TRƯỜNG ĐẠI HỌC BÁCH KHOA HÀ NỘI
HANOI UNIVERSITY OF SCIENCE AND TECHNOLOGY

Introduction to Data Science

Capstone Project Name: FIFA 2021 Clustering
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Group number: 14

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Introduction

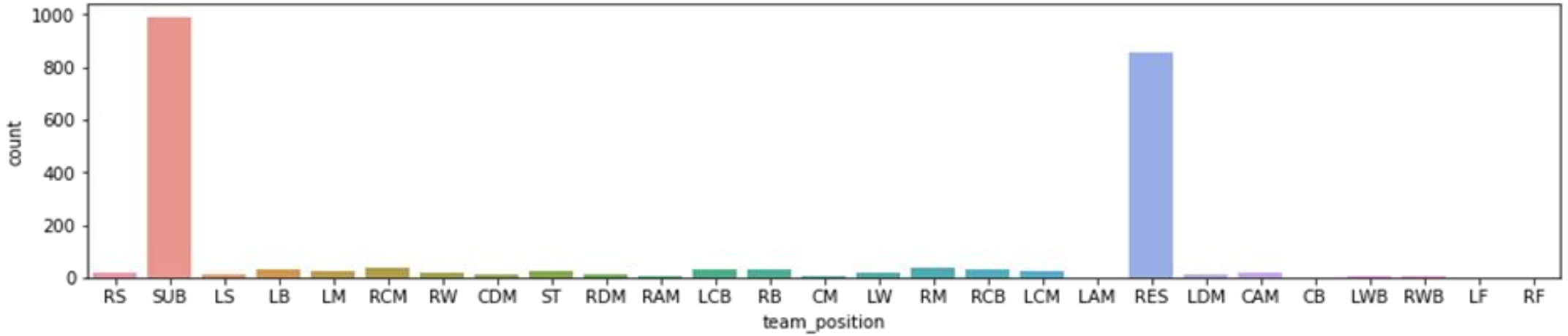
Our subject for the Capstone is applying Machine Learning to cluster football players. In detail, our project aims to look at the class of young players and cluster them into groups based on their attributes. We will filter and consider the players who are promised based on their potential attributes.

DATASET

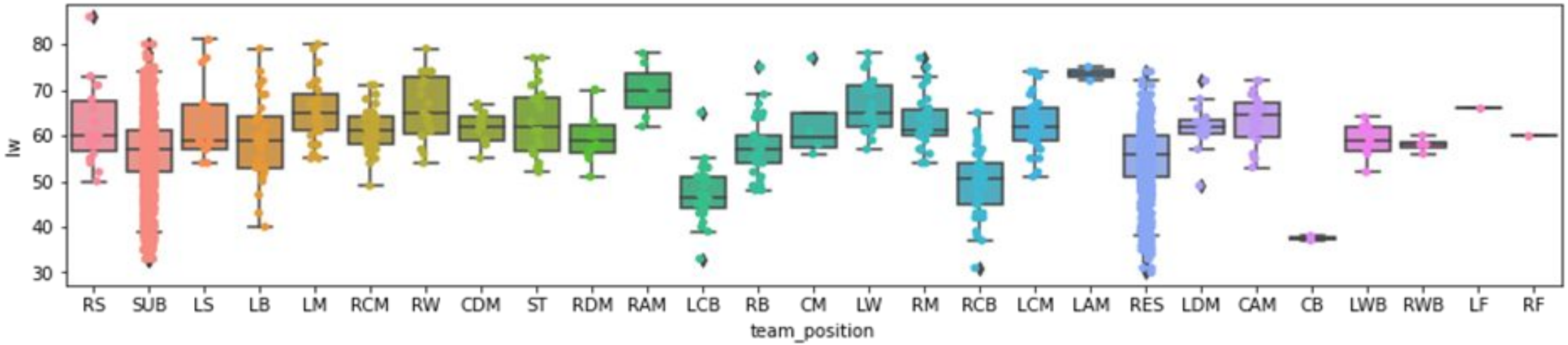
- Name
- Age
- Height
- Overall: Rated between 1-99
- Potential: Rated between 1-99
- Work Rate: Effort the player puts in the game
- Team Position
- Playing Skills: Rated between 1-99
- Positional Skills: Rating of a player when play at a specified position rated between 1-99

