Single Layer Perceptron 6var

November 30, 2022

```
[38]: import pandas as pd
      import numpy as np
      import math
      np.set_printoptions(suppress=True)
      df = pd.read_csv('LogRes_Train.csv')
      df = df.sample(frac = 1)
      df.head()
[38]:
           Latitude Longitude Altitude min Temo Max Temp Sunshine Hour \
              21.09
                         79.07
                                               13.9
                                                         27.1
      110
                                      311
                                                                     0.650000
              22.39
      89
                         88.27
                                        6
                                               24.2
                                                         30.6
                                                                     0.600000
              23.35
                         85.33
                                               23.4
                                                         27.7
      49
                                      654
                                                                     3.500000
              17.36
                                               18.8
      164
                         78.46
                                      536
                                                         31.4
                                                                     0.728571
              12.57
                         77.38
                                               17.1
      201
                                      897
                                                         29.3
                                                                     0.742857
           Solar Radiation
      110
      89
                         0
                         0
      49
      164
                         1
      201
[39]: def normalize(X):
          X = X.copy()
          if len(X.shape) < 2:</pre>
              raise Exception("Enter a dataset or 2D array into normalize function")
          for col in range(X.shape[1]):
              X[:,col] = (X[:,col] - min(X[:,col]))/(max(X[:,col])-min(X[:,col]))
          return X
[40]: x_train = df.iloc[:,:6]
      x_train = x_train.to_numpy()
      \# x_train = normalize(x_train, axis = 1)
      x_train = normalize(x_train)
      x train
```

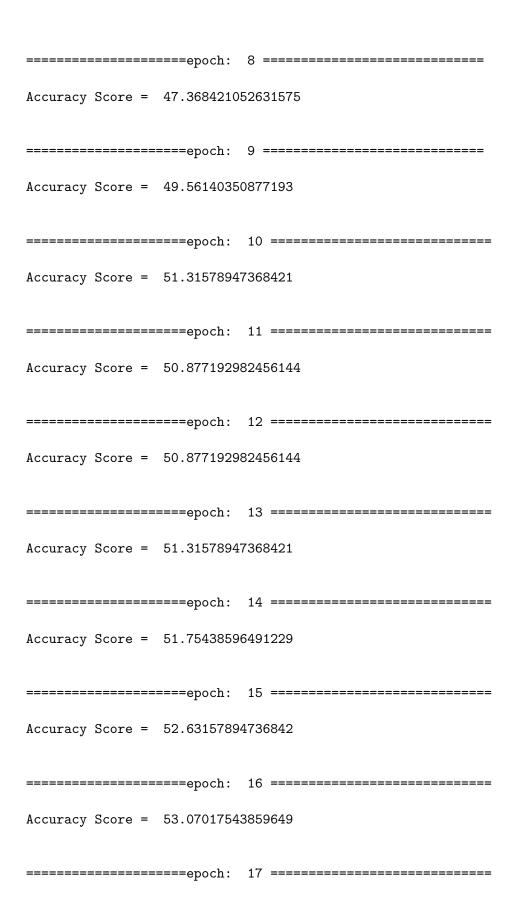
```
[40]: array([[0.49651163, 0.33641225, 0.17881944, 0.48366013, 0.61956522,
              0.02094595],
             [0.54689922, 0.78366553, 0.00231481, 0.82026144, 0.71467391,
              0.01621622],
             [0.58410853, 0.64073894, 0.37731481, 0.79411765, 0.63586957,
             0.29054054],
             [0.72248062, 0.17841517, 0.24826389, 0.83333333, 0.77717391,
             0.02297297],
             [0.27945736, 0.08070005, 0.00289352, 0.91503268, 0.76086957,
             0.02635135],
             [0.58410853, 0.64073894, 0.37731481, 0.37908497, 0.48097826,
              0.80135135]])
[41]: y_train = df.iloc[:,6]
      y_train = y_train.to_numpy()
      y_train
[41]: array([0, 0, 0, 1, 1, 0, 0, 0, 0, 1, 1, 1, 1, 0, 0, 1, 0, 0, 1, 0, 0,
             0, 1, 1, 1, 1, 0, 0, 1, 0, 1, 0, 1, 0, 1, 0, 0, 0, 1, 0, 0, 1, 0,
             0, 1, 0, 1, 1, 0, 0, 1, 1, 0, 0, 1, 1, 0, 1, 1, 1, 1, 1, 0, 1, 1, 0,
             1, 0, 0, 1, 1, 0, 1, 1, 0, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 0, 0, 1,
             0, 1, 1, 1, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 1, 0, 1, 1, 0, 0,
             0, 0, 1, 0, 1, 1, 0, 0, 1, 0, 1, 0, 1, 0, 0, 1, 0, 1, 1, 0, 0, 1,
             1, 1, 0, 0, 0, 1, 0, 1, 0, 1, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1, 1,
             1, 0, 1, 1, 1, 0, 1, 1, 1, 1, 0, 0, 1, 1, 0, 0, 1, 0, 0, 1, 0, 0,
             0, 0, 0, 1, 1, 1, 0, 1, 1, 0, 1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 0, 0,
             0, 0, 0, 1, 0, 0, 1, 0, 1, 0, 0, 0, 0, 1, 1, 0, 1, 1, 1, 0, 1, 1,
             1, 0, 0, 1, 1, 1, 1, 0])
[42]: class Node:
          def __init__(self, weight):
              self.weight = weight
          def fire (self, input_val):
              return input_val*self.weight
[43]: class slp:
          def __init__(self, thresh, c, nodes=[]):
              self.thresh = thresh
              self.c = c
              self.n1 = nodes[0]
              self.n2 = nodes[1]
              self.n3 = nodes[2]
              self.n4 = nodes[3]
              self.n5 = nodes[4]
              self.n6 = nodes[5]
```

```
self.weight_log = []
       self.accuracies=[]
   def test (self, ins=[]):
        print(ins)
      nout1 = self.n1.fire(ins[0])
      nout2 = self.n2.fire(ins[1])
      nout3 = self.n3.fire(ins[2])
      nout4 = self.n4.fire(ins[3])
      nout5 = self.n5.fire(ins[4])
      nout6 = self.n6.fire(ins[5])
      sum = nout1 + nout2 + nout3 + nout4 + nout5 + nout6
        print(sum)
       if sum >= thresh:
           return 1
       else:
          return 0
   def train (self, desired, actual, vals=[]):
       error = desired - actual
       delw1 = self.c * error * vals[0]
       delw2 = self.c * error * vals[1]
      delw3 = self.c * error * vals[2]
      delw4 = self.c * error * vals[3]
      delw5 = self.c * error * vals[4]
      delw6 = self.c * error * vals[5]
      self.n1 = Node(self.n1.weight + delw1)
      self.n2 = Node(self.n2.weight + delw2)
       self.n3 = Node(self.n3.weight + delw3)
       self.n4 = Node(self.n4.weight + delw4)
       self.n5 = Node(self.n5.weight + delw5)
       self.n6 = Node(self.n6.weight + delw6)
       return
   def training (self, nepoch, inputs=[], outputs=[]):
      for i in range(nepoch):
          print("\n\n======epoch: ", i+1, u
→"======\n")
          flag = 1
           wrong = 0
           for x in range(len(inputs)):
               out = self.test(inputs[x])
              \#print("inputs: ", inputs[x], "\n/", "actual = ", out, "/
\rightarrow desired = ", outputs[x], "/ weights: ",
                     #round((self.n1.weight), 2), round(self.n2.weight, 2),
\rightarrow round(self.n3.weight, 2),
```

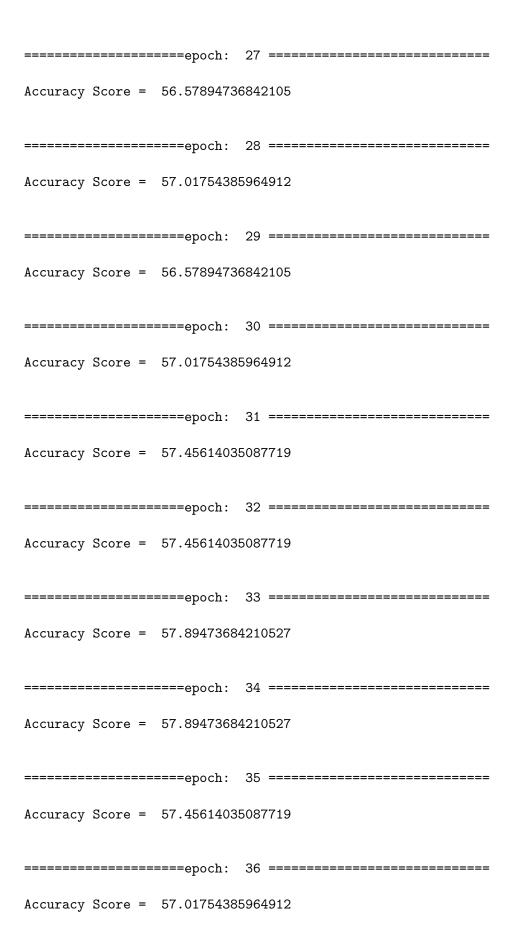
```
#round(self.n4.weight, 2), round(self.n5.weight, 2),
\rightarrow round(self.n6.weight, 2))
               if out != outputs[x]:
                  self.train(outputs[x], out, inputs[x])
                  flag = 0
                  wrong += 1
           if flag:
             Ш
→print("-----\nTraining complete
→with 100% accuracy!")
              percent_error = (len(y_train) - wrong)/len(y_train)*100
              self.accuracies.append(percent error)
              self.weight_log.append([self.n1.weight,self.n2.weight,self.n3.
→weight,self.n4.weight,self.n5.weight,self.n6.weight])
              return
           else:
              percent_error = ((len(y_train) - wrong)/len(y_train))*100
              print("Accuracy Score = ", percent_error)
              self.accuracies.append(percent_error)
              self.weight_log.append([self.n1.weight,self.n2.weight,self.n3.
→weight,self.n4.weight,self.n5.weight,self.n6.weight])
  def testing (self, inputs=[], outputs=[]):
      correct = 0
      y_pred = []
      for x in range(len(inputs)):
           out = self.test(inputs[x])
          y_pred.append(out)
           print('desired = ', outputs[x], "| predicted = ", out)
           if out == outputs[x]:
              correct+=1
      acc = correct/len(outputs)
      print("accuracy = ", acc, "correct: ", correct)
      return y_pred
  def metrics (self, inputs=[], outputs=[]):
      correct = 0
      FN = 0
      FP = 0
      TN = 0
      TP = 0
      y_pred = []
      for x in range(len(inputs)):
          out = self.test(inputs[x])
          y_pred.append(out)
           if out == outputs[x]:
              correct+=1
```

