# **Dimensionality Reduction**

# **Principal Component Analysis**

Out of the m independent variables, PCA selects n < m new independent variables that explain the most variance **regardless of the dependent variable**. Since dependent variable is not considered, PCA is an unsupervised model.

```
In [1]:

1   import pandas as pd
2   import matplotlib.pyplot as plt
3   import numpy as np
4   find the pd.read_csv('datasets/diabetes.csv')
6   find the pd.read_csv('datasets/diabetes.csv')
```

## Out[1]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	ВМІ	DiabetesPedigreeFunction
0	6	148	72	35	0	33.6	0.62
1	1	85	66	29	0	26.6	0.35
2	8	183	64	0	0	23.3	0.67;
3	1	89	66	23	94	28.1	0.16 <sup>-</sup>
4	0	137	40	35	168	43.1	2.28
4							<b>)</b>

```
In [2]:

1 df.shape
```

# Out[2]:

(768, 9)

In [3]:

1 df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 768 entries, 0 to 767
Data columns (total 9 columns):
                            768 non-null int64
Pregnancies
Glucose
                            768 non-null int64
BloodPressure
                            768 non-null int64
                            768 non-null int64
SkinThickness
                            768 non-null int64
Insulin
                            768 non-null float64
BMI
DiabetesPedigreeFunction
                            768 non-null float64
                            768 non-null int64
Age
Outcome
                            768 non-null int64
```

dtypes: float64(2), int64(7)

memory usage: 54.1 KB

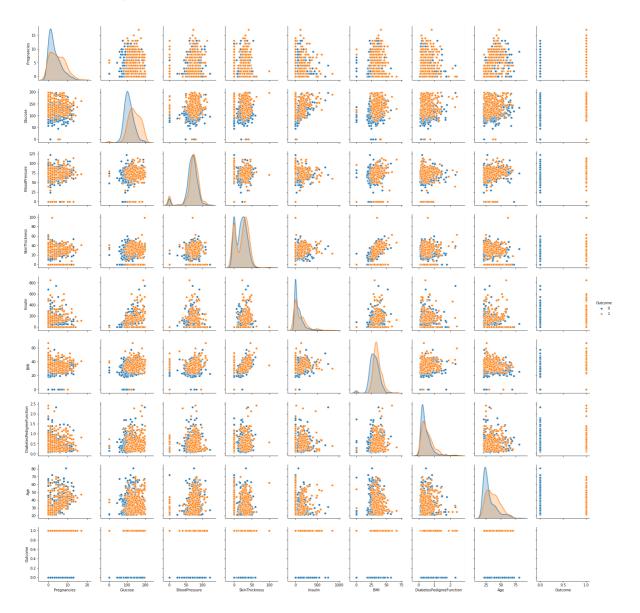
In [4]: 
▶

```
import seaborn as sns
sns.pairplot(df, hue = 'Outcome')
```

C:\Users\Jesus\Anaconda3\lib\site-packages\statsmodels\nonparametric\kde.py:
488: RuntimeWarning: invalid value encountered in true\_divide
 binned = fast\_linbin(X, a, b, gridsize) / (delta \* nobs)
C:\Users\Jesus\Anaconda3\lib\site-packages\statsmodels\nonparametric\kdetool
s.py:34: RuntimeWarning: invalid value encountered in double\_scalars
 FAC1 = 2\*(np.pi\*bw/RANGE)\*\*2

# Out[4]:

<seaborn.axisgrid.PairGrid at 0x1c588ebb240>



```
1 | X = df.drop('Outcome',axis = 1)
   y = df['Outcome']
In [6]:
                                                                                          M
 1 | # Split in training and testing
   from sklearn.model_selection import train_test_split
 3 | X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3,random_state =
In [7]:
                                                                                          M
    from sklearn.linear_model import LogisticRegression
 2 clf = LogisticRegression()
   clf.fit(X_train, y_train)
C:\Users\Jesus\Anaconda3\lib\site-packages\sklearn\linear_model\logistic.py:
433: FutureWarning: Default solver will be changed to 'lbfgs' in 0.22. Speci
fy a solver to silence this warning.
  FutureWarning)
Out[7]:
LogisticRegression(C=1.0, class_weight=None, dual=False, fit_intercept=True,
          intercept_scaling=1, max_iter=100, multi_class='warn',
          n_jobs=None, penalty='12', random_state=None, solver='warn',
          tol=0.0001, verbose=0, warm_start=False)
In [8]:
                                                                                          H
   y pred = clf.predict(X test)
In [9]:
                                                                                          H
    from sklearn.metrics import confusion_matrix
 2
    cm = confusion_matrix(y_test, y_pred)
 3
    cm
Out[9]:
array([[130, 17],
       [ 38, 46]], dtype=int64)
```

M

In [5]:

In [10]: ▶

```
1 clf.score(X_test, y_test)
```

### Out[10]:

0.7619047619047619

In [11]: ▶

- 1 # Split in training and testing
  - 2 | from sklearn.model\_selection import train\_test\_split
  - 3 X\_train, X\_test, y\_train, y\_test = train\_test\_split(X, y, test\_size=3,random\_state = 4

In [12]:

1 X\_train.head()

#### Out[12]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	DiabetesPedigreeFunct
690	8	107	80	0	0	24.6	0.
473	7	136	90	0	0	29.9	0
204	6	103	72	32	190	37.7	0.
97	1	71	48	18	76	20.4	0.
336	0	117	0	0	0	33.8	0.
4							<b>•</b>

In [13]:

- 1 # Always scale data for good results on PCA
- 2 | from sklearn.preprocessing import StandardScaler
- 3 X\_sca = StandardScaler()
- 4 | X\_train = X\_sca.fit\_transform(X\_train)
- 5 X\_test = X\_sca.transform(X\_test)
- C:\Users\Jesus\Anaconda3\lib\site-packages\sklearn\preprocessing\data.py:64
- 5: DataConversionWarning: Data with input dtype int64, float64 were all converted to float64 by StandardScaler.

return self.partial\_fit(X, y)

C:\Users\Jesus\Anaconda3\lib\site-packages\sklearn\base.py:464: DataConversi onWarning: Data with input dtype int64, float64 were all converted to float6 4 by StandardScaler.

return self.fit(X, \*\*fit\_params).transform(X)

C:\Users\Jesus\Anaconda3\lib\site-packages\ipykernel\_launcher.py:5: DataConversionWarning: Data with input dtype int64, float64 were all converted to float64 by StandardScaler.

....

```
In [14]:
                                                                                         H
   X_train[0:5,:]
Out[14]:
array([[ 1.23168349, -0.43604825, 0.56160593, -1.28775566, -0.69294238,
       -0.93586599, 1.15656737, 0.06294811],
       [0.93510248, 0.47024091, 1.07773825, -1.28775566, -0.69294238,
        -0.26437369, -0.79299869, 1.4242287],
       [0.63852147, -0.56105365, 0.14870008, 0.71926405, 0.95559722,
        0.72386024, -0.44895762, 1.84962888],
       [-0.84438358, -1.56109686, -1.09001747, -0.15880707, -0.03352654,
        -1.46799195, -0.45197552, -0.95801234],
       [-1.14096459, -0.12353475, -3.56745259, -1.28775566, -0.69294238,
        0.22974327, 1.38592809, 0.91374848])
                                                                                         H
In [15]:
 1 from sklearn.decomposition import PCA
 2
    pca = PCA()
 3
    pca.fit_transform(X_train)
 4 | explained_variance_ratio = pca.explained_variance_ratio_
    print(explained_variance_ratio)
[0.26171944 0.21677862 0.128557
                                 0.10958539 0.09494264 0.08534802
0.05253743 0.05053145]
In [16]:
                                                                                         M
 1 # Extract top 2 principal components
    pca = PCA(n_components=2)
    X_train = pca.fit_transform(X_train)
 4 X_test = pca.transform(X_test)
   explained_variance = pca.explained_variance_ratio_
    print(explained_variance)
[0.26171944 0.21677862]
In [17]:
                                                                                         H
    X_train
Out[17]:
array([[-0.78397591, 1.34563612],
       [-0.23714761, 2.44484824],
       [ 1.21249591, 0.96614483],
       [ 1.85090101, 0.80663508],
       [-1.74219774, -0.89044531],
       [-1.38214951, -0.25137068]])
```

```
In [18]:
    plt.scatter(X_train[:,0],X_train[:,1])
    plt.scatter(explained_variance[0],explained_variance[1],c = 'r')
 2
 3
    plt.show()
  4
  3
  2
  1
  0
 -1
 -2
In [19]:
                                                                                           H
    from sklearn.linear_model import LogisticRegression
 2 clf = LogisticRegression()
    clf.fit(X_train, y_train)
C:\Users\Jesus\Anaconda3\lib\site-packages\sklearn\linear_model\logistic.py:
433: FutureWarning: Default solver will be changed to 'lbfgs' in 0.22. Speci
fy a solver to silence this warning.
  FutureWarning)
```

#### Out[19]:

```
In [20]:

1 y_pred = clf.predict(X_test)
```

```
In [21]:
```

```
from sklearn.metrics import confusion_matrix
cm = confusion_matrix(y_test, y_pred)
cm
```

### Out[21]:

```
array([[3]], dtype=int64)
```

```
In [22]:
    from sklearn.metrics import accuracy_score
    accuracy_score(y_test, y_pred)
Out[22]:
1.0
                                                                                            M
In [23]:
 1 | clf.score(X_test, y_test)
Out[23]:
1.0
In [24]:
                                                                                            M
 1 # Split in training and testing
    from sklearn.model_selection import train_test_split
 3 | X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=3,random_state = 7
In [25]:
                                                                                            H
    from sklearn.discriminant_analysis import LinearDiscriminantAnalysis
    lda = LinearDiscriminantAnalysis(n_components=2)
    X_train = lda.fit_transform(X_train, y_train)
 4
    X_test = lda.transform(X_test)
 5
In [26]:
                                                                                            H
   X_train
       [ 8.93942725e-01],
       [ 3.91899748e-01],
       [-4.06773438e-01],
       [-6.04159603e-03],
       [ 8.65074171e-02],
       [-8.62645714e-01],
       [ 3.73737798e-01],
       [-1.16915125e+00],
       [ 5.38354761e-02],
       [ 4.39553771e-02],
       [ 1.73063543e+00],
       [-1.23637148e+00],
       [ 1.72140836e+00],
       [-9.21589191e-01],
       [-9.85956059e-01],
       [ 1.25719080e+00],
       [ 1.18119464e-01],
       [-1.18018629e+00],
       [-1.42486210e-01],
       [ 7.17493164e-01].
```

```
H
In [27]:
    from sklearn.linear_model import LogisticRegression
    from sklearn.metrics import confusion_matrix
    clf = LogisticRegression(random_state=0)
    clf.fit(X_train, y_train)
 5
    y_pred = clf.predict(X_test)
 7
    cm = confusion_matrix(y_test, y_pred)
 8
C:\Users\Jesus\Anaconda3\lib\site-packages\sklearn\linear_model\logistic.py:
433: FutureWarning: Default solver will be changed to 'lbfgs' in 0.22. Speci
fy a solver to silence this warning.
  FutureWarning)
Out[27]:
array([[1, 0],
       [0, 2]], dtype=int64)
In [28]:
                                                                                          M
 1 clf.score(X_test, y_test)
Out[28]:
```

1.0