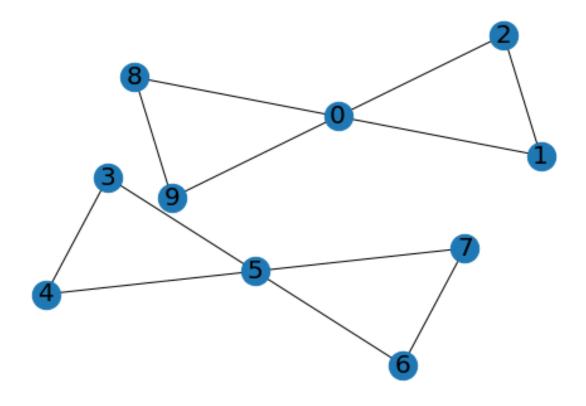
spectral_clustering-2

May 21, 2022

```
[]: import numpy as np
     from PIL import Image
     import matplotlib.pyplot as plt
     import networkx as nx
     from sklearn.cluster import KMeans
[]: def BuildGraph (n, adj):
         g = nx.Graph()
         for i in range(n):
             for j in range(n):
                 if (adj[i][j] != 0) :
                     g.add_edge(i, j)
         return g
[]: adj = np.array([
                     [0, 1, 1, 0, 0, 0, 0, 0, 1, 1],
                     [1, 0, 1, 0, 0, 0, 0, 0, 0, 0],
                     [1, 1, 0, 0, 0, 0, 0, 0, 0, 0],
                     [0, 0, 0, 0, 1, 1, 0, 0, 0, 0],
                     [0, 0, 0, 1, 0, 1, 0, 0, 0, 0],
                     [0, 0, 0, 1, 1, 0, 1, 1, 0, 0],
                     [0, 0, 0, 0, 0, 1, 0, 1, 0, 0],
                     [0, 0, 0, 0, 0, 1, 1, 0, 0, 0],
                     [1, 0, 0, 0, 0, 0, 0, 0, 0, 1],
                     [1, 0, 0, 0, 0, 0, 0, 0, 1, 0]
             ])
     n = len(adj[0])
     graph = BuildGraph(n, adj)
     nx.draw_kamada_kawai(graph, font_size = 20, width = 1, with_labels=True,_
      ⊖node size = 500, arrowsize = 100)
     plt.show()
```

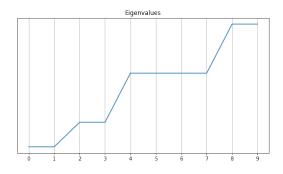


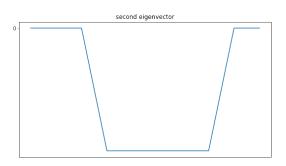
```
[]: deg = np.diag(adj.sum(axis=1))
     laplacian = deg - adj
     eigenvalues, eigenvectors = np.linalg.eig(laplacian)
     print("Laplacian: ")
     print (laplacian)
     fig, axs = plt.subplots(1, 2, figsize=(20, 5), subplot_kw=dict(xticks=[],__

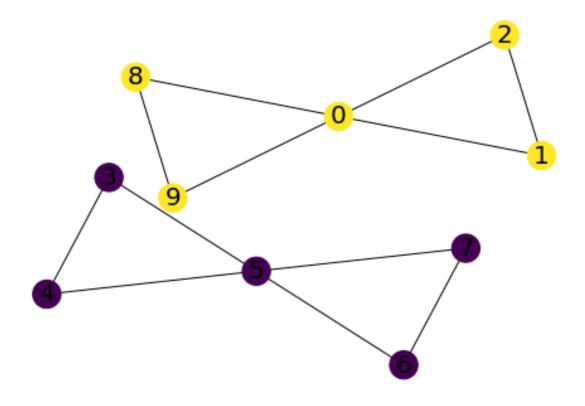
yticks=[]))
     eigenvectors = eigenvectors[:, np.argsort(eigenvalues)]
     eigenvalues = eigenvalues[np.argsort(eigenvalues)]
     axs[0].plot(eigenvalues)
     axs[0].set_xticks(range(len(eigenvalues)))
     axs[0].grid()
     axs[0].set_title("Eigenvalues")
     axs[1].plot(eigenvectors[:,1])
     axs[1].set_yticks([0])
     axs[1].set_title("second eigenvector")
     plt.show()
     print ("Fiedler value(second eigenvelue): ", eigenvalues[1])
```

Laplacian:

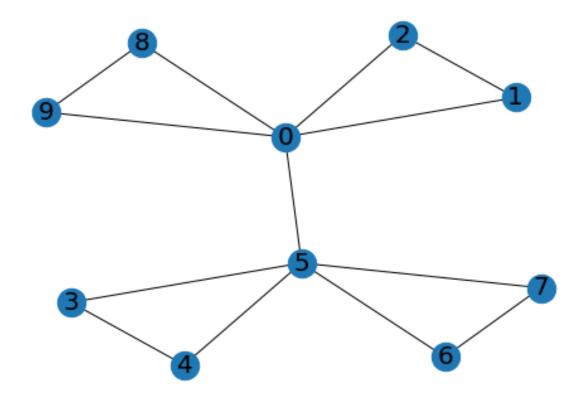
```
[[ 4 -1 -1 0 0 0 0 0 0 -1 -1]
[-1 2 -1 0 0 0 0 0 0 0 0 0]
[-1 -1 2 0 0 0 0 0 0 0 0]
[0 0 0 2 -1 -1 0 0 0 0 0]
[0 0 0 -1 2 -1 0 0 0 0]
[0 0 0 -1 -1 4 -1 -1 0 0]
[0 0 0 0 0 0 -1 2 -1 0 0]
[0 0 0 0 0 0 -1 -1 2 0 0]
[-1 0 0 0 0 0 0 0 0 0 2 -1]
[-1 0 0 0 0 0 0 0 0 0 0 1 2]
```







```
[]: adj = np.array([
                     [0, 1, 1, 0, 0, 1, 0, 0, 1, 1],
                     [1, 0, 1, 0, 0, 0, 0, 0, 0, 0],
                     [1, 1, 0, 0, 0, 0, 0, 0, 0, 0],
                     [0, 0, 0, 0, 1, 1, 0, 0, 0, 0],
                     [0, 0, 0, 1, 0, 1, 0, 0, 0, 0],
                     [1, 0, 0, 1, 1, 0, 1, 1, 0, 0],
                     [0, 0, 0, 0, 0, 1, 0, 1, 0, 0],
                     [0, 0, 0, 0, 0, 1, 1, 0, 0, 0],
                     [1, 0, 0, 0, 0, 0, 0, 0, 0, 1],
                     [1, 0, 0, 0, 0, 0, 0, 0, 1, 0]
             ])
     n = len(adj[0])
     graph = BuildGraph(n, adj)
     nx.draw_kamada_kawai(graph, font_size = 20, width = 1, with_labels=True,_
      →node_size = 500, arrowsize = 100)
     plt.show()
```