Exploring Data Analysis

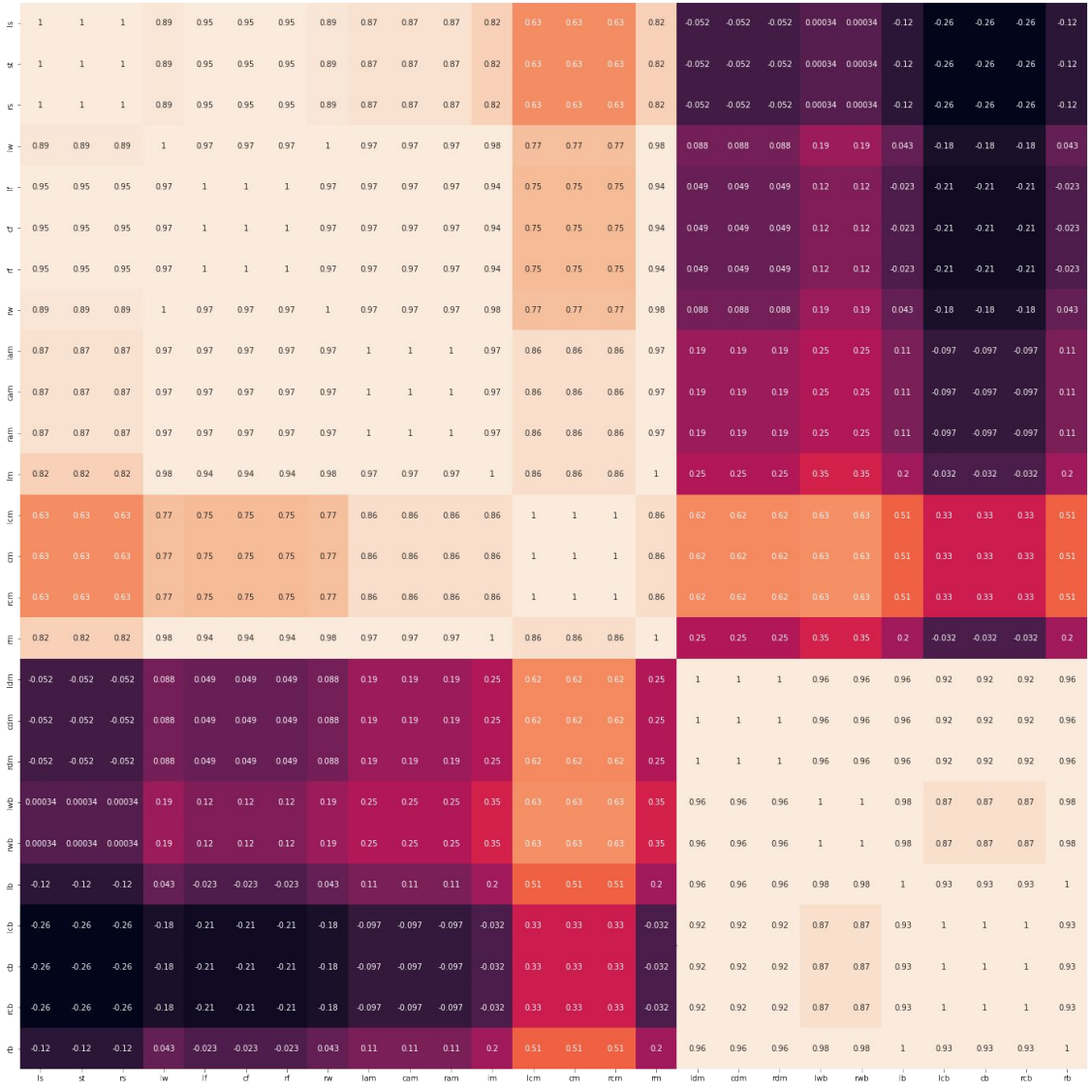
Plot the number of players in each position:

