ffa ANN without threading MLP $\rm V0.1$

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
    import time
    from ffaAnn_V_I import *
    class MultiLayerPerceptron():
        # accepts a vector or list and returns a list after performing
     →corresponding function on all elements
        Ostaticmethod
        def sigmoid(vectorSig):
            """returns 1/(1+exp(-x)), where the output values lies between zero and
     ⇒one"""
            sig = expit(vectorSig)
            return sig
```

```
@staticmethod
   def binaryStep(x):
       """ It returns '0' is the input is less then zero otherwise it returns_{\sqcup}
⇔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_{\sqcup}
\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}
⇔the given input"""
       x1 = []
       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 = \prod
       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))
```

```
[2]: 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=1, dimensions = [100,10],
 input_values=input_val, output_values_expected=output_val, iterations = 100)
fit, b, weights, dim = a.main()
end_time = time.time()
print("Time taken : ", end_time - start_time)
print("\n Fitness : ", fit, "\n Best Weights : ", weights, "\n Dimensions : ", \u00c4
 →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: 0.016080141067504883
======= Calling FFA to get best weights ========
-----GENERATION O-----
hi
hi
hi
hi
hi
```

-----GENERATION 1-----

-----GENERATION 2-----

hi hi hi hi

hi

hi		
hi		
hi		
	GENERATION	3
hi		
hi h:		
hi hi		
hi		
	GENERATION	4
hi		
hi		
hi		
	GENERATION	5
hi		
hi 	APAPD ATTON	C
hi	GENERATION	0
hi		
	GENERATION	7
hi		
hi		_
	GENERATION	8
hi		
hi hi		
hi		
	GENERATION	9
hi		
hi		
hi		
	GENERATION	10
hi		
hi		4.4
	GENERATION	11

hi		
hi		
hi		
	GENERATION	12
hi		
hi hi		
	GENERATION	13
hi	GENERALION	10
hi		
hi		
	GENERATION	14
hi		
hi		
hi		
	GENERATION	15
hi		
	GENERATION	16
hi		
hi		
	GENERATION	17
hi	APAIPD ATTON	10
hi	GENERATION	18
hi		
	GENERATION	19
hi		10
	GENERATION	20
hi		
hi		
	GENERATION	21
hi		
	GENERATION	22
	GENERATION	23
hi		
	GENERATION	24
hi		
hi		
	GENERATION	25
hi		
hi	GENERATION	26
hi	GENEKATIUN	ZU
hi		
	GENERATION	27
	QD11D111111111111111111111111111111111	

GENERATION	28
GENERATION	29
hi	
GENERATION	30
hi	
GENERATION	31
hi	
GENERATION	32
hi	
GENERATION	33
hi	
GENERATION	34
hi	
GENERATION	35
hi	
GENERATION	36
hi	30
GENERATION	37
hi	31
GENERATION	38
GENERATION	39
GENERATION	40
GENERATION	41
GENERATION	42
GENERATION	43
GENERATION	44
GENERATION	45
GENERATION	46
GENERATION	47
GENERATION	48
GENERATION	49
GENERATION	50
GENERATION	51
GENERATION	52
GENERATION	53
GENERATION	54
GENERATION	55
