

```
import pandas as pd
#nazwy kolumn
column_names = [
    "ID", "Diagnosis",
    "radius1", "texture1", "perimeter1", "area1", "smoothness1",
    "compactness1", "concavity1", "concave_points1",
    "symmetry1", "fractal_dimension1",
    "radius2", "texture2", "perimeter2", "area2", "smoothness2",
    "compactness2", "concavity2", "concave_points2",
    "symmetry2", "fractal_dimension2",
    "radius3", "texture3", "perimeter3", "area3", "smoothness3",
    "compactness3", "concavity3", "concave_points3",
    "symmetry3", "fractal_dimension3"
]
```

```
# Ścieżka do pliku CSV
sciezka = 'wdbc.csv'
```

```
# Wczytywanie pliku CSV do DataFrame
df = pd.read_csv(sciezka, header=None, names=column_names)
```

```
# Wyświetlenie pierwszych 5 wierszy
print(df.head())
```

	ID	Diagnosis	radius1	texture1	perimeter1	area1
smoothness1 \						
0	842302	M	17.99	10.38	122.80	1001.0
0.11840						
1	842517	M	20.57	17.77	132.90	1326.0
0.08474						
2	84300903	M	19.69	21.25	130.00	1203.0
0.10960						
3	84348301	M	11.42	20.38	77.58	386.1
0.14250						
4	84358402	M	20.29	14.34	135.10	1297.0
0.10030						

	compactness1	concavity1	concave_points1	...	radius3	
texture3 \						
0	0.27760	0.3001	0.14710	...	25.38	17.33
1	0.07864	0.0869	0.07017	...	24.99	23.41
2	0.15990	0.1974	0.12790	...	23.57	25.53
3	0.28390	0.2414	0.10520	...	14.91	26.50
4	0.13280	0.1980	0.10430	...	22.54	16.67

	perimeter3	area3	smoothness3	compactness3	concavity3
concave_points3 \					
0	184.60	2019.0	0.1622	0.6656	0.7119
0.2654					
1	158.80	1956.0	0.1238	0.1866	0.2416
0.1860					
2	152.50	1709.0	0.1444	0.4245	0.4504
0.2430					
3	98.87	567.7	0.2098	0.8663	0.6869
0.2575					
4	152.20	1575.0	0.1374	0.2050	0.4000
0.1625					

	symmetry3	fractal_dimension3
0	0.4601	0.11890
1	0.2750	0.08902
2	0.3613	0.08758
3	0.6638	0.17300
4	0.2364	0.07678

[5 rows x 32 columns]

```
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.preprocessing import StandardScaler
from sklearn.cluster import KMeans
from sklearn.metrics import silhouette_score

# Przygotowanie danych
X = df.drop(columns=["ID", "Diagnosis"])
X_scaled = StandardScaler().fit_transform(X)

# Testowanie różnych liczby skupień w KMeans
silhouette_scores = []
k_range = range(2, 11)

for k in k_range:
    kmeans = KMeans(n_clusters=k, random_state=0)
    labels = kmeans.fit_predict(X_scaled)
    score = silhouette_score(X_scaled, labels)
    silhouette_scores.append(score)
    print(f"Liczba skupień: {k}, Silhouette Score: {score:.2f}")

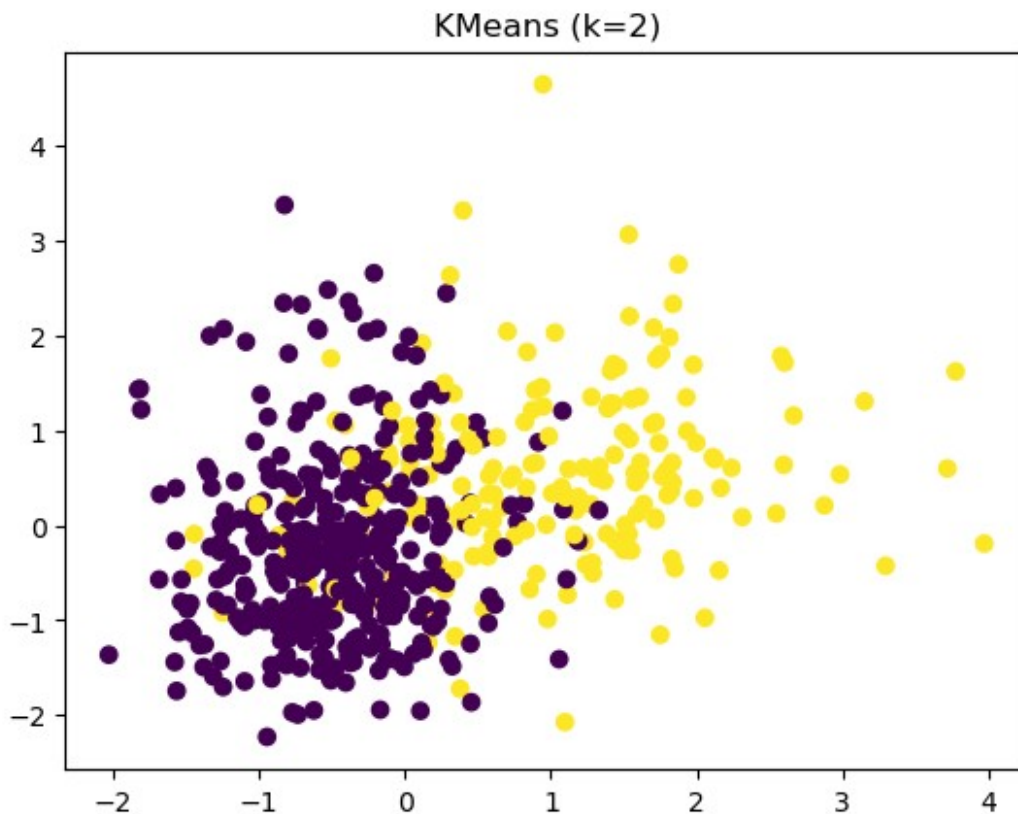
    plt.scatter(X_scaled[:, 0], X_scaled[:, 1], c=labels,
cmap='viridis')
    plt.title(f"KMeans (k={k})")
    plt.show()

# Wykres Silhouette Score vs liczba skupień
plt.plot(k_range, silhouette_scores, marker='o')
```

```
plt.title("Wpływ liczby skupień na Silhouette Score")
plt.xlabel("Liczba skupień (k)")
plt.ylabel("Silhouette Score")
plt.grid(True)
plt.show()
```

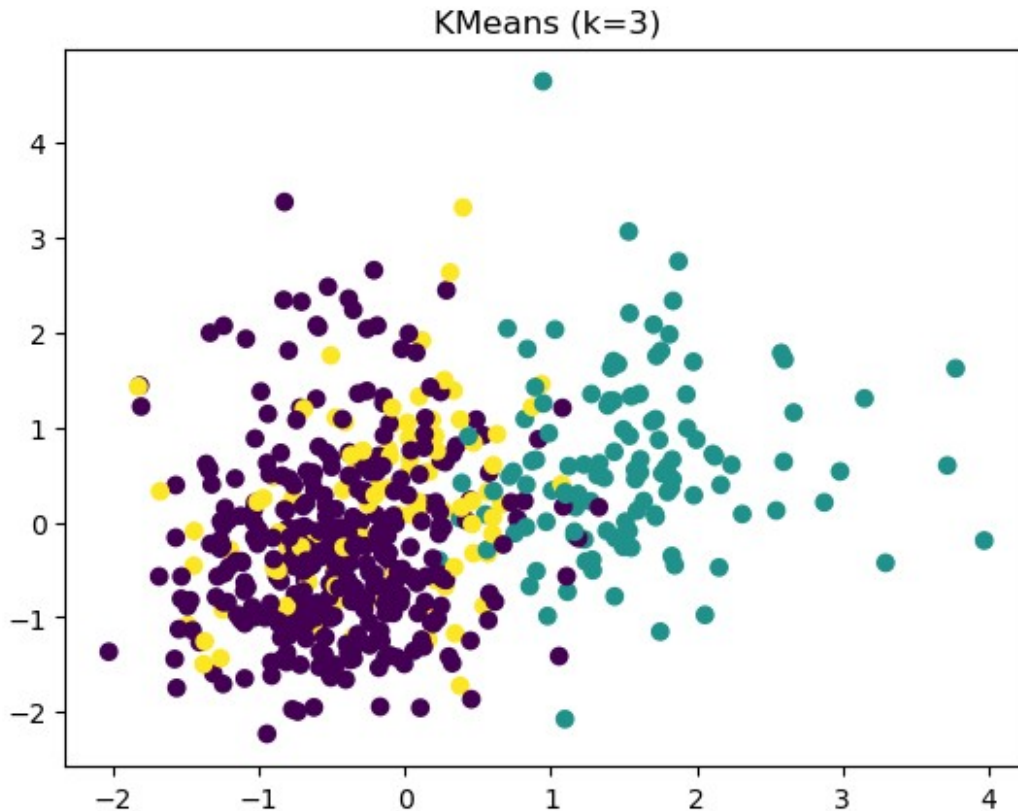
```
C:\Users\48664\anaconda3\Lib\site-packages\sklearn\cluster\
_kmeans.py:1429: UserWarning: KMeans is known to have a memory leak on
Windows with MKL, when there are less chunks than available threads.
You can avoid it by setting the environment variable
OMP_NUM_THREADS=3.
  warnings.warn(
```

Liczba skupień: 2, Silhouette Score: 0.34



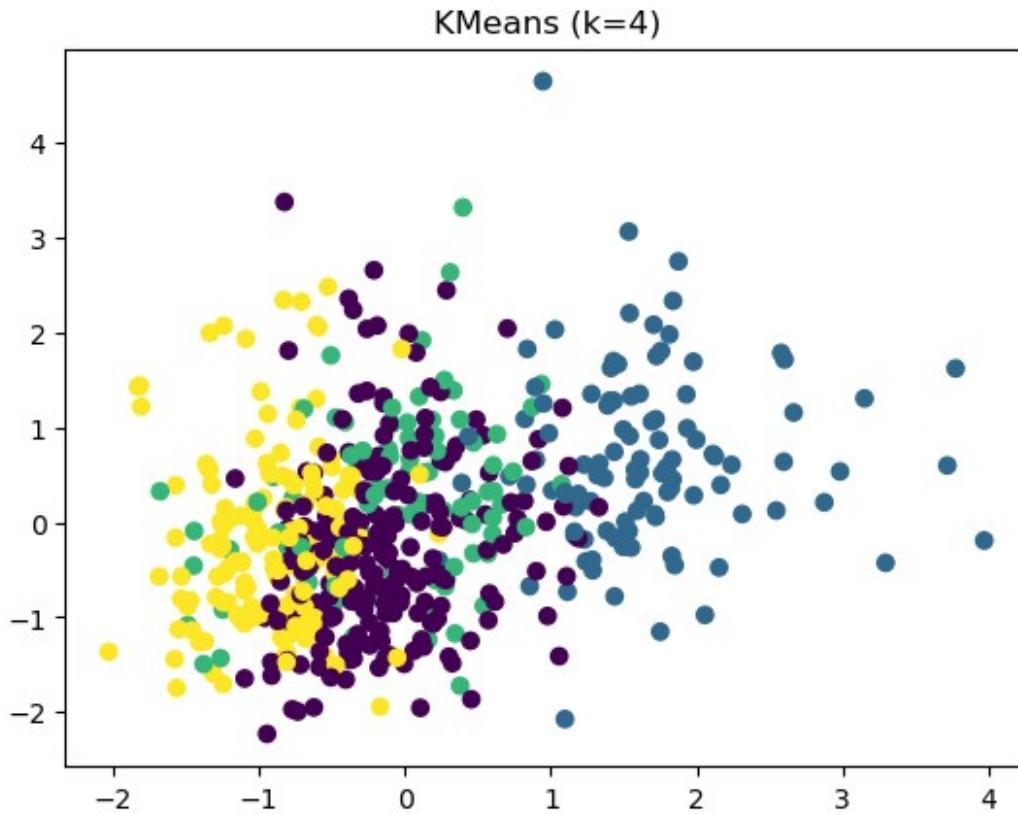
```
C:\Users\48664\anaconda3\Lib\site-packages\sklearn\cluster\
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```

Liczba skupień: 3, Silhouette Score: 0.32



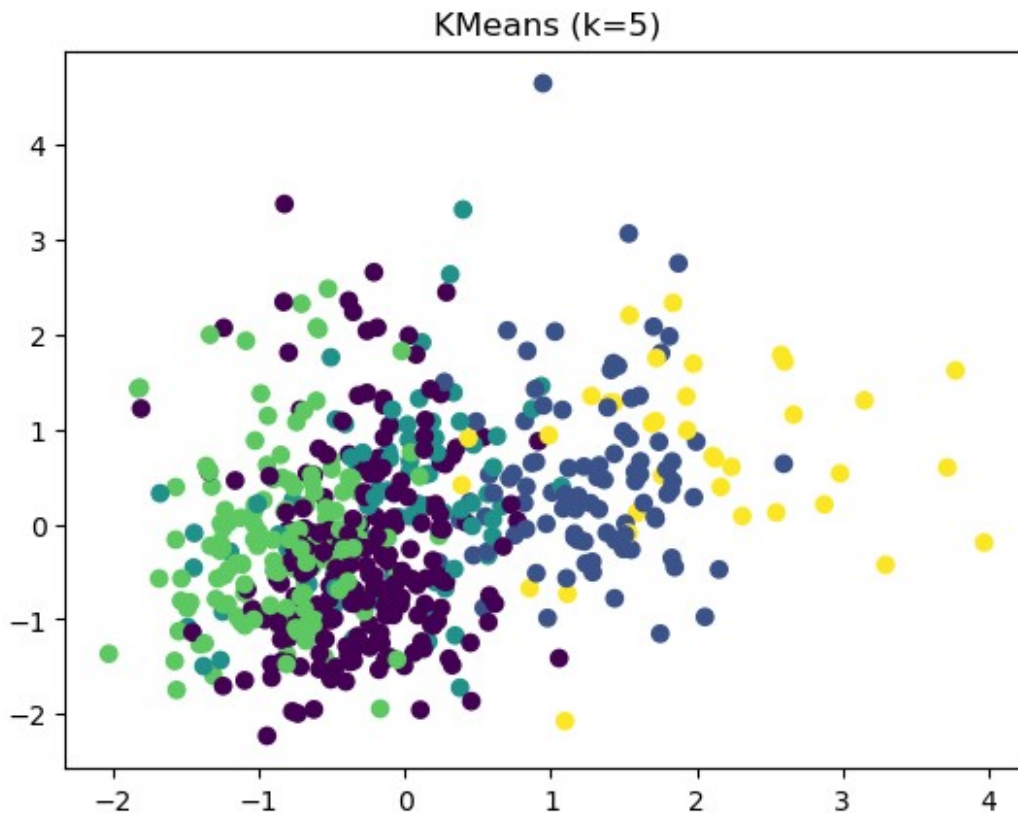
```
C:\Users\48664\anaconda3\Lib\site-packages\sklearn\cluster\  
_kmeans.py:1429: UserWarning: KMeans is known to have a memory leak on  
Windows with MKL, when there are less chunks than available threads.  
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```

Liczba skupień: 4, Silhouette Score: 0.16



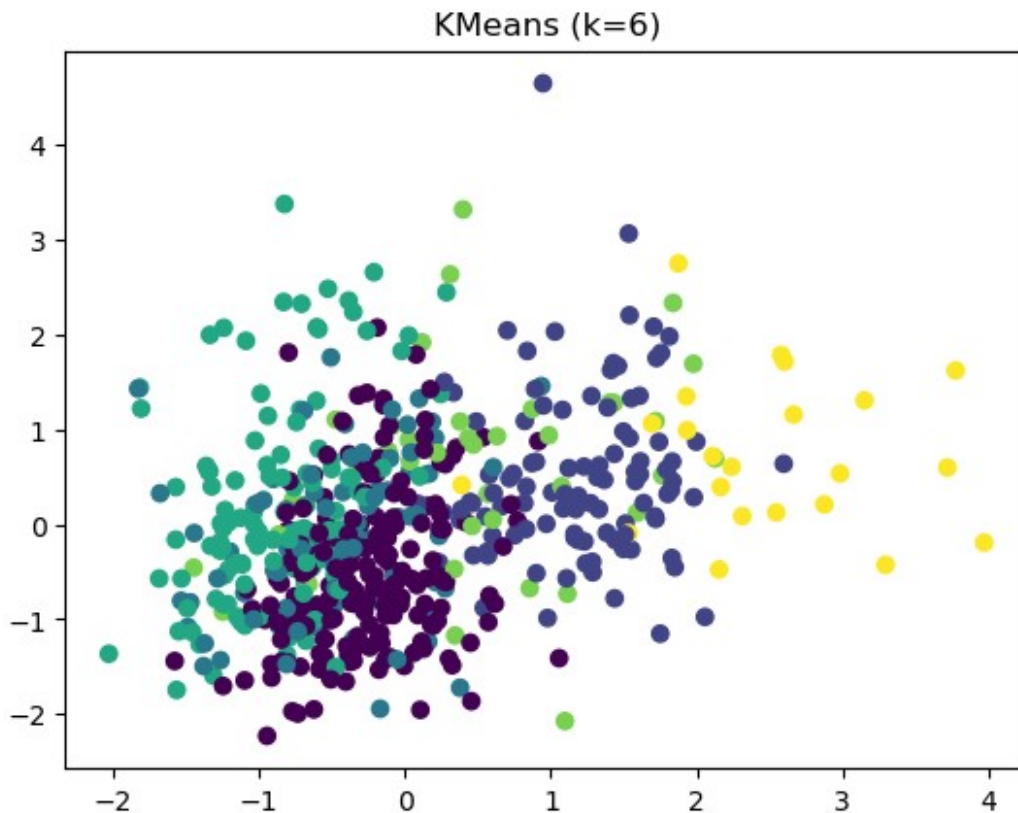
```
C:\Users\48664\anaconda3\Lib\site-packages\sklearn\cluster\  
_kmeans.py:1429: UserWarning: KMeans is known to have a memory leak on  
Windows with MKL, when there are less chunks than available threads.  
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warnings.warn(
```

Liczba skupień: 5, Silhouette Score: 0.17



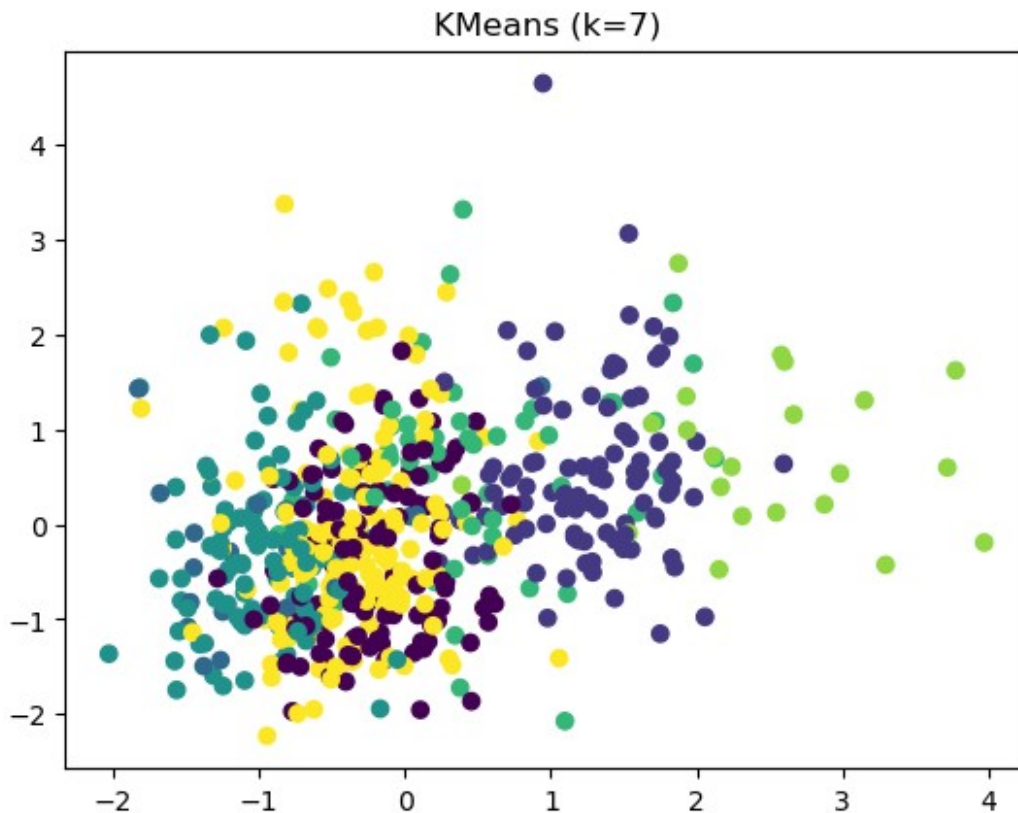
```
C:\Users\48664\anaconda3\Lib\site-packages\sklearn\cluster\  
_kmeans.py:1429: UserWarning: KMeans is known to have a memory leak on  
Windows with MKL, when there are less chunks than available threads.  
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warnings.warn(
```

Liczba skupień: 6, Silhouette Score: 0.16



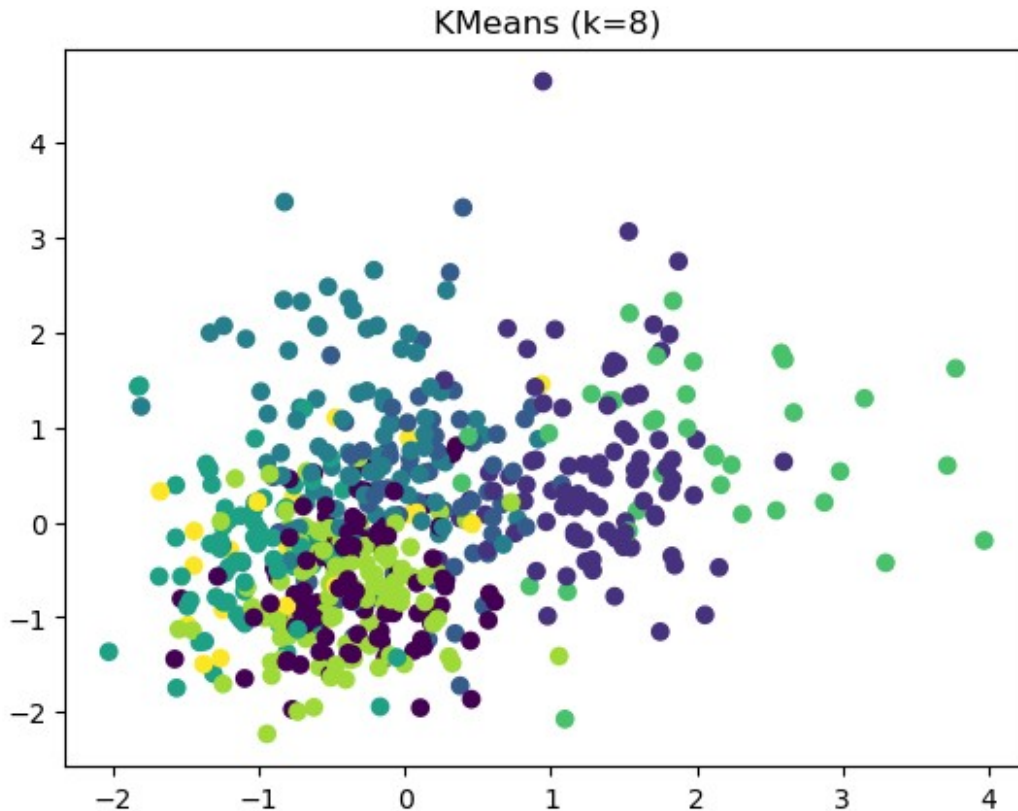
```
C:\Users\48664\anaconda3\Lib\site-packages\sklearn\cluster\  
_kmeans.py:1429: UserWarning: KMeans is known to have a memory leak on  
Windows with MKL, when there are less chunks than available threads.  
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warnings.warn(
```

Liczba skupień: 7, Silhouette Score: 0.14



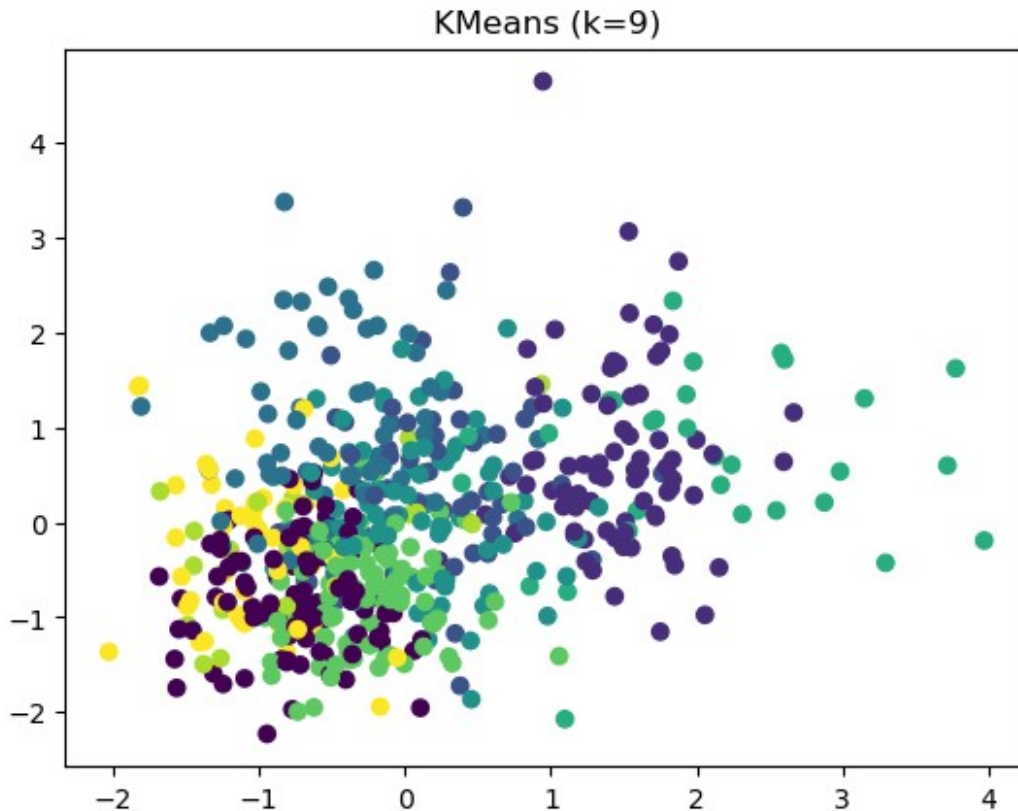
```
C:\Users\48664\anaconda3\Lib\site-packages\sklearn\cluster\  
_kmeans.py:1429: UserWarning: KMeans is known to have a memory leak on  
Windows with MKL, when there are less chunks than available threads.  
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  warnings.warn(
```

Liczba skupień: 8, Silhouette Score: 0.13



```
C:\Users\48664\anaconda3\Lib\site-packages\sklearn\cluster\  
_kmeans.py:1429: UserWarning: KMeans is known to have a memory leak on  
Windows with MKL, when there are less chunks than available threads.  
You can avoid it by setting the environment variable  
OMP_NUM_THREADS=3.  
  warnings.warn(
```

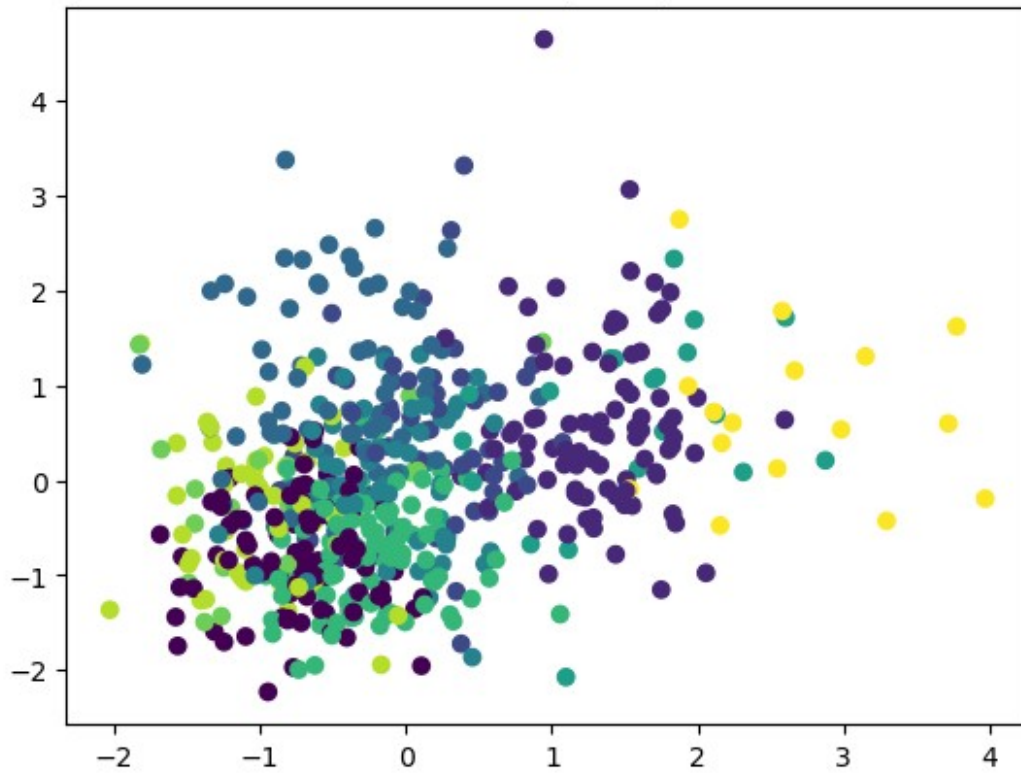
Liczba skupień: 9, Silhouette Score: 0.12

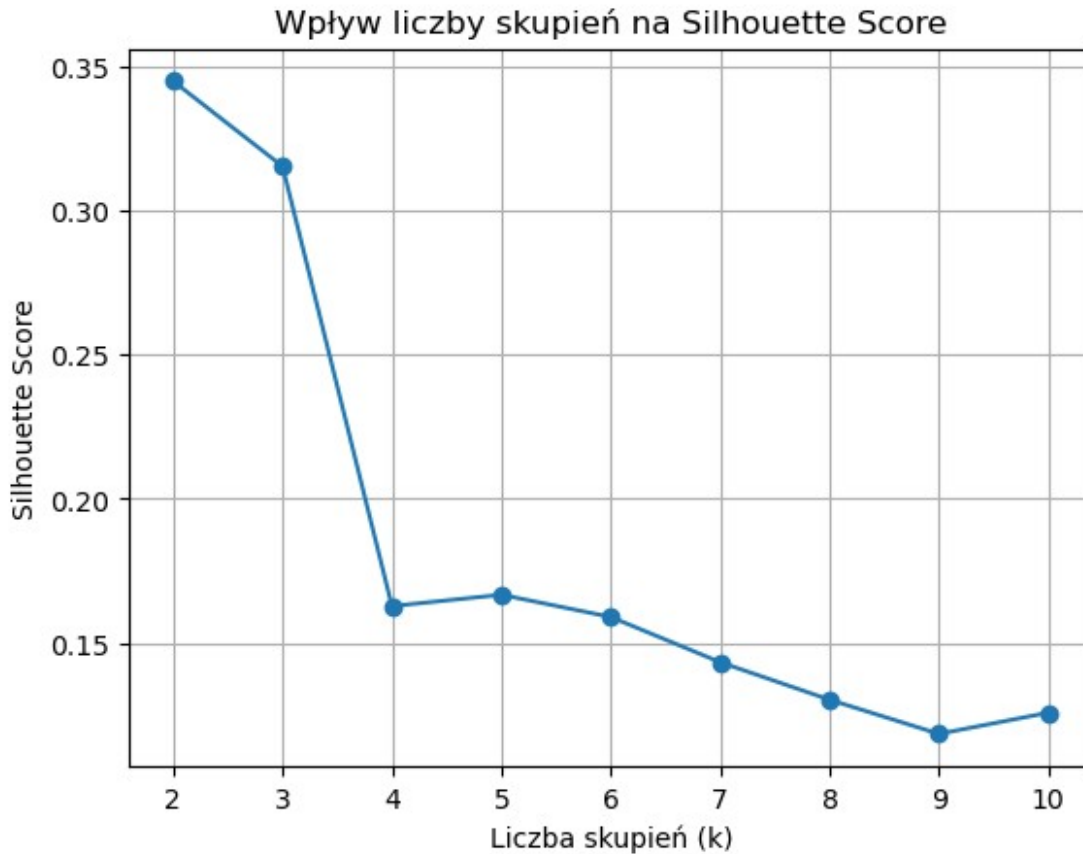


```
C:\Users\48664\anaconda3\Lib\site-packages\sklearn\cluster\  
_kmeans.py:1429: UserWarning: KMeans is known to have a memory leak on  
Windows with MKL, when there are less chunks than available threads.  
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warnings.warn(
```

Liczba skupień: 10, Silhouette Score: 0.13

KMeans (k=10)





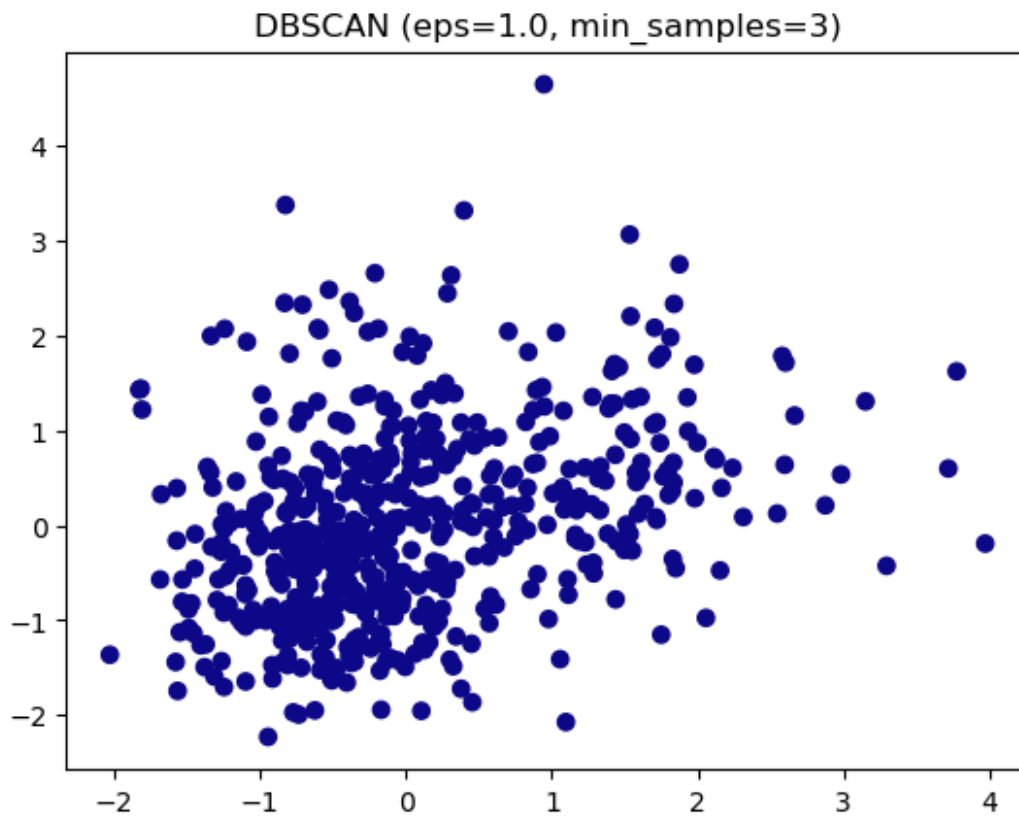
```
# --- Zadanie 2: DBSCAN ---
eps_values = [1.0, 2.0, 3.0]
min_samples_values = [3, 5, 10]

for eps in eps_values:
    for min_samples in min_samples_values:
        dbscan = DBSCAN(eps=eps, min_samples=min_samples)
        labels = dbscan.fit_predict(X_scaled)
        n_clusters = len(set(labels)) - (1 if -1 in labels else 0)

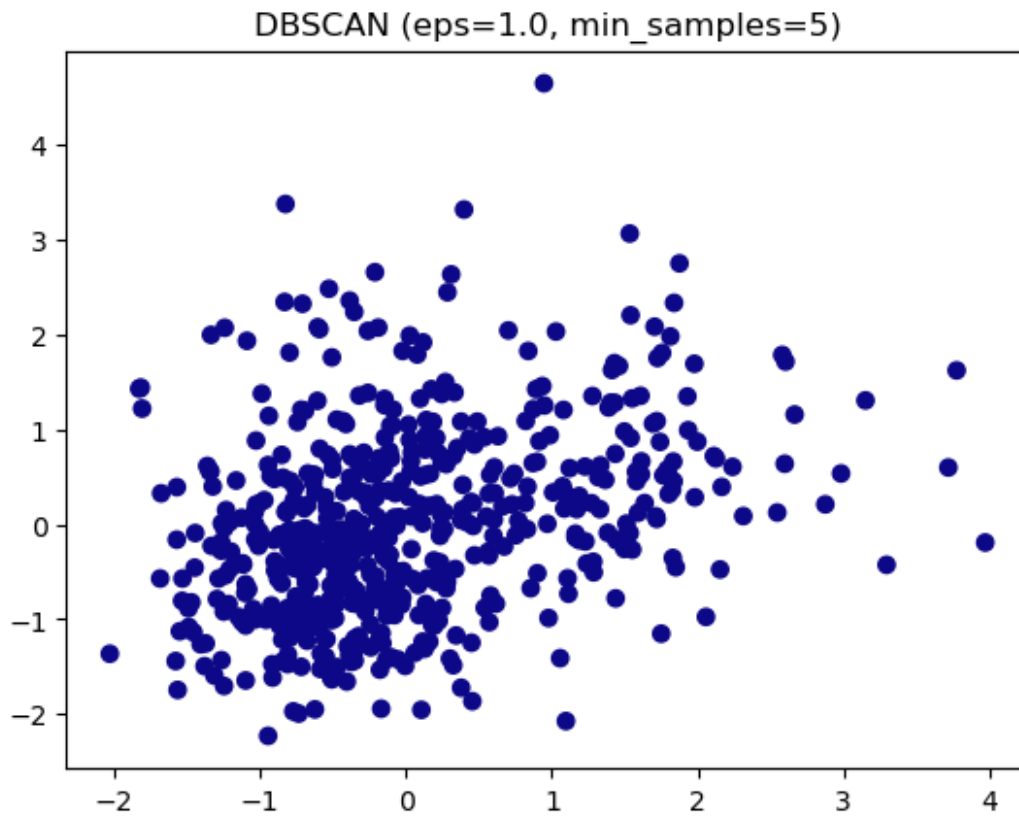
        if n_clusters > 1:
            score = silhouette_score(X_scaled, labels)
            print(f"DBSCAN (eps={eps}, min_samples={min_samples}):
Silhouette Score = {score:.2f}, klastry = {n_clusters}")
        else:
            print(f"DBSCAN (eps={eps}, min_samples={min_samples}):
zbyt mało klastrów do obliczenia Silhouette")

plt.scatter(X_scaled[:, 0], X_scaled[:, 1], c=labels,
cmap='plasma')
plt.title(f"DBSCAN (eps={eps}, min_samples={min_samples})")
plt.show()
```

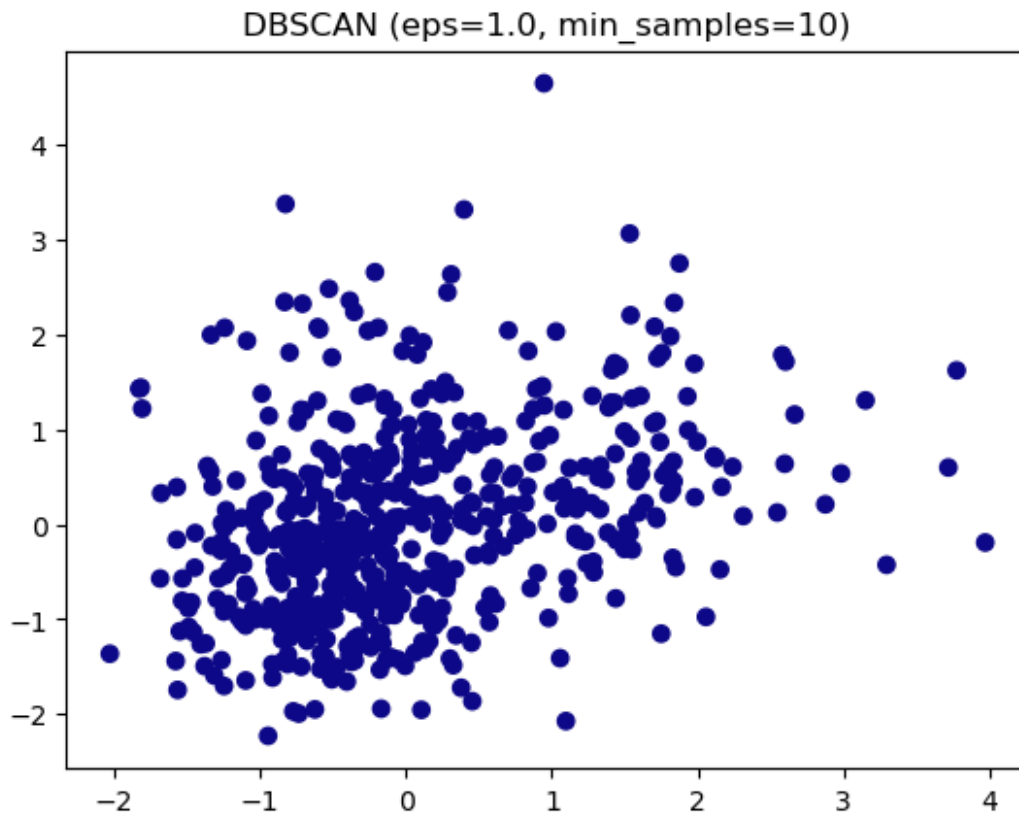
DBSCAN (eps=1.0, min_samples=3): zbyt mało klastrów do obliczenia Silhouette



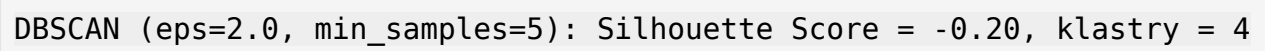
DBSCAN (eps=1.0, min_samples=5): zbyt mało klastrów do obliczenia Silhouette