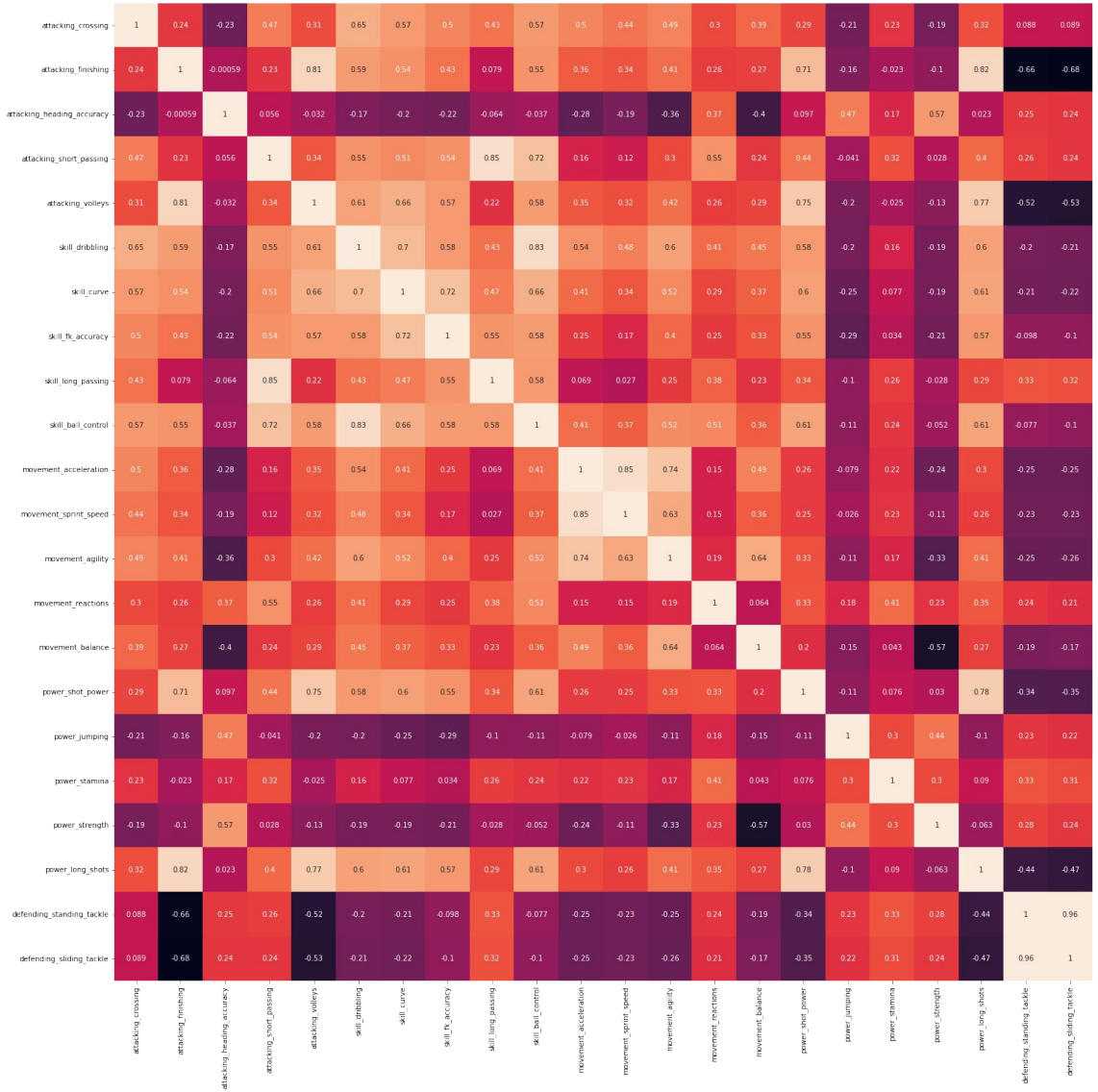


Plot the relationship between team positions and positional skills.



Correlation between the position and playing skills features





Data Modeling

Step 1: Drop unused attributes for clustering

Step 2: Reduce Dimension

Step 3: Standardized data.

```
Explained variance ratio for Striker PCA application:  [1.]
Explained variance ratio for Forward PCA application:  [1.]
Explained variance ratio for Attack Winger PCA application:  [1.]
Explained variance ratio for Midfielder PCA application:  [0.70388172]
Explained variance ratio for Defensive Winger PCA application:  [1.]
Explained variance ratio for Back PCA application:  [1.]
```

```
Explained variance ratio for defensive skills PCA application:  [0.98299723]
Explained variance ratio for movement PCA application:  [0.82825812]
Explained variance ratio for control skills PCA application:  [0.91608148]
Explained variance ratio for passing skills PCA application:  [0.92329027]
Explained variance ratio for defending skills PCA application:  [0.91549397]
```

Principal Component Analysis (PCA)

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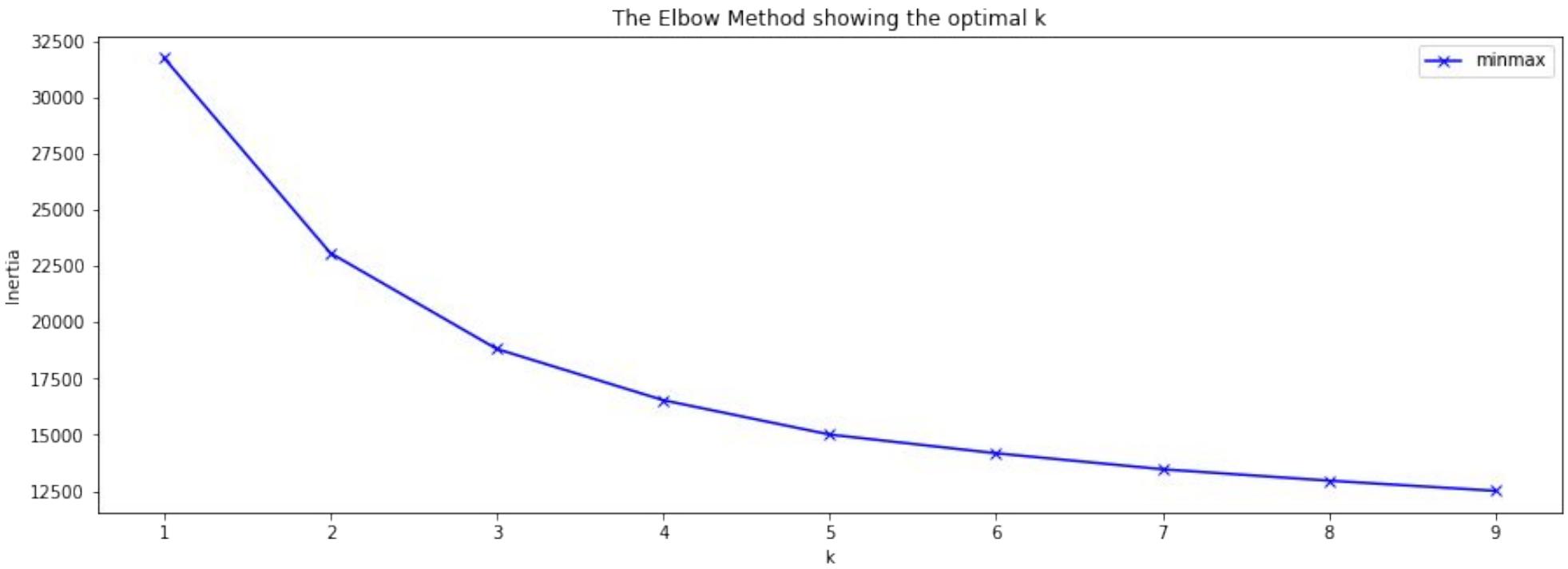
Explained variance ratio for defending skills PCA application:  [0.91549397]
```

Discover the relationship between Work Rate and other skills

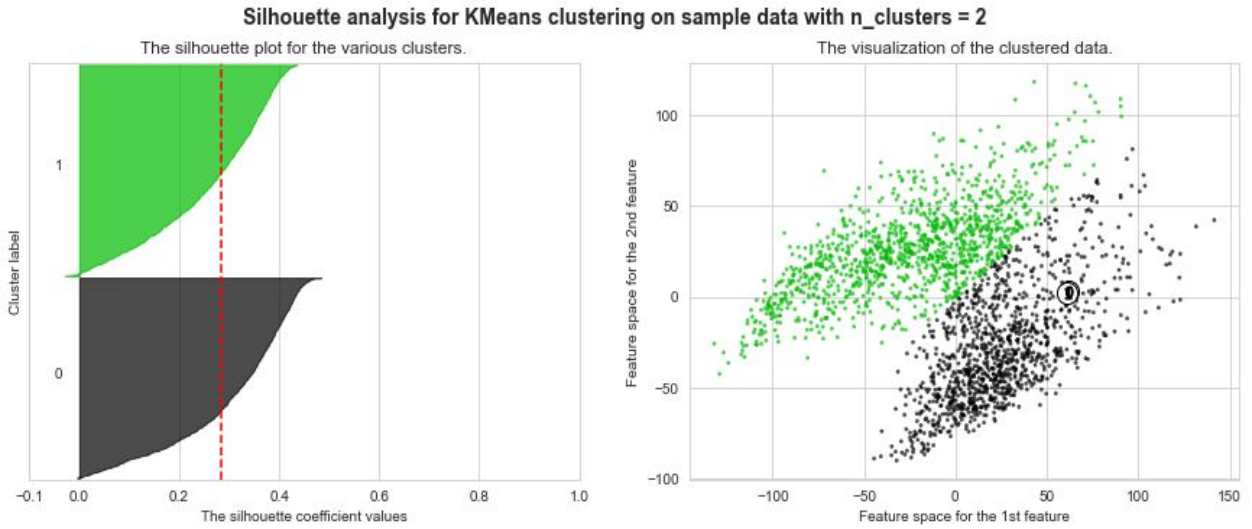
attack_work_rate	attacking_crossing	attacking_finishing	attacking_short_passing	attacking_volleys
0.0	37	26	56	29
0.5	49	48	59	42
1.0	55	56	60	47

defense_work_rate	defending_standing_tackle	defending_sliding_tackle
0.0	28	26
0.5	52	49
1.0	62	59

Method to find optimal K

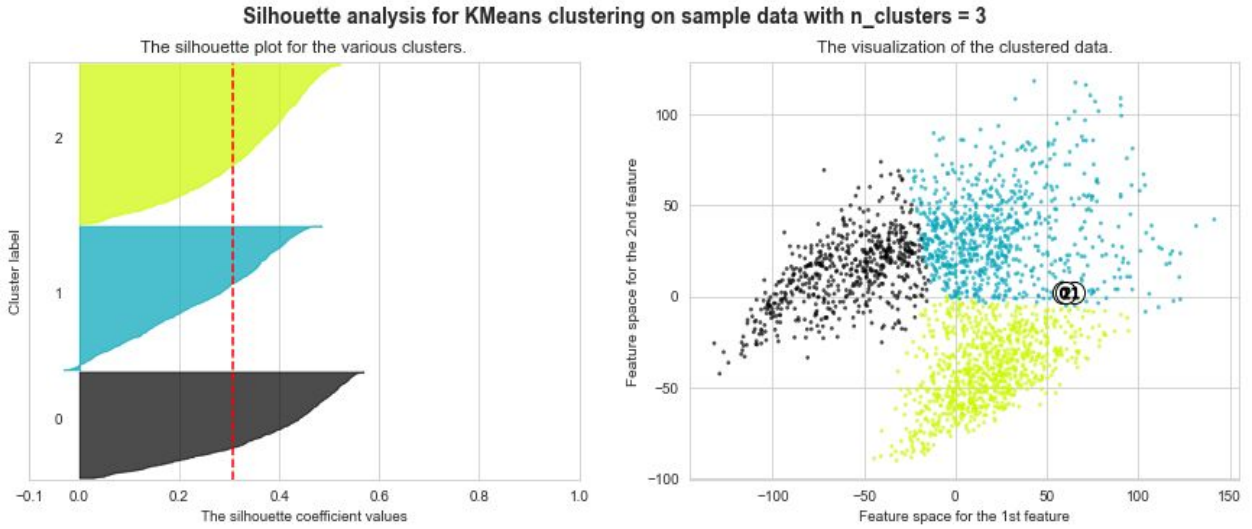


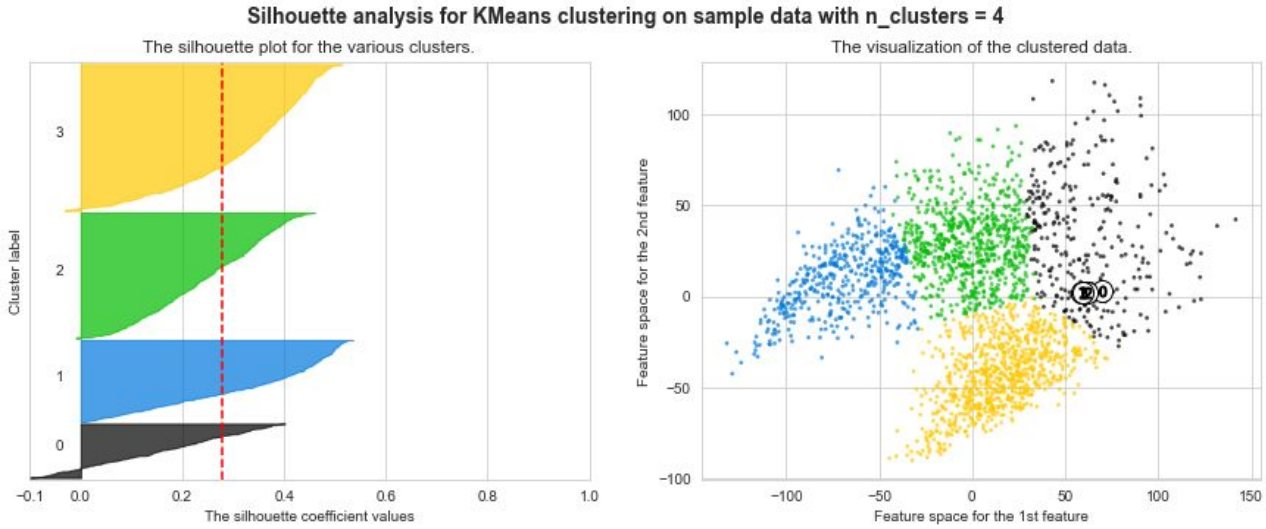
Silhouette method



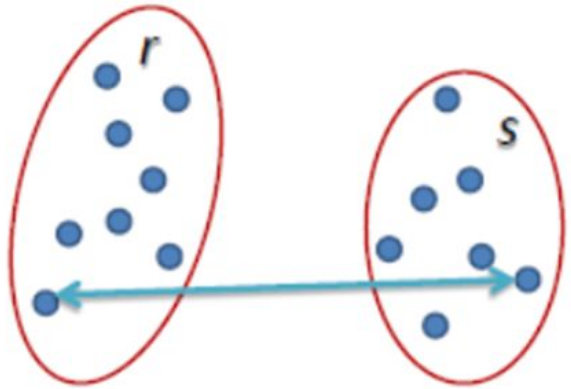

```
For n_clusters = 2 The average silhouette_score is : 0.28272602345704606
For n_clusters = 3 The average silhouette_score is : 0.3084938723509853
For n_clusters = 4 The average silhouette_score is : 0.2795438505249029
For n_clusters = 5 The average silhouette_score is : 0.22999870816740864
For n_clusters = 6 The average silhouette_score is : 0.21964100225379676
For n_clusters = 7 The average silhouette_score is : 0.20744387774940753
For n_clusters = 8 The average silhouette_score is : 0.19222045902464063
For n_clusters = 9 The average silhouette_score is : 0.1860086216190746
```

r of b(i) and a(i)

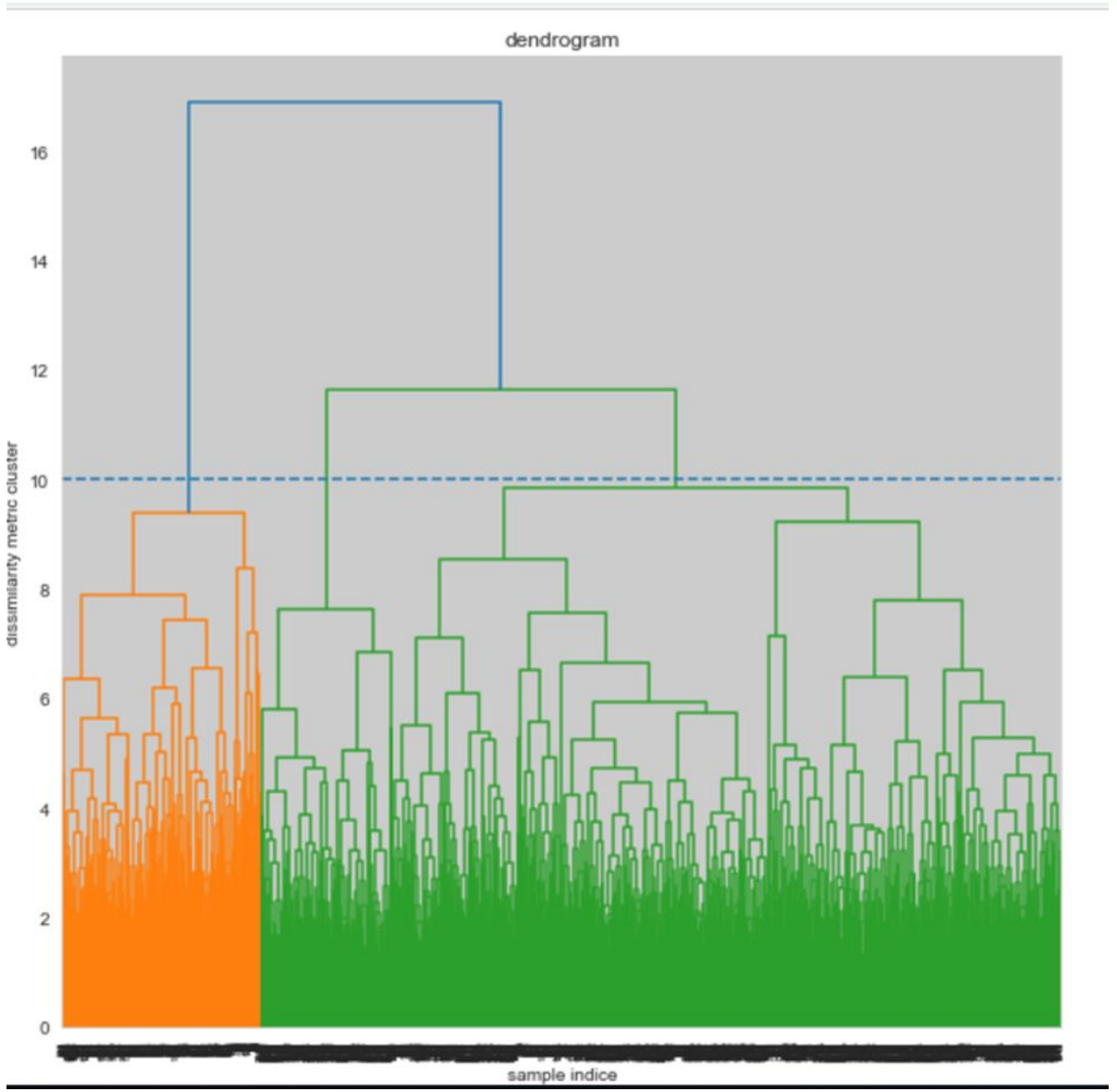




Hierarchical Clustering



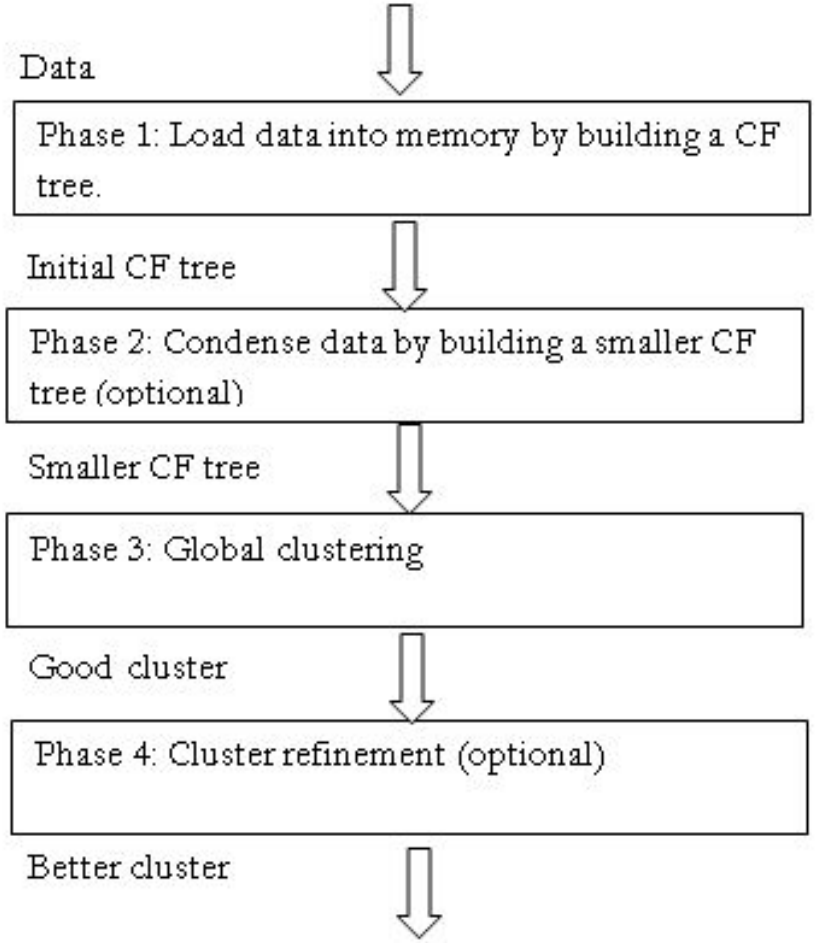
$$L(r,s) = \max(D(x_{ri}, x_{sj}))$$



Birch Algorithm

Birch is a Hierarchical clustering algorithm using top-down

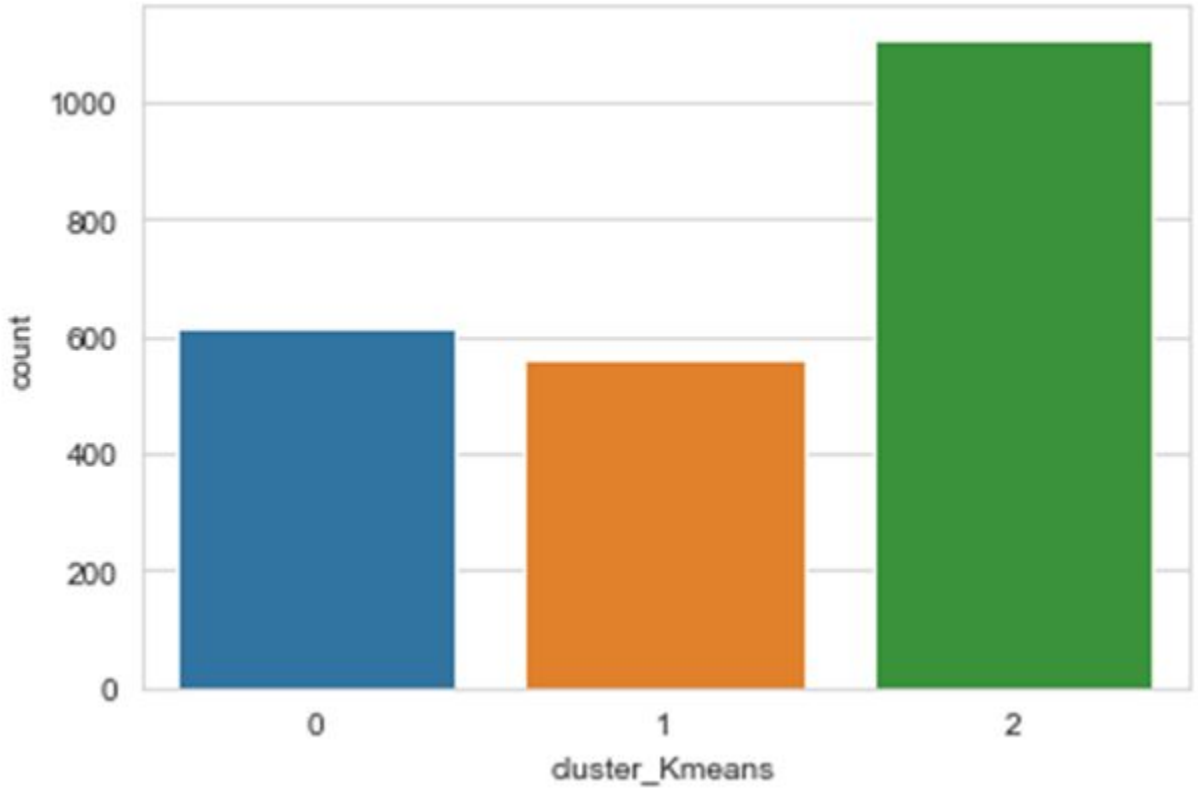
- Idea: Each successful iteration, a cluster is split into smaller clusters according to the value of some similarity measure until each object is a cluster or until the stopping condition is satisfied.
- This approach into using divide and conquer strategy in the clustering process



Comparision three Algorithms

	K cluster	kmean_silhouette_scores	birch_silhouette_scores	agg_silhouette_scores
0	2	0.282726	0.238964	0.238964
1	3	0.308363	0.236685	0.236685
2	4	0.279544	0.212797	0.212797
3	5	0.229999	0.193872	0.193872
4	6	0.219155	0.185722	0.185722
5	7	0.206945	0.154607	0.154607
6	8	0.191965	0.155710	0.155710
7	9	0.185837	0.155961	0.155961

Result



cluster_Kmeans	potential	overall
2	79.738211	67.549593
1	74.443060	59.176157
0	74.352570	58.725879

```
2    1109
0     615
1     562
Name: cluster_Kmeans, dtype: int64
running time of Kmeans  70.96637034416199
```

Result

cluster_Kmeans	striker	forward	attack_winger	midfielder	defensive_winger	back

cluster_Kmeans	defensive	movement	control	passing	finishing

cluster_Kmeans	attack_work_rate	defense_work_rate	defensive
0	0.494662	0.588968	21.025102
1	0.638864	0.458070	-13.352598
2	0.704878	0.518699	4.864916

3
3
7
23.755704
5.300393
12.150520

THANK YOU FOR LISTENING