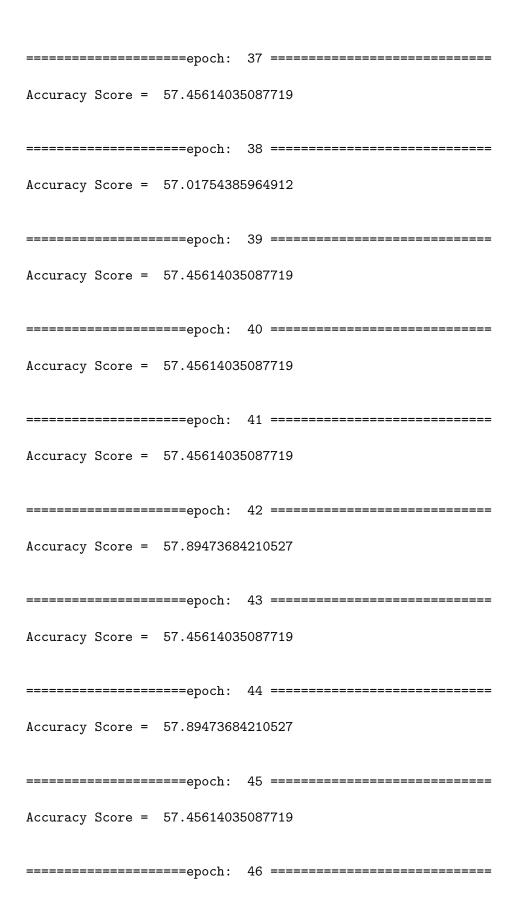
```
if out == outputs[x] == 1:
                      TP += 1
                  elif out == outputs[x] == 0:
                      TN += 1
                  elif out != outputs[x] == 0:
                      FP += 1
                  elif out != outputs[x] == 1:
                      FN += 1
              acc = correct/len(outputs)
              metrics = {'FN':FN, 'FP':FP, 'TN':TN, 'TP':TP, 'acc':acc}
              conf_mat = [[TP, FP],[FN, TN]]
              accuracy = (TN+TP)/(TN+TP+FN+FP)
              precision = (TP)/(TP+FP)
              recall = (TP)/(TP+FN)
              specificity = (TN)/(TN+FP)
              f1_score = 2*((precision*recall)/precision+recall)
              print('Confusion Matrix: \n[', conf mat[0],'\n', conf mat[1], ']')
              print('accuracy: ', accuracy, '\nprecision: ', precision, '\nrecall: ', 
       →recall, '\nspecificity: ', specificity,
                    '\nf1 score: ', f1_score)
              return metrics
[44]: import random
      1 = []
      for x in range(6):
          l.append(random.uniform(0,1))
      n1 = Node(1[0])
     n2 = Node(1[1])
     n3 = Node(1[2])
     n4 = Node(1[3])
     n5 = Node(1[4])
     n6 = Node(1[5])
      thresh = float(input("enter the threshold:\t"))
      c = float(input("enter the learning rate:\t"))
      inputs = x_train
      outputs = y_train
      print('weights: ', 1)
     enter the threshold:
                             0.3
     enter the learning rate:
                                     0.001
     weights: [0.27156831598480313, 0.2108244260300134, 0.0303630762871695,
     0.6298024925880238, 0.04773123060731865, 0.8162547692323535
[45]: myslp = slp(thresh, c, [n1, n2, n3, n4, n5, n6])
      myslp.training(int(input("Enter the number of epochs:\t")), inputs, outputs)
```

```
import matplotlib.pyplot as plt
plt.plot(myslp.accuracies)
plt.title('accuracies with time')
plt.show()
max_acc = max(myslp.accuracies)
max_index = myslp.accuracies.index(max_acc)
best_weights = myslp.weight_log[max_index]
print("best weights: ", best_weights)
Enter the number of epochs: 500
=======epoch: 1 ===================
Accuracy Score = 50.43859649122807
Accuracy Score = 52.19298245614035
Accuracy Score = 51.75438596491229
=======epoch: 4 =========================
Accuracy Score = 49.56140350877193
=======epoch: 5 ==============
Accuracy Score = 46.49122807017544
======epoch: 6 ==================
Accuracy Score = 44.29824561403509
Accuracy Score = 46.05263157894737
```

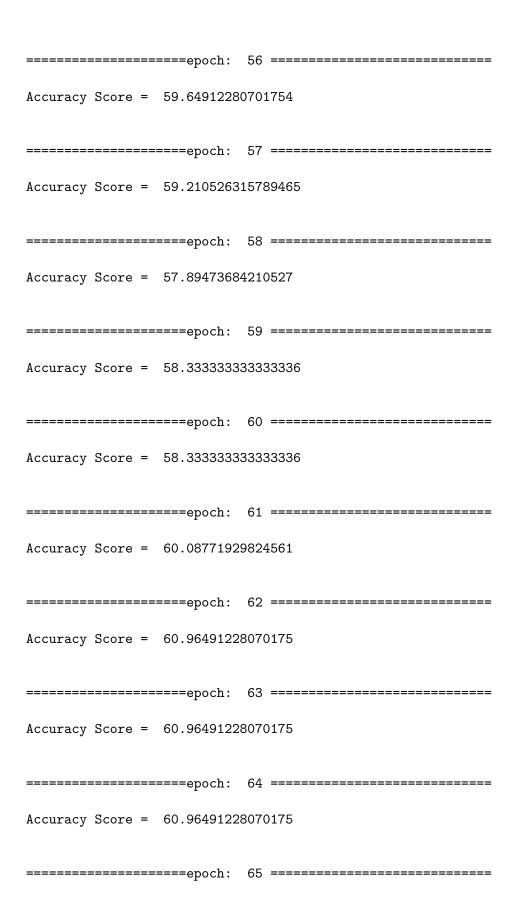


| Accuracy | Score | = | 53.9473684 | 2105 | 5263 | | | | |
|----------|--------|------|------------|------|-------|-------|-------|-------|------|
| ====== | ===== | | ===epoch: | 18 | | | | ===== | ==== |
| Accuracy | Score | = | 53.5087719 | 2982 | 2456 | | | | |
| | | | ===epoch: | 19 | | ===== | ===== | ===== | ==== |
| Accuracy | Score | = | 53.5087719 | 2982 | 2456 | | | | |
| | | === | ===epoch: | 20 | ===== | | ===== | | ==== |
| Accuracy | Score | = | 54.8245614 | 0350 | 8774 | | | | |
| ====== | | | ===epoch: | 21 | ===== | ===== | ===== | ===== | ==== |
| Accuracy | Score | = | 53.5087719 | 2982 | 2456 | | | | |
| ====== | ===== | : | ===epoch: | 22 | | | | ===== | |
| Accuracy | Score | = | 54.3859649 | 1228 | 30706 | | | | |
| ====== | ===== | ===: | ===epoch: | 23 | ===== | ===== | ===== | ===== | ==== |
| Accuracy | Score | = | 56.1403508 | 7719 | 9298 | | | | |
| | =====: | ===: | ===epoch: | 24 | | | ===== | ===== | ==== |
| Accuracy | Score | = | 56.1403508 | 7719 | 9298 | | | | |
| | ===== | === | ===epoch: | 25 | ===== | ===== | ===== | ===== | ==== |
| Accuracy | Score | = | 57.0175438 | 5964 | 1912 | | | | |
| ====== | =====: | ===: | ===epoch: | 26 | | ===== | ===== | ===== | ==== |
| Accuracy | Score | = | 57.0175438 | 5964 | 1912 | | | | |

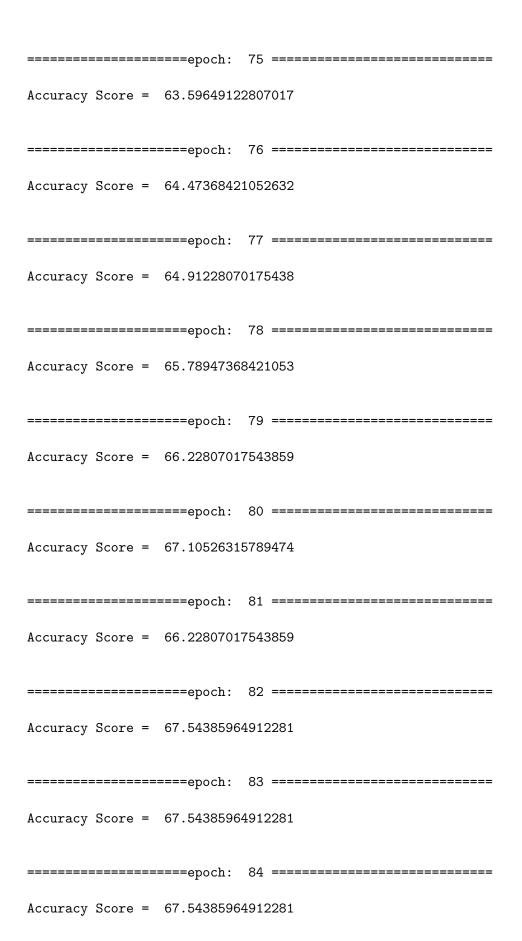


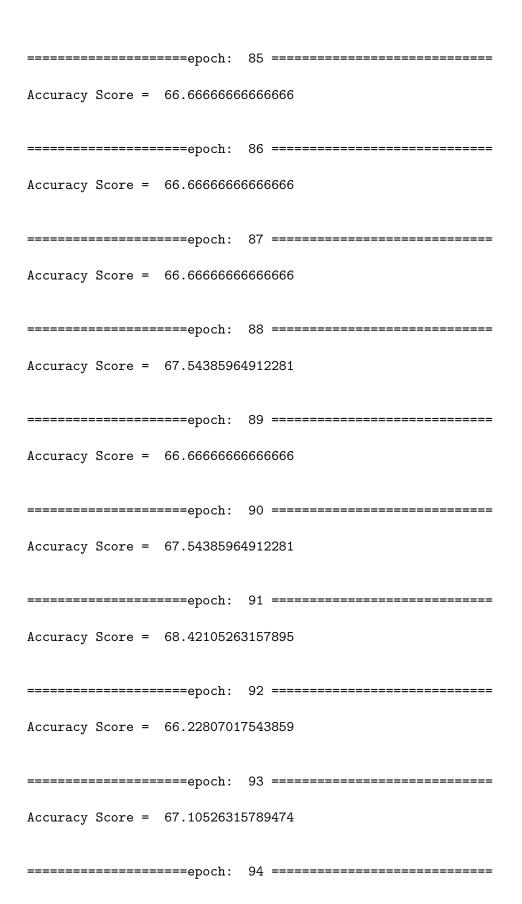


| Accuracy Score = 58.3333333333333 | |
|-------------------------------------|-----|
| epoch: 47 | |
| Accuracy Score = 57.89473684210527 | |
| epoch: 48 | === |
| Accuracy Score = 58.77192982456141 | |
| epoch: 49 | |
| Accuracy Score = 59.210526315789465 | |
| epoch: 50 | === |
| Accuracy Score = 59.210526315789465 | |
| epoch: 51 | -== |
| Accuracy Score = 59.210526315789465 | |
| epoch: 52 | |
| Accuracy Score = 59.210526315789465 | |
| epoch: 53 | -== |
| Accuracy Score = 59.210526315789465 | |
| epoch: 54 | -== |
| Accuracy Score = 59.210526315789465 | |
| 55 | === |
| Accuracy Score = 59.210526315789465 | |

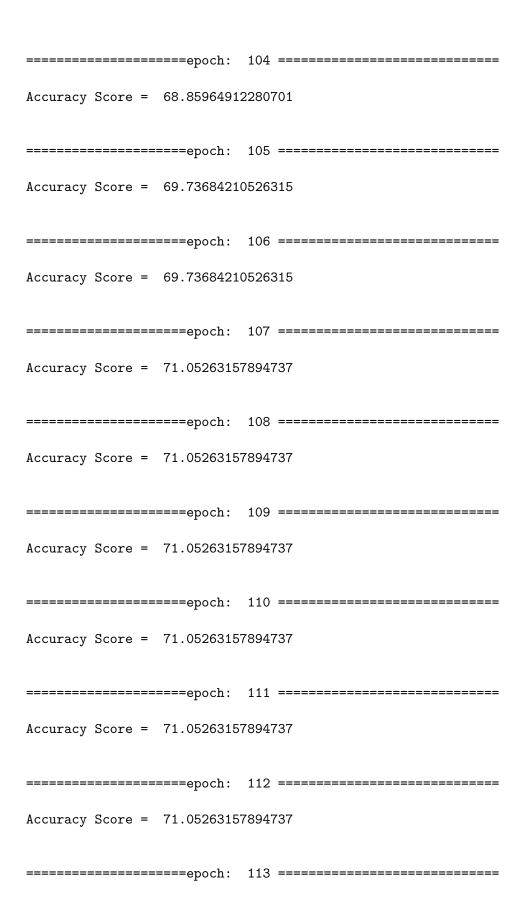


| Accuracy | Score | = | 61.8421052 | 6315 | 579 |
|----------|-------|------|------------|------|---|
| ====== | ===== | | ===epoch: | 66 | |
| Accuracy | Score | = | 62.2807017 | 5438 | 3597 |
| ====== | | ==== | ===epoch: | 67 | |
| Accuracy | Score | = | 60.9649122 | 8070 | 0175 |
| | | | ===epoch: | 68 | |
| Accuracy | Score | = | 60.9649122 | 8070 | 0175 |
| ====== | | -=== | ===epoch: | 69 | |
| Accuracy | Score | = | 62.2807017 | 5438 | 3597 |
| ====== | | ==== | ===epoch: | 70 | |
| Accuracy | Score | = | 61.8421052 | 6315 | 579 |
| ====== | ===== | | ===epoch: | 71 | ======================================= |
| Accuracy | Score | = | 62.7192982 | 4561 | 1403 |
| ====== | | | ===epoch: | 72 | |
| Accuracy | Score | = | 62.2807017 | 5438 | 3597 |
| | | -=== | ===epoch: | 73 | |
| Accuracy | Score | = | 63.1578947 | 3684 | 121 |
| ====== | | | ===epoch: | 74 | |
| Accuracy | Score | = | 63.1578947 | 3684 | 121 |

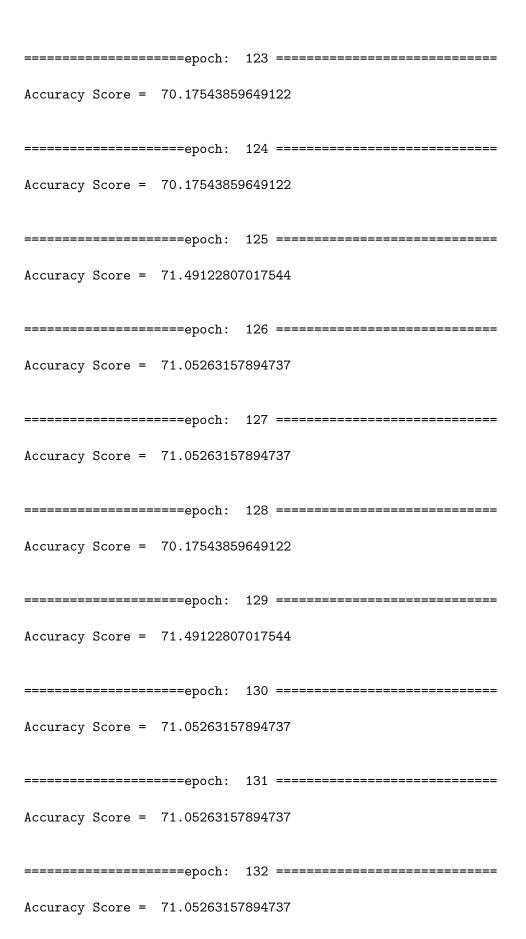


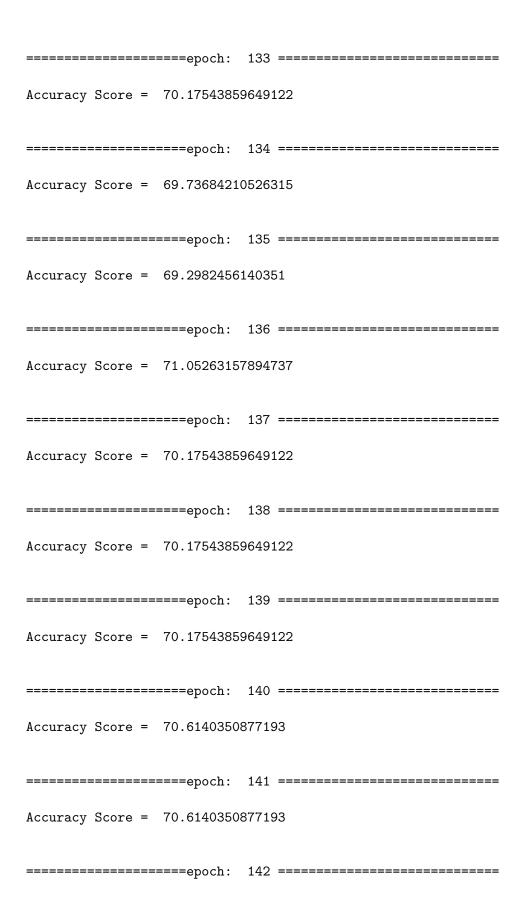


| Accuracy | Score | = | 67.54385964 | 4912281 |
|----------|-------|------|-------------|----------------|
| | ===== | -== | ===epoch: | 95 |
| Accuracy | Score | = | 67.54385964 | 4912281 |
| | | | ===epoch: | 96 ====== |
| Accuracy | Score | = | 67.54385964 | 4912281 |
| | | | ===epoch: | 97 ======== |
| Accuracy | Score | = | 68.42105263 | 3157895 |
| | | -==- | ===epoch: | 98 =========== |
| Accuracy | Score | = | 68.42105263 | 3157895 |
| ====== | | -=== | ===epoch: | 99 ======= |
| Accuracy | Score | = | 68.42105263 | 3157895 |
| ====== | ===== | | ===epoch: | 100 |
| Accuracy | Score | = | 67.54385964 | 4912281 |
| ====== | | -=== | ===epoch: | 101 |
| Accuracy | Score | = | 68.42105263 | 3157895 |
| | ===== | | ===epoch: | 102 |
| Accuracy | Score | = | 68.42105263 | 3157895 |
| | ===== | | ===epoch: | 103 |
| Accuracy | Score | = | 69.2982456 | 140351 |



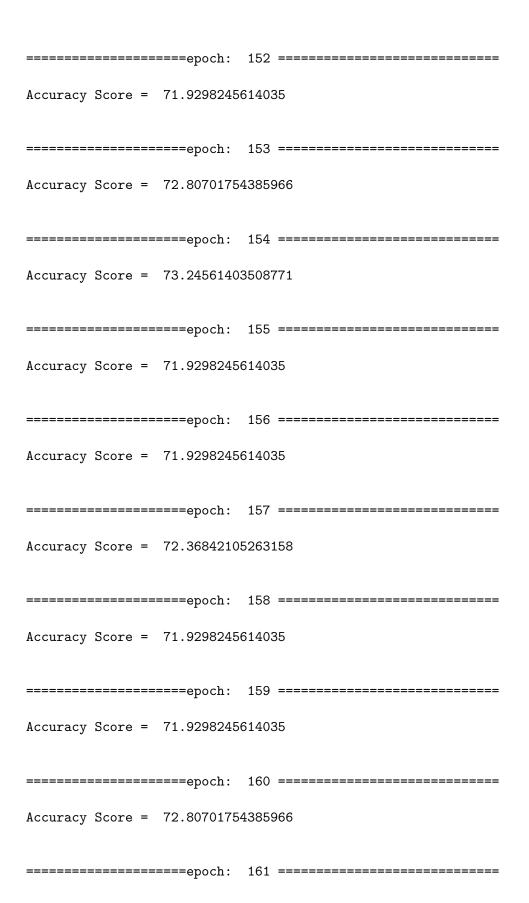
| Accuracy | Score | = | 71.0526315 | 78947 | 37 |
|----------|-------|------|-------------|-------|---|
| | ===== | ==== | ===epoch: | 114 | |
| Accuracy | Score | = | 71.0526315 | 78947 | 37 |
| ====== | ===== | | ===epoch: | 115 | |
| Accuracy | Score | = | 70.61403508 | 87719 | 3 |
| ====== | | | ===epoch: | 116 | |
| Accuracy | Score | = | 71.0526315 | 78947 | 37 |
| | ===== | ==== | ===epoch: | 117 | |
| Accuracy | Score | = | 71.0526315 | 78947 | 37 |
| ====== | | ===: | ===epoch: | 118 | ======================================= |
| Accuracy | Score | = | 71.0526315 | 78947 | 737 |
| ====== | | === | ===epoch: | 119 | |
| Accuracy | Score | = | 70.6140350 | 87719 | 3 |
| ====== | | ===: | ===epoch: | 120 | |
| Accuracy | Score | = | 71.0526315 | 78947 | 737 |
| ====== | | === | ===epoch: | 121 | |
| Accuracy | Score | = | 71.0526315 | 78947 | 737 |
| ====== | ===== | ===: | ===epoch: | 122 | |
| Accuracy | Score | = | 71.0526315 | 78947 | 37 |



