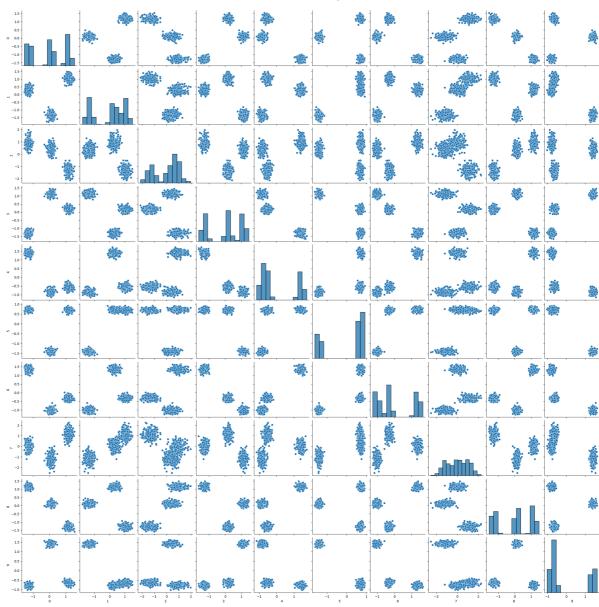
```
import numpy as np
In [20]:
          import pandas as pd
          import seaborn as sns
          from sklearn.cluster import DBSCAN
          from sklearn.datasets import make_blobs
          from sklearn.preprocessing import StandardScaler
          from sklearn.decomposition import PCA
          from IPython.display import display
In [13]: X, _ = make_blobs(
               n_samples=750, n_features=10, cluster_std=0.4, random_state=0
          X = StandardScaler().fit_transform(X)
In [14]:
          df = pd.DataFrame(X)
          display(df)
                       0
                                 1
                                           2
                                                                                             7
                                                     3
                                                               4
                                                                         5
                                                                                   6
            0 -1.412178
                          0.138881
                                    1.030354 -1.374328
                                                        1.277320
                                                                  0.716573
                                                                            1.352784 -0.072073
                                                                                                1.1181
                1.120948
                          1.053943
                                   -1.291701
                                              0.250096
                                                        -0.622285
                                                                  0.671510 -0.185737
                                                                                       0.496576 -1.3574
            2 -1.473422
                          0.555454
                                    0.518343 -1.198269
                                                        1.419566
                                                                  0.741096
                                                                            1.278301
                                                                                       0.156885
                                                                                                 0.97550
                1.148823
                          1.054616 -1.359783
                                              0.234991
                                                        -0.427253
                                                                  0.757168
                                                                            -0.342457
                                                                                       1.161739
                                                                                               -1.3782
               -0.233203
                        -1.327316
                                    0.666572
                                              1.119386
                                                        -0.706356
                                                                 -1.231405
                                                                           -1.040749
                                                                                      -0.681332
                                                                                               -0.0407
          745
                1.532491
                          0.955039
                                   -0.995627
                                              0.251636
                                                        -0.703416
                                                                  0.668838
                                                                            -0.438386
                                                                                       1.397128 -1.5740!
          746
              -1.313371
                          0.409123
                                    1.169887
                                             -1.131298
                                                         1.336661
                                                                  0.653766
                                                                             1.441933
                                                                                      -0.313999
                                                                                                 1.25639
          747 -1.460007
                          0.411918
                                    1.076572 -1.169049
                                                        1.268149
                                                                  0.726909
                                                                             1.590668
                                                                                       0.275679
                                                                                                 1.1709
               -1.454470
                          0.153634
                                    1.031319
                                             -1.409967
                                                         1.515520
                                                                  0.826626
                                                                             1.075929
                                                                                       0.356608
                                                                                                 0.9122
          749 -1.174188 -0.013734
                                    1.146523 -1.503986
                                                        1.126015
                                                                  0.579320
                                                                             1.567498 -0.068226
                                                                                                 1.1694
         750 rows × 10 columns
In [15]:
          dbscan = DBSCAN()
          clustering = dbscan.fit_predict(X)
          np.unique(clustering)
In [24]:
          array([-1, 0, 1, 2])
Out[24]:
In [16]:
          sns.pairplot(df)
          <seaborn.axisgrid.PairGrid at 0x7fafa7ff1d50>
Out[16]:
```



```
In [23]: df_pca = pd.DataFrame(PCA(n_components=0.9).fit_transform(df))
    display(df_pca)
    df_with_labels = df_pca.copy()
    df_with_labels["clusters"] = clustering
    display(df_with_labels)
    sns.scatterplot(df_with_labels, x=0, y=1, hue="clusters")

# # %%
# clustering.labels_
```

	0	1
0	2.879354	-1.528204
1	-0.237821	2.665336
2	2.934311	-1.010308
3	-0.149849	2.941895
4	-2.541960	-1.539241
•••		
745	-0.335449	3.151280
746	2.846703	-1.641857
747	3.071572	-1.350743
748	2.986945	-1.256748
749	2.817957	-1.611575

750 rows × 2 columns

	0	1	clusters
0	2.879354	-1.528204	0
1	-0.237821	2.665336	1
2	2.934311	-1.010308	0
3	-0.149849	2.941895	1
4	-2.541960	-1.539241	2
•••			•••
745	-0.335449	3.151280	1
746	2.846703	-1.641857	0
747	3.071572	-1.350743	0
748	2.986945	-1.256748	0
749	2.817957	-1.611575	0

750 rows × 3 columns

Out[23]: <AxesSubplot: xlabel='0', ylabel='1'>

19/09/2024, 15:42 Plots of DBSCAN data

