```
In [1]:
             import pandas as pd
              import numpy as np
             import seaborn as sns
             from sklearn.preprocessing import StandardScaler
             from sklearn.decomposition import PCA
             from sklearn.manifold import TSNE
             import matplotlib.pyplot as plt
          M df
                  = pd.read_csv("TSNE_data.csv")
In [2]:
             df
    Out[2]:
                   diagnosis radius_mean texture_mean perimeter_mean area_mean smoothness_mean
                0
                          Μ
                                    17.99
                                                  10.38
                                                                122.80
                                                                           1001.0
                                                                                             0.11840
                1
                          Μ
                                    20.57
                                                                132.90
                                                                           1326.0
                                                                                             0.08474
                                                  17.77
                 2
                          Μ
                                    19.69
                                                  21.25
                                                                130.00
                                                                            1203.0
                                                                                             0.10960
                3
                                    11.42
                                                  20.38
                                                                 77.58
                                                                            386.1
                                                                                             0.14250
                          Μ
                                                                                             0.10030
                 4
                          Μ
                                    20.29
                                                  14.34
                                                                135.10
                                                                            1297.0
                ...
                                                                    ...
                                                                               ...
                                    21.56
                                                  22.39
                                                                142.00
                                                                            1479.0
                                                                                             0.11100
               564
                          Μ
               565
                                                  28.25
                                                                131.20
                                                                            1261.0
                                                                                             0.09780
                          Μ
                                    20.13
               566
                                    16.60
                                                  28.08
                                                                108.30
                                                                            858.1
                                                                                             0.08455
                          Μ
               567
                          Μ
                                    20.60
                                                  29.33
                                                                140.10
                                                                            1265.0
                                                                                             0.11780
               568
                          В
                                     7.76
                                                  24.54
                                                                 47.92
                                                                             181.0
                                                                                             0.05263
              569 rows × 31 columns
          | df1 = df.drop(['diagnosis'],axis=1)
In [3]:
In [4]:

    df['diagnosis'].unique()

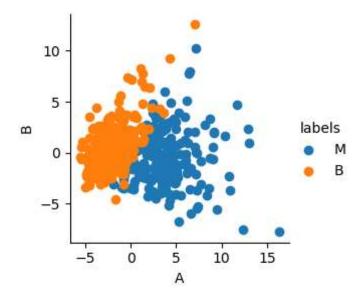
    Out[4]: array(['M', 'B'], dtype=object)

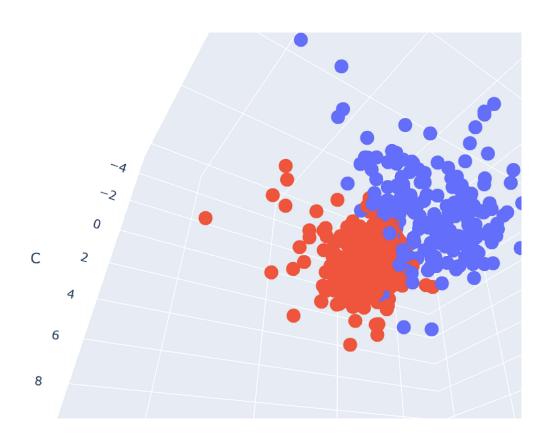
    df_labels = df.diagnosis

In [5]:
```

```
■ scaler = StandardScaler()

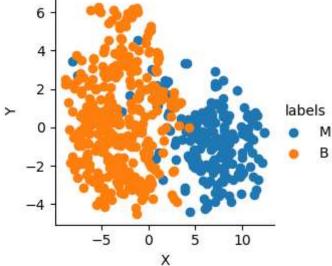
 In [6]:
             x scaled = scaler.fit transform(df1)
            x scaled
    Out[6]: array([[ 1.09706398, -2.07333501, 1.26993369, ..., 2.29607613,
                     2.75062224, 1.93701461],
                    [ 1.82982061, -0.35363241, 1.68595471, ..., 1.0870843 ,
                    -0.24388967, 0.28118999],
                   [1.57988811, 0.45618695, 1.56650313, ..., 1.95500035,
                     1.152255 , 0.20139121],
                    . . . ,
                    [0.70228425, 2.0455738, 0.67267578, ..., 0.41406869,
                    -1.10454895, -0.31840916],
                   [ 1.83834103, 2.33645719, 1.98252415, ..., 2.28998549,
                     1.91908301, 2.21963528],
                   [-1.80840125, 1.22179204, -1.81438851, ..., -1.74506282,
                     -0.04813821, -0.75120669]])
 In [7]: ▶ from sklearn.decomposition import PCA
            pca=PCA(n_components=3,random_state = 42)
             pca_data1=pca.fit_transform(x_scaled)
 In [8]:
          pca_data1
    Out[8]: array([[ 9.19283683, 1.94858333, -1.12316483],
                   [2.3878018, -3.76817147, -0.52929058],
                   [ 5.73389628, -1.07517381, -0.55174723],
                    [ 1.25617928, -1.90229684, 0.56272972],
                   [10.37479406, 1.67200998, -1.87702989],
                   [-5.4752433 , -0.6706363 , 1.49044698]])
          pca_data = np.column_stack((pca_data1, df_labels))
 In [9]:
            pca_df = pd.DataFrame(data=pca_data, columns=("A","B","C","labels"))
         ▶ print(pca df.head(10))
In [10]:
                      Α
                                 В
                                           C labels
             0 9.192837
                          1.948583 -1.123165
             1 2.387802 -3.768171 -0.529291
                                                  Μ
             2 5.733896 -1.075174 -0.551747
                                                  Μ
             3 7.122953 10.275589 -3.232788
                                                  Μ
            4 3.935302 -1.948071 1.389769
                                                  Μ
             5
               2.380247
                          3.949929 -2.934877
                                                  Μ
             6 2.238883 -2.690031 -1.639912
                                                  Μ
             7 2.143299 2.340244 -0.871947
                                                  Μ
             8 3.174924 3.391813 -3.119986
                                                  Μ
            9 6.351747 7.727174 -4.341915
                                                  Μ
```

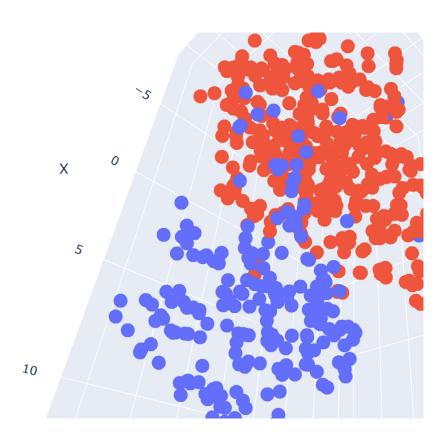




T-SNE

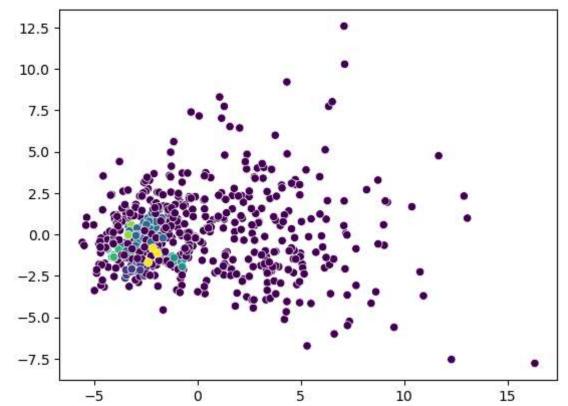
```
In [15]:
         ▶ t data1
   Out[15]: array([[10.08635 , -2.3840137 , 0.3878798 ],
                  [ 5.8013787 , 0.67179567, 4.693045 ],
                  [ 8.852475 , -1.3143523 , 1.7740929 ],
                  [ 3.9376345 , 3.347712 , 3.1025887 ],
                  [10.872616, -0.5450061, 0.23185869],
                  [-7.1865973 , 3.4996672 , 3.3574657 ]], dtype=float32)
        In [16]:
            tsne_df = pd.DataFrame(data=t_data, columns=("X", "Y","Z" ,"labels"))
In [17]:
         ▶ print(tsne_df.head(10))
                     Χ
                                       Z labels
              10.08635 -2.384014
                                  0.38788
            1 5.801379 0.671796 4.693045
                                              Μ
            2 8.852475 -1.314352 1.774093
                                              Μ
            3 8.106648 -2.284597 -4.001067
                                              Μ
            4 6.580023 -2.768508 3.192085
                                              Μ
            5 5.744499 -0.276084 -3.567405
                                              Μ
            6 5.474233 0.426576 3.133465
            7 5.112227 -1.646257 -3.011196
                                              Μ
            8 6.100016 -0.293798 -2.979045
                                              Μ
              7.99544 -0.751353 -3.964393
                                              Μ
         ▶ sns.FacetGrid(tsne_df, hue="labels" ).map(plt.scatter, 'X', 'Y').add_legender
In [18]:
            plt.show()
                 6
                 4
```

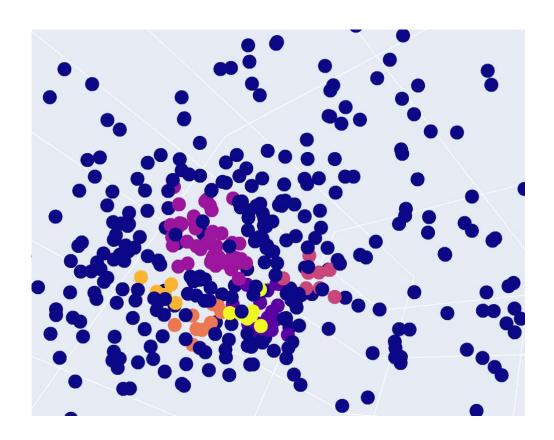




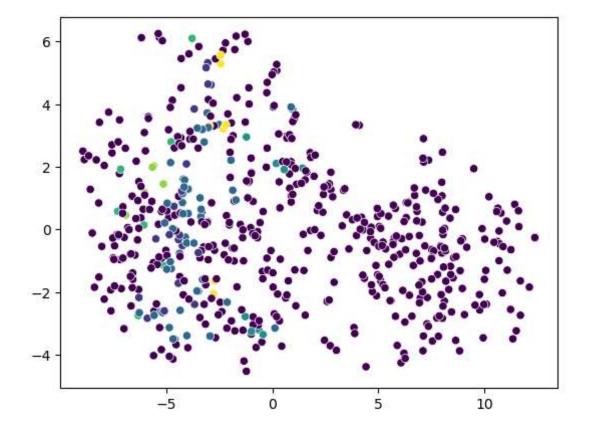
USE DBSCAN FOR CLUSTERING

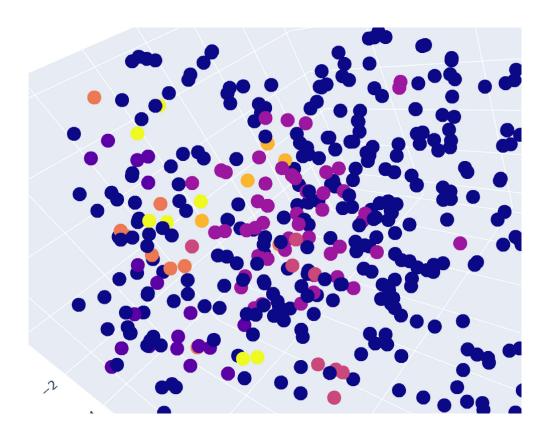
FOR PCA





Out[23]: <Axes: >





In []: **M**