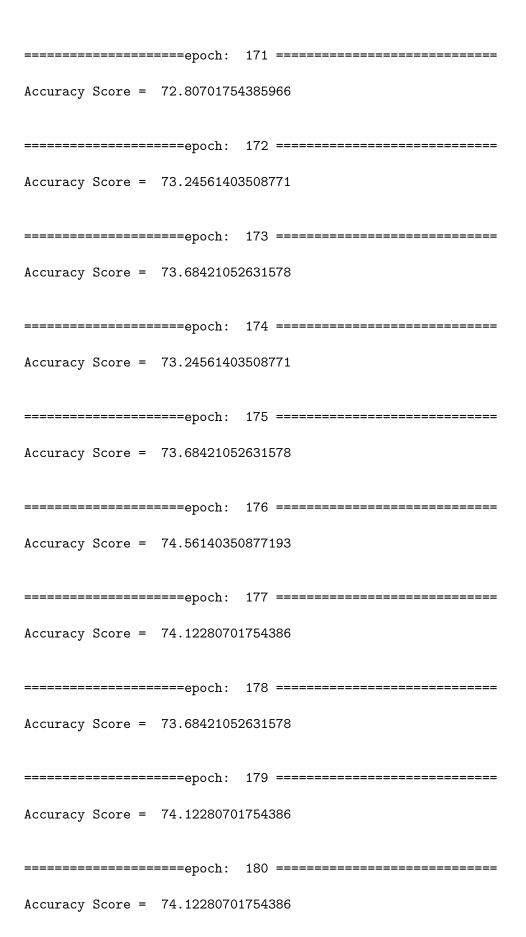


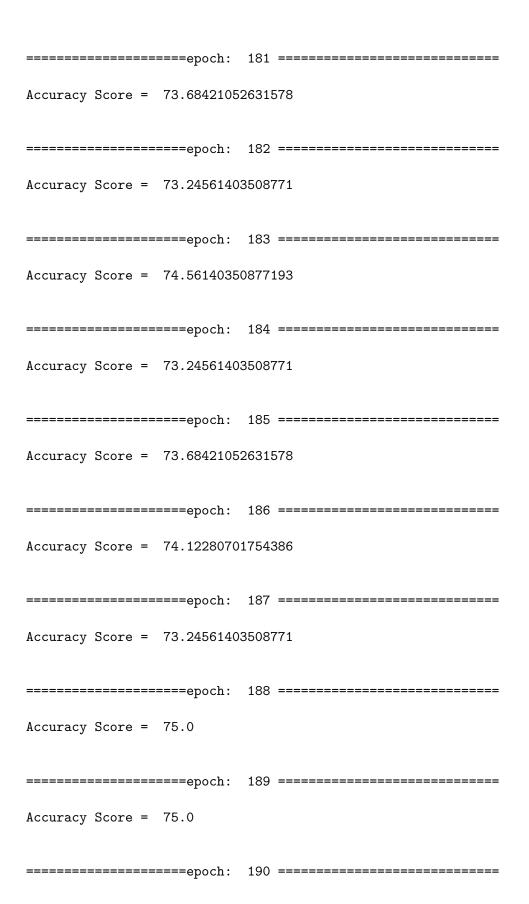
| Accuracy | Score | = | 70.61403508 | 37719 | 93 |
|----------|-------|------|--------------------------|-------|---|
| | | | - | | ======================================= |
| · | | | 70.17543859 | | |
| | | | ===epoch: 70.17543859 | | 122 |
| | | | _ | | |
| Accuracy | Score | = | 70.17543859 | 96491 | 122 |
| | | | - | | ======================================= |
| Accuracy | Score | = | 70.17543859 | 96491 | 122 |
| | | | _ | | |
| Accuracy | Score | = | 70.61403508 | 37719 | 93 |
| | | | - | | ======================================= |
| Accuracy | Score | = | 70.61403508 | 37719 | 93 |
| | | === | ===epoch: | 149 | |
| Accuracy | Score | = | 71.05263157 | 78947 | 737 |
| ====== | | ==== | ===epoch: | 150 | |
| Accuracy | Score | = | 71.05263157 | 78947 | 737 |
| ====== | | -== | ===epoch: | 151 | |

Accuracy Score = 71.9298245614035



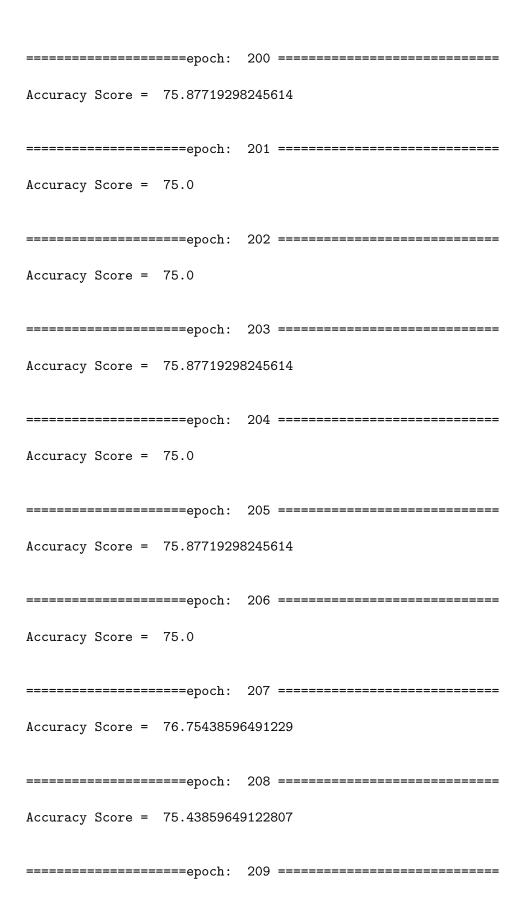
| Accuracy | Score | = | 73.2456140 | 35087 | 771 |
|----------|-------|------|-------------|-------|---|
| ====== | ===== | | ===epoch: | 162 | |
| Accuracy | Score | = | 71.92982456 | 61403 | 85 |
| | | | ===epoch: | 163 | |
| Accuracy | Score | = | 71.92982456 | 61403 | 35 |
| ====== | | -=== | ===epoch: | 164 | |
| Accuracy | Score | = | 73.24561403 | 35087 | 771 |
| ====== | ===== | -=== | ===epoch: | 165 | |
| Accuracy | Score | = | 72.8070175 | 43859 | 966 |
| ====== | | | ===epoch: | 166 | |
| Accuracy | Score | = | 72.3684210 | 52631 | .58 |
| ====== | | ==== | ===epoch: | 167 | ======================================= |
| Accuracy | Score | = | 73.6842105 | 26315 | 578 |
| ====== | ===== | | ===epoch: | 168 | |
| Accuracy | Score | = | 71.9298245 | 51403 | 35 |
| | | | | | |
| Accuracy | Score | = | 73.24561403 | 35087 | 771 |
| ====== | ===== | | ===epoch: | 170 | |
| Accuracy | Score | = | 72.8070175 | 43859 | 966 |



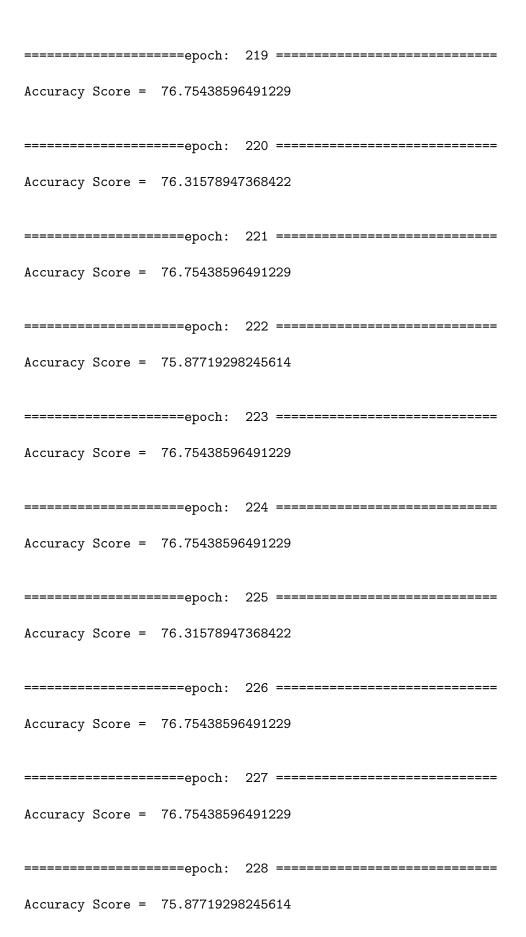


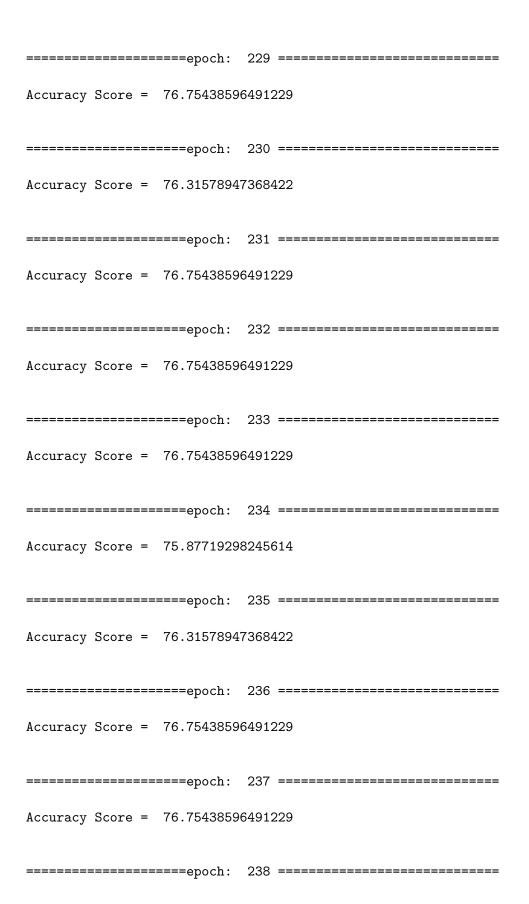
| Accuracy | Score | = | 75.0 | | |
|----------|-------|---|-------------|-------|---|
| ====== | | | ===epoch: | 191 | |
| Accuracy | Score | = | 75.0 | | |
| Accuracy | | | _ | 192 | |
| | | | ===epoch: | 193 | ======================================= |
| Accuracy | Score | = | 75.0 | | |
| ====== | | | ===epoch: | 194 | |
| Accuracy | Score | = | 74.56140350 |)877: | 193 |
| Accuracy | | | _ | 195 | |
| | | | _ | 196 | |
| Accuracy | | | | | |
| Accuracy | | | - | 197 | |
| ======= | | | ===epoch: | 198 | |
| Accuracy | Score | = | 75.0 | | |
| ====== | | | ===epoch: | 199 | ======================================= |

Accuracy Score = 74.12280701754386

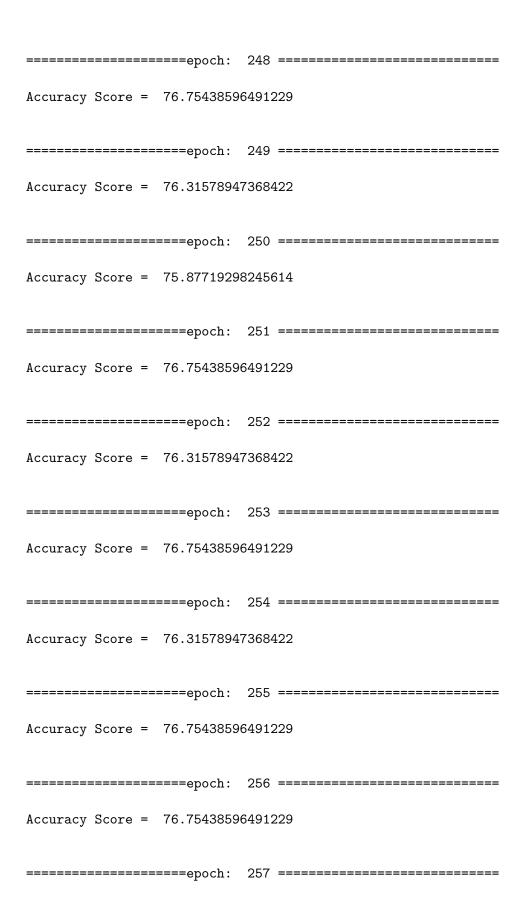


| Accuracy | Score | = | 76.7543859 | 64912 | 229 |
|----------|-------|------|-------------|-------|-----|
| ====== | ===== | ==== | ===epoch: | 210 | |
| Accuracy | Score | = | 76.7543859 | 64912 | 229 |
| | | | ===epoch: | 211 | |
| Accuracy | Score | = | 76.3157894 | 73684 | 122 |
| ====== | ===== | | ===epoch: | 212 | |
| Accuracy | Score | = | 75.8771929 | 82456 | 314 |
| ====== | ===== | ==== | ===epoch: | 213 | |
| Accuracy | Score | = | 76.7543859 | 64912 | 229 |
| | ===== | ==== | ===epoch: | 214 | |
| Accuracy | Score | = | 76.7543859 | 64912 | 229 |
| ====== | ===== | ==== | ===epoch: | 215 | |
| Accuracy | Score | = | 75.87719298 | 82456 | 514 |
| ====== | | ==== | ===epoch: | 216 | |
| Accuracy | Score | = | 76.3157894 | 73684 | 122 |
| | ===== | ==== | ===epoch: | 217 | |
| Accuracy | Score | = | 76.7543859 | 64912 | 229 |
| | ===== | | ===epoch: | 218 | |
| Accuracy | Score | = | 76.7543859 | 64912 | 229 |

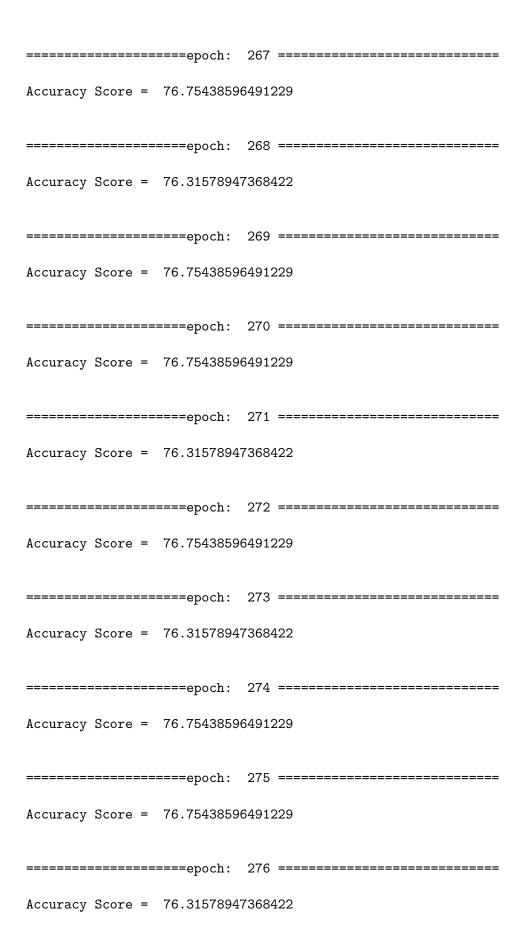


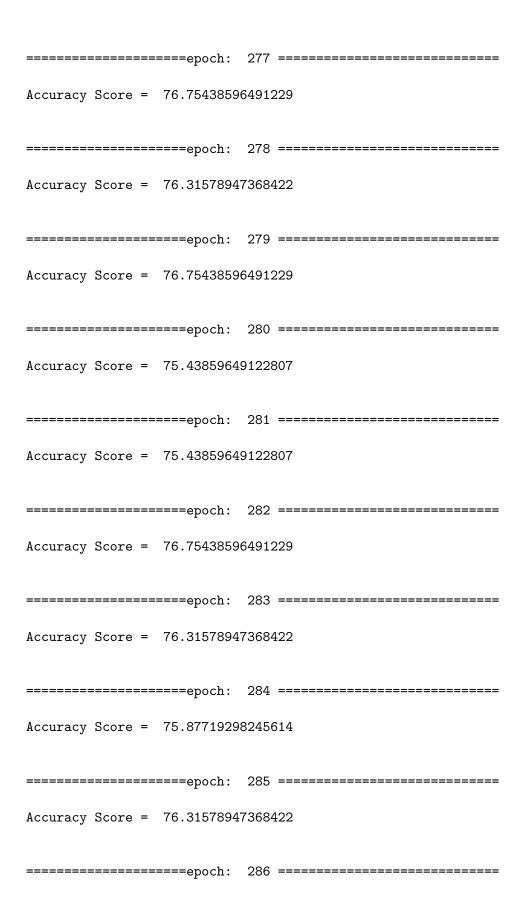


| Accuracy | Score | = | 76.75438596 | 64912 | 229 | | | | | |
|----------|-------|------|-------------|-------|-------|-------|-------|------|-------|------|
| ====== | ===== | | ===epoch: | 239 | ====: | ===== | -==== | ==== | ====: | |
| Accuracy | Score | = | 76.75438596 | 64912 | 229 | | | | | |
| | ===== | ==== | ===epoch: | 240 | ====: | ===== | ===== | ==== | ====: | |
| Accuracy | Score | = | 76.75438596 | 64912 | 229 | | | | | |
| | ===== | ==== | ===epoch: | 241 | ====: | ===== | ===== | ==== | ====: | |
| Accuracy | Score | = | 76.75438596 | 64912 | 229 | | | | | |
| | ===== | | ===epoch: | 242 | ====: | ===== | -==== | ==== | ====: | ==== |
| Accuracy | Score | = | 76.3157894 | 73684 | 122 | | | | | |
| | ===== | ==== | ===epoch: | 243 | ====: | | | ==== | ====: | ==== |
| Accuracy | Score | = | 76.75438596 | 64912 | 229 | | | | | |
| | ===== | ==== | ===epoch: | 244 | ====: | | | ==== | ====: | ==== |
| Accuracy | Score | = | 76.3157894 | 73684 | 122 | | | | | |
| | ===== | ===: | ===epoch: | 245 | ====: | ===== | -==== | ==== | ====: | |
| Accuracy | Score | = | 75.87719298 | 82456 | 814 | | | | | |
| | ===== | | ===epoch: | 246 | ====: | | | ==== | ====: | ==== |
| Accuracy | Score | = | 76.75438596 | 64912 | 229 | | | | | |
| | ===== | | ===epoch: | 247 | | | | | | ==== |
| Accuracy | Score | = | 76.3157894 | 73684 | 122 | | | | | |



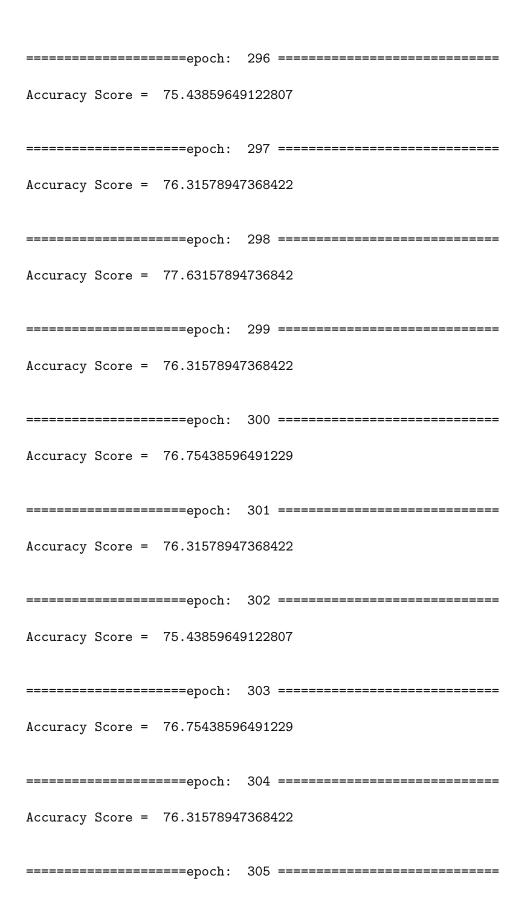
| Accuracy | Score | = | 76.3157894 | 73684 | 122 |
|----------|--------|------|------------|-------|-----|
| ====== | ===== | | ===epoch: | 258 | |
| Accuracy | Score | = | 76.7543859 | 64912 | 229 |
| | | | - | | |
| Accuracy | Score | = | 76.7543859 | 64912 | 229 |
| | | | | | |
| Accuracy | Score | = | 76.3157894 | 73684 | 122 |
| ====== | ===== | ==== | ===epoch: | 261 | |
| Accuracy | Score | = | 76.7543859 | 64912 | 229 |
| ====== | ===== | ===: | ===epoch: | 262 | |
| Accuracy | Score | = | 76.7543859 | 64912 | 229 |
| ====== | =====: | ==== | ===epoch: | 263 | |
| Accuracy | Score | = | 76.3157894 | 73684 | 122 |
| ====== | ===== | ==== | ===epoch: | 264 | |
| Accuracy | Score | = | 76.7543859 | 64912 | 229 |
| ====== | ===== | ==== | ===epoch: | 265 | |
| Accuracy | Score | = | 76.7543859 | 64912 | 229 |
| ====== | =====: | ==== | ===epoch: | 266 | |
| Accuracy | Score | = | 76.3157894 | 73684 | 122 |



