accuracy - ffaANN MLP V0.1

May 20, 2022

```
[]:
```

```
[1]: import math
    import numpy as np
    import pandas as pd
    # import seaborn as sns
    # import matplotlib.pyplot as plt
    # from sklearn.preprocessing import LabelEncoder
    # from sklearn.preprocessing import OneHotEncoder
    from sklearn.preprocessing import LabelBinarizer
    from sklearn import preprocessing
    from scipy.special import expit
    from numpy.random import default_rng
    from sklearn.metrics import classification_report
    from sklearn.metrics import confusion_matrix
    from sklearn.metrics import accuracy_score
    import time
    from accffaAnn_thread_V_I import *
    class MultiLayerPerceptron():
       # accepts a vector or list and returns a list after performing \Box
     →corresponding function on all elements
       Ostaticmethod
       def sigmoid(vectorSig):
           ⇔one"""
           sig = expit(vectorSig)
```

```
return sig
   Ostaticmethod
   def binaryStep(x):
        """ It returns '0' is the input is less then zero otherwise it returns \Box
⇒one """
       return np.heaviside(x, 1)
   Ostaticmethod
   def linear(x):
        """ y = f(x) It returns the input as it is"""
       return x
   Ostaticmethod
   def tanh(x):
        """ It returns the value (1-exp(-2x))/(1+exp(-2x)) and the value (1-exp(-2x))/(1+exp(-2x))
\hookrightarrow returned will be lies in between -1 to 1"""
       return np.tanh(x)
   Ostaticmethod
   def relu(x): # Rectified Linear Unit
        """ It returns zero if the input is less than zero otherwise it returns_{\sqcup}
\hookrightarrow the given input"""
       x1 = \Gamma
       for i in x:
            if i < 0:
                x1.append(0)
            else:
                x1.append(i)
       return x1
   Ostaticmethod
   def leakyRelu(x):
        """ It returns zero if the input is less than zero otherwise it returns_{\sqcup}
\hookrightarrow the given input"""
       x1 = []
       for i in x:
            if i < 0:
                x1.append((0.01 * i))
            else:
                x1.append(i)
       return x1
   Ostaticmethod
   def parametricRelu(self, a, x):
```

```
""" It returns zero if the input is less than zero otherwise it returns_{\sqcup}
\hookrightarrow the given input"""
       x1 = []
       for i in x:
           if i < 0:
               x1.append((a * i))
           else:
               x1.append(i)
       return x1
   Ostaticmethod
   def softmax(self, x):
       """ Compute softmax values for each sets of scores in x"""
       return np.exp(x) / np.sum(np.exp(x), axis=0)
   # ======= Activation Functions Part Ends ======= #
   # ====== Distance Calculation ======= #
   @staticmethod
   def chebishev(self, cord1, cord2, exponent_h):
       dist = 0.0
       if ((type(cord1) == int and type(cord2) == int) or ((type(cord1) ==_{\sqcup}
→float and type(cord2) == float))):
           dist = math.pow((cord1 - cord2), exponent_h)
       else:
           for i, j in zip(cord1, cord2):
               dist += math.pow((i - j), exponent_h)
       dist = math.pow(dist, (1.0 / exponent_h))
       return dist
   Ostaticmethod
   def minimum_distance(self, cord1, cord2):
       # min(|x1-y1|, |x2-y2|, |x3-y3|, ...)
       dist = float('inf')
       if ((type(cord1) == int and type(cord2) == int) or ((type(cord1) ==_{\sqcup}
→float and type(cord2) == float))):
           dist = math.fabs(cord1 - cord2)
       else:
           for i, j in zip(cord1, cord2):
               temp_dist = math.fabs(i - j)
               if (temp_dist < dist):</pre>
                   dist = temp_dist
       return dist
   Ostaticmethod
```

```
def maximum_distance(self, cord1, cord2):
       \# \max(|x1-y1|, |x2-y2|, |x3-y3|, ...)
       dist = float('-inf')
       if ((type(cord1) == int and type(cord2) == int) or ((type(cord1) ==__
→float and type(cord2) == float))):
           dist = math.fabs(cord1 - cord2)
       else:
           for i, j in zip(cord1, cord2):
               temp_dist = math.fabs(i - j)
               if (temp_dist > dist):
                   dist = temp_dist
       return dist
   Ostaticmethod
   def manhattan(self, cord1, cord2):
       \# |x1-y1| + |x2-y2| + |x3-y3| + \dots
       dist = 0.0
       if ((type(cord1) == int and type(cord2) == int) or ((type(cord1) == __
→float and type(cord2) == float))):
           dist = math.fabs(cord1 - cord2)
           for i, j in zip(cord1, cord2):
               dist += math.fabs(i - j)
       return dist
   Ostaticmethod
   def eucledian(self, cord1, cord2):
       dist = 0.0
       if ((type(cord1) == int and type(cord2) == int) or ((type(cord1) == __
→float and type(cord2) == float))):
           dist = math.pow((cord1 - cord2), 2)
       else:
           for i, j in zip(cord1, cord2):
               dist += math.pow((i - j), 2)
       return math.pow(dist, 0.5)
   # ====== Distance Calculation Ends ======== #
   def __init__(self, dimensions=(8, 5), all_weights=(0.1, 0.2),_

→fileName="iris"):
       11 11 11
       Arqs:
           dimensions: dimension of the neural network
           all_weights: the optimal weights we get from the bio-algoANN models
```

```
self.allPop_Weights = []
       self.allPopl_Chromosomes = []
