

Experiment 3 - Multicategory Single Layered Classifier

Aim:

To implement RDPTA and MCPTA training algorithms for single layered neural networks

Problem Statement:

Implement RDPTA (R-Category Discrete Perceptron Training Algorithm) and MCPTA (Multi-Category Perceptron Training Algorithm) for the given problem:

Class = 1	for $\mathbf{x} = [1 \ 0 \ 0]^t$,	$[1 \ 1 \ 0]^t$
Class = 2	for $\mathbf{x} = [1 \ 0 \ 1]^t$,	$[1 \ 1 \ 1]^t$
Class = 3	for $\mathbf{x} = [0 \ 1 \ 0]^t$,	$[0 \ 1 \ 1]^t$
Class = 4	for $\mathbf{x} = [0 \ 0 \ 0]^t$,	$[0 \ 0 \ 1]^t$

Tool/ Language:

Python, numpy

RCPTA Code:

class RCPTA:

def __init__(self):

4 x 4 weight matrix of random values self.W =
np.zeros((4,4))

def predict(self, X_test):

res = np.dot(X_test, self.W) return
np.where(res >= 0, 1, 0)

def fit(self, X_train, d_train, learning_rate, epochs):

print("Initial Weights: ", self.W)

```

for k in range(epochs):
    print("##### Epoch ", k, "
#####")
for i in range(0,8): E = 0 print("##### X ", i, "
#####") print("X_train: ", X_train[i])
    for j in range(0,4):

        print("Starting Weights: ", str(self.W[j])) z =

        np.dot(self.W[j].T, X_train[i])

        print("Z: ", z) y_net =
        uni_activation(z)

        print("Y: ", str(y_net))

        e = d_train[i][j] - y_net print("Error:
        ", E)

        E = E + 0.5*(e)**2

        self.W[j] = self.W[j] + learning_rate * e * X_train[i] print("New

        Weights: ", str(self.W[j]))

    if(E == 0):
        print("Model fit complete")
        print("Num Epochs: ", (k+1))
        print("Final Error: ", E) break
print("Final Weights: ", str(self.W))

```

Output:

```

Num Epochs: 4
Final Error: 0.0
Final Weights: [[ 1. -1. -2. -1.]
 [ 2. -1.  1. -2.]
 [-1.  1.  0. -1.]
 [-1. -1.  0.  0.]]

```

MCPTA Code:

class MCPTA:

def __init__(self):

4 x 4 weight matrix of random values self.W =
np.zeros((4,4))

def predict(self, X_test):

res = np.dot(X_test, self.W) return
np.where(res >= 0, 1, 0)

def fit(self, X_train, d_train, learning_rate, epochs):

print("Initial Weights: ", self.W)

for k in range(epochs):

print("##### Epoch ", (k+1), "

#####")

for i in range(0,8): E = 0 print("##### X ", i, "

#####") print("X_train: ", X_train[i])

for j in range(0,4):

print("Starting Weights: ", str(self.W[j])) z =

np.dot(self.W[j].T, X_train[i])

print("Z: ", z) y_net

= sigmoid(z)

```
print("Y: ", str(y_net)) e
= d_train[i][j] - y_net E =
E + 0.5*(e)**2
print("Error: ", E)

self.W[j] = self.W[j] + (learning_rate * e * y_net * (1 - y_net) * X[i]) print("New
Weights: ", str(self.W[j]))

if(E < 0.25):
    print("Num Epochs: ", (k+1))
    print("Final Error: ", E) break
print("Final Weights: ", str(self.W))
```

Output:

```
Num Epochs: 8
Final Error: 0.24645643966221747
Final Weights: [[ 0.98527061 -0.26500754 -1.55818758 -0.75591037]
 [ 1.03784349 -0.50210392  0.78387578 -1.43160128]
 [-1.43941956  1.26018238 -0.16700673 -0.75365022]
 [-1.32841085 -1.20028229  0.17104602  0.12224247]]
```

Conclusion:

Thus, we have solved the given problem using MCPTA and RCPTA. We have found the final weights which help in solving the problem.