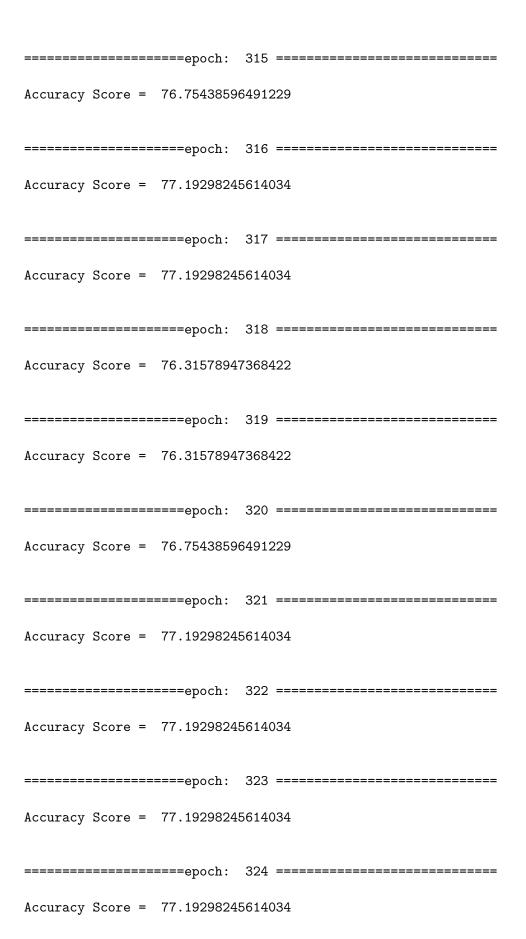


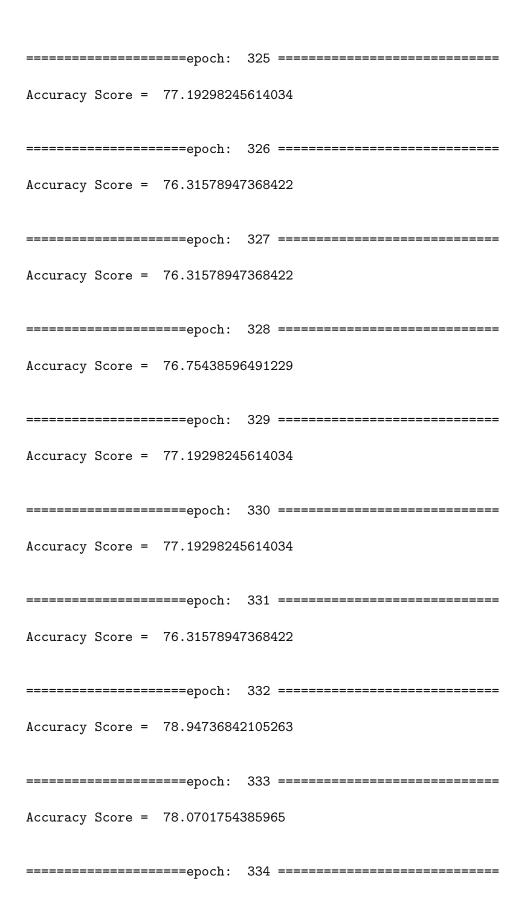
| Accuracy | Score | = | 76.7543859 | 64912 | 229 | |
|----------|-------|------|------------|-------|------|---|
| ====== | | ===: | ===epoch: | 287 | ===: | |
| Accuracy | Score | = | 76.3157894 | 73684 | 122 | |
| ====== | | | ===epoch: | 288 | ===: | ======================================= |
| Accuracy | Score | = | 76.7543859 | 64912 | 229 | |
| ====== | ===== | | ===epoch: | 289 | ===: | |
| Accuracy | Score | = | 75.4385964 | 91228 | 307 | |
| ====== | | | ===epoch: | 290 | ===: | ======================================= |
| Accuracy | Score | = | 76.7543859 | 64912 | 229 | |
| | ===== | ===: | ===epoch: | 291 | ===: | |
| Accuracy | Score | = | 76.3157894 | 73684 | 122 | |
| | ===== | | ===epoch: | 292 | ===: | |
| Accuracy | Score | = | 75.8771929 | 82456 | 814 | |
| ====== | ===== | ===: | ===epoch: | 293 | ===: | |
| Accuracy | Score | = | 76.3157894 | 73684 | 122 | |
| ====== | | | ===epoch: | 294 | ===: | |
| Accuracy | Score | = | 75.8771929 | 82456 | 814 | |
| ====== | | | ===epoch: | 295 | ===: | |

Accuracy Score = 76.31578947368422



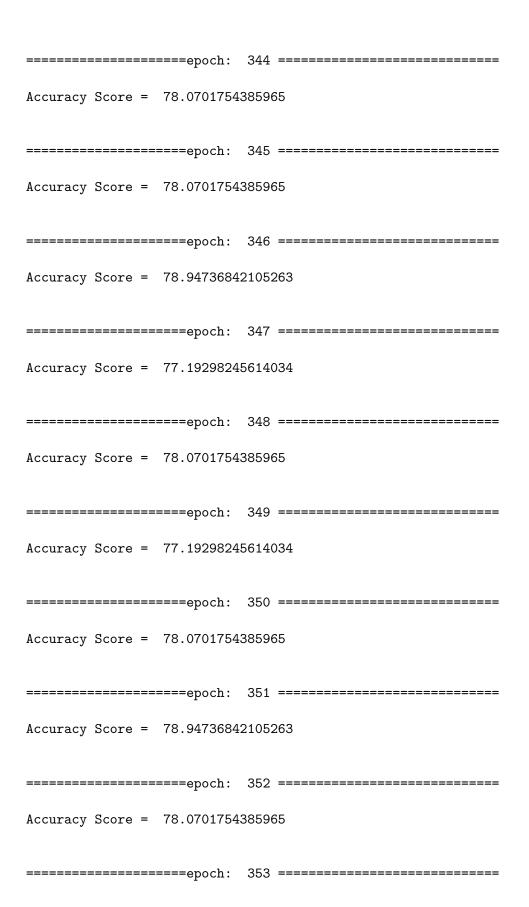
| Accuracy | Score | = | 76.3157894 | 73684 | 122 |
|----------|-------|------|------------|-------|-----|
| ====== | | ==== | ===epoch: | 306 | |
| Accuracy | Score | = | 76.7543859 | 64912 | 229 |
| ====== | | ==== | ===epoch: | 307 | |
| Accuracy | Score | = | 77.1929824 | 56140 | 034 |
| | ===== | ==== | ===epoch: | 308 | |
| Accuracy | Score | = | 76.3157894 | 73684 | 22 |
| ====== | ===== | ==== | ===epoch: | 309 | |
| Accuracy | Score | = | 76.3157894 | 73684 | 122 |
| | ===== | ==== | ===epoch: | 310 | |
| Accuracy | Score | = | 76.3157894 | 73684 | 122 |
| ====== | | ==== | ===epoch: | 311 | |
| Accuracy | Score | = | 76.7543859 | 64912 | 229 |
| ====== | | ==== | ===epoch: | 312 | |
| Accuracy | Score | = | 77.1929824 | 56140 | 934 |
| | | ==== | ===epoch: | 313 | |
| Accuracy | Score | = | 77.1929824 | 56140 | 034 |
| ====== | ===== | -=== | ===epoch: | 314 | |
| Accuracy | Score | = | 76.3157894 | 73684 | 122 |



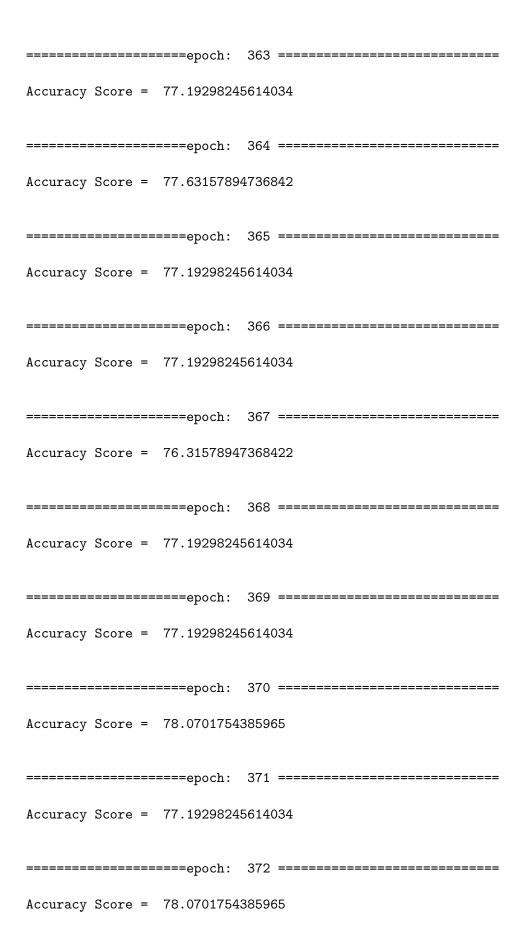


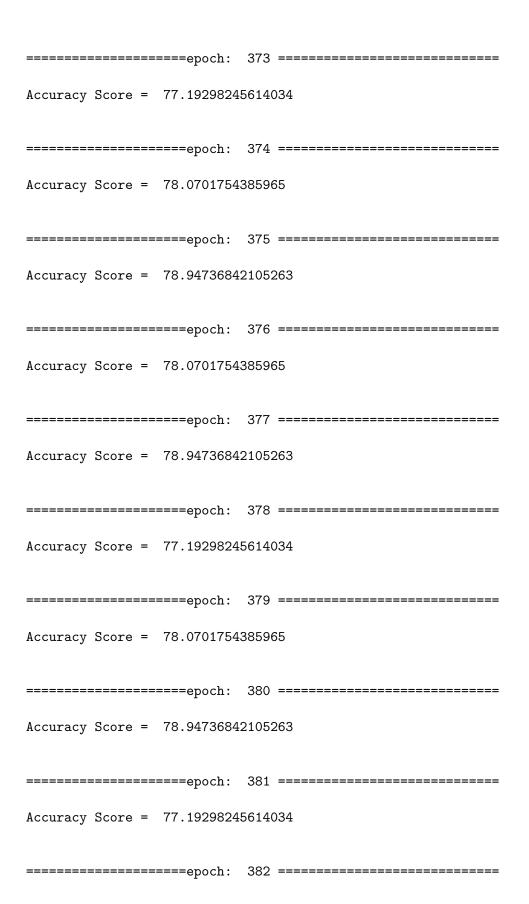
| Accuracy | Score | = | 77.1929824 | 56140 | 034 |
|----------|-------|------|-------------------------|-------|-----|
| ====== | | | ===epoch: | 335 | |
| Accuracy | Score | = | 76.7543859 | 64912 | 229 |
| | | | ===epoch: 77.1929824 | | 034 |
| | | | | | |
| | | | 77.1929824 | | |
| ====== | | | ===epoch: | 338 | |
| Accuracy | Score | = | 77.1929824 | 56140 | 034 |
| | | | ===epoch: 77.1929824 | | |
| · | | | | | |
| | | | epoch: 77.1929824 | | 034 |
| ====== | | | ===epoch: | 341 | |
| Accuracy | Score | = | 77.1929824 | 56140 | 034 |
| | | | | | |
| · | | | 77.1929824 | | |
| ====== | | ==== | ===epoch: | 343 | |

Accuracy Score = 78.94736842105263



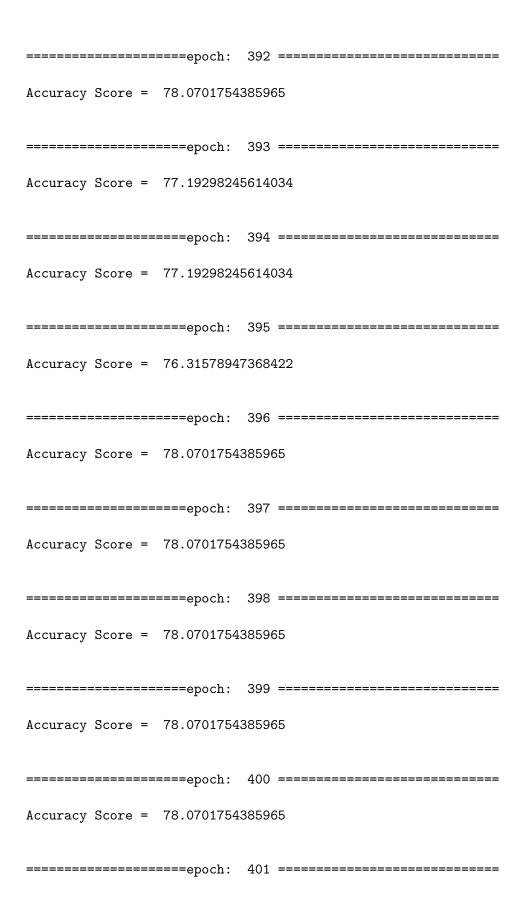
| Accuracy | Score | = | 77.1929824 | 56140 | 034 |
|----------|-------|------|------------|-------|---|
| ====== | ===== | | ===epoch: | 354 | |
| Accuracy | Score | = | 78.0701754 | 38596 | 35 |
| ====== | ===== | | ===epoch: | 355 | |
| Accuracy | Score | = | 78.0701754 | 38596 | 65 |
| ====== | | ==== | ===epoch: | 356 | |
| Accuracy | Score | = | 77.1929824 | 56140 | 034 |
| ====== | ===== | | ===epoch: | 357 | |
| Accuracy | Score | = | 76.3157894 | 73684 | 122 |
| ====== | ===== | ===: | ===epoch: | 358 | ======================================= |
| Accuracy | Score | = | 78.0701754 | 38596 | 55 |
| ====== | : | | ===epoch: | 359 | |
| Accuracy | Score | = | 78.0701754 | 38596 | 55 |
| | ===== | | ===epoch: | 360 | |
| Accuracy | Score | = | 77.1929824 | 56140 | 034 |
| | | | | | |
| Accuracy | Score | = | 78.9473684 | 21052 | 263 |
| | | | | | |
| Accuracy | Score | = | 78.0701754 | 38596 | 35 |



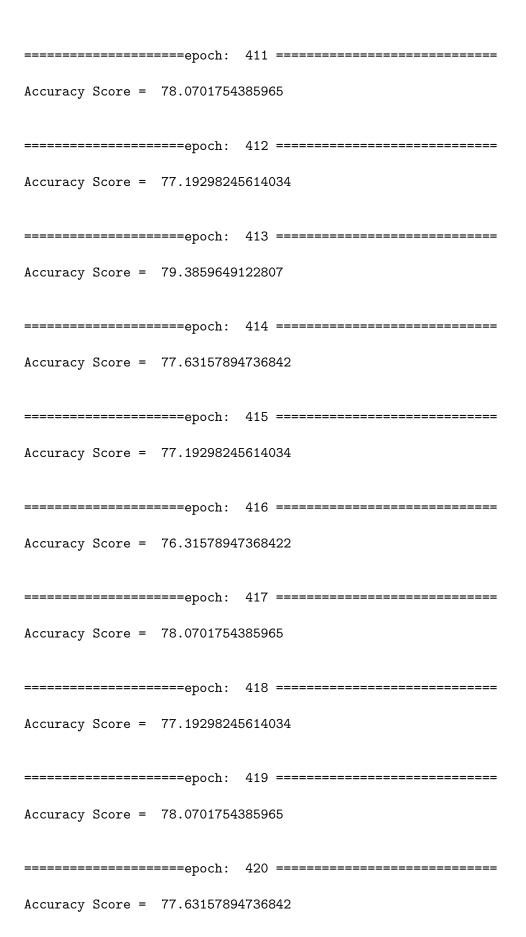


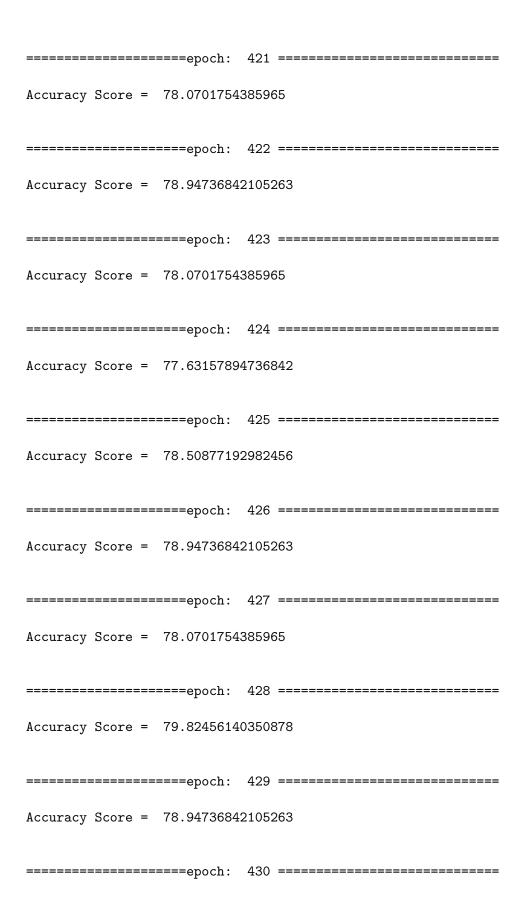
| Accuracy | Score | = | 78.0701754 | 38596 | 55 |
|----------|-------|------|-------------------------|-------|---|
| | | | - | | ======================================= |
| · | | | 78.0701754 | | ======================================= |
| | | | 76.3157894 | | |
| | | | -==epoch: 78.0701754 | | ======================================= |
| · | | | | | ======================================= |
| Accuracy | Score | = | 78.0701754 | 38596 | 65 |
| | | | ===epoch: 77.1929824 | |)34 |
| | | | epoch: 78.0701754 | | |
| ====== | | ==== | ===epoch: | 389 | |
| · | | | 78.0701754 | | ======================================= |
| | | | 78.0701754 | | |
| ====== | : | | ===epoch: | 391 | ======================================= |

Accuracy Score = 78.0701754385965

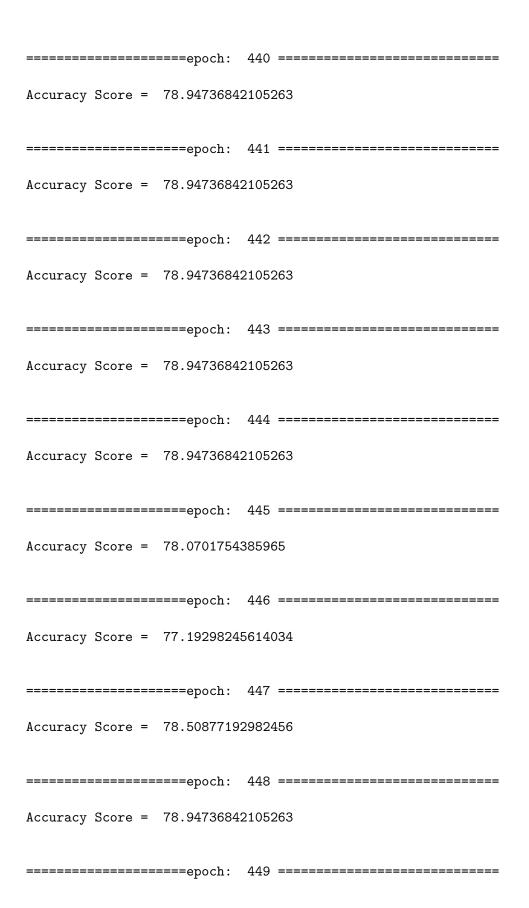


| Accuracy | Score | = | 78.9473684 | 21052 | 263 |
|----------|-------|------|-------------|-------|-----|
| ====== | ===== | -=== | ===epoch: | 402 | |
| Accuracy | Score | = | 77.1929824 | 56140 | 034 |
| ====== | | -=== | ===epoch: | 403 | |
| Accuracy | Score | = | 78.07017543 | 38596 | 55 |
| ====== | | -=== | ===epoch: | 404 | |
| Accuracy | Score | = | 78.07017543 | 38596 | 55 |
| ====== | | -=== | ===epoch: | 405 | |
| Accuracy | Score | = | 77.1929824 | 56140 | 034 |
| ====== | | | ===epoch: | 406 | |
| Accuracy | Score | = | 78.07017543 | 38596 | 55 |
| ====== | ===== | ==== | ===epoch: | 407 | |
| Accuracy | Score | = | 78.07017543 | 38596 | 65 |
| ====== | | | ===epoch: | 408 | |
| Accuracy | Score | = | 78.07017543 | 38596 | 65 |
| | | | ===epoch: | 409 | |
| Accuracy | Score | = | 78.94736842 | 21052 | 263 |
| ====== | | -=== | ===epoch: | 410 | |
| Accuracy | Score | = | 77.1929824 | 56140 | 034 |

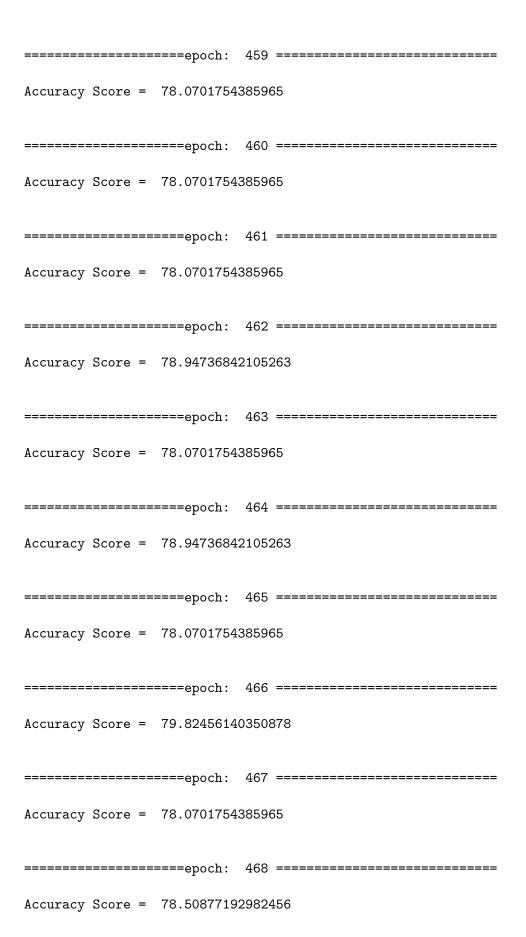


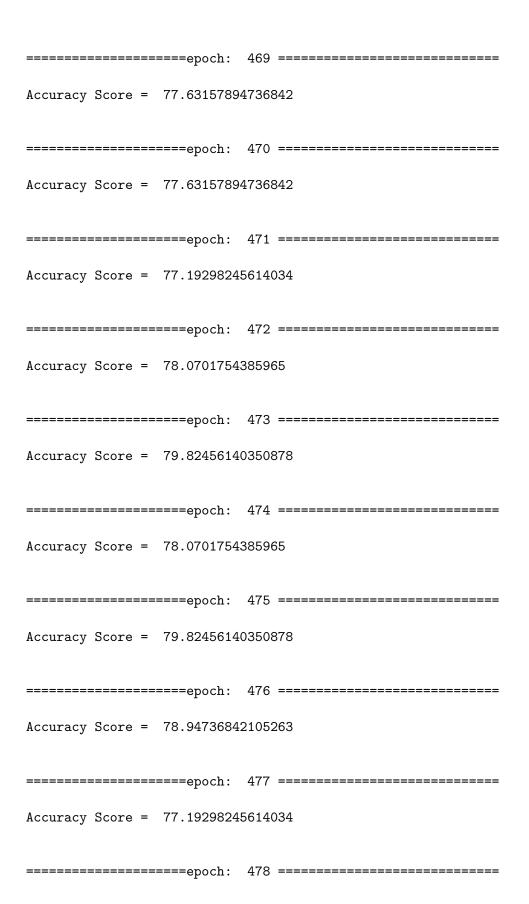


| Accuracy S | Score = | 78.0701754 | 38596 | 65 |
|------------|---------|------------|-------|-----|
| | | ====epoch: | 431 | |
| Accuracy S | Score = | 77.6315789 | 47368 | 342 |
| | | ====epoch: | 432 | |
| Accuracy S | Score = | 77.6315789 | 47368 | 342 |
| ======= | | ====epoch: | 433 | |
| Accuracy S | Score = | 78.5087719 | 29824 | 156 |
| | | ====epoch: | 434 | |
| Accuracy S | Score = | 78.0701754 | 38596 | 65 |
| | | ====epoch: | 435 | |
| Accuracy S | Score = | 77.6315789 | 47368 | 342 |
| ======= | | ====epoch: | 436 | |
| Accuracy S | Score = | 77.6315789 | 47368 | 342 |
| | | ====epoch: | 437 | |
| Accuracy S | Score = | 77.1929824 | 56140 | 034 |
| | | ====epoch: | 438 | |
| Accuracy S | Score = | 78.0701754 | 38596 | 65 |
| ======= | | ====epoch: | 439 | |
| Accuracy S | Score = | 78.9473684 | 21052 | 263 |

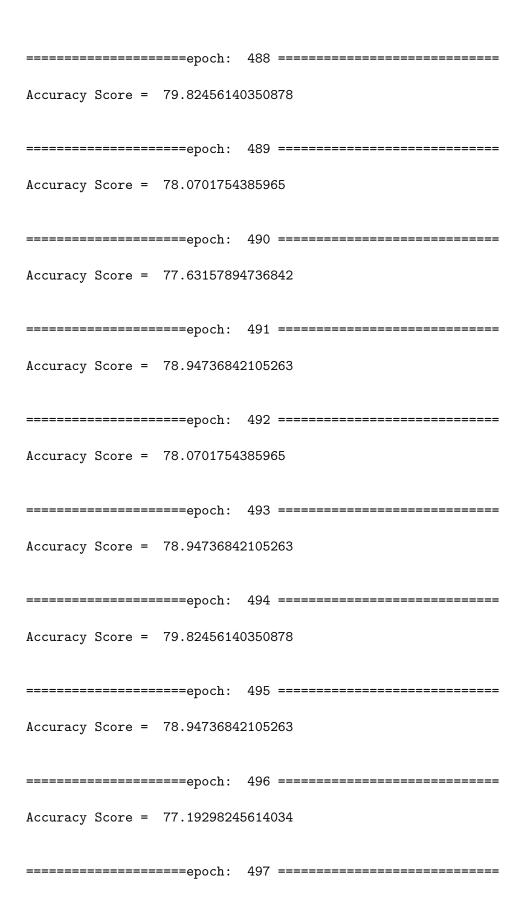


| Accuracy | Score | = | 78.0701754 | 385965 | |
|----------|-------|------|------------|---------|---|
| ====== | ===== | ==== | ===epoch: | 450 == | |
| Accuracy | Score | = | 79.8245614 | 035087 | 8 |
| | ===== | ==== | ===epoch: | 451 = | |
| Accuracy | Score | = | 78.9473684 | 210526 | 3 |
| ====== | : | ==== | ===epoch: | 452 == | |
| Accuracy | Score | = | 78.0701754 | 385965 | |
| | | | ===epoch: | 453 == | |
| Accuracy | Score | = | 77.1929824 | 561403 | 4 |
| | ===== | ==== | ===epoch: | 454 == | |
| Accuracy | Score | = | 78.0701754 | 385965 | |
| ====== | ===== | ==== | ===epoch: | 455 == | |
| Accuracy | Score | = | 78.0701754 | 385965 | |
| | | ==== | ===epoch: | 456 == | |
| Accuracy | Score | = | 77.6315789 | 473684: | 2 |
| | | ==== | ===epoch: | 457 == | |
| Accuracy | Score | = | 77.6315789 | 473684: | 2 |
| | ===== | ==== | ===epoch: | 458 == | |
| Accuracy | Score | = | 79.3859649 | 122807 | |





| Accuracy | Score | = | 78.0701754 | 38596 | 55 |
|----------|-------|------|------------|-------|-----|
| | | | ===epoch: | 479 | |
| Accuracy | Score | = | 78.0701754 | 38596 | 65 |
| | | | ===epoch: | 480 | |
| Accuracy | Score | = | 78.0701754 | 38596 | 65 |
| ====== | | | ===epoch: | 481 | |
| Accuracy | Score | = | 78.0701754 | 38596 | 55 |
| | | | ===epoch: | 482 | |
| Accuracy | Score | = | 78.5087719 | 29824 | 156 |
| | | | ===epoch: | 483 | |
| Accuracy | Score | = | 77.6315789 | 47368 | 342 |
| | | :=== | ===epoch: | 484 | |
| Accuracy | Score | = | 77.6315789 | 47368 | 342 |
| | | :=== | ===epoch: | 485 | |
| Accuracy | Score | = | 78.9473684 | 21052 | 263 |
| | | -== | ===epoch: | 486 | |
| Accuracy | Score | = | 78.9473684 | 21052 | 263 |
| | | | ===epoch: | 487 | |
| Accuracy | Score | = | 78.0701754 | 38596 | S5 |



Accuracy Score = 78.0701754385965

============epoch: 498 =====================

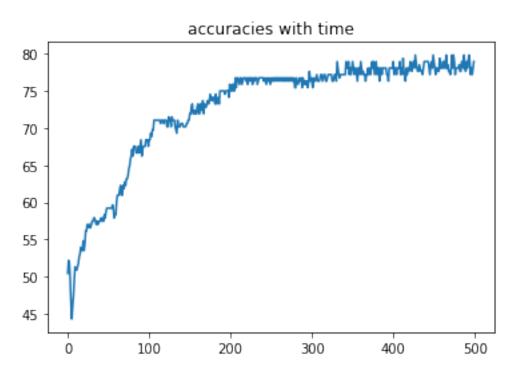
Accuracy Score = 77.19298245614034

===============epoch: 499 ======================

Accuracy Score = 78.0701754385965

==========epoch: 500 ==================

Accuracy Score = 78.94736842105263

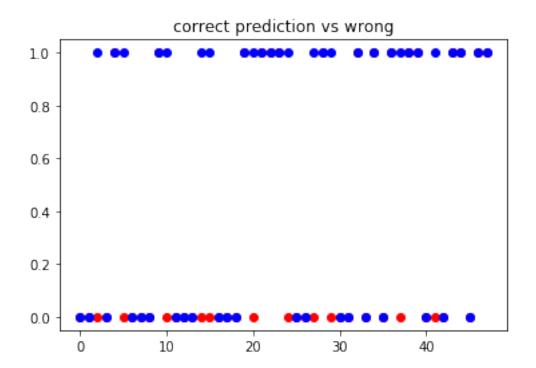


best weights: [-0.023625211146979477, 0.018778728412125, 0.09687985869456792, -0.01600469695446354, 0.4360274262594645, -0.002196244230574364]

[46]: best_weights = [-0.04474501103384887, 0.019309246725076966, 0.09594388284299246, -0.03536095737989012, 0.46637396064940045, 0.021523510357524298]
bestn1 = Node(best_weights[0])
bestn2 = Node(best_weights[1])
bestn3 = Node(best_weights[2])

```
bestn4 = Node(best_weights[3])
     bestn5 = Node(best_weights[4])
     bestn6 = Node(best_weights[5])
     best model = slp(thresh, c, [bestn1, bestn2, bestn3, bestn4, bestn5, bestn6])
[47]: testdf = pd.read_csv('LogRes_Test.csv')
     testdf = testdf.sample(frac = 1)
     testdf.head()
[47]:
         Latitude Longitude Altitude min Temo Max Temp Sunshine Hour \
                                            19.1
     2
             25.34
                       91.53
                                                      20.2
                                                                 3.100000
                                  1598
     1
            28.35
                       77.12
                                   216
                                             9.1
                                                      19.9
                                                                 0.621429
     26
             8.50
                       76.90
                                    64
                                            24.6
                                                      28.7
                                                                 0.550000
     3
            25.34
                       91.53
                                  1598
                                             9.4
                                                      10.5
                                                                 5.800000
     42
             8.50
                       76.90
                                    64
                                            25.4
                                                      32.2
                                                                 0.642857
         Solar Radiation
     2
                       0
     1
     26
                       1
                       0
     3
     42
                       1
[48]: x_test = testdf.iloc[:,:6]
     x_test = x_test.to_numpy()
     x_test = normalize(x_test)
     y_test = testdf.iloc[:,6]
     y_test = y_test.to_numpy()
[50]: | y_pred = best_model.testing(x_test, y_test)
     desired = 0 | predicted = 0
     desired = 0 | predicted = 0
     desired = 1 | predicted = 0
     desired = 0 | predicted = 0
     desired = 1 | predicted = 1
     desired = 1 | predicted =
     desired = 0 | predicted = 0
     desired = 0 | predicted = 0
     desired = 0 | predicted = 0
     desired = 1 | predicted = 1
     desired = 1 | predicted = 0
     desired = 0 | predicted = 0
     desired = 0 | predicted = 0
     desired = 0 | predicted = 0
     desired = 1 | predicted = 0
```

```
desired = 1 | predicted = 0
     desired = 0 | predicted = 0
     desired = 0 | predicted = 0
     desired = 0 | predicted = 0
     desired = 1 | predicted = 1
     desired = 1 | predicted = 0
     desired = 1 | predicted = 1
     desired = 1 | predicted = 1
     desired = 1 | predicted =
     desired = 1 | predicted = 0
     desired = 0 | predicted = 0
     desired = 0 | predicted = 0
     desired = 1 | predicted = 0
     desired = 1 | predicted =
     desired = 1 | predicted = 0
     desired = 0 | predicted =
     desired = 0 | predicted = 0
     desired = 1 | predicted =
     desired = 0 | predicted = 0
     desired = 1 | predicted = 1
     desired = 0 | predicted = 0
     desired = 1 | predicted =
     desired = 1 | predicted = 0
     desired = 1 | predicted = 1
     desired = 1 | predicted =
     desired = 0 | predicted = 0
     desired = 1 | predicted = 0
     desired = 0 | predicted = 0
     desired = 1 | predicted = 1
     desired = 1 | predicted =
     desired = 0 | predicted = 0
     desired = 1 | predicted = 1
     desired = 1 | predicted = 1
     accuracy = 0.770833333333334 correct: 37
[51]: import matplotlib.pyplot as plt
[52]: plt.plot(y_pred, 'ro')
     plt.plot(y_test, 'bo')
     plt.title('correct prediction vs wrong')
     plt.show()
```



[53]: mets = best_model.metrics(x_test, y_test)

Confusion Matrix:

[[16, 0] [11, 21]]

accuracy: 0.7708333333333333

precision: 1.0

recall: 0.5925925925925926

specificity: 1.0

f1 score: 2.3703703703703702

[]: