## ML - EXP - 8

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## ▼ PCA : Principal Component Analysis

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
np.random.seed(23)
mu\_vec1 = np.array([0,0,0])
cov_mat1 = np.array([[1,0,0],[0,1,0],[0,0,1]])
class1 sample = np.random.multivariate normal(mu vec1, cov mat1, 20)
df = pd.DataFrame(class1_sample,columns=['feature1','feature2','feature3'])
df['target'] = 1
mu_vec2 = np.array([1,1,1])
cov_mat2 = np.array([[1,0,0],[0,1,0],[0,0,1]])
class2_sample = np.random.multivariate_normal(mu_vec2, cov_mat2, 20)
df1 = pd.DataFrame(class2_sample,columns=['feature1','feature2','feature3'])
df1['target'] = 0
df = df.append(df1,ignore_index=True)
df = df.sample(40)
     <ipython-input-24-b42d3f200777>:23: FutureWarning:
     The frame.append method is deprecated and will be removed from pandas in a future version. Use pandas.concat instead.
df.head()
          feature1 feature2 feature3 target
      2 -0.367548 -1.137460 -1.322148
```

```
        feature1
        feature2
        feature3
        target

        2
        -0.367548
        -1.137460
        -1.322148
        1

        34
        0.177061
        -0.598109
        1.226512
        0

        14
        0.420623
        0.411620
        -0.071324
        1

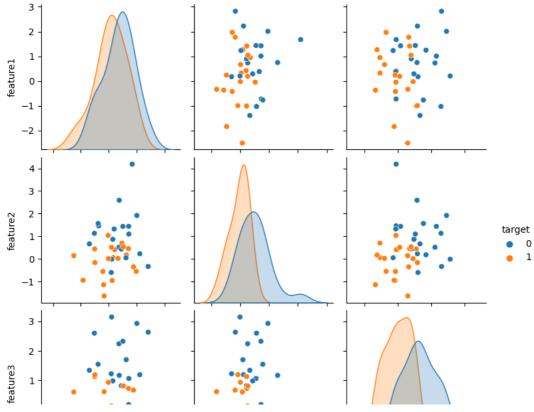
        11
        1.968435
        -0.547788
        -0.679418
        1

        12
        -2.506230
        0.146960
        0.606195
        1
```

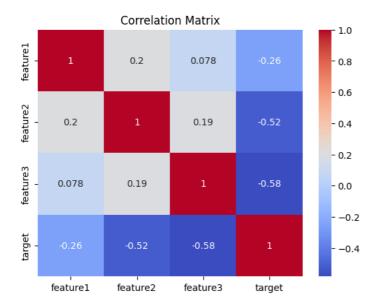
```
# Summary statistics
print("Summary Statistics:")
print(df.describe())
```

```
Summary Statistics:
       feature1 feature2
                            feature3
                                        target
count 40.000000 40.000000
                           40.000000
                                     40.00000
       0.433721 0.460790
                           0.667670
                                      0.50000
mean
std
       1.157915
                 1.060976
                           1.152079
                                      0.50637
      -2.506230 -1.632386 -1.322148
                                      0.00000
min
      -0.340600 -0.048988 -0.107669
                                      0.00000
25%
50%
       0.402744 0.423790
                           0.699508
                                      0.50000
75%
       1.254864
                 1.055595
                            1.200931
                                      1.00000
max
       2.823378
                 4.187503
                            3.150780
                                      1.00000
```

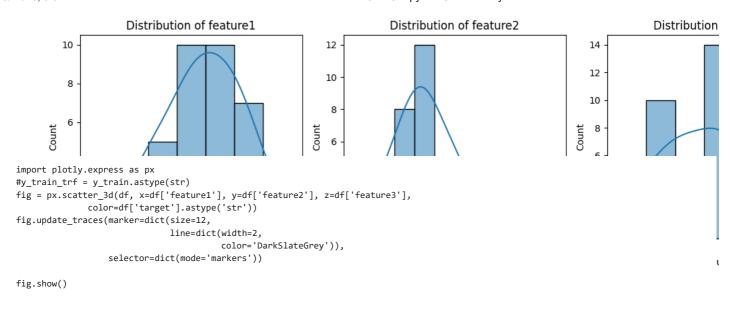
```
# Pairplot
sns.pairplot(df, hue='target', diag_kind='kde')
plt.show()
```



# Correlation matrix
correlation\_matrix = df.corr()
sns.heatmap(correlation\_matrix, annot=True, cmap='coolwarm')
plt.title("Correlation Matrix")
plt.show()



```
# Distribution of each feature
plt.figure(figsize=(12, 4))
for i, feature in enumerate(['feature1', 'feature2', 'feature3']):
    plt.subplot(1, 3, i + 1)
    sns.histplot(df[feature], kde=True)
    plt.title(f'Distribution of {feature}')
plt.tight_layout()
plt.show()
```



```
# Step 1 - Apply standard scaling
from \ sklearn.preprocessing \ import \ StandardScaler
scaler = StandardScaler()
df.iloc[:,0:3] = scaler.fit_transform(df.iloc[:,0:3])
# Step 2 - Find Covariance Matrix
covariance_matrix = np.cov([df.iloc[:,0],df.iloc[:,1],df.iloc[:,2]])
print('Covariance Matrix:\n', covariance_matrix)
     Covariance Matrix:
      [[1.02564103 0.20478114 0.080118 ]
      [0.20478114 1.02564103 0.19838882]
      [0.080118    0.19838882    1.02564103]]
# Step 3 - Finding EV and EVs
eigen_values, eigen_vectors = np.linalg.eig(covariance_matrix)
                                                        — + Code — + Text -
print("eigen_values : ", eigen_values , "\n", "eigen_vectors : ", eigen_vectors , "\n")
     eigen_values : [1.3536065 0.94557084 0.77774573]
      eigen_vectors : [[-0.53875915 -0.69363291 0.47813384]
      [-0.65608325 -0.01057596 -0.75461442]
      [-0.52848211 0.72025103 0.44938304]]
pc = eigen_vectors[0:2]
```

2 -0.546230 -0.2264953 -1.005401 -0.772965

selector=dict(mode='markers'))

1