       self.allPop_ReceivedOut = []
       self.allPop_ErrorVal = []
      self.all_weights = all_weights
      self.fitness = []
       \# ========= Input dataset and corresponding output
self.fileName = fileName
       self.fileName += ".csv"
       data = pd.read_csv(self.fileName)
      classes = []
      output_values_expected = []
       input values = []
       # ~~~~ encoding ~~~~#
       # labelencoder = LabelEncoder()
       # data[data.columns[-1]] = labelencoder.fit_transform(data[data.
\rightarrow columns [-1]])
       # one hot encoding - for multi-column
       # enc = OneHotEncoder(handle unknown='ignore')
       # combinedData = np.vstack((data[data.columns[-2]], data[data.
\hookrightarrow columns [-1])). T
       # print(combinedData)
       # y = enc.fit_transform(combinedData).toarray()
       # y = OneHotEncoder().fit_transform(combinedData).toarray()
       y = LabelBinarizer().fit_transform(data[data.columns[-1]])
       # print(y)
       # ~~~~ encoding ends~~~~#
      for j in range(len(data)):
           output_values_expected.append(y[j])
       # print(output_values_expected)
       input_values = []
       for j in range(len(data)):
```

```
b = []
          for i in range(1, len(data.columns) - 1):
              b.append(data[data.columns[i]][j])
          input_values.append(b)
      self.X = input_values[:]
      self.Y = output_values_expected[:]
       # input and output
      self.X = input_values[:]
      self.Y = output_values_expected[:]
      self.dimension = dimensions
       # print(self.dimension)
       # ======= Finding Initial Weights ======= #
      self.pop = [] # weights
      reshaped_all_weights = []
      start = 0
      for i in range(len(self.dimension) - 1):
          end = start + self.dimension[i + 1] * self.dimension[i]
          temp_arr = self.all_weights[start:end]
          w = np.reshape(temp_arr[:], (self.dimension[i + 1], self.
→dimension[i]))
          reshaped_all_weights.append(w)
          start = end
      self.pop.append(reshaped_all_weights)
      self.init_pop = self.all_weights
   # ======= Initial Weights Part Ends ======== #
  def Predict(self, chromo):
      # X, Y and pop are used
      self.fitness = []
      total_error = 0
      m arr = []
      k1 = 0
      for i in range(len(self.dimension) - 1):
          p = self.dimension[i]
          q = self.dimension[i + 1]
          k2 = k1 + p * q
          m_temp = chromo[k1:k2]
          m_arr.append(np.reshape(m_temp, (p, q)))
          k1 = k2
```

```
y_predicted = []
       for x, y in zip(self.X, self.Y):
           yo = x
           for mCount in range(len(m_arr)):
               yo = np.dot(yo, m_arr[mCount])
               yo = self.sigmoid(yo)
           # converting to sklearn acceptable form
           max_yo = max(yo)
           for y_vals in range(len(yo)):
               if(yo[y_vals] == max_yo):
                   yo[y_vals] = 1
               else:
                   yo[y_vals] = 0
           y_predicted.append(yo)
       return (y_predicted, self.Y)
   def main(self):
       Y_PREDICT, Y_ACTUAL = self.Predict(self.init_pop)
       Y_PREDICT = np.array(Y_PREDICT)
       Y_ACTUAL = np.array(Y_ACTUAL)
       n classes = 3
       label_binarizer = LabelBinarizer()
       label_binarizer.fit(range(n_classes))
       Y PREDICT = label_binarizer.inverse transform(np.array(Y_PREDICT))
       Y_ACTUAL = label_binarizer.inverse_transform(np.array(Y_ACTUAL))
       # find error
       \#print("\n Actual / Expected", Y_ACTUAL)
       #print("\n Predictions", Y_PREDICT)
       \#print("\n\nConfusion Matrix")
       #print(confusion_matrix(Y_ACTUAL, Y_PREDICT))
       \#print("\n\nClassification Report")
       target names = ['class 0', 'class 1', 'class 2']
       #print(classification_report(Y_ACTUAL, Y_PREDICT,__
→ target_names=target_names))
       #print("\n\n")
       return accuracy_score(Y_ACTUAL, Y_PREDICT)
```

```
[6]: start_time = time.time()
    i = InputData(fileName="iris")
    input_val, output_val = i.main()
    end_time = time.time()
    print("Time for inputting data : ", end_time - start_time)
    print("======= Calling FFA to get best weights ========")
    start_time = time.time()
    a = ffaAnn(initialPopSize=100, m=10, dimensions = [100,10],
     input_values=input_val, output_values_expected=output_val, iterations = 100)
    fit, b, weights, dim, all_gen_best_weight = a.main()
    end_time = time.time()
    print("Time taken : ", end_time - start_time)
    print("\n Fitness: ", fit, "\n Best Weights: ", weights, "\n Dimensions: ", u
     →dim)
    import matplotlib.pyplot as plt
    x=b[:]
    z=[i for i in range(0,100)]
    plt.plot(z,x)
    plt.title("Firefly Algorithm")
    plt.ylabel("Fitness")
    plt.xlabel("Iterations")
    end_time = time.time()
    print("Time Taken : ", end_time - start_time)
    Time for inputting data: 0.028992414474487305
    ======= Calling FFA to get best weights ==========
    -----GENERATION O-----
    Initial worst fitness = 939.1697997791503
    Initial best fitness = 368.3201652603337
    -----GENERATION 1-----
    -----GENERATION 2-----
    -----GENERATION 3-----
    hi
    hi
    hi
    hi
    hi
```