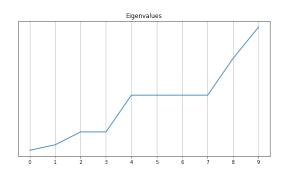


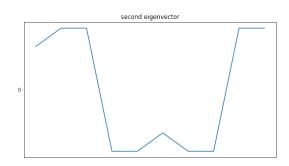
```
[]: deg = np.diag(adj.sum(axis=1))
     laplacian = deg - adj
     eigenvalues, eigenvectors = np.linalg.eig(laplacian)
     fig, axs = plt.subplots(1, 2, figsize=(20, 5), subplot_kw=dict(xticks=[],_u

yticks=[]))
     eigenvectors = eigenvectors[:, np.argsort(eigenvalues)]
     eigenvalues = eigenvalues[np.argsort(eigenvalues)]
     axs[0].plot(eigenvalues)
     axs[0].set_xticks(range(len(eigenvalues)))
     axs[0].grid()
     axs[0].set_title("Eigenvalues")
     axs[1].plot(eigenvectors[:,1])
     axs[1].set_yticks([0])
     axs[1].set_title("second eigenvector")
     plt.show()
     print ("Fiedler value(second eigenvelue): ", eigenvalues[1])
     kmeans = KMeans(n_clusters=2)
     kmeans.fit(eigenvectors[:,1:2])
```

```
colors = kmeans.labels_
print ("Fiedler vector(second eigenvector): ", np.round(eigenvectors[:,1], 3))
print ("2-Cluster centers: \n", kmeans.cluster_centers_)
print ("2-Cluster labels: ", colors)
nx.draw_kamada_kawai(graph, font_size = 20, width = 1, with_labels=True,_
 anode_size = 500, arrowsize = 100, node_color=colors, nodelist=range(0,n))
plt.title("2-clustering")
plt.show()
fig, axs = plt.subplots(1, 2, figsize=(20, 5), subplot_kw=dict(xticks=[],__

yticks=[]))
eigenvectors = eigenvectors[:, np.argsort(eigenvalues)]
eigenvalues = eigenvalues[np.argsort(eigenvalues)]
axs[0].plot(eigenvalues)
axs[0].set_xticks(range(len(eigenvalues)))
axs[0].grid()
axs[0].set_title("Eigenvalues")
axs[1].plot(eigenvectors[:,1])
axs[1].plot(eigenvectors[:,2])
axs[1].plot(eigenvectors[:,3])
axs[1].set_title("second eigenvector")
axs[1].set_yticks([0])
plt.show()
print ("First* 3 eigenvectors:")
print (np.round(eigenvectors[:,1], 3))
print (np.round(eigenvectors[:,2], 3))
print (np.round(eigenvectors[:,3], 3))
kmeans = KMeans(n clusters=4)
kmeans.fit(eigenvectors[:,1:4])
colors = kmeans.labels
print ("Cluster centers: \n", kmeans.cluster_centers_)
print ("Cluster labels: ", colors)
nx.draw_kamada_kawai(graph, font_size = 20, width = 1, with_labels=True,_
 →node_size = 500, arrowsize = 100, node_color=colors, nodelist=range(0,n))
plt.title("Optimal clustering")
plt.show()
```





Fiedler value(second eigenvelue): 0.29843788128357546

Fiedler vector(second eigenvector): [0.234 0.334 0.334 -0.334 -0.334 -0.234

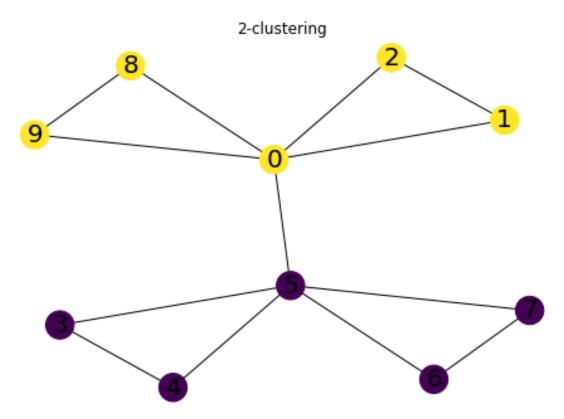
-0.334 -0.334 0.334 0.334]

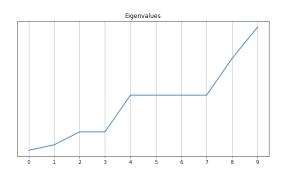
2-Cluster centers:

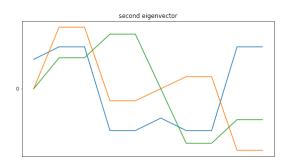
[[-0.31370984]

[0.31370984]]

2-Cluster labels: [1 1 1 0 0 0 0 0 1 1]







First* 3 eigenvectors:

 $[\ 0.234 \ \ 0.334 \ \ 0.334 \ \ -0.334 \ \ -0.334 \ \ -0.334 \ \ \ 0.334 \ \ 0.334]$

[-0. 0.491 0.491 -0.096 -0.096 0. 0.096 0.096 -0.491 -0.491]

 $\begin{bmatrix} -0. & 0.247 & 0.247 & 0.435 & 0.435 & -0. & -0.435 & -0.435 & -0.247 & -0.247 \end{bmatrix}$

Cluster centers:

[[-0.33362299 -0.09621369 0.43490268]

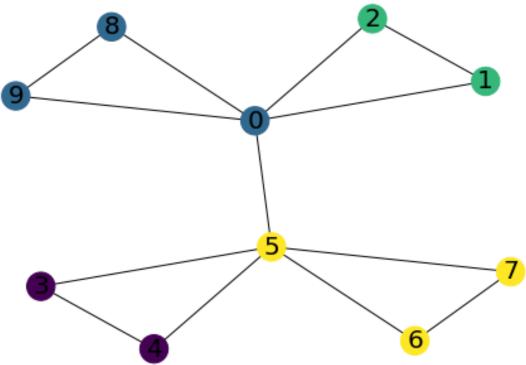
[0.30043441 -0.32710374 -0.16446501]

[0.33362299 0.49065561 0.24669751]

[-0.30043441 0.06414246 -0.28993512]]

Cluster labels: [1 2 2 0 0 3 3 3 1 1]

Optimal clustering

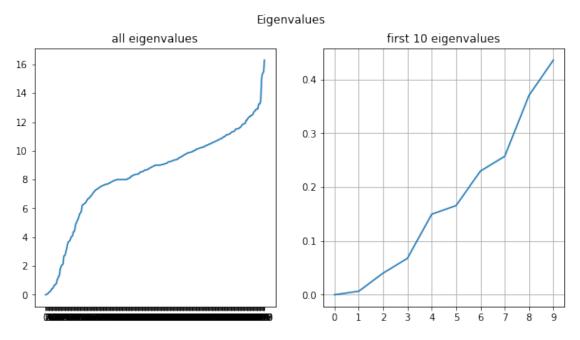


```
[]: def spectral_cluster(n, adj, k= 2):
             graph = BuildGraph(n, adj)
             deg = np.diag(adj.sum(axis=1))
             laplacian = deg - adj
             eigenvalues, eigenvectors = np.linalg.eigh(laplacian)
             eigenvectors = eigenvectors[:, np.argsort(eigenvalues)]
             eigenvalues = eigenvalues[np.argsort(eigenvalues)]
             fig, axs = plt.subplots(1, 2, figsize=(10, 5))
             axs[0].plot(eigenvalues)
             axs[0].set_xticks(range(len(eigenvalues)))
             axs[0].set_title("all eigenvalues")
             axs[1].plot(eigenvalues[:10])
             axs[1].set_xticks(range(10))
             axs[1].grid()
             axs[1].set_title("first 10 eigenvalues")
             fig.suptitle("Eigenvalues")
             plt.show()
             kmeans = KMeans(n_clusters=k)
             kmeans.fit(eigenvectors[:,1:k])
             colors = kmeans.labels_
             return colors
```

```
fig, axs = plt.subplots(1, 3, figsize=(27, 9), subplot_kw=dict(xticks=[],_u

yticks=[]))
nx.draw(graph, font_size = 20, width = 1, with_labels=False, node_size = 50, __
⇒arrowsize = 100, nodelist=range(0,n), pos= X, ax= axs[0])
axs[1].scatter(X[:,0], X[:,1], c=colors)
nx.draw(graph, font_size = 20, width = 1, with_labels=False, node_size = 50, u
 →arrowsize = 100, nodelist=range(0,n), pos= X, node_color=colors, ax = axs[2])
fig.suptitle("Spectral", fontsize=16)
plt.show()
kmeans = KMeans(n_clusters=2)
kmeans.fit(X)
colors = kmeans.labels_
print("K-means clustering accuracy:\n", classification_report(labels, colors))
fig, axs = plt.subplots(1, 3, figsize=(30, 10), subplot_kw=dict(xticks=[],_u

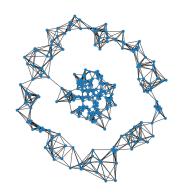
yticks=[]))
axs[0].scatter(X[:,0], X[:,1])
axs[1].scatter(X[:,0], X[:,1], c=colors)
nx.draw(graph, font_size = 20, width = 1, with_labels=False, node_size = 50,__
 arrowsize = 100, nodelist=range(0,n), pos= X, node_color=colors, ax = axs[2])
fig.suptitle("K-means", fontsize=16)
plt.show()
```



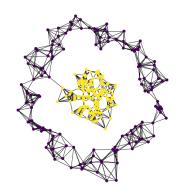
Spectral clustering accuracy:

	precision	recall	f1-score	support
0	1.00	0.99	1.00	150
1	0.99	1.00	1.00	150
accuracy			1.00	300
macro avg	1.00	1.00	1.00	300
weighted avg	1.00	1.00	1.00	300

Spectral



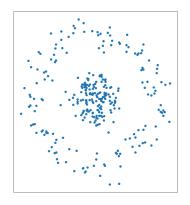


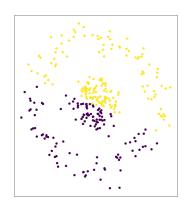


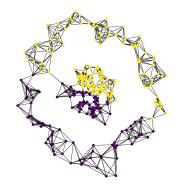
K-means clustering accuracy:

	precision	recall	f1-score	support
0 1	0.53 0.52	0.47 0.59	0.50 0.55	150 150
accuracy macro avg	0.53	0.53	0.53 0.52	300 300
weighted avg	0.53	0.53	0.52	300

K-means



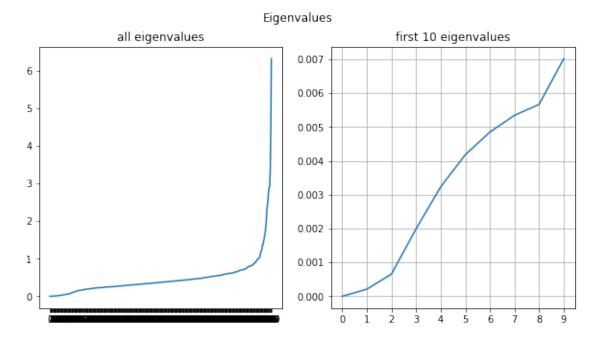




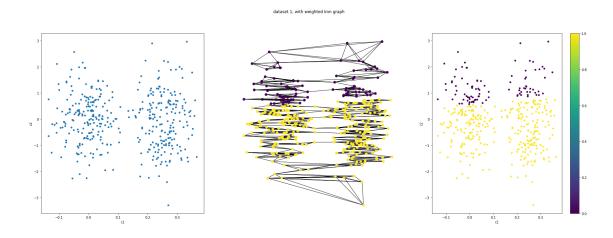
```
[]: import pandas as pd
     df1 = pd.read_csv("./first_clustering_dataset.csv", names=['c1', 'c2'])
     df2 = pd.read_csv("./second_clustering_dataset.csv", names=['c1', 'c2'])
     df3 = pd.read_csv("./third_clustering_dataset.csv", names=['c1', 'c2'])
     from sklearn.neighbors import kneighbors_graph
     adj1 = kneighbors_graph(df1, n_neighbors=5, ).toarray()
     n = len(adj1)
     for i in range(n):
             for j in range(i, n):
                     if (adj1[i][j] != adj1[j][i]):
                             adj1[i][j] = 1
                             adj1[j][i] = 1
     graph1 = BuildGraph(len(adj1), adj1)
     adj1_2 = kneighbors_graph(df1, n_neighbors=5, mode='distance').toarray()
     n = len(adj1_2)
     for i in range(n):
             for j in range(i, n):
                     if (adj1_2[i][j] != adj1_2[j][i]):
                             adj1_2[i][j] = max(adj1_2[i][j], adj1_2[j][i])
                             adj1_2[j][i] = adj1_2[i][j]
     graph1_2 = BuildGraph(len(adj1_2), adj1_2)
     adj2 = kneighbors_graph(df2, n_neighbors=7, ).toarray()
     n = len(adj2)
     for i in range(n):
             for j in range(i, n):
                     if (adj2[i][j] != adj2[j][i]):
                             adj2[i][j] = 1
                             adj2[j][i] = 1
     graph2 = BuildGraph(len(adj2), adj2)
     adj3 = kneighbors_graph(df3, n_neighbors=7, ).toarray()
     n = len(adj3)
     for i in range(n):
             for j in range(i, n):
                     if (adj3[i][j] != adj3[j][i]):
                             adj3[i][j] = 1
                             adi3[i][i] = 1
     graph3 = BuildGraph(len(adj3), adj3)
```

```
[]: colors1_2 = spectral_cluster(len(adj1_2), adj1_2, 2)

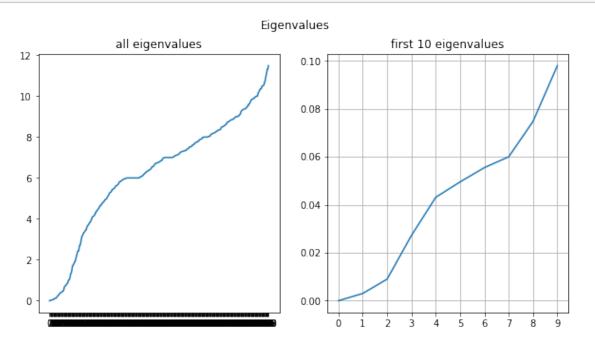
fig, axs = plt.subplots(1, 3, figsize=(30, 10))
    df1.plot.scatter(ax= axs[0], x = 'c1', y = 'c2')
    nx.draw(graph1_2, font_size = 20, width = 1, with_labels=False, node_size = 50, arrowsize = 100, nodelist=range(0,len(adj1_2)), pos= df1.to_numpy(), arrowsize = 100, ax = axs[1])
    df1.plot.scatter(ax= axs[2], x = 'c1', y = 'c2', c=colors1_2, arrowsize = colormap='viridis')
    fig.suptitle("dataset 1, with weighted knn graph")
```



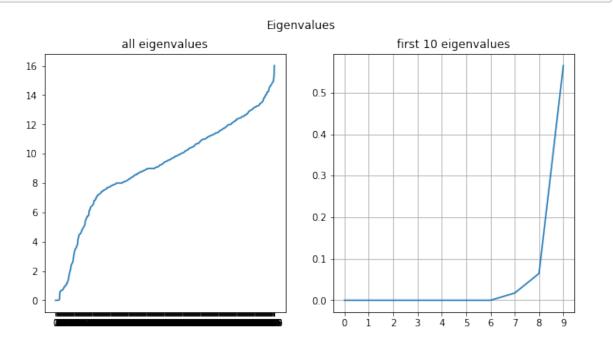
[]: Text(0.5, 0.98, 'dataset 1, with weighted knn graph')



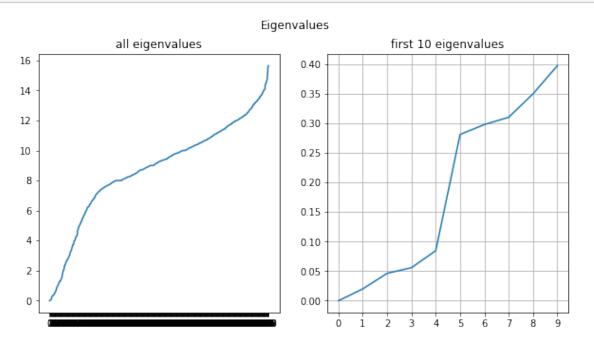
[]: colors1 = spectral_cluster(len(adj1), adj1, 3)



[]: colors2 = spectral_cluster(len(adj2), adj2, 9)



[]: colors3 = spectral_cluster(len(adj3), adj3, 5)



```
[]: fig, axs = plt.subplots(3, 3, figsize=(30, 20))
     df1.plot.scatter(ax= axs[0][0], x = 'c1', y = 'c2')
     df2.plot.scatter(ax= axs[0][1], x = 'c1', y = 'c2')
     df3.plot.scatter(ax= axs[0][2], x = 'c1', y = 'c2')
     axs[0][0].set_title("dataset 1")
     axs[0][1].set_title("dataset 2")
     axs[0][2].set_title("dataset 3")
     nx.draw(graph1, font_size = 20, width = 1, with_labels=False, node_size = 50,__
      →arrowsize = 100, nodelist=range(0,len(adj1)), pos= df1.to_numpy(),
      →node_color=colors1, ax = axs[1][0])
     nx.draw(graph2, font_size = 20, width = 1, with_labels=False, node_size = 50,__
      Garrowsize = 100, nodelist=range(0,len(adj2)), pos= df2.to_numpy(),⊔
      →node_color=colors2, ax = axs[1][1])
     nx.draw(graph3, font_size = 20, width = 1, with_labels=False, node_size = 50,__
      →arrowsize = 100, nodelist=range(0,len(adj3)), pos= df3.to_numpy(),__
      →node_color=colors3, ax = axs[1][2])
     df1.plot.scatter(ax= axs[2][0], x = 'c1', y = 'c2', c=colors1, \Box
      ⇔colormap='viridis')
     df2.plot.scatter(ax= axs[2][1], x = 'c1', y = 'c2', c=colors2,
      ⇔colormap='viridis')
     df3.plot.scatter(ax= axs[2][2], x = 'c1', y = 'c2', c=colors3,
      ⇔colormap='viridis')
```

[]: <AxesSubplot:xlabel='c1', ylabel='c2'>

