y_train.shape

→ (712,)

Single Layer Perceptron

```
!pip install tensorflow
 Requirement already satisfied: tensorflow in /usr/local/lib/python3.10/dist-packages (2.17.0)
       Requirement already satisfied: absl-py>=1.0.0 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (1.4.0)
       Requirement already satisfied: astunparse>=1.6.0 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (1.6.3)
       Requirement already satisfied: flatbuffers>=24.3.25 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (24.3.25)
       Requirement already satisfied: gast!=0.5.0,!=0.5.1,!=0.5.2,>=0.2.1 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (0.6.0)
       Requirement already satisfied: google-pasta>=0.1.1 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (0.2.0)
       Requirement already satisfied: h5py>=3.10.0 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (3.11.0)
       Requirement already satisfied: libclang>=13.0.0 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (18.1.1)
       Requirement already \ satisfied: \ ml-dtypes < 0.5.0, >= 0.3.1 \ in \ /usr/local/lib/python 3.10/dist-packages \ (from \ tensorflow) \ (0.4.1)
       Requirement already satisfied: opt-einsum>=2.3.2 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (3.4.0)
       Requirement already satisfied: packaging in /usr/local/lib/python3.10/dist-packages (from tensorflow) (24.1)
       Requirement already satisfied: protobuf!=4.21.0,!=4.21.1,!=4.21.2,!=4.21.3,!=4.21.4,!=4.21.5,<5.0.0dev,>=3.20.3 in /usr/local/lib/python
       Requirement already satisfied: requests<3,>=2.21.0 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (2.32.3)
       Requirement already satisfied: setuptools in /usr/local/lib/python3.10/dist-packages (from tensorflow) (71.0.4)
       Requirement already satisfied: six>=1.12.0 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (1.16.0)
       Requirement already satisfied: termcolor>=1.1.0 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (2.5.0)
       Requirement already satisfied: typing-extensions>=3.6.6 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (4.12.2)
       Requirement already satisfied: wrapt>=1.11.0 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (1.16.0)
       Requirement already satisfied: grpcio<2.0,>=1.24.3 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (1.64.1)
       Requirement already satisfied: tensorboard<2.18,>=2.17 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (2.17.0)
       Requirement already satisfied: keras>=3.2.0 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (3.4.1)
       Requirement already satisfied: tensorflow-io-gcs-filesystem>=0.23.1 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (0.37.1
       Requirement already satisfied: numpy<2.0.0,>=1.23.5 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (1.26.4)
       Requirement already satisfied: wheel<1.0,>=0.23.0 in /usr/local/lib/python3.10/dist-packages (from astunparse>=1.6.0->tensorflow) (0.44.
       Requirement already satisfied: rich in /usr/local/lib/python3.10/dist-packages (from keras>=3.2.0->tensorflow) (13.9.2)
       Requirement already satisfied: namex in /usr/local/lib/python3.10/dist-packages (from keras>=3.2.0->tensorflow) (0.0.8)
       Requirement already satisfied: optree in /usr/local/lib/python3.10/dist-packages (from keras>=3.2.0->tensorflow) (0.13.0)
       Requirement already satisfied: charset-normalizer<4,>=2 in /usr/local/lib/python3.10/dist-packages (from requests<3,>=2.21.0->tensorflow
       Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.10/dist-packages (from requests<3,>=2.21.0->tensorflow) (3.10)
       Requirement already satisfied: urllib3<3,>=1.21.1 in /usr/local/lib/python3.10/dist-packages (from requests<3,>=2.21.0->tensorflow) (2.2
       Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.10/dist-packages (from requests<3,>=2.21.0->tensorflow) (202
       Requirement already satisfied: markdown>=2.6.8 in /usr/local/lib/python3.10/dist-packages (from tensorboard<2.18,>=2.17->tensorflow) (3.
       Requirement already satisfied: tensorboard-data-server < 0.8.0, >= 0.7.0 in /usr/local/lib/python 3.10/dist-packages (from tensorboard < 2.18, >= 0.7.0 in /usr/local/lib/python >= 0.7.0 in /usr/local/lib/pyth
       Requirement already satisfied: werkzeug>=1.0.1 in /usr/local/lib/python3.10/dist-packages (from tensorboard<2.18,>=2.17->tensorflow) (3.
       Requirement already satisfied: MarkupSafe>=2.1.1 in /usr/local/lib/python3.10/dist-packages (from werkzeug>=1.0.1->tensorboard<2.18,>=2.
       Requirement already satisfied: markdown-it-py>=2.2.0 in /usr/local/lib/python3.10/dist-packages (from rich->keras>=3.2.0->tensorflow) (3
       Requirement already satisfied: pygments<3.0.0,>=2.13.0 in /usr/local/lib/python3.10/dist-packages (from rich->keras>=3.2.0->tensorflow)
       Requirement already satisfied: mdurl~=0.1 in /usr/local/lib/python3.10/dist-packages (from markdown-it-py>=2.2.0->rich->keras>=3.2.0->te
# Import necessary libraries
import numpy as np
import pandas as pd
import tensorflow as tf
import matplotlib.pyplot as plt
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense
from sklearn.datasets import load_iris
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
df = pd.read csv("/content/titanic cleaned.csv")
X = df.drop('Survived',axis=1)
y = df['Survived']
X_train,X_test,y_train,y_test = train_test_split(X,y,test_size=0.2,random_state=42)
X train.shape
 \rightarrow (712, 3)
X test.shape

→ (179, 3)
```

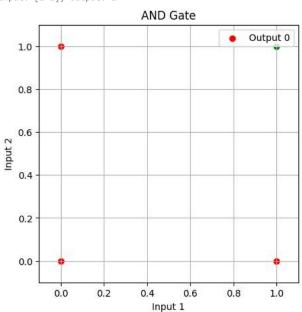
```
y_test.shape

→ (179,)

scaler = StandardScaler()
X_train = scaler.fit_transform(X_train)
X test = scaler.transform(X test)
model = Sequential([Dense(1, input_dim=X_train.shape[1], activation='sigmoid')])
🚁 /usr/local/lib/python3.10/dist-packages/keras/src/layers/core/dense.py:87: UserWarning: Do not pass an `input_shape`/`input_dim` argumen
       super().__init__(activity_regularizer=activity_regularizer, **kwargs)
    4
model.compile(optimizer='adam', loss='binary_crossentropy', metrics=['accuracy'])
# Train the model
model.fit(X_train, y_train, epochs=100, batch_size=10, verbose=1)
                              — 0s 2ms/step - accuracy: 0.7874 - loss: 0.4411
    72/72
     Epoch 73/100
                               - 0s 2ms/step - accuracy: 0.8109 - loss: 0.4251
     72/72
     Epoch 74/100
     72/72 ·
                              – 0s 2ms/step - accuracy: 0.7857 - loss: 0.4457
     Epoch 75/100
     72/72
                               - 0s 3ms/step - accuracy: 0.7857 - loss: 0.4665
     Epoch 76/100
     72/72 •
                               - 0s 3ms/step - accuracy: 0.8005 - loss: 0.4517
     Epoch 77/100
     72/72
                               Os 3ms/step - accuracy: 0.7694 - loss: 0.4903
     Epoch 78/100
     72/72 -
                              - 0s 3ms/step - accuracy: 0.7694 - loss: 0.4935
     Epoch 79/100
                               - 0s 3ms/step - accuracy: 0.8048 - loss: 0.4444
     72/72
     Epoch 80/100
     72/72
                               - 0s 3ms/step - accuracy: 0.8061 - loss: 0.4464
     Epoch 81/100
     72/72
                               - Os 3ms/step - accuracy: 0.7666 - loss: 0.4704
     Epoch 82/100
     72/72
                               Os 2ms/step - accuracy: 0.7793 - loss: 0.4676
     Epoch 83/100
     72/72
                               - 0s 3ms/step - accuracy: 0.7779 - loss: 0.4657
     Epoch 84/100
     72/72
                              — 0s 2ms/step - accuracy: 0.7893 - loss: 0.4534
     Epoch 85/100
     72/72
                               - 0s 1ms/step - accuracy: 0.8144 - loss: 0.4454
     Epoch 86/100
     72/72
                               - 0s 1ms/step - accuracy: 0.7650 - loss: 0.4899
     Epoch 87/100
     72/72
                               - 0s 2ms/step - accuracy: 0.7831 - loss: 0.4611
     Epoch 88/100
     72/72
                               - 0s 1ms/step - accuracy: 0.7719 - loss: 0.4975
     Epoch 89/100
     72/72
                               - 0s 2ms/step - accuracy: 0.7917 - loss: 0.4498
     Epoch 90/100
     72/72
                              - 0s 1ms/step - accuracy: 0.7938 - loss: 0.4673
     Epoch 91/100
     72/72
                               - 0s 1ms/step - accuracy: 0.7957 - loss: 0.4452
     Epoch 92/100
     72/72
                               - 0s 2ms/step - accuracy: 0.7725 - loss: 0.4714
     Epoch 93/100
     72/72
                               - 0s 1ms/step - accuracy: 0.8002 - loss: 0.4536
     Epoch 94/100
     72/72
                               - 0s 2ms/step - accuracy: 0.7982 - loss: 0.4128
     Epoch 95/100
     72/72 -
                               - 0s 1ms/step - accuracy: 0.7841 - loss: 0.4563
     Epoch 96/100
     72/72
                               - 0s 2ms/step - accuracy: 0.7939 - loss: 0.4470
     Epoch 97/100
     72/72
                               - 0s 1ms/step - accuracy: 0.8129 - loss: 0.4280
     Epoch 98/100
     72/72
                               Os 2ms/step - accuracy: 0.7833 - loss: 0.4763
     Epoch 99/100
     72/72
                               - 0s 1ms/step - accuracy: 0.8021 - loss: 0.4610
     Fnoch 100/100
     72/72 ·
                               - 0s 2ms/step - accuracy: 0.7813 - loss: 0.4707
     <keras.src.callbacks.history.History at 0x7adcbc3d4460>
```

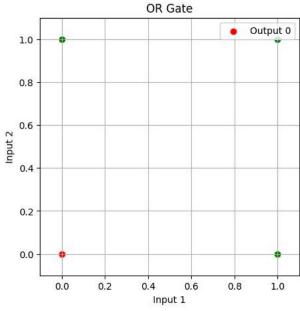
```
# Evaluate the model on test data
loss, accuracy = model.evaluate(X_test, y_test)
print(f'Test Accuracy: {accuracy*100:.2f}%')
→ 6/6 -
                             — 0s 3ms/step - accuracy: 0.8217 - loss: 0.4159
     Test Accuracy: 81.01%
def sigmoid(x):
    return 1 / (1 + np.exp(-x))
class Perceptron:
    def __init__(self, weights, bias):
        self.weights = weights
        self.bias = bias
    def predict(self, inputs):
        # Calculate weighted sum
        z = np.dot(inputs, self.weights) + self.bias
        # Apply sigmoid activation function
        return sigmoid(z)
# Input for AND, OR gates (binary input pairs)
inputs = np.array([[0, 0], [0, 1], [1, 0], [1, 1]])
# Input for NOT gate (single binary input)
inputs_not = np.array([[0], [1]])
# AND gate weights and bias (both inputs need to be 1 to output 1)
weights_and = np.array([1, 1])
bias_and = 1 # Adjusting for threshold
# OR gate weights and bias (one input being 1 is enough for output 1)
weights_or = np.array([0, 1])
bias_or = 1 # Adjusting for threshold
# NOT gate weights and bias (single input)
weights_not = np.array([1])
bias_not = 1 # Adjusting for threshold
perceptron and = Perceptron(weights and, bias and)
perceptron_or = Perceptron(weights_or, bias_or)
perceptron_not = Perceptron(weights_not, bias_not)
# AND gate results
print("AND Gate")
for x in inputs:
    output = perceptron_and.predict(x)
    print(f"Input: {x}, Output: {round(output)}")
 → AND Gate
     Input: [0 0], Output: 0
     Input: [0 1], Output: 0
     Input: [1 0], Output: 0
Input: [1 1], Output: 1
# OR gate results
print("\nOR Gate")
for x in inputs:
    output = perceptron_or.predict(x)
    print(f"Input: {x}, Output: {round(output)}")
     OR Gate
     Input: [0 0], Output: 0
     Input: [0 1], Output: 1
Input: [1 0], Output: 1
     Input: [1 1], Output: 1
```

```
# NOT gate results
print("\nNOT Gate")
for x in inputs_not:
   output = perceptron_not.predict(x)
   print(f"Input: {x}, Output: {round(output)}")
     NOT Gate
     Input: [0], Output: 1
     Input: [1], Output: 0
def plot_logic_gate(inputs, outputs, gate_name):
   # Define figure
   plt.figure(figsize=(5, 5))
   # Plot the points: Red for 0 output, Green for 1 output
   for i, point in enumerate(inputs):
        if outputs[i] == 0:
           plt.scatter(point[0], point[1] if len(point) > 1 else 0, color='red', label='Output 0' if i == 0 else "")
           plt.scatter(point[0], point[1] if len(point) > 1 else 0, color='green', label='Output 1' if i == 0 else "")
   # Set plot details
   plt.title(f'{gate_name} Gate')
   plt.xlabel('Input 1')
   if len(inputs[0]) > 1:
       plt.ylabel('Input 2')
   else:
        plt.yticks([0])
   plt.xlim(-0.1, 1.1)
   plt.ylim(-0.1, 1.1)
   plt.grid(True)
   plt.legend()
   plt.show()
# AND Gate
print("AND Gate")
and outputs = []
for x in inputs:
   output = round(perceptron_and.predict(x))
   and_outputs.append(output)
   print(f"Input: {x}, Output: {output}")
plot_logic_gate(inputs, and_outputs, "AND")
→ AND Gate
     Input: [0 0], Output: 0
    Input: [0 1], Output: 0
Input: [1 0], Output: 0
     Input: [1 1], Output: 1
                                 AND Gate
```



```
# OR Gate
print("\nOR Gate")
or_outputs = []
for x in inputs:
    output = round(perceptron_or.predict(x))
    or_outputs.append(output)
    print(f"Input: {x}, Output: {output}")
plot_logic_gate(inputs, or_outputs, "OR")

OR Gate
    Input: [0 0], Output: 0
    Input: [0 1], Output: 1
    Input: [1 0], Output: 1
    Input: [1 0], Output: 1
```



```
# NOT Gate
print("\nNOT Gate")
not_outputs = []
for x in inputs_not:
    output = round(perceptron_not.predict(x))
    not_outputs.append(output)
    print(f"Input: {x}, Output: {output}")
# For NOT gate, we only plot single input
plot_logic_gate(inputs_not, not_outputs, "NOT")
```

₹

NOT Gate

Input: [0], Output: 1
Input: [1], Output: 0

NOT Gate