hi		
hi		
	GENERATION	4
	GENERATION	5
	GENERATION	6
	GENERATION	7
	GENERATION	8
	GENERATION	9
	GENERATION	10
	GENERATION	11
	GENERATION	12
	GENERATION	13
	GENERATION	14
	GENERATION	15
	GENERATION	16
	GENERATION	17
	GENERATION	18
hi	GENERALION	10
hi		
hi		
hi		
hihi		
1. 2		
hi	251155 A 15 T C 15	4.0
	GENERATION	19
hi		
hihi		
hi		
hi		
	GENERATION	20
hi		
hihi		
hi		
hi		
hihi		

hi		
hi		
hi		
	-GENERATION	21
	-GENERATION	22
hi		
hihi		
hi		
hi		
	-GENERATION	23
	-GENERATION	24
	-GENERATION	25
	-GENERATION	26
	-GENERATION	27
	-GENERATION	28
	-GENERATION	29
	-GENERATION	30
	-GENERATION	31
	-GENERATION	32
	-GENERATION	33
	-GENERATION	34
	-GENERATION	35
	-GENERATION	36
	-GENERATION	37
	-GENERATION	38
	-GENERATION	39
	-GENERATION	40
	-GENERATION	41
	-GENERATION	42
	-GENERATION	43
	-GENERATION	44
	-GENERATION	45
	-GENERATION	46
	-GENERATION	47
	-GENERATION	48
	-GENERATION	49
		50
	-GENERATION	51
	-GENERATION	52
		53
		54

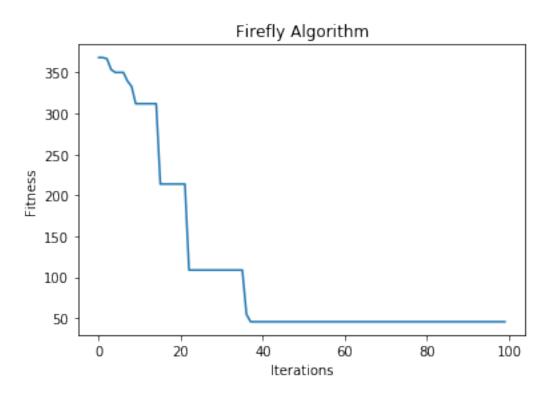
GENERATION	55
GENERATION	56
GENERATION	57
GENERATION	58
GENERATION	59
GENERATION	60
GENERATION	61
GENERATION	62
GENERATION	63
GENERATION	64
GENERATION	65
GENERATION	66
GENERATION	67
GENERATION	68
GENERATION	69
GENERATION	70
GENERATION	71
GENERATION	72
GENERATION	73
GENERATION	74
GENERATION	75
GENERATION	76
GENERATION	77
GENERATION	78
GENERATION	79
GENERATION	80
GENERATION	81
GENERATION	82
GENERATION	83
GENERATION	84
GENERATION	85
GENERATION	86
GENERATION	87
GENERATION	88
GENERATION	89
GENERATION	90
GENERATION	91
GENERATION	92
GENERATION	93
GENERATION	94
GENERATION	95
GENERATION	96
GENERATION	97
GENERATION	98
GENERATION	99
Fitness : 45 5221176084	

Fitness: 45.52211760841647 Time taken: 168.88895511627197 Fitness: 45.52211760841647

Best Weights: [12, 1, 19, 12, 0, 18, 25, -3, 2, -7, -13, 0, 0, -5, -18, 0, 15, 2, 4, -17, -11, -13, -10, -6, 3, 10, 2, 6, 7, -1, 17, 9, -3, -7, -20, 14, -18, 0, -20, 0, 11, 15, -14, -5, 13, 0, 6, -2, 2, -23, 0, 3, -21, 7, 0, 3, 1, 1, -9, 19, -10, 0, 0, 15, 1, 0, -1, 11, -6, -5, 3, 13, 0, 7, -1, -9, 3, -15, -17, 13, -24, 9, 0, 0, 13, -18, -18, -10, -32, -4, -2, 2, -1, 24, 10, -3, -17, 1, 10, 4, 27, 10, 10, 1, 8, 13, 9, -14, 0, -4, -1, 7, 5, 15, -14, 1, 16, 1, 16, -4, 14, 8, 0, 16, -4, 24, 16, 0, 2, -10, 26, 1, 21, 0, -5, 3, 16, 14, 5, 0, -16, -25,-9, -4, 0, 12, 17, -15, -1, -12, 0, 0, 6, -7, 4, -4, 0, -13, -5, 14, -6, -9, 0, -22, 0, -11, 10, 0, -13, -3, -13, -6, 0, -25, 16, -24, 15, 3, -4, 9, 0, 1, 0, 0, 16, -2, 0, 6, -7, -11, -4, -5, 18, -5, -6, -16, -2, -3, -8, -1, 12, 6, -4, 23,5, -12, -14, -12, 0, -5, 16, -4, -3, -17, -2, -1, 0, -23, -1, 0, 12, 0, -6, -1, 20, 18, -7, -20, 14, -15, -26, 18, -13, -1, 12, -1, 22, 4, 17, 9, 5, -3, 3, 21, 13, 22, -2, -19, 3, 1, -10, 3, 0, -24, 0, -25, 0, 0, 15, 13, 12, -3, -11, 4,-12, 7, -1, 2, 0, 0, 13, 35, 6, 4, -4, -7, 6, -1, 4, -3, 0, -2, -24, -1, -15, -1, -11, 12, 12, -1, 29, 13, -5, -15, -10, -5, -15, 21, -4, 0, 2, -8, -11, 3, -6, -2, 0, -10, -20, 10, 1, -7, 0, 13, 2, -21, -24, 9, -5, 2, 9, -4, 14, 8, 0, -3, 28, -21, 8, -11, 0, -8, -2, -7, -5, 8, -17, -1, -1, -30, 4, -11, 3, 0, 1, -13, 4, 10, 16, 6, -19, 24, 8, 22, 5, 1, 0, 0, 0, -20, -16, 25, -15, -2, -2, -9, 3, 8, 0, -2, -10, -18, -8, -14, 4, 0, -5, 8, 13, -1, -17, 7, 15, 7, 21, 4, -22, -15, -18, -23, 1, 10, 24, 12, -5, 3, 1, -8, 4, 16, 14, 3, -6, 29, -12, -32, 5, 0, -13, -2, -21, -15, 26, 11, 7, 2, 7, -7, -1, -17, -18, 1, -21, -19, 3, -26, 10, -1, 4, -20, 5, 11, 12, -1, -11, 20, 0, 0, 0, -11, 13, -22, 7, -15, -8, 0, 0, 12, 9, 15, -13, -11, 3, -5, -6, 22, 19, 17, 1, 5, -1, -1, -8, -13, 1, 31, 5, -7, -6, 8, 10, 5, 21, -14, -15, 10, -2, -16, 8, -27, -21, 22, 23, 3, 4, 0, 14, -4, 11, 9, 8, -9, 7, 4, 0, 6, -16, 20, -13, 0, 17, -15, 0, 0, 9, 22, 0, 20, -1, -7,-6, -4, -11, -17, 0, -19, 1, 8, -3, 0, -3, 8, 0, 9, -4, -2, 14, 7, 14, -6, 22, 15, -4, -4, 0, -7, 13, -1, 11, -11, -16, -34, 2, -16, 5, 1, 24, -15, -14, 14,-4, -4, 7, 8, 6, 14, 2, -4, -19, 0, -18, -9, -14, 0, 0, -2, 5, 0, 8, 6, 9, 2, 6, -19, -10, -6, 0, 15, 14, -12, -15, 5, 9, -9, -6, -1, 30, 1, -14, -3, -17, -6, 21, -3, -26, -1, 11, 11, 0, 14, 16, 17, -12, -26, 6, 0, 0, 13, 16, -16, 11, -14, 0, 3, 4, 17, -4, 0, 24, 8, -4, -9, -20, 15, 15, 14, 0, 3, 0, -10, -10, -19, -19,-18, -1, -6, 14, -36, -5, 0, -20, -20, -13, -11, -1, -19, -4, -2, -16, -13, 14, -1, 3, 0, -10, 0, 9, -18, -5, -11, -7, 6, 12, -1, -4, 16, -21, 11, -4, 11, -5, 7, -1, 15, -2, 2, -14, 5, 4, -6, -12, 9, 3, -1, 13, -11, -12, -23, -3, 10, -18,-9, 0, 0, 16, 0, -12, -17, -12, 22, -3, -7, 0, -6, -19, 1, -2, -6, -15, -9, 9, 3, -2, 4, -7, 1, 11, 4, -24, 26, 17, 15, 6, 12, -10, 30, 1, -18, 10, 0, 4, 8, 6,0, -19, -7, -1, 0, 0, -1, 0, -10, 0, -10, -10, 4, 21, 18, 25, 0, 1, -16, -4, 5,11, -26, 0, 1, 7, -1, -5, 8, -8, -3, 19, -1, 12, 0, -5, -2, 3, 2, -16, -15, 0,2, 2, 2, 22, 8, -8, -21, -5, 16, 17, 0, 29, 9, -11, -13, -2, 4, -3, 7, -8, 4,-8, 1, 0, -14, 15, 16, 2, -6, 2, 0, -8, -1, 3, -24, -18, -3, 0, 8, -3, -16, 21, 3, 0, -10, 14, 10, -13, 1, 18, -4, 8, -18, 19, -1, 8, -19, -2, 0, -16, 31, -17,31, 0, -5, -5, -1, -2, 15, -18, -9, 3, 16, 0, -3, -6, 1, -2, 12, 13, 4, 19, -1,2, -7, 11, 4, -4, -16, 0, 2, -7, 3, -1, 0, -11, 18, 11, 8, 6, 4, 2, 0, 0, 9,-15, -10, -9, -8, 9, -2, 2, -11, -14, 18, 11, 8, -7, -4, 6, 9, 0, 10, -13, 0, 0, -3, 21, 24, 5, 2, 13, 2, 15, 28, -17, 9, 10, 17, 1, 0, 0, 10, 10, -14, -2, -20, -14, 4, 24, 0, 0, -17, 0, -5, -13, 7, 16, 0, -2, 17, 17, -7, 23, 12, 0, -12, -20, -1, 8, 3, 1, 6, 3, 7, -23, 17, -5, 14, -5, -7, 11, 21, 9, -15, -17, 4, 7,

