

psoANN MLP V0.1

May 20, 2022

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```
[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 psoAnn_thread_V_I import *

class MultiLayerPerceptron():
    # ===== Activation Functions ===== #

    # accepts a vector or list and returns a list after performing
    ↪ corresponding function on all elements

    @staticmethod
    def sigmoid(vectorSig):
        """returns 1/(1+exp(-x)), where the output values lies between zero and
        ↪ one"""
        sig = expit(vectorSig)
        return sig
```

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    @staticmethod
    def binaryStep(x):
        """ It returns '0' if the input is less than zero otherwise it returns 1
        ↪ one """
        return np.heaviside(x, 1)

    @staticmethod
    def linear(x):
        """  $y = f(x)$  It returns the input as it is """
        return x

    @staticmethod
    def tanh(x):
        """ It returns the value  $(1 - \exp(-2x)) / (1 + \exp(-2x))$  and the value
        ↪ returned will be lies in between -1 to 1 """
        return np.tanh(x)

    @staticmethod
    def relu(x): # Rectified Linear Unit
        """ It returns zero if the input is less than zero otherwise it returns
        ↪ the given input """
        x1 = []
        for i in x:
            if i < 0:
                x1.append(0)
            else:
                x1.append(i)

        return x1

    @staticmethod
    def leakyRelu(x):
        """ It returns zero if the input is less than zero otherwise it returns
        ↪ the given input """
        x1 = []
        for i in x:
            if i < 0:
                x1.append((0.01 * i))
            else:
                x1.append(i)

        return x1

    @staticmethod
    def parametricRelu(self, a, x):

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        """ It returns zero if the input is less than zero otherwise it returns
        ↳ the given input """
        x1 = []
        for i in x:
            if i < 0:
                x1.append((a * i))
            else:
                x1.append(i)

        return x1

    @staticmethod
    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) ==
        ↳ 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

    @staticmethod
    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) ==
        ↳ 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

    @staticmethod

```

```

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

@staticmethod
def manhattan(self, cord1, cord2):
    # |x1-y1| + |x2-y2| + |x3-y3| + ...
    dist = 0.0
    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):
            dist += math.fabs(i - j)
    return dist

@staticmethod
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"):

    """
    Args:
        dimensions : dimension of the neural network
        all_weights : the optimal weights we get from the bio-algoANN models
    """

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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.
→ columns[-1]])

# one hot encoding - for multi-column
# enc = OneHotEncoder(handle_unknown='ignore')
# combinedData = np.vstack((data[data.columns[-2]], data[data.
→ 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)):

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        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

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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))

```

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[2]: start_time = time.time()
i = InputData(fileName="iris")
input_val, output_val = i.main()

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end_time = time.time()
print("Time for inputting data : ", end_time - start_time)

print("===== Calling PSO to get best weights =====")

start_time = time.time()

a = psoAnn(initialPopSize=100, m=10, input_values=input_val,
    ↪output_values_expected=output_val, iterations = 100, dimensions = [100,10])

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 : ",
    ↪dim)

import matplotlib.pyplot as plt
x=b[:]
z=[i for i in range(0,100)]
plt.plot(z,x)

plt.title("PSO")
plt.ylabel("Fitness")
plt.xlabel("Time")
end_time = time.time()
print("Time Taken : ", end_time - start_time)

```

```

Time for inputting data : 0.015381574630737305
===== Calling PSO to get best weights =====
-----GENERATION 0-----
-248.51308995240814
-----GENERATION 1-----
-249.9974025700157
-----GENERATION 2-----
-249.1693744959391
-----GENERATION 3-----
-249.9974467734275
-----GENERATION 4-----
-249.00000000001853
-----GENERATION 5-----
-237.96758065438723
-----GENERATION 6-----
-215.99999999999508

```