GENERATION	56
GENERATION	57
GENERATION	58
GENERATION	59
hi	
GENERATION	60
GENERATION	61
GENERATION	62
GENERATION	63
GENERATION	64
GENERATION	65
OLIVLIANI TON	

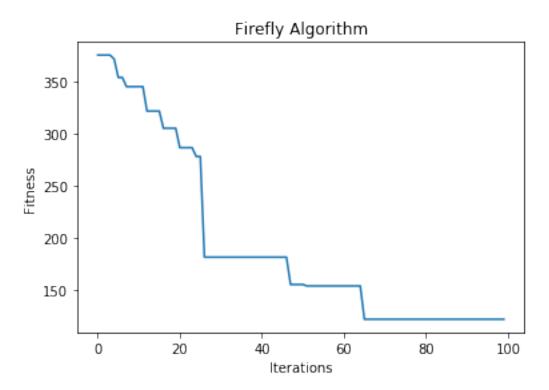
	-GENERATION	66
hi		
	-GENERATION	67
	-GENERATION	68
	-GENERATION	69
	-GENERATION	70
	-GENERATION	71
	-GENERATION	72
	-GENERATION	73
	-GENERATION	74
	-GENERATION	75
	-GENERATION	76
	-GENERATION	77
	-GENERATION	78
	-GENERATION	79
	-GENERATION	80
	-GENERATION	81
	-GENERATION	82
	-GENERATION	83
	-GENERATION	84
	-GENERATION	85
	-GENERATION	86
	-GENERATION	87
	-GENERATION	88
	-GENERATION	89
	-GENERATION	90
	-GENERATION	91
	-GENERATION	92
	-GENERATION	93
hi	GENED ATTON	0.4
	-GENERATION	94
hi	GENER ARTON	0.5
	-GENERATION	95
hi		
	-GENERATION	96
hi		
hi		
	-GENERATION	97
hi		
hi		
	-GENERATION	98
hi		
hi		
	-GENERATION	99
hi		
hi		
hi		
Fitness: 12	1.8640537866	66149

Time taken: 84.98340249061584

Fitness: 121.86405378666149

Best Weights: [0, 4, 15, 0, 24, 0, 0, 0, -7, 1, 0, 0, 0, 0, 0, 5, -2, 0, 0, 0, -2, 0, -5, 0, 0, 1, 8, 0, 4, 2, 2, 67, 0, -7, 0, 0, -2, -5, -2, 1, 0, -1,-29, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 3, 0, 44, 0, 1, 0, 0, -36, 2, 10, 0, 0, 0, 2, 0, 2, 0, 2, 0, 0, 0, -2, 0, 28, 0, -6, 0, 0, 0, 0, 0, -18, 0, 0, 0, 0, 0, 3, 1, 0, 0, -7, -13, -3, 0, 0, 0, -3, 0, -5, 0, 0, 0, -5, -6, 0, 0, 0, -29, 0, 1, 0, 0, 0, 0, 0, 13, 51, 1, 0, 0, 0, -6, 0, 0, 0, 0, 0, 1, -3, 0, 0, 0, -21,0, 0, 0, 0, -28, -1, 2, 1, 0, 0, 0, -2, 0, 7, 2, 0, 0, -8, -28, 2, 2, 0, -5, -8, -2, 4, -3, 0, 0, 0, 0, 0, 0, 0, 2, 1, 2, 0, 0, 0, 3, -5, 3, -6, 2, 0, 0, 0, -14, -1, 0, 0, 0, 0, 0, -2, 0, 0, -4, -8, 0, 10, 0, 0, 0, 5, 0, 2, 0, 0, -2, 0, -2, 0, 4, 0, 0, -6, 0, 0, 0, 1, 0, 0, 0, 0, 19, 34, 0, 24, 0, 11, -5, -5, 0, 0, 10, -5, 12, 14, 0, 0, -2, 0, 0, 0, -6, 3, 1, 3, 0, -8, 19, 0, 0, -34, 34, 0, 3, 0, 0, 14, 0, 2, 0, 2, -46, 0, 0, 0, 0, 0, 8, 0, -1, 0, 0, 0, 0, 1, 1, 0, 38, 0, 0, 28, 0, 0, 0, -42, -2, 1, 1, 0, 0, 0, 0, -12, 0, 0, 0, 0, 12, 1, -5, 0, -16, 0, 9, -7, 0, 0, 0, -2, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 4, 0, 0, -5, 0, 1, 0, -2, 0, 0, 24, 0, 34, 0, 0, 0, 0, -8, 0, 0, -13, 0, -2, 0, 20, 0, 4, 32, 0, 2, 16, -5, 0, 4, 16, 0, -2, 1, 0, 0, 0, -5, 0, 0, 0, 0, 0, 0, 1, 0, -22, 0, 0, 0, 0, 0, 0, 34, -8, -3, 0, 48, 1, 0, 0, -2, 0, -7, 0, 0, 0, 15, 0, 30, 0, 0, -5, 0, 0, 0, 0, -2, 12, 0, -11, 0, 8, 0, 0, 1, 0, -37, 2, 6, -4, 1, 0, 0, -2, 43, 0, 0, -35, 0, 7, 0, 0, 11, 8, 0, 2, 4, 0, 5, 0, -28, 0, 6, 6, 0, 0, 0, 2, 0, 0, 2, 0, 0, 0, 0, 0, 3, 0, 0, 0, 0, 0, 0, 0, 3, -1, 0, 0, 0, 0, 0, 0, 20, 0, 0, 9, 0, 27, 0, -2, 0, -44, 1, 0, 10, 5, -32, 0, 0, 0, 0, 0, -3, 5, 0, 0, 8, -23, -37, 0, -28, -19, 0, -18, -40, 0, 24, 0, 0, 7, -2, 0, -11, 0, 1, -27, -11, 4, 0, 0, 8, -2, 0, 0, 0, 3, 0, -3, 0, 0, 0, 0, 0, 0, 14, -2, -45, 17, 0, -7, 0, -3, 0, 0, 0, 0, -1, 2, 0, -1, 0, 0, 0, 2, 0, -2, -3, 0, 0, 0, 0, 2, 0, -1, -2, 0, 13, 2, 5,-2, 2, 0, 0, 0, 23, 0, 0, -36, 0, 0, 0, 0, -2, 0, 2, 8, 0, -3, 0, -33, 0, -8, 0, 0, 0, -24, -2, 2, 0, -21, 2, 0, -4, -34, 0, -29, 0, 0, 0, 0, 43, 0, 13, 4, 0, 0, 0, 2, 8, 0, 0, 15, 0, -28, 0, 0, 1, 0, 18, 0, -24, 0, 2, -6, 0, 0, 0, -2, 0, 0, -23, 0, 17, 0, 0, -2, 2, 6, 0, -3, 1, 2, -17, 0, 3, 6, 0, 0, 2, -4, 0, 12, 0, 7, 0, 0, -6, 0, 0, 0, 3, 0, 1, 0, 0, 0, 0, -14, 0, 0, 0, 0, -21, 0, 0, -5, 0, 8, 0, 0, 2, 8, 0, 2, 0, 0, 2, 0, 1, -5, 0, 0, 0, -5, -47, -18, 0, 0, -18, 0, 1, -3,-18, -3, 0, 2, 0, 0, 0, 0, 8, 2, 0, 2, 7, 0, 0, 6, 4, 0, 13, 0, 0, 0, 2, -2, 0, 8, 0, 0, 0, 0, 27, 0, 0, 8, -4, 0, 0, 0, 0, 0, -4, 0, 0, 0, 0, 0, 2, -30, 0, 1, -4, 0, 0, 0, 0, 1, 0, 39, 0, 0, 5, 0, 0, -4, 0, 0, 0, 28, 0, 8, 0, 0, 2, 0, 0, 0, 0, 0, -19, 0, -3, 0, -30, 0, 0, 0, -2, 0, -31, 0, 0, 0, 0, 29, -11, -18, -3, 0, 0, -6, -25, 0, 0, 1, 0, 0, 1, -2, 0, 0, -3, 0, -2, 2, 0, 17, 1, 11, 0, 0, -17, 0, 0, 0, 1, 0, 12, 0, 20, 0, 0, -3, 0, -6, 1, 0, 0, 0, 0, -5, 0, 0, 0, -1, 0, -10, -2, 4, 2, 0, 3, 0, 0, 0, 0, -4, 0, 0, 0, 1, -2, 0, 0, 0, 2, 0, -16, 0,0, -5, 0, -9, 0, 0, 0, 0, 21, -9, 0, 0, 0, 0, 2, 1, 0, 3, 2, 0, 31, 0, 0, 0, 0, 35, 0, 0, 36, 0, 4, 0, 5, 0, -12, 0, -21, 0, 2, 0, 4, 0, 0, 9, 0, 0, 0, -20, 2, 0, 0, 0, -2, 2, 14, 0, 0, 7, 0, 0, 1, 4, 0, 11, 0, 0, -3, 0, 0, 0, 0, -6,34, 33, 16, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 2, 0, -2, 0, 0, 0, -9, 2, 2, 0, 1, 0, 0, 0, 0, 0, 0, 0, 2, 0, 0, 0, 13, 0, 41, -6, 0, 0, 0, 0, 0, -2, 0, 0, 5, 0, 13, 0, 0, 0, 0, -29, 0, -6, 0, 0, 20, 0, 0, 5, 0, 0, 0, 0, 0, 1, 0, 0, -3, 2, 20, 7, 1, 0, 0, -1, 10, 3, 0, 13, 0, -5, 0, 0, 0, -31, 0, 0, -33, 0,

Dimensions: [4, 100, 10, 3] Time Taken: 85.01589250564575



```
m.main()
end_time = time.time()
print("Time taken = ", end_time - start_time)
```

```
====== MLP Program Begins =======
```

Training

Confusion Matrix

[[40 0 0] [0 36 4] [0 4 36]]

Classification Report

	precision	recall	f1-score	support
class 0	1.00	1.00	1.00	40
class 1	0.90	0.90	0.90	40
class 2	0.90	0.90	0.90	40
accuracy			0.93	120
macro avg	0.93	0.93	0.93	120
weighted avg	0.93	0.93	0.93	120

Time taken = 0.020166397094726562

```
[4]: 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

Confusion Matrix

[[9 0 1]

[0 10 0]

[0 0 10]]

Classification Report

	precision	recall	f1-score	support
class 0	1.00	0.90	0.95	10
Class U	1.00	0.90	0.95	10
class 1	1.00	1.00	1.00	10
class 2	0.91	1.00	0.95	10
accuracy			0.97	30
macro avg	0.97	0.97	0.97	30
weighted avg	0.97	0.97	0.97	30

 $Time\ taken = 0.010457515716552734$

[]: