<u>עבור בדיקת הפרמטרים הכי טובים עבור K-Means</u>

*בדיקת הפרמטרים מסומנת בצהוב

```
from sklearn.cluster import KMeans
from sklearn.metrics import silhouette score
best kmeans score = -1
best kmeans params = {}
kmeans_results = []
for n clusters in range(2, 11):
 for init method in ['k-means++', 'random']:
   kmeans = KMeans(n clusters=n clusters, init=init method, random state=42)
   score = silhouette_score(X, kmeans.labels_)
   kmeans_results.append((n_clusters, init_method, score))
   if score > best_kmeans_score:
     best_kmeans_score = score
print('All KMeans Results:')
for result in kmeans results:
 print(f'n_clusters: {result[0]}, init: {result[1]}, silhouette score: {result[2]:}')
                                 n_clusters: מספר הקבוצות (נבדקו ערכים בין 2 ל-10).
                            init: שיטת האתחול (נבדקו ערכים '++k-means').
                                                                   -התוצאות
All KMeans Results:
n clusters: 2, init: k-means++, silhouette score:
0.5672585793914866
n clusters: 2, init: random, silhouette score: 0.5672585793914866
n clusters: 3, init: k-means++, silhouette score:
0.5857904307157648
n clusters: 3, init: random, silhouette score: 0.5857904307157648
n clusters: 4, init: k-means++, silhouette score:
0.6813192037113843
n clusters: 4, init: random, silhouette score: 0.6813192037113843
n clusters: 5, init: k-means++, silhouette score:
0.6432758294253328
n clusters: 5, init: random, silhouette score: 0.6976139874193659
n clusters: 6, init: k-means++, silhouette score:
0.6605985626907828
n clusters: 6, init: random, silhouette score: 0.7251181630059241
n clusters: 7, init: k-means++, silhouette score:
0.7026041646015059
n_clusters: 7, init: random, silhouette score: 0.640026695635202
n clusters: 8, init: k-means++, silhouette score:
0.716259254869025
n clusters: 8, init: random, silhouette score: 0.6178145880678158
n clusters: 9, init: k-means++, silhouette score:
0.7117502750033986
n clusters: 9, init: random, silhouette score: 0.6267846061018335
```

```
n_clusters: 10, init: k-means++, silhouette score:
0.7105122316647566
n clusters: 10, init: random, silhouette score: 0.559886632470254
```

–Agglomerative Clustering עבור בדיקת הפרמטרים הכי טובים עבור

*בדיקת הפרמטרים מסומנת בצהוב

```
from sklearn.cluster import AgglomerativeClustering
best agg score = -1
best_agg_params = {}
agg_results = []
for n_clusters in range(2, 11):
  for linkage in ['ward', 'complete', 'average', 'single']:
    agg = AgglomerativeClustering(n_clusters=n_clusters, linkage=linkage)
    agg.fit(X)
    score = silhouette_score(X, agg.labels_)
    agg_results.append((n_clusters, linkage, score))
    if score > best_agg_score:
      best_agg_score = score
print('All Agglomerative Clustering Results:')
for result in agg_results:
  print(f'n_clusters: {result[0]}, linkage: {result[1]}, silhouette score: {result[2]:}')
                                          n_clusters: מספר הקבוצות (נבדקו ערכים בין 2 ל-10).
```

linkage: שיטת הקישור (נבדקו ערכים 'ward', 'complete', 'average', 'single').

-התוצאות

```
All Agglomerative Clustering Results:
n clusters: 2, linkage: ward, silhouette score:
0.5698462725885549
n clusters: 2, linkage: complete, silhouette score:
0.5090246272851084
n clusters: 2, linkage: average, silhouette score:
0.5698462725885549
n clusters: 2, linkage: single, silhouette score:
0.3587824456286678
n clusters: 3, linkage: ward, silhouette score:
0.5806226231752926
n_clusters: 3, linkage: complete, silhouette score:
0.47295414538916447
n clusters: 3, linkage: average, silhouette score:
0.5806226231752926
n clusters: 3, linkage: single, silhouette score:
0.5866220981494794
n clusters: 4, linkage: ward, silhouette score:
0.6812638781424561
```

```
n clusters: 4, linkage: complete, silhouette score:
0.6812638781424561
n clusters: 4, linkage: average, silhouette score:
0.6812638781424561
n clusters: 4, linkage: single, silhouette score:
0.5180398014280129
n clusters: 5, linkage: ward, silhouette score: 0.69778623369389
n clusters: 5, linkage: complete, silhouette score:
0.6923413939127293
n clusters: 5, linkage: average, silhouette score:
0.6923413939127293
n clusters: 5, linkage: single, silhouette score:
0.6923413939127293
n clusters: 6, linkage: ward, silhouette score:
0.7256900714288006
n clusters: 6, linkage: complete, silhouette score:
0.6594689980695821
n clusters: 6, linkage: average, silhouette score:
0.6680458764033119
n clusters: 6, linkage: single, silhouette score:
0.613327782899459
n clusters: 7, linkage: ward, silhouette score:
0.7014960182059224
n clusters: 7, linkage: complete, silhouette score:
0.668466237784836
n clusters: 7, linkage: average, silhouette score:
0.6765904629884553
n clusters: 7, linkage: single, silhouette score:
0.604522825115808
n clusters: 8, linkage: ward, silhouette score:
0.7085758473975964
n clusters: 8, linkage: complete, silhouette score:
0.7018771832746021
n clusters: 8, linkage: average, silhouette score:
0.7099391405045268
n clusters: 8, linkage: single, silhouette score:
0.44573935523939057
n clusters: 9, linkage: ward, silhouette score:
0.7088190473478537
n clusters: 9, linkage: complete, silhouette score:
0.7046018505552566
n clusters: 9, linkage: average, silhouette score:
0.7093447499958512
n clusters: 9, linkage: single, silhouette score:
0.35175198194067037
n clusters: 10, linkage: ward, silhouette score:
0.7088165601086022
n clusters: 10, linkage: complete, silhouette score:
0.7078385268925678
n clusters: 10, linkage: average, silhouette score:
0.7073095644071425
n clusters: 10, linkage: single, silhouette score:
0.25184685533953083
```

```
from sklearn.cluster import DBSCAN
best dbscan score = -1
best dbscan params = {}
dbscan_results = []
eps_values = [0.5, 1.0, 1.5, 2.0, 2.5]
min samples values = [5, 10, 15, 20]
for eps in eps_values:
  for min_samples in min_samples_values:
    dbscan = DBSCAN(eps=eps, min samples=min samples)
    if len(set(dbscan.labels_)) > 1: # Ensure there is more than one cluster
      score = silhouette score(X, dbscan.labels )
      dbscan_results.append((eps, min_samples, score))
      if score > best_dbscan_score:
         best_dbscan_score = score
print('All DBSCAN Results:')
for result in dbscan_results:
  print(f'eps: {result[0]}, min_samples: {result[1]}, silhouette score: {result[2]:}')
```

eps: רדיוס השכנות (נבדקו ערכים 0.5, 1.0, 1.5, 2.0, 2.5).

min_samples: מספר הדוגמאות המינימלי בתוך קבוצת הליבה (נבדקו ערכים 5, 10, 15, 20).

-התוצאות

```
All DBSCAN Results:
eps: 0.5, min samples: 5, silhouette score: 0.28865209433880534
eps: 0.5, min samples: 10, silhouette score: 0.44913770731330993
eps: 0.5, min samples: 15, silhouette score: 0.30549274312867614
eps: 0.5, min samples: 20, silhouette score: 0.17396195043369547
eps: 1.0, min samples: 5, silhouette score: 0.7070123298589694
eps: 1.0, min samples: 10, silhouette score: 0.6190125761523078
eps: 1.0, min samples: 15, silhouette score: 0.5102933416053745
eps: 1.0, min samples: 20, silhouette score: 0.6612622555185036
eps: 1.5, min samples: 5, silhouette score: 0.6894016420828084
eps: 1.5, min samples: 10, silhouette score: 0.7222593187874933
eps: 1.5, min samples: 15, silhouette score: 0.7135727033307426
eps: 1.5, min samples: 20, silhouette score: 0.6884017793929554
eps: 2.0, min samples: 5, silhouette score: 0.613327782899459
eps: 2.0, min samples: 10, silhouette score: 0.613327782899459
eps: 2.0, min samples: 15, silhouette score: 0.6809842207458477
eps: 2.0, min_samples: 20, silhouette score: 0.6809842207458477
eps: 2.5, min_samples: 5, silhouette score: 0.6923413939127293
eps: 2.5, min_samples: 10, silhouette score: 0.6923413939127293
eps: 2.5, min samples: 15, silhouette score: 0.6923413939127293
```