

# nrcm-hierarchical-clustering-1

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## 0.1 project title

Analysis and prediction of “mall\_customers.csv” of American mall market called as Phonix Mall, to find out requirements of dendrogram using seipy library with the help of ”seipy.cluster.hierarchy”, to see the no of linkage of the clustering to predict

## 0.2 PROJECT TITLE :The american finance market client as per the rate of gdp of 2011 found as highest number of growth in there business market.

As a data science engineer find out which hierarchy gives upcoming feature

## 0.3 TASK

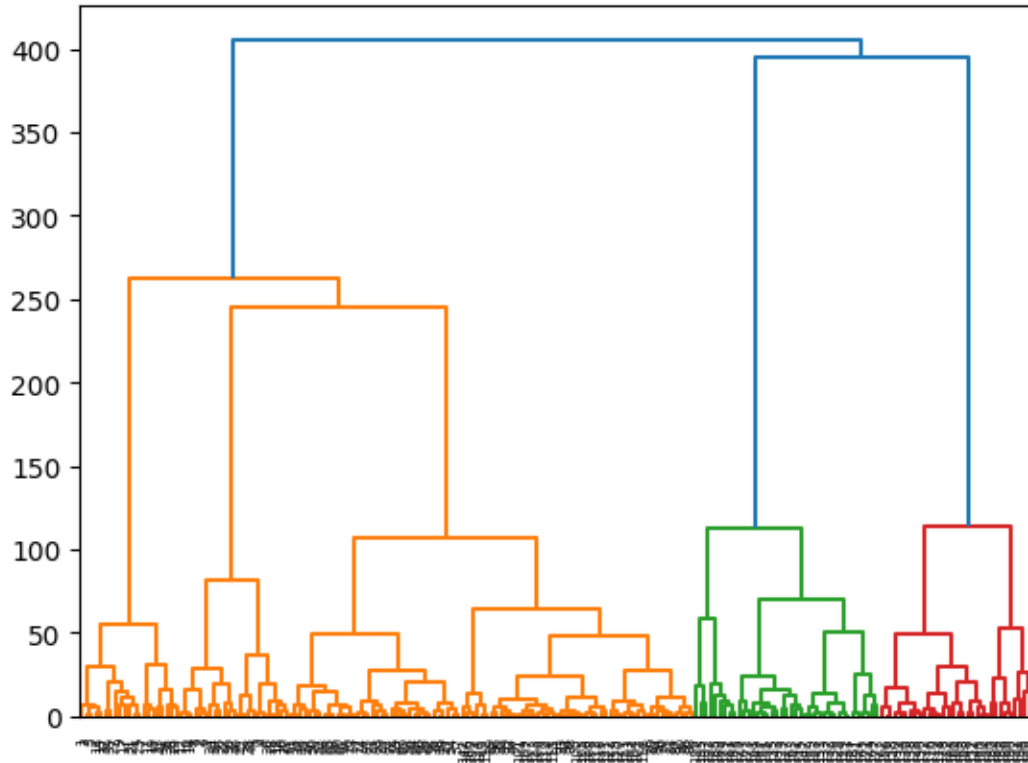
1. with the help of seipy import the libraries and datasets
2. using the dendrogram to find the optimal number of the clusters
3. create the hierarchy model and visualize the cluster with the help of matplotlib libraries
- 4.

```
[3]: #Import the numpy, pandas , matplotlib, seaborn library's
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

```
[10]: #Assign variable name "dataset" and the input variable as "X" including select
      ↪ all the row and index columns which you want [column_index, Column_index].
dataset=pd.read_csv("Mall_Customers.csv")
X=dataset.iloc[:,[3,4]].values
```

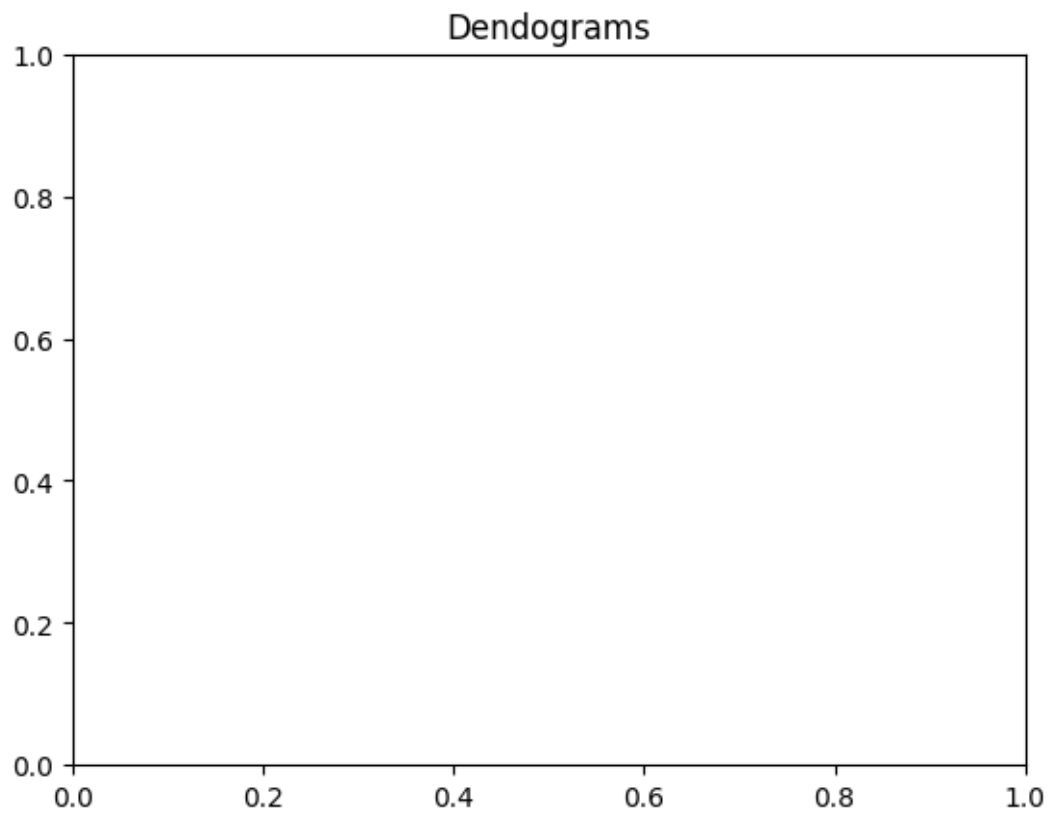
```
[11]: #import scipy cluster using attribute "scipy.cluster.hierarchy" as sch alias
import scipy.cluster.hierarchy as sch
```

```
[13]: #Using the dendrogram to find the optimal number of clusters
# Assign a variable as dendrogram and declares the "sch.dendrogram(sch.
      ↳linkage(X, method = 'ward'))"
dendrogram = sch.dendrogram(sch.linkage(X,method='ward'))
```



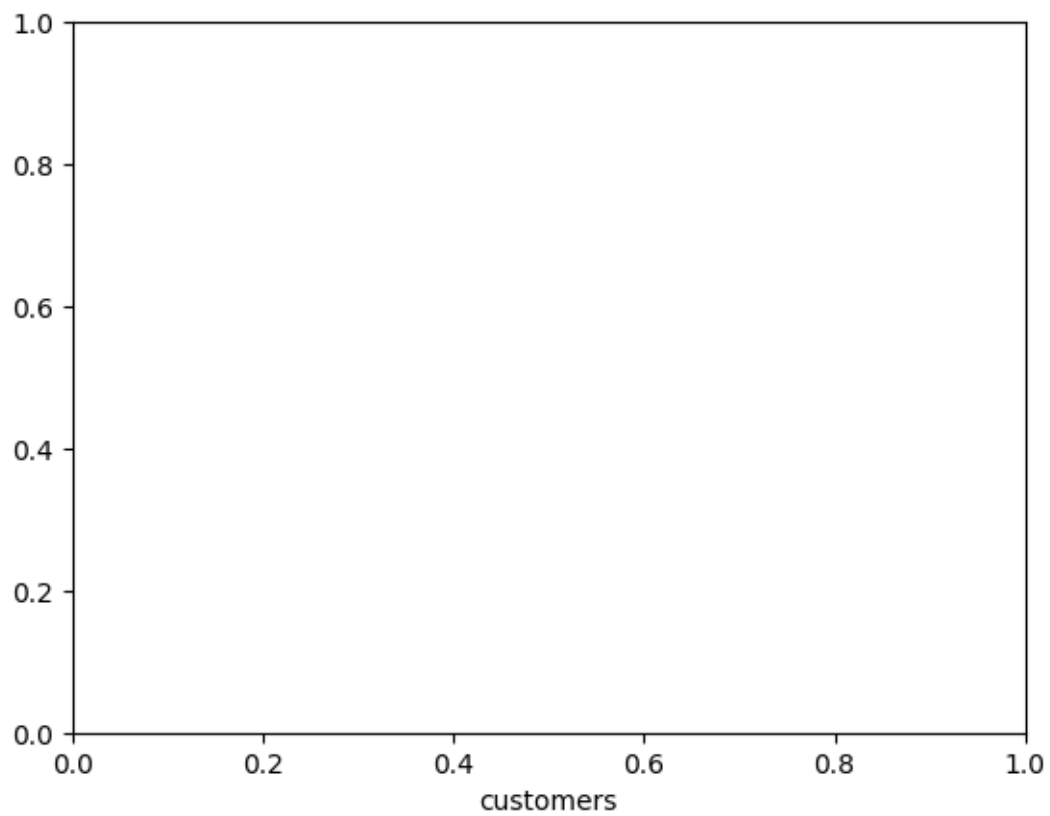
```
[14]: #Assign the title as "Dendograms"
plt.title("Dendograms")
```

```
[14]: Text(0.5, 1.0, 'Dendograms')
```



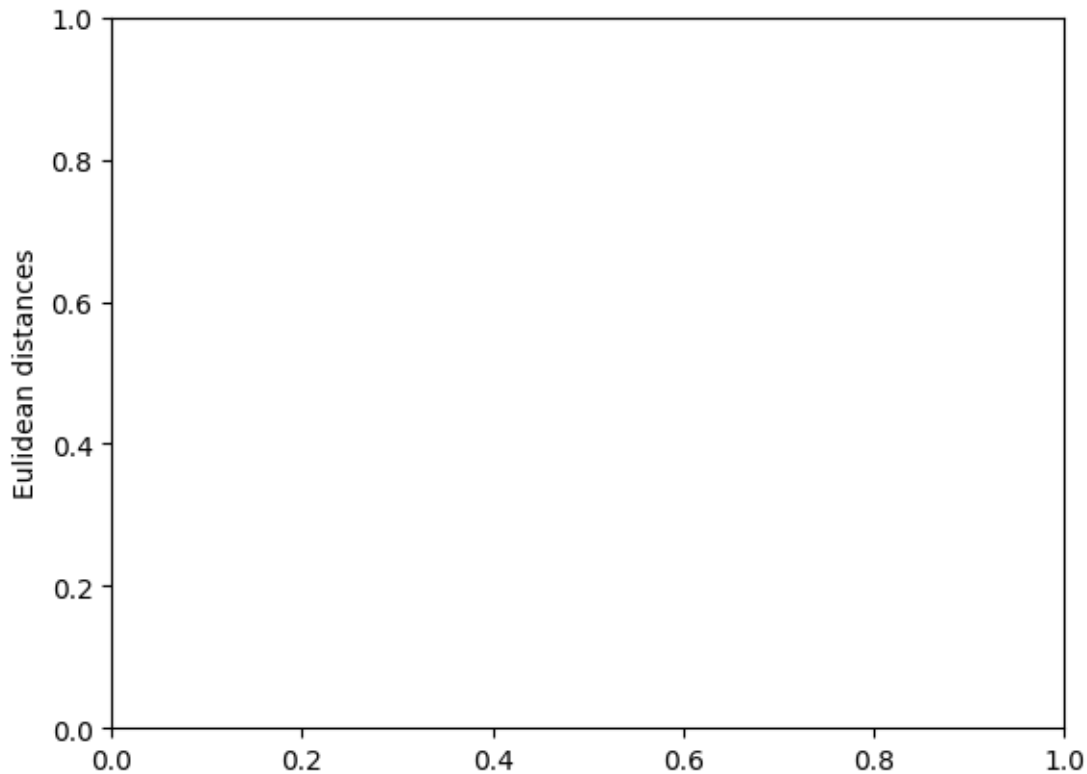
```
[15]: #Label X axis as "Customers"  
plt.xlabel("customers")
```

```
[15]: Text(0.5, 0, 'customers')
```



```
[16]: #Label Y axis as 'Euclidean distances'  
plt.ylabel("Euclidean distances")
```

```
[16]: Text(0, 0.5, 'Euclidean distances')
```



```
[18]: # from "sklearn.cluster" attribute import "AgglomerativeClustering" default
      ↪ argument.
      from sklearn.cluster import AgglomerativeClustering
```

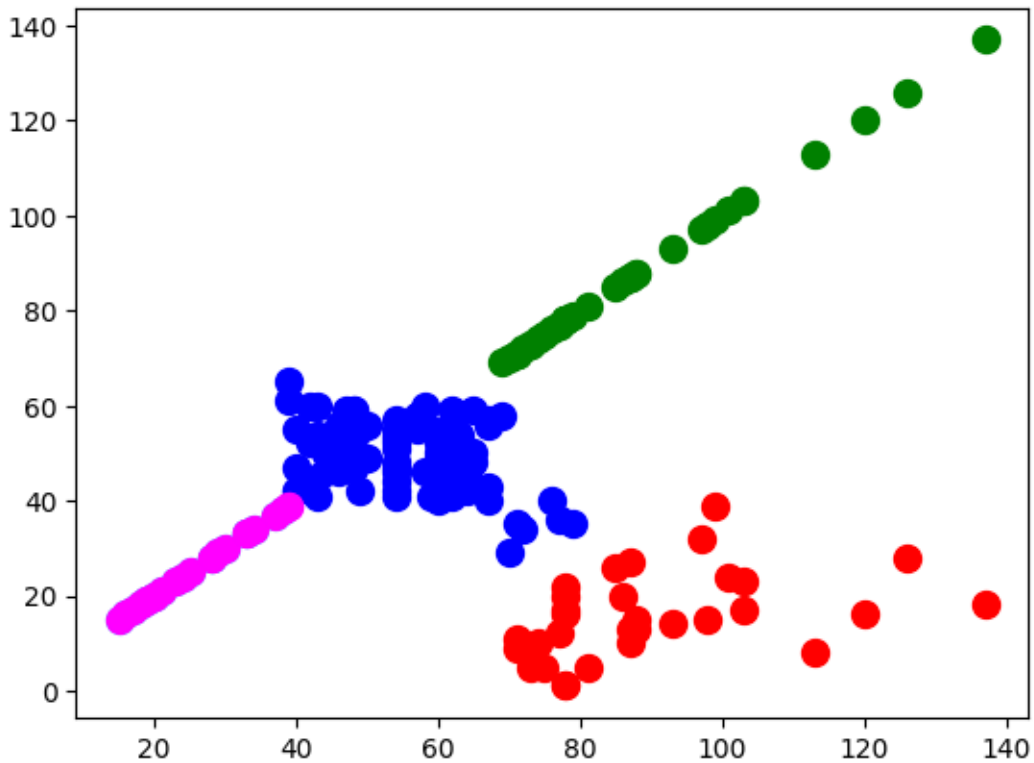
```
[19]: #Create a cluster for five or nth cluster which you want.
      hc = AgglomerativeClustering(n_clusters = 5, affinity = 'euclidean', linkage =
      ↪ 'ward')
      y_hc = hc.fit_predict(X)
```

```
/usr/local/lib/python3.10/dist-packages/sklearn/cluster/_agglomerative.py:983:
FutureWarning: Attribute `affinity` was deprecated in version 1.2 and will be
removed in 1.4. Use `metric` instead
  warnings.warn(
```

```
[23]: #Plot the scatter plot for scatter visualization.
      plt.scatter(X[y_hc == 0,0 ], X[y_hc == 0,1 ], s = 100, c = 'red', label =
      ↪ 'Cluster 1')
      plt.scatter(X[y_hc == 1,0], X[y_hc == 1,1], s = 100, c = 'blue', label =
      ↪ 'Cluster 2')
      plt.scatter(X[y_hc == 2,0], X[y_hc == 2,0 ], s = 100, c = 'green', label =
      ↪ 'Cluster 3')
```

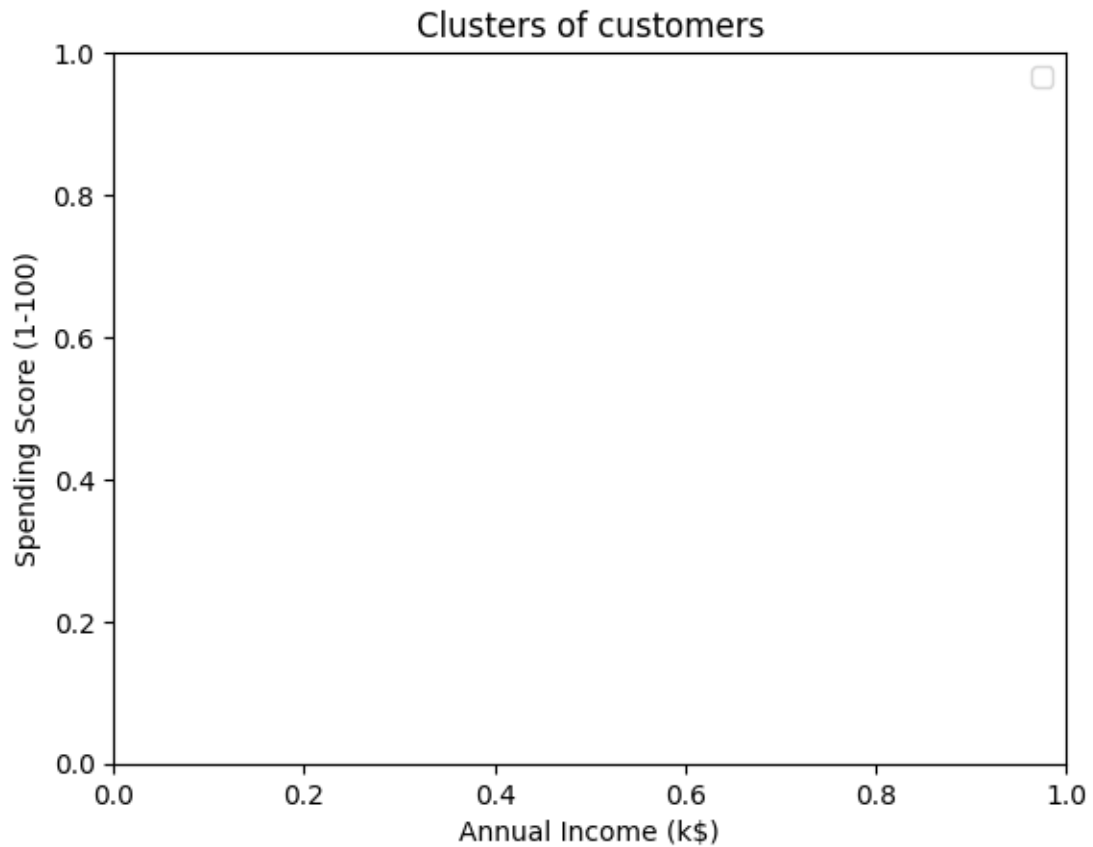
```
plt.scatter(X[y_hc == 3,0], X[y_hc ==3,0 ], s = 100, c = 'cyan', label =_
↳'Cluster 4')
plt.scatter(X[y_hc ==4,0 ], X[y_hc == 4,0], s = 100, c = 'magenta', label =_
↳'Cluster 5')
```

[23]: <matplotlib.collections.PathCollection at 0x7927c4c34b80>



```
[25]: plt.title('Clusters of customers')
plt.xlabel('Annual Income (k$)')
plt.ylabel('Spending Score (1-100)')
plt.legend()
plt.show()
```

WARNING:matplotlib.legend:No artists with labels found to put in legend. Note that artists whose label start with an underscore are ignored when legend() is called with no argument.



#### 0.4 CONCLUSION

According to the model building as a engineer my prediction is cluster number 3 heighest number linkage.

[ ]: