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CODE:
import numpy as np
import matplotlib.pyplot as plt
# Define the perceptron class
class Perceptron:
  def init (self, learning rate=0.1, num epochs=100):
     self.learning rate = learning rate
     self.num epochs = num epochs
  # Train the perceptron on the input data
  def train(self, X, y):
     num_samples, num features = X.shape
     self.weights = np.zeros(num features + 1)
     X = np.hstack((X, np.ones((num samples, 1))))
     for epoch in range(self.num epochs):
        for i in range(num samples):
          y pred = np.sign(np.dot(X[i], self.weights))
          if y pred != y[i]:
             self.weights += self.learning rate * v[i] * X[i]
  # Predict the output for a given input
  def predict(self, X):
     X = np.hstack((X, np.ones((X.shape[0], 1))))
     return np.sign(np.dot(X, self.weights))
# Define the input data
X = \text{np.array}([[1, 2], [2, 3], [3, 1], [4, 3], [5, 2], [6, 3]))
y = np.array([-1, -1, -1, 1, 1, 1])
# Create and train the perceptron
perceptron = Perceptron(learning rate=0.1, num epochs=100)
perceptron.train(X, y)
# Plot the input data and the decision boundary
plt.scatter(X[y == -1, 0], X[y == -1, 1], color='blue', marker='o', label='-1')
plt.scatter(X[y == 1, 0], X[y == 1, 1], color='red', marker='x', label='1')
x \min_{x} x \max = X[:, 0].\min() - 1, X[:, 0].\max() + 1
y \min_{x \in X} y \max_{x \in X} = X[:, 1].\min() - 1, X[:, 1].\max() + 1
xx, yy = np.meshgrid(np.arange(x min, x max, 0.1), np.arange(y min, y max, 0.1))
Z = perceptron.predict(np.c_[xx.ravel(), yy.ravel()])
Z = Z.reshape(xx.shape)
plt.contourf(xx, yy, Z, alpha=0.3, levels=[-1, 0, 1], colors=['blue', 'red'])
plt.xlim(x min, x max)
plt.ylim(y min, y max)
plt.xlabel('X1')
plt.ylabel('X2')
plt.legend(loc='upper left')
plt.show()
```

## **OUTPUT:**