```

-----GENERATION 7-----
-204.00001653324261
-----GENERATION 8-----
-203.92682667652082
-----GENERATION 9-----
-206.82436412821843
-----GENERATION 10-----
-202.99999722948934
-----GENERATION 11-----
-150.999935747535
-----GENERATION 12-----
-149.99999999999915
-----GENERATION 13-----
-150.0
-----GENERATION 14-----
-150.0
-----GENERATION 15-----
-150.0
-----GENERATION 16-----
-150.0
-----GENERATION 17-----
-149.99999999996624
-----GENERATION 18-----
-145.96337158842206
-----GENERATION 19-----
-149.99971511840522
-----GENERATION 20-----
-148.390944310847
-----GENERATION 21-----
-137.23216815581006
-----GENERATION 22-----
-128.00746423932443
-----GENERATION 23-----
-137.99963219007728
-----GENERATION 24-----
-143.40372811150294
-----GENERATION 25-----
-167.00697245747614
-----GENERATION 26-----
-167.00927856274234
-----GENERATION 27-----
-170.00902514703466
-----GENERATION 28-----
-168.72655680769233
-----GENERATION 29-----
-168.86044327348242
-----GENERATION 30-----
-170.64790427755884

```

```

-----GENERATION 31-----
-160.52724784196366
-----GENERATION 32-----
-146.9452066865872
-----GENERATION 33-----
-141.741666415709
-----GENERATION 34-----
-141.6810335614381
-----GENERATION 35-----
-141.6810335614381
-----GENERATION 36-----
-140.87211432141098
-----GENERATION 37-----
-140.94593889116712
-----GENERATION 38-----
-143.78533709558195
-----GENERATION 39-----
-143.95371299177813
-----GENERATION 40-----
-144.0518908321367
-----GENERATION 41-----
-144.35142773322832
-----GENERATION 42-----
-145.50954748756055
-----GENERATION 43-----
-145.50954748756055
-----GENERATION 44-----
-145.50954748756055
-----GENERATION 45-----
-145.50954748756055
-----GENERATION 46-----
-145.50954748756055
-----GENERATION 47-----
-145.50954748756055
-----GENERATION 48-----
-145.50954748756055
-----GENERATION 49-----
-145.50954748756055
-----GENERATION 50-----
-145.50954748756055
-----GENERATION 51-----
-145.50954748756055
-----GENERATION 52-----
-140.40489298712
-----GENERATION 53-----
-140.40489298712
-----GENERATION 54-----
-140.40489298712

```

```

-----GENERATION 55-----
-140.40489298712
-----GENERATION 56-----
-140.40489298712
-----GENERATION 57-----
-140.40489298705023
-----GENERATION 58-----
-140.40489293311717
-----GENERATION 59-----
-143.9915981335279
-----GENERATION 60-----
-143.99159790831382
-----GENERATION 61-----
-143.99159790831382
-----GENERATION 62-----
-143.99159813350943
-----GENERATION 63-----
-143.99159813350943
-----GENERATION 64-----
-143.99159813350943
-----GENERATION 65-----
-143.99159813350943
-----GENERATION 66-----
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-----GENERATION 69-----
-143.99159813350943
-----GENERATION 70-----
-143.99159813350943
-----GENERATION 71-----
-143.99159813350943
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-----GENERATION 73-----
-143.99159813350943
-----GENERATION 74-----
-143.99159813350943
-----GENERATION 75-----
-143.99159813350943
-----GENERATION 76-----
-143.99159813350943
-----GENERATION 77-----
-143.99159813350943
-----GENERATION 78-----
-143.99159813350943

```

```

-----GENERATION 79-----
-140.40489293311003
-----GENERATION 80-----
-140.40489293311003
-----GENERATION 81-----
-140.40489293311003
-----GENERATION 82-----
-140.40489293311003
-----GENERATION 83-----
-140.40489293311003
-----GENERATION 84-----
-140.40489293311003
-----GENERATION 85-----
-140.40489293311003
-----GENERATION 86-----
-140.40489293311003
-----GENERATION 87-----
-140.40489293311003
-----GENERATION 88-----
-140.40489293311003
-----GENERATION 89-----
-140.40489293311003
-----GENERATION 90-----
-143.99159813353864
-----GENERATION 91-----
-143.99159813353864
-----GENERATION 92-----
-143.99159813353864
-----GENERATION 93-----
-143.99159813353864
-----GENERATION 94-----
-143.99159813353864
-----GENERATION 95-----
-143.99159813353864
-----GENERATION 96-----
-143.9916052272588
-----GENERATION 97-----
-143.9916052272588
-----GENERATION 98-----
-143.9916052272588
-----GENERATION 99-----
-143.9916052272588
Global : 41.80745683206267
Time taken : 72.68625950813293

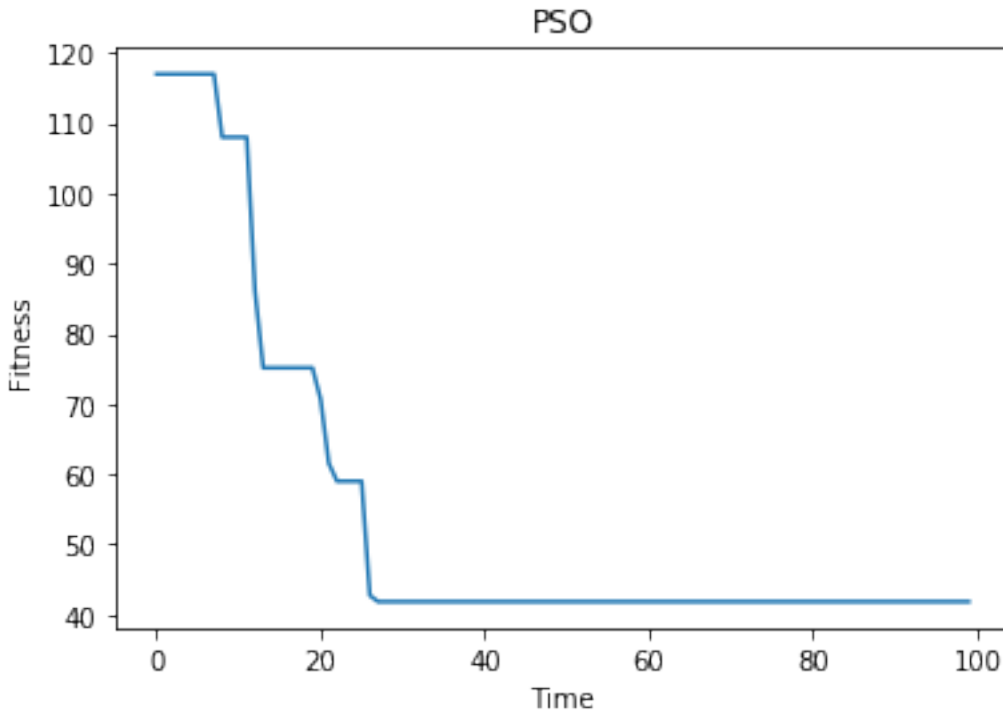
```

```

Fitness : [-41.80745683206267]
Best Weights : [-86.23843964 -58.80409054 54.30552104 ... 0.22068293
47.92238716

```

```
57.78789971]
Dimensions : [4, 100, 10, 3]
Time Taken : 72.72134065628052
```



```
[3]: 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

[illegible]

```

Predictions [0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0
0 0 0 1 1 2 1 1 1 2 1 1 2 1 2 1 1 2 1 2 2 1 1 2 1 1 2 1 1 1 1 2 1 1 1 2
2 2 1 1 2 2 2 2 2 2 2 2 2 2 2 1 2 1 2 2 2 2 2 2 2 2 2 2 1 2 2 1 2 2 2 2
2 2 2 2 2 2 2 2 1]

```

```

Confusion Matrix
[[40  0  0]
 [ 0 25 15]
 [ 0  5 35]]

```

```

Classification Report
              precision    recall  f1-score   support

   class 0       1.00      1.00      1.00        40
   class 1       0.83      0.62      0.71        40
   class 2       0.70      0.88      0.78        40

 accuracy              0.83        120
 macro avg       0.84      0.83      0.83        120
weighted avg       0.84      0.83      0.83        120

```

Time taken = 0.01929616928100586

```

[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

Actual / Expected [0 0 0 0 0 0 0 0 0 0 0 1 1 1 1 1 1 1 1 1 2 2 2 2 2 2 2 2 2 2]

Predictions [0 0 0 0 0 0 0 0 0 0 0 2 2 1 1 2 1 2 1 2 1 2 2 2 2 2 1 2 1 2 2]

```

Confusion Matrix
[[10  0  0]
 [ 0  5  5]
 [ 0  2  8]]

```

Classification Report

	precision	recall	f1-score	support
class 0	1.00	1.00	1.00	10
class 1	0.71	0.50	0.59	10
class 2	0.62	0.80	0.70	10
accuracy			0.77	30
macro avg	0.78	0.77	0.76	30
weighted avg	0.78	0.77	0.76	30

Time taken = 0.010092973709106445

[]: