```
# Load the diabetes dataset
from sklearn import tree
from pandas import read_csv
import os
import numpy as np
df = read_csv("diabetes.csv")
x = np.array(df.drop(["Outcome"], 1))
y = np.array(df["Outcome"])
     <ipython-input-5-33647d10cb48>:8: FutureWarning: In a future version of pandas all arguments of DataFrame.drop except for the argument
       x = np.array(df.drop(["Outcome"], 1))
У
     array([1, 0, 1, 0, 1, 0, 1, 0, 1, 1, 0, 1, 0, 1, 1, 1, 1, 1, 1, 0, 1, 0, 0,
            1, 1, 1, 1, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 1, 1, 0, 0, 0, 1,
            0, 1, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 1, 0,
           1, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0,
           1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 1,
           1, 1, 0, 0, 1, 1, 1, 0, 0, 0, 1, 0, 0, 0, 1, 1, 0, 0, 1, 1, 1,
            1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0,
            1, 1, 0, 0, 0, 1, 0, 0, 0, 1, 1, 0, 0, 0, 0, 1, 1, 0, 0, 0, 1,
           0, 1, 0, 1, 0, 0, 0, 0, 1, 1, 1, 1, 1, 0, 0, 1, 1, 0, 1, 0, 1,
            1, 1, 0, 0, 0, 0, 0, 0, 1, 1, 0, 1, 0, 0, 0, 1, 1, 1, 1, 0, 1, 1,
            1, 1, 0, 0, 0, 0, 0, 1, 0, 0, 1, 1, 0, 0, 0, 1, 1, 1, 1, 0, 0, 0,
           1, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 1, 0, 1, 0, 0,
            1, 0, 1, 0, 0, 1, 1, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1, 1,
            0, 1, 0, 0, 0, 1, 1, 1, 0, 0, 1, 0, 1, 0, 1, 1, 0, 1, 0, 0, 1, 0,
            1, 1, 0, 0, 1, 0, 1, 0, 0, 1, 0, 1, 0, 1, 1, 1, 0, 0, 1, 0, 1, 0,
            0, 0, 1, 0, 0, 0, 0, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0,
           0, 0, 0, 1, 1, 1, 0, 1, 1, 0, 0, 1, 0, 0, 1, 0, 0, 1, 1, 0, 0, 0,
            0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 1, 1, 0, 0, 1, 0, 0, 1, 0,
           0, 1, 0, 1, 1, 0, 1, 0, 1, 0, 1, 0, 1, 1, 0, 0, 0, 0, 1, 1, 0, 1,
           0, 1, 0, 0, 0, 0, 1, 1, 0, 1, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0,
            1, 0, 0, 1, 1, 1, 0, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 0,
            0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0,
           1, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0,
            1, 0, 0, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0,
            0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 1, 1, 1, 0, 0, 1, 1, 0, 0, 0,
           0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0,
            0, 0, 0, 0, 0, 1, 0, 1, 1, 0, 0, 0, 1, 0, 1, 0, 1, 0, 1, 0,
            0, 1, 0, 0, 1, 0, 0, 0, 0, 1, 1, 0, 1, 0, 0, 0, 0, 1, 1, 0, 1, 0,
            0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0,
            1, 0, 0, 0, 1, 0, 0, 0, 1, 1, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1,
            0, 1, 1, 1, 1, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 1, 0, 0, 1,
            0, 1, 0, 0, 0, 0, 0, 1, 0, 1, 0, 1, 0, 1, 1, 0, 0, 0, 0, 1, 1, 0,
           0, 0, 1, 0, 1, 1, 0, 0, 1, 0, 0, 1, 1, 0, 0, 1, 0, 0, 1, 0, 0, 0,
            0, 0, 0, 0, 1, 1, 1, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 1, 0, 0, 1, 0,
            1, 1, 1, 0, 0, 1, 1, 1, 0, 1, 0, 1, 0, 1, 0, 0, 0, 0, 1, 0])
# Display the first few rows of the dataset to inspect the data
print(df.head(10))
        Pregnancies Glucose BloodPressure SkinThickness Insulin
                                                                    BMI \
    a
               6.0
                        148
                                        72
                                                     35.0
                                                               0.0 33.6
    1
                NaN
                         85
                                        66
                                                     29.0
                                                               0.0
                                                                    26.6
                8.0
                                                      0.0
     2
                        183
                                        64
                                                               0.0
                                                                    23.3
     3
               1.0
                         89
                                        66
                                                     23.0
                                                              94.0
                                                                    28.1
     4
               0.0
                        137
                                        40
                                                     35.0
                                                             168.0
                                                                    43.1
     5
                5.0
                        116
                                        74
                                                      NaN
                                                               0.0 25.6
     6
                3.0
                         78
                                        50
                                                     32.0
                                                              88.0
                                                                    31.0
              10.0
                                         0
                                                                    35.3
     7
                        115
                                                      0.0
                                                               0.0
     8
               2.0
                        197
                                        70
                                                     45.0
                                                             543.0 30.5
     9
                8.0
                        125
                                        96
                                                      0.0
                                                               0.0
```

```
DiabetesPedigreeFunction
                                     Outcome
                               Age
0
                        0.627
1
                        0.351
                                31
                                           0
2
                        0.672
                                32
                                           1
3
                        0.167
                                21
                                           0
4
                        2.288
                                33
                                           1
5
                        0.201
                                           0
                                30
6
                        0.248
                                26
                                           1
7
                        0.134
                                29
                                           0
8
                        0.158
                                53
                                           1
                        0.232
                                54
                                           1
```

```
# finding the missing values
print(df.isnull().sum())
     Pregnancies
    Glucose
                                a
    BloodPressure
                                0
     SkinThickness
    Insulin
                                1
    BMT
     DiabetesPedigreeFunction
                                0
    Age
    Outcome
                                 0
     dtype: int64
# Handle missing values (replace missing values with the mean of the column)
from sklearn.impute import SimpleImputer
imputer = SimpleImputer(strategy='mean')
meanx = imputer.fit_transform(x)
meanx
                       , 148.
                                                     , ..., 33.6
     array([[ 6.
                                         72.
                         , 50.
              0.627
            [ 3.84876141, 85.
                                         66.
                                                     , ..., 26.6
                       , 31.
              0.351
                                      ],
                        , 183.
                                                     , ..., 23.3
              8.
                                          64.
                                      ,
],
                        , 32.
              0.672
                                                     , ..., 26.2
                        , 121.
                                        72.
                        , 30.
              0.245
                                      ],
                        , 126.
                                                     , ..., 30.1
            [ 1.
                                         60.
              0.349
                        , 47.
                                       ],
                        , 93.
                                                     , ..., 30.4
              1.
                                          70.
            [
                        , 23.
              0.315
                                       ]])
# Min-Max Scaling (Normalization to a range of [0, 1])
from sklearn.preprocessing import MinMaxScaler, StandardScaler
minmax_scaler = MinMaxScaler()
x_minmax = minmax_scaler.fit_transform(x)
x_minmax
     array([[0.35294118, 0.74371859, 0.59016393, ..., 0.50074516, 0.23441503,
            0.48333333],
                   nan, 0.42713568, 0.54098361, ..., 0.39642325, 0.11656704,
            0.16666667],
            [0.47058824, 0.91959799, 0.52459016, ..., 0.34724292, 0.25362938,
            0.18333333],
            [0.29411765, 0.6080402, 0.59016393, ..., 0.390462, 0.07130658,
            0.15
                     ],
            [0.05882353, 0.63316583, 0.49180328, ..., 0.4485842, 0.11571307,
             0.43333333],
            [0.05882353, 0.46733668, 0.57377049, ..., 0.45305514, 0.10119556,
            0.03333333]])
# Z-score Normalization (Standardization with mean=0 and std=1)
zscore_scaler = StandardScaler()
z_zscore = zscore_scaler.fit_transform(x)
z_zscore
    \verb"array" ([[ \ 0.63872696, \ \ 0.84832379, \ \ 0.14964075, \ \ldots, \ \ 0.20401277,
             0.46849198, 1.4259954],
                   nan, -1.12339636, -0.16054575, ..., -0.68442195,
            -0.36506078, -0.19067191],
            [ 1.23254948, 1.94372388, -0.26394125, ..., -1.10325546,
             0.60439732, -0.10558415],
            [ 0.3418157 , 0.00330087, 0.14964075, ..., -0.73518964,
             -0.68519336, -0.27575966],
            [-0.84582934, 0.1597866, -0.47073225, ..., -0.24020459,
             -0.37110101, 1.17073215],
            [-0.84582934, -0.8730192, 0.04624525, ..., -0.20212881,
             -0.47378505, -0.87137393]])
# Robust Scaler (Robust to outliers)
from sklearn.preprocessing import RobustScaler
robust_scaler = RobustScaler()
x_robust = robust_scaler.fit_transform(x)
x_robust
```

```
[ nan, -0.77575758, -0.333333333, ..., -0.58064516, -0.05620915, 0.11764706], [ 1. , 1.6 , -0.44444444, ..., -0.93548387, 0.78300654, 0.17647059],
           [ 0.4 , 0.0969697 , 0. , ..., -0.62365591, -0.333333333 , 0.05882353], [-0.4 , 0.21818182, -0.666666667, ..., -0.20430108,
           -0.06143791, 1.05882353],

[-0.4 , -0.58181818, -0.111111111, ..., -0.17204301,

-0.1503268 , -0.35294118]])
```

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