13, 1, 0, -3, -10, -7, 21, 23, -2, -9, 12, 4, -1, -17, 0, 20, 3, 2, -24, 29, -4, 0, -1, -19, -13, 22, 0, 0, 1, 18, 0, -22, 3, -8, -18, 3, 0, 2, -6, -7, 18, 0, 0,17, -13, -4, -15, 2, 10, -3, 7, 2, 18, -2, 0, 2, 8, 1, -10, 15, 16, -11, 6, 5,16, -3, 16, 0, 0, -5, 19, 1, -6, 11, -4, 7, -2, -20, -6, -9, 0, 9, -11, 0, 15, 5, 9, -6, 20, -20, -25, 0, 4, -8, -7, 15, 9, -12, -19, -1, -12, -24, 9, 6, 1, -26, -7, -25, 9, 5, 13, 5, -12, 18, 2, -1, -9, 0, 9, 4, -3, 0, 6, 5, 8, 4, -19, 4, -5, 17, -18, -33, -1, -17, 0, 28, -15, -3, -10, -34, 0, 6, 3, -12, 7, -14, 0,1, 16, -21, -5, -1, 18, 12, 0, -13, 0, 3, 9, 1, 0, 9, -17, 19, 2, 15, 6, 4, 4, 0, 1, -12, 10, 5, -3, 3, -15, 13, 4, 20, -19, 7, 3, 21, 0, 0, -5, -16, -4, 19,0, 0, 10, 1, 2, -27, -14, -1, 16, 9, -4, -9, 4, 0, -1, 1, 5, 27, -2, 10, -12, 2,0, -2, 19, 0, -1, -10, -8, -8, -15, 26, 4, -15, -4, -15, 2, 7, 1, 13, -8, -8,-1, 9, -1, -14, 8, 3, -1, 8, 11, -13, 0, -14, -10, -20, 0, -11, 20, -17, 18, 16, -5, -15, 20, 22, -10, 0, -9, 0, -12, -11, 0, -20, -1, 18, -7, 15, -5, 11, 2, 4, -6, 13, 0, 8, 4, -1, -16, -2, 4, 25, 8, 11, -19, -12, -1, 9, 13, -11, 5, -28, -16, 13, 6, 16, 4, 4, 4, 0, -14, 0, -6, -8, -2, 22, -4, 0, -3, -4, -1, -1, -8, 0, -12, -2, -3, 7, -8, -16, -2, 5, -14, 2, 4, -14, 19, 5, 2, 0, 0, -8, 0, -18,0, -11, -10, 14, 0, 25, 8, -35, 4, 4, 0, 0, 0, 0, 18, 9, 18, 0, 4, -20, 1, 18, -22, 5, 9, -2, 0, 0, 0, 6, 8, 9, 33, -5, -1, -6, 3, -2, -3, 8, 1, 27, 5, -18, 9, -9, -1, -1, -18, 0, -17, 10, 6, -13, 0, 17, 0, 20, 6, 16, -7, 3, -18, 9, 20, -15, -19, 5, 1, -2, 4, 1, -13, 5, 0, -4, -3, -10, -19, -1, -2, 1, 8, -2, 0, 0, -9, 12, -8, -6, 1, -14, -1, 5, 7, -4, 11, -29, 9, 25, -1, -15, -6, -6, 5, 13, 8, 0, -8, -14, 4, -25, 8, 14, -15, -15, -3, 9, 5, -12, 19, -10, 0, -19, -9, 11, 5,11, -20, 15, -10, -1, 7]

Dimensions: [4, 100, 10, 3] Time Taken: 168.90278959274292



```
[7]: | print("\n\n========== MLP Program Begins ========")
    start_time = time.time()
    print("Training")
    m = MultiLayerPerceptron(fileName="iris_train", dimensions=dim,__
     →all_weights=weights)
    m.main()
    end time = time.time()
    print("Time taken = ", end_time - start_time)
    ====== MLP Program Begins ========
    Training
    Time taken = 0.019589900970458984
[8]: start_time = time.time()
    print("Testing")
    m = MultiLayerPerceptron(fileName="iris_test", dimensions=dim,__
     →all_weights=weights)
    m.main()
    end_time = time.time()
    print("Time taken = ", end_time - start_time)
    Testing
    Time taken = 0.008346796035766602
[9]: all_accuracy = []
    for weights in all_gen_best_weight:
        m = MultiLayerPerceptron(fileName="iris_train", dimensions=[4,100,10,3],__
     →all_weights=weights)
        accuracy val = m.main()
        print(accuracy_val)
        all_accuracy.append(accuracy_val)
    import matplotlib.pyplot as plt
    x=all_accuracy[:]
    z=[i for i in range(len(x))]
    plt.plot(z,x)
    plt.title("Firefly Algorithm")
    plt.ylabel("Accuracy")
    plt.xlabel("Iterations")
```

0.5583333333333333

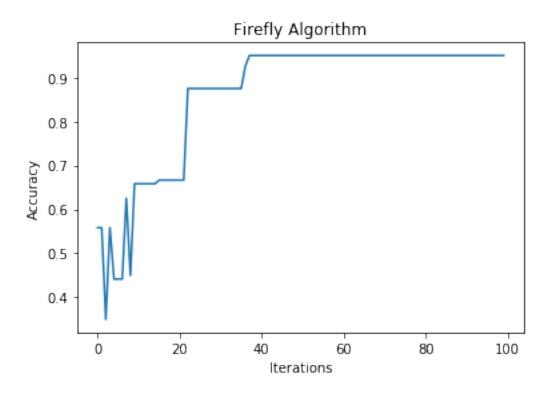
- 0.5583333333333333
- 0.35
- 0.5583333333333333
- 0.441666666666665
- 0.441666666666665
- 0.441666666666665
- 0.625
- 0.45
- 0.6583333333333333
- 0.6583333333333333
- 0.6583333333333333
- 0.6583333333333333
- 0.6583333333333333
- 0.65833333333333333
- 0.66666666666666
- 0.66666666666666
- 0.666666666666666
- 0.66666666666666
- 0.66666666666666
- 0.66666666666666
- 0.66666666666666
- 0.875
- 0.875
- 0.875
- 0.875
- 0.875
- 0.875
- 0.875
- 0.875
- 0.875
- 0.875
- 0.875
- 0.875
- 0.875
- 0.875
- 0.925
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95 0.95
- 0.95

- 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95
- 0.95 0.95
- 0.95
- 0.95

0.95 0.95

0.95

[9]: Text(0.5,0,'Iterations')



[]: