.877838	0.441814
P6	0.580517	0.662722	0.458039	0.877838	0.000000	0.588727
P7	0.101980	0.758024	0.322490	0.441814	0.588727	0.000000
P8	0.674759	0.471699	0.411461	0.829277	0.214709	0.650308
P9	0.715891	0.210000	0.339559	0.680074	0.453982	0.655515
(P5,P10)	0.853815	0.353412	0.510098	0.872812	0.449444	0.808084

	P8	P9	(P5,P10)
P1	0.674759	0.715891	0.853815
P2	0.471699	0.210000	0.353412
P3	0.411461	0.339559	0.510098
P4	0.829277	0.680074	0.872812
P6	0.214709	0.453982	0.449444
P7	0.650308	0.655515	0.808084
P8	0.000000	0.263059	0.236008
P9	0.263059	0.000000	0.193132
(P5,P10)	0.236008	0.193132	0.050000

Merging clusters P1 and P7 with distance 0.10198039027185567

	P2	P3	P4	P6	P8	
P9 \						
P2	0.000000	0.439318	0.650077	0.662722	0.471699	0.210000
P3	0.439318	0.000000	0.425206	0.458039	0.411461	0.339559
P4	0.650077	0.425206	0.000000	0.877838	0.829277	0.680074
P6	0.662722	0.458039	0.877838	0.000000	0.214709	0.453982

P8	0.471699	0.411461	0.829277	0.214709	0.000000	0.263059
P9	0.210000	0.339559	0.680074	0.453982	0.263059	0.000000
(P5,P10)	0.353412	0.510098	0.872812	0.449444	0.236008	0.193132
(P1,P7)	0.836301	0.396989	0.540370	0.588727	0.674759	0.715891

	(P5,P10)	(P1,P7)
P2	0.353412	0.836301
P3	0.510098	0.396989
P4	0.872812	0.540370
P6	0.449444	0.588727
P8	0.236008	0.674759
P9	0.193132	0.715891
(P5,P10)	0.050000	0.853815
(P1,P7)	0.853815	0.101980

Merging clusters P9 and (P5,P10) with distance 0.1931320791582796

	P2	P3	P4	P6	P8
(P1,P7) \					
P2	0.000000	0.439318	0.650077	0.662722	0.471699
0.836301					
P3	0.439318	0.000000	0.425206	0.458039	0.411461
0.396989					
P4	0.650077	0.425206	0.000000	0.877838	0.829277
0.540370					
P6	0.662722	0.458039	0.877838	0.000000	0.214709
0.588727					
P8	0.471699	0.411461	0.829277	0.214709	0.000000
0.674759					
(P1,P7)	0.836301	0.396989	0.540370	0.588727	0.674759
0.101980					
(P9,(P5,P10))	0.353412	0.510098	0.872812	0.453982	0.263059
0.853815					

	(P9,(P5,P10))
P2	0.353412
P3	0.510098
P4	0.872812
P6	0.453982
P8	0.263059
(P1,P7)	0.853815
(P9,(P5,P10))	0.193132

Merging clusters P6 and P8 with distance 0.21470910553583897

	P2	P3	P4	(P1,P7)	(P9,(P5,P10))
(P6,P8)					

P2	0.000000	0.439318	0.650077	0.836301	0.353412
0.662722					
P3	0.439318	0.000000	0.425206	0.396989	0.510098
0.458039					
P4	0.650077	0.425206	0.000000	0.540370	0.872812
0.877838					
(P1,P7)	0.836301	0.396989	0.540370	0.101980	0.853815
0.674759					
(P9,(P5,P10))	0.353412	0.510098	0.872812	0.853815	0.193132
0.453982					
(P6,P8)	0.662722	0.458039	0.877838	0.674759	0.453982
0.214709					

Merging clusters P2 and (P9,(P5,P10)) with distance 0.35341194094144585

	P3	P4	(P1,P7)	(P6,P8)	(P2,(P9,(P5,P10)))
P3	0.000000	0.425206	0.396989	0.458039	
0.510098					
P4	0.425206	0.000000	0.540370	0.877838	
0.872812					
(P1,P7)	0.396989	0.540370	0.101980	0.674759	
0.853815					
(P6,P8)	0.458039	0.877838	0.674759	0.214709	
0.662722					
(P2,(P9,(P5,P10)))	0.510098	0.872812	0.853815	0.662722	
0.353412					

Merging clusters P3 and (P1,P7) with distance 0.3969886648255841

	P4	(P6,P8)	(P2,(P9,(P5,P10)))	(P3,(P1,P7))
P4	0.000000	0.877838		0.872812
0.540370				
(P6,P8)	0.877838	0.214709		0.662722
0.674759				
(P2,(P9,(P5,P10)))	0.872812	0.662722		0.353412
0.853815				
(P3,(P1,P7))	0.540370	0.674759		0.853815
0.396989				

Merging clusters P4 and (P3,(P1,P7)) with distance 0.5403702434442518

	(P6,P8)	(P2,(P9,(P5,P10)))	(P4,(P3,(P1,P7)))
(P6,P8)	0.214709	0.662722	0.877838
(P2,(P9,(P5,P10)))	0.662722	0.353412	0.872812
(P4,(P3,(P1,P7)))	0.877838	0.872812	0.540370

Merging clusters (P6,P8) and (P2,(P9,(P5,P10))) with distance 0.6627216610312356

	(P4,(P3,(P1,P7)))	((P6,P8),(P2,(P9,(P5,P10))))
(P4,(P3,(P1,P7)))		
((P6,P8),(P2,(P9,(P5,P10))))		

```

(P4, (P3, (P1, P7)))                                0.540370
0.877838
((P6, P8), (P2, (P9, (P5, P10))))                  0.877838
0.662722

Merging clusters (P4, (P3, (P1, P7))) and ((P6, P8), (P2, (P9, (P5, P10))))
with distance 0.8778382538941898
((P4, (P3, (P1, P7))), ((P6, P8), (P2, (P9, (P5, P10))))),
((P6, P8), (P2, (P9, (P5, P10))))
((P4, (P3, (P1, P7))), ((P6, P8), (P2, (P9, (P5, P10)))))
0.877838

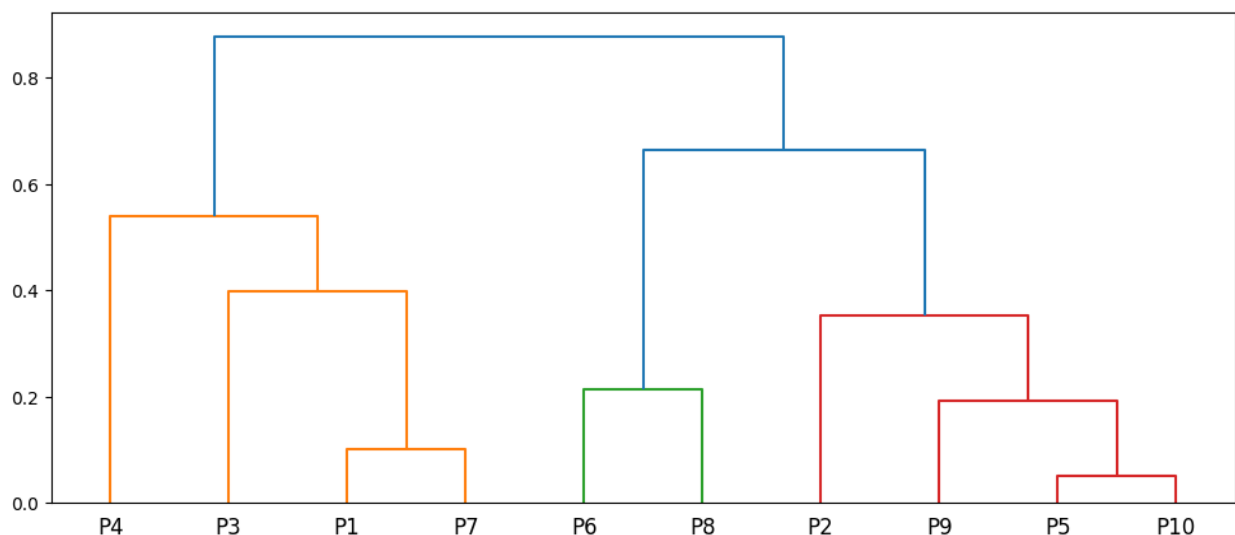
```

## Visualize Dendrogram for Complete-Link Cluster

```

plt.figure(figsize=(12,5))
dend = shc.dendrogram(shc.linkage(data[['X', 'Y']],
method='complete'), labels=data.index)

```



## Model Interpretation

### Single-Linkage:

- The single-link clustering algorithm merges clusters based on the minimum distance between individual data points in the clusters.
- As the merging progresses, clusters with smaller distances are merged first, indicating high similarity or proximity between the data points within those clusters
- The process continues, gradually merging clusters with larger distances, until all data points are included in a single cluster

### Complete-Linkage:

- Complete-link clustering merges clusters based on the maximum distance between any two points in the clusters.

- The merging process starts with clusters having smaller maximum distances, indicating high similarity or proximity between the data points within those clusters
- As the process continues, clusters with larger maximum distances are merged, suggesting either dissimilarity between some points within those clusters or the presence of outliers.