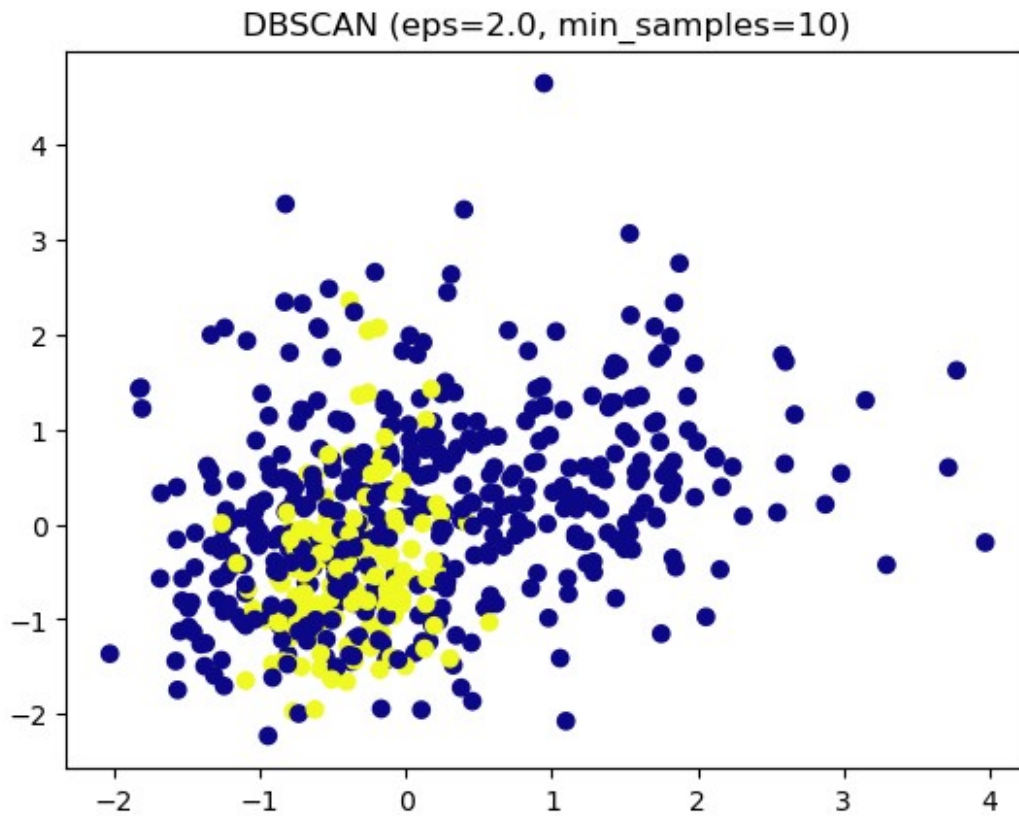


DBSCAN (eps=1.0, min_samples=10): zbyt mało klastrów do obliczenia Silhouette

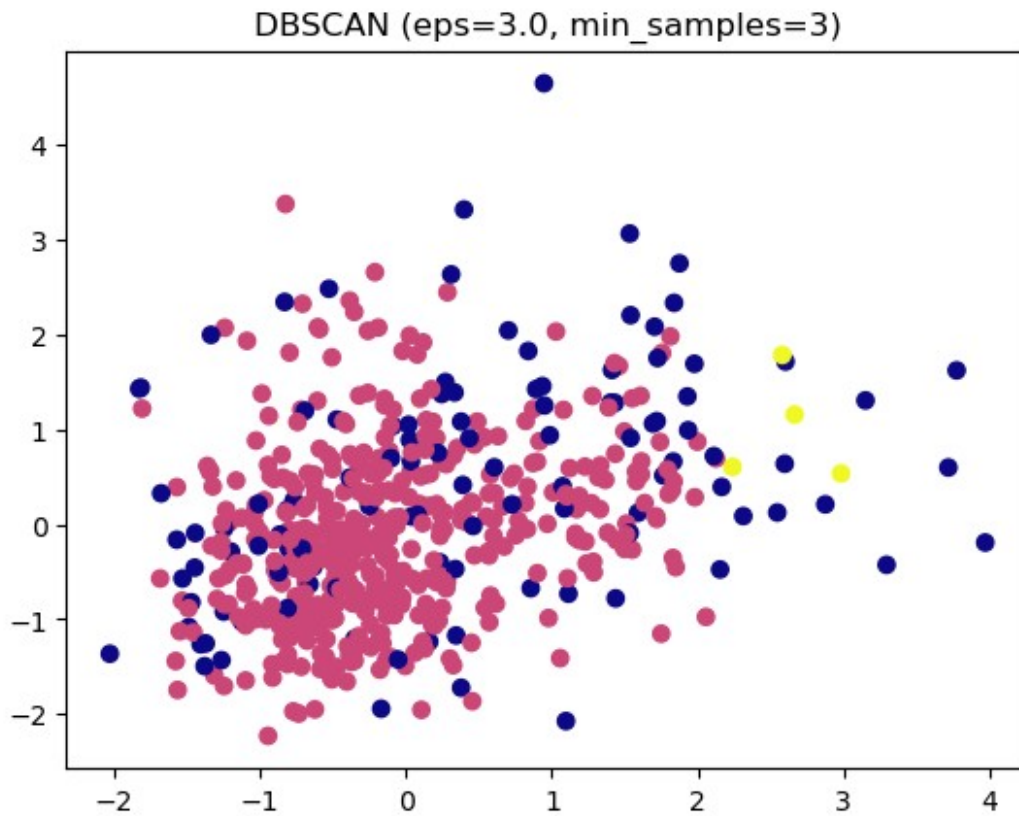


DBSCAN (eps=2.0, min_samples=3): Silhouette Score = -0.20, klastry = 7

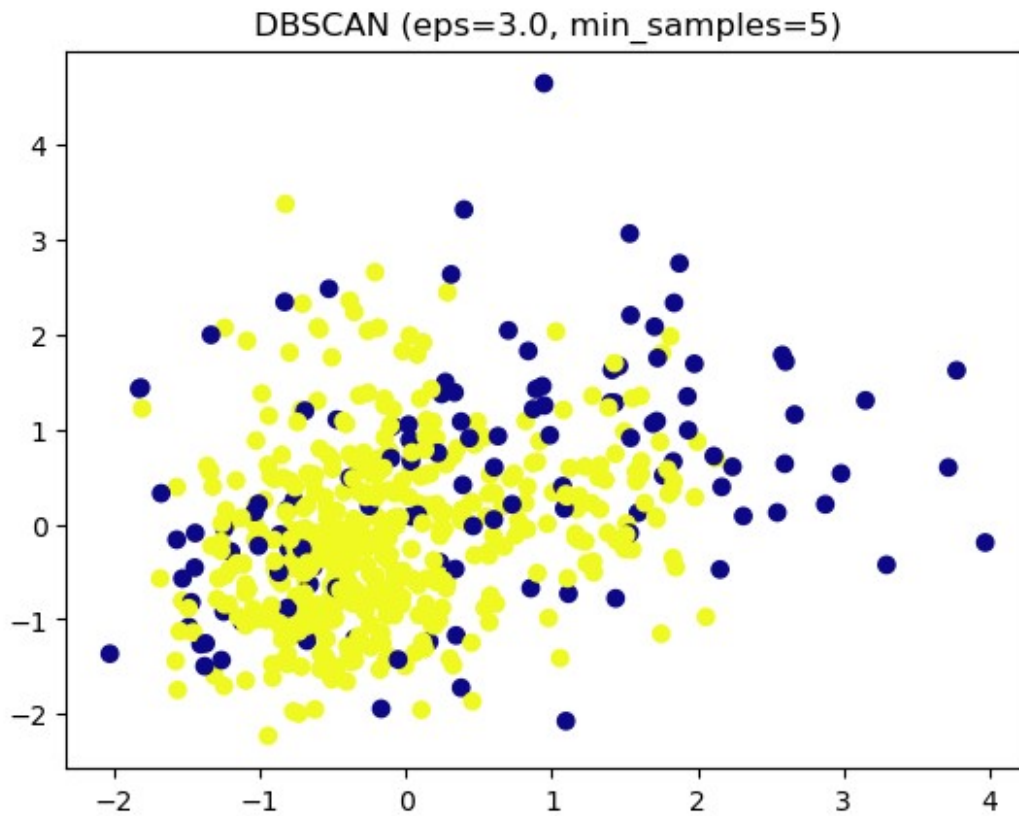




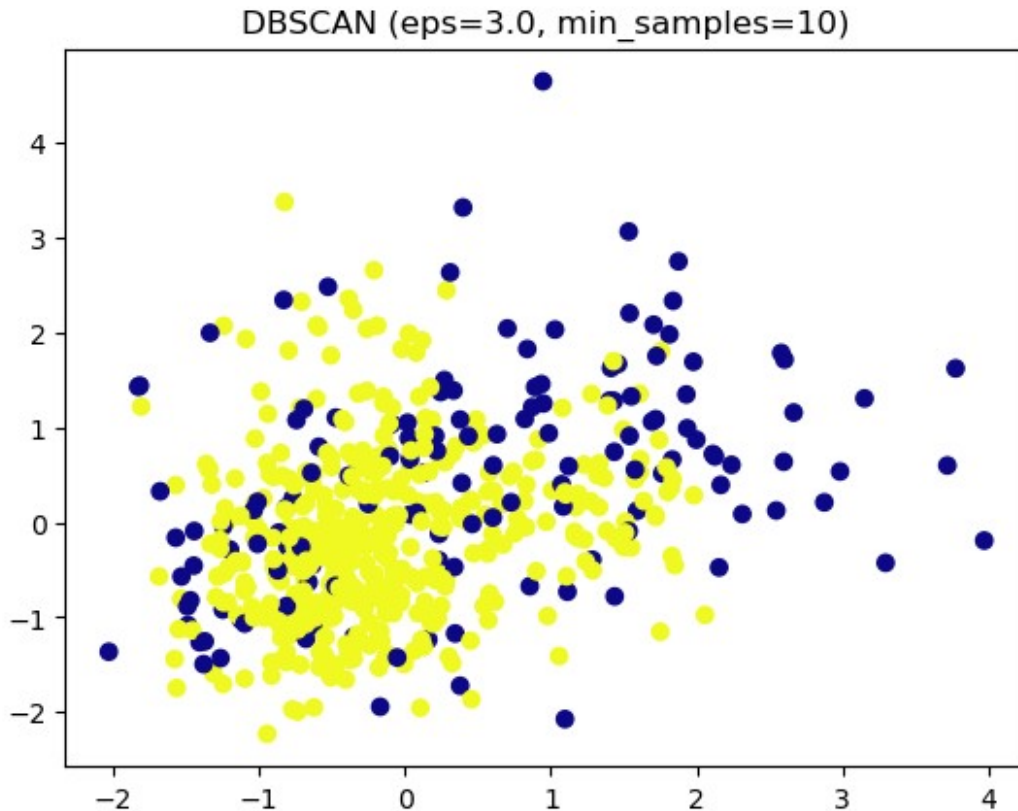
DBSCAN (eps=3.0, min_samples=3): Silhouette Score = 0.26, klastry = 2



DBSCAN (eps=3.0, min_samples=5): zbyt mało klastrów do obliczenia Silhouette



DBSCAN (eps=3.0, min_samples=10): zbyt mało klastrów do obliczenia Silhouette



```
# --- Zadanie 3: Porównanie metod ---
# Wybór konkretnych konfiguracji
kmeans = KMeans(n_clusters=3, random_state=0)
kmeans_labels = kmeans.fit_predict(X_scaled)
kmeans_score = silhouette_score(X_scaled, kmeans_labels)

agg = AgglomerativeClustering(n_clusters=3)
agg_labels = agg.fit_predict(X_scaled)
agg_score = silhouette_score(X_scaled, agg_labels)

dbscan = DBSCAN(eps=1.5, min_samples=5)
dbscan_labels = dbscan.fit_predict(X_scaled)
dbscan_score = silhouette_score(X_scaled, dbscan_labels) if
len(set(dbscan_labels)) > 1 else -1

# Wyświetlenie wyników
print("\nPorównanie metod (Silhouette Score):")
print(f"- KMeans: {kmeans_score:.2f}")
print(f"- Agglomerative: {agg_score:.2f}")
print(f"- DBSCAN: {dbscan_score:.2f}")

fig, axs = plt.subplots(1, 3, figsize=(18, 5))
axs[0].scatter(X_scaled[:, 0], X_scaled[:, 1], c=kmeans_labels,
cmap='viridis')
```

```

axs[0].set_title("KMeans")

axs[1].scatter(X_scaled[:, 0], X_scaled[:, 1], c=agg_labels,
cmap='rainbow')
axs[1].set_title("Agglomerative")

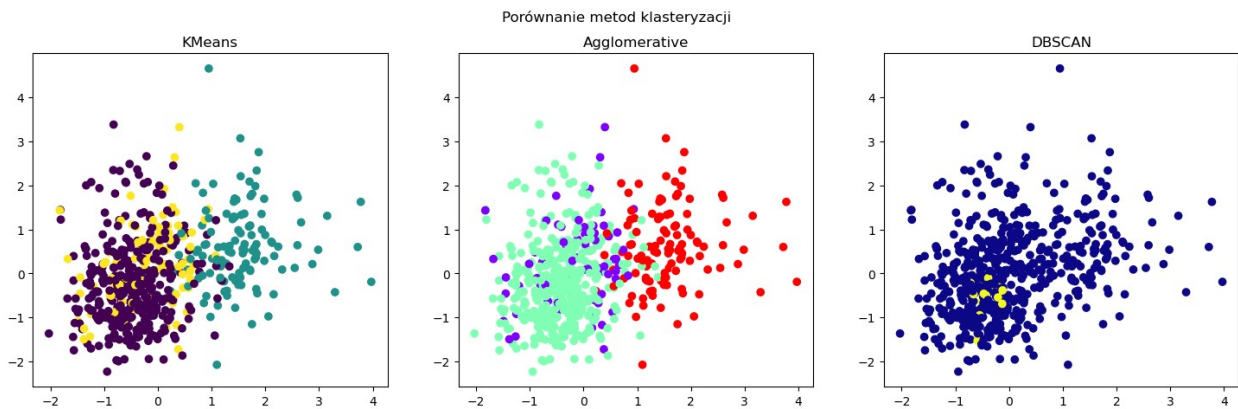
axs[2].scatter(X_scaled[:, 0], X_scaled[:, 1], c=dbscan_labels,
cmap='plasma')
axs[2].set_title("DBSCAN")

plt.suptitle("Porównanie metod klasteryzacji")
plt.show()

C:\Users\48664\anaconda3\Lib\site-packages\sklearn\cluster\
_kmeans.py:1429: UserWarning: KMeans is known to have a memory leak on
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OMP_NUM_THREADS=3.
  warnings.warn(

Porównanie metod (Silhouette Score):
- KMeans: 0.32
- Agglomerative: 0.33
- DBSCAN: -0.22

```



```

# --- Zadanie 4: Zastosowanie KMeans do danych rzeczywistych
(oryginalny DataFrame) ---
kmeans_real = KMeans(n_clusters=3, random_state=0)
kmeans_real_labels = kmeans_real.fit_predict(X_scaled)
kmeans_real_score = silhouette_score(X_scaled, kmeans_real_labels)

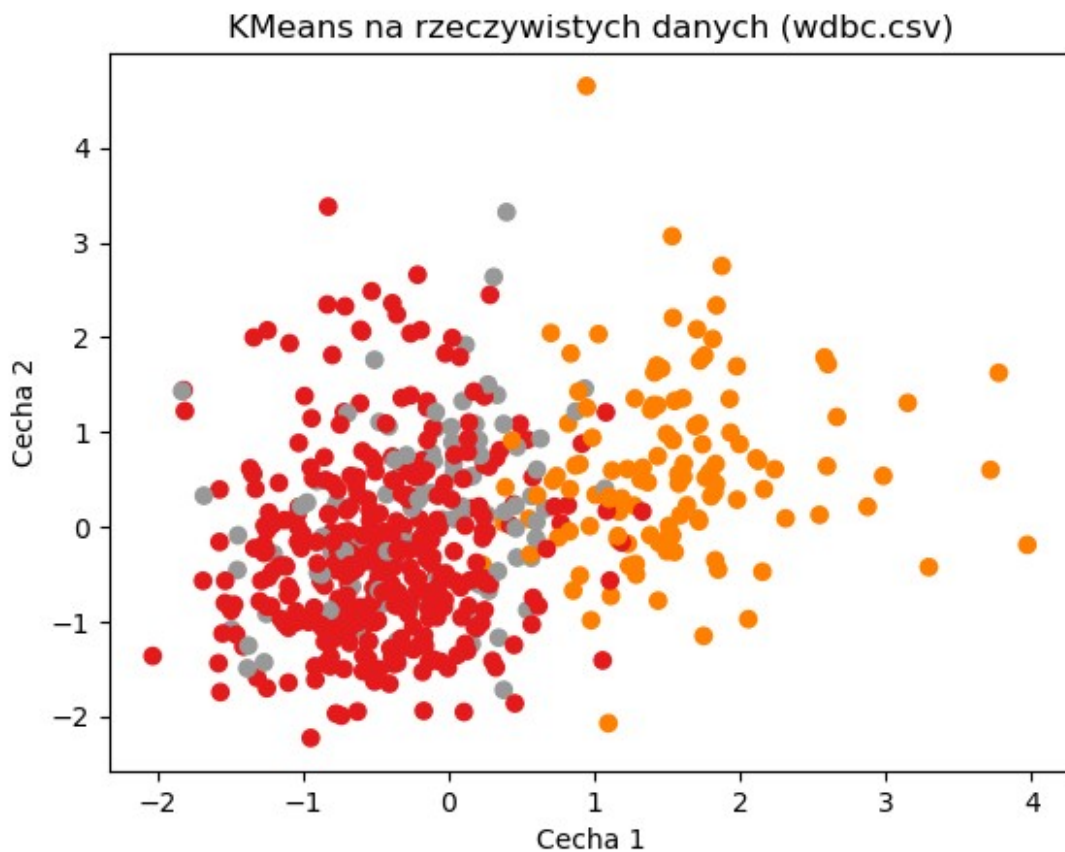
plt.scatter(X_scaled[:, 0], X_scaled[:, 1], c=kmeans_real_labels,
cmap='Set1')
plt.title("KMeans na rzeczywistych danych (wdbc.csv)")
plt.xlabel("Cecha 1")
plt.ylabel("Cecha 2")

```

```
plt.show()

print(f"Silhouette Score (wdbc.csv, KMeans): {kmeans_real_score:.2f}")

C:\Users\48664\anaconda3\Lib\site-packages\sklearn\cluster\_kmeans.py:1429: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP_NUM_THREADS=3.
  warnings.warn(
```



```
Silhouette Score (wdbc.csv, KMeans): 0.32
```