1. Implement the perceptron learning single layer algorithm by initializing the weights and threshold. Execute the code and check, how many iterations are needed, until the network coverage.

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Explanation:

Perceptron consist of four parts-

- a. Input values or one input layer: The input layer of a perceptron is made of artificial input neurons and brings the initial data into the system for further processing.
- b. Weights: Weight represents the strength or dimension of the connection between units. If the weight from node 1 to node 2 has the greater quantity, then neuron 1 has greater influence over neuron 2. How much influence of the input will have on the output, is determined by weight.
- c. Bias is similar to the intercept added in a linear equation. It is an additional parameter which task is to adjust the output along with the weighted sum of the inputs to the neuron.
- d. Activation Function: A neuron should be activated or not, determined by an activation function. It calculates a weighted sum and further adds bias to the given result.

```
In [2]: | import numpy as np
        theta = 1
        epoch = 3
        class Perceptron(object):
            def init (self, input size, learning rate=0.2):
                self.learning rate = learning rate
                self.weights = np.zeros(input size + 1) # zero init for weights and bias
            def predict(self, x):
                 return (np.dot(x, self.weights[1:]) + self.weights[0]) # X.W + B
            def train(self, x, y, weights):
                for inputs, label in zip(x, y):
                    net in = self.predict(inputs)
                    if net in > theta:
                        y out = 1
                    elif net in < -theta:</pre>
                        y \text{ out} = -1
                    else:
                        y out = 0
                    if y out != label: # updating the net on incorrect prediction
                        self.weights[1:] += self.learning_rate * label * inputs # W = alpha * Y * X
                         self.weights[0] += self.learning_rate * label # B = alpha * Y
                    print(inputs, net_in, label, y_out, self.weights)
        if __name__
                    == "__main__":
            x = []
            x.append(np.array([1, 1]))
            x.append(np.array([1, -1]))
            x.append(np.array([-1, 1]))
            x.append(np.array([-1, -1]))
            y = np.array([1, -1, -1, -1])
            perceptron = Perceptron(2)
            for i in range(epoch):
                print("Epoch",i)
                print("X1 X2 ", " Net ", " T ", " Y ", " B Weights")
                weights = perceptron.weights
                print("Initial Weights", weights)
                perceptron.train(x, y, weights)
        Epoch 0
```

```
X1 X2 Net T Y B Weights
Initial Weights [0. 0. 0.]
[1 1] 0.0 1 0 [0.2 0.2 0.2]
[1 -1] 0.2 -1 0 [0. 0. 0.4]
[-1 \ 1] \ 0.4 \ -1 \ 0 \ [-0.2 \ 0.2 \ 0.2]
Epoch 1
X1 X2 Net T Y B Weights
Initial Weights [-0.4 0.4 0.4]
[1 1] 0.4 1 0 [-0.2 0.6 0.6]
[ 1 -1] -0.2 -1 0 [-0.4 0.4 0.8]
[-1 1] -5.551115123125783e-17 -1 0 [-0.6 0.6 0.6]
Epoch 2
X1 X2  Net T  Y  B  Weights
Initial Weights [-0.6 0.6 0.6]
[1 1] 0.600000000000001 1 0 [-0.4 0.8 0.8]
[-1 1] -0.20000000000000018 -1 0 [-0.8 0.8 0.8]
```