

MACHINE LEARNING

LAB ASSIGNMENT 2

GITHUB LINK:

<https://github.com/Vishakha1909/ML-Lab-Assignments/tree/main/Assignment%202>

DATASET:

jupyter dataset.csv ✓ 2 hours ago	
File	Edit View Language
<pre>1 15.26,14.84,0.871,5.763,3.312,2.221,5.22,1 2 14.88,14.57,0.8811,5.554,3.333,1.018,4.956,1 3 14.29,14.09,0.905,5.291,3.337,2.699,4.825,1 4 13.84,13.94,0.8955,5.324,3.379,2.259,4.805,1 5 16.14,14.99,0.9034,5.658,3.562,1.355,5.175,1 6 14.38,14.21,0.8951,5.386,3.312,2.462,4.956,1 7 14.69,14.49,0.8799,5.563,3.259,3.586,5.219,1 8 14.11,14.1,0.8911,5.42,3.302,2.70,5.00,1 9 16.63,15.46,0.8747,6.053,3.465,2.04,5.877,1 10 16.44,15.25,0.888,5.884,3.505,1.969,5.533,1 11 15.26,14.85,0.8696,5.714,3.242,4.543,5.314,1 12 14.03,14.16,0.8796,5.438,3.201,1.717,5.001,1 13 13.89,14.02,0.888,5.439,3.199,3.986,4.738,1 14 13.78,14.06,0.8759,5.479,3.156,3.136,4.872,1 15 13.74,14.05,0.8744,5.482,3.114,2.932,4.825,1 16 14.59,14.28,0.8993,5.351,3.333,4.185,4.781,1 17 13.99,13.83,0.9183,5.119,3.383,5.234,4.781,1 18 15.69,14.75,0.9058,5.527,3.514,1.599,5.046,1 19 14.7,14.21,0.9153,5.205,3.466,1.767,4.649,1 20 12.72,13.57,0.8686,5.226,3.049,4.102,4.914,1 21 14.16,14.4,0.8584,5.658,3.129,3.072,5.176,1 22 14.11,14.26,0.8722,5.52,3.168,2.688,5.219,1 23 15.88,14.9,0.8988,5.618,3.507,0.7651,5.091,1 24 12.08,13.23,0.8664,5.099,2.936,1.415,4.961,1 25 15.01,14.76,0.8657,5.789,3.245,1.791,5.001,1 26 16.19,15.16,0.8849,5.833,3.421,0.903,5.307,1 27 13.02,13.76,0.8641,5.395,3.026,3.373,4.825,1 28 12.74,13.67,0.8564,5.395,2.956,2.504,4.869,1 29 14.11,14.18,0.882,5.541,3.221,2.754,5.038,1 30 13.45,14.02,0.8604,5.516,3.065,3.531,5.097,1</pre>	

CODE:

```
#importing required libraries

from random import seed

from random import randrange

from random import random

from csv import reader

from math import exp


# Load a CSV file

def load_csv(filename):

    dataset = list()

    with open(filename, 'r') as file:

        csv_reader = reader(file)

        for row in csv_reader:
```

```
        if not row:
            continue
        dataset.append(row)
    return dataset
```

Convert string column to float

```
def str_column_to_float(dataset, column):
    for row in dataset:
        row[column] = float(row[column].strip())
```

Convert string column to integer

```
def str_column_to_int(dataset, column):
    class_values = [row[column] for row in dataset]
    unique = set(class_values)
    lookup = dict()
    for i, value in enumerate(unique):
        lookup[value] = i
    for row in dataset:
        row[column] = lookup[row[column]]
    return lookup
```

Find the min and max values for each column

```
def dataset_minmax(dataset):
    minmax = list()
    stats = [[min(column), max(column)] for column in zip(*dataset)]
    return stats
```

Rescale dataset columns to the range 0-1

```
def normalize_dataset(dataset, minmax):
    for row in dataset:
        for i in range(len(row)-1):
```

```
row[i] = (row[i] - minmax[i][0]) / (minmax[i][1] - minmax[i][0])
```

```
# Split a dataset into k folds
```

```
def cross_validation_split(dataset, n_folds):
```

```
    dataset_split = list()
```

```
    dataset_copy = list(dataset)
```

```
    fold_size = int(len(dataset) / n_folds)
```

```
    for i in range(n_folds):
```

```
        fold = list()
```

```
        while len(fold) < fold_size:
```

```
            index = randrange(len(dataset_copy))
```

```
            fold.append(dataset_copy.pop(index))
```

```
        dataset_split.append(fold)
```

```
    return dataset_split
```

```
# Calculate accuracy percentage
```

```
def accuracy_metric(actual, predicted):
```

```
    correct = 0
```

```
    for i in range(len(actual)):
```

```
        if actual[i] == predicted[i]:
```

```
            correct += 1
```

```
    return correct / float(len(actual)) * 100.0
```

```
# Evaluate an algorithm using a cross validation split
```

```
def evaluate_algorithm(dataset, algorithm, n_folds, *args):
```

```
    folds = cross_validation_split(dataset, n_folds)
```

```
    scores = list()
```

```
    for fold in folds:
```

```
        train_set = list(folds)
```

```
        train_set.remove(fold)
```

```
        train_set = sum(train_set, [])
```

```
test_set = list()

for row in fold:

    row_copy = list(row)

    test_set.append(row_copy)

    row_copy[-1] = None

predicted = algorithm(train_set, test_set, *args)

actual = [row[-1] for row in fold]

accuracy = accuracy_metric(actual, predicted)

scores.append(accuracy)

return scores


# Calculate neuron activation for an input
def activate(weights, inputs):

    activation = weights[-1]

    for i in range(len(weights)-1):

        activation += weights[i] * inputs[i]

    return activation


# Transfer neuron activation
def transfer(activation):

    return 1.0 / (1.0 + exp(-activation))


# Forward propagate input to a network output
def forward_propagate(network, row):

    inputs = row

    for layer in network:

        new_inputs = []

        for neuron in layer:

            activation = activate(neuron['weights'], inputs)

            neuron['output'] = transfer(activation)

            new_inputs.append(neuron['output'])
```

```
        inputs = new_inputs
    return inputs

# Calculate the derivative of an neuron output
def transfer_derivative(output):
    return output * (1.0 - output)

# Backpropagate error and store in neurons
def backward_propagate_error(network, expected):
    for i in reversed(range(len(network))):
        layer = network[i]
        errors = list()
        if i != len(network)-1:
            for j in range(len(layer)):
                error = 0.0
                for neuron in network[i + 1]:
                    error += (neuron['weights'][j] * neuron['delta'])
                errors.append(error)
            er.append(error)
        else:
            for j in range(len(layer)):
                neuron = layer[j]
                errors.append(neuron['output'] - expected[j])
                er.append(neuron['output'] - expected[j])
        for j in range(len(layer)):
            neuron = layer[j]
            neuron['delta'] = errors[j] * transfer_derivative(neuron['output'])

# Update network weights with error
def update_weights(network, row, l_rate):
    for i in range(len(network)):
```

```
inputs = row[:-1]

if i != 0:

    inputs = [neuron['output'] for neuron in network[i - 1]]

for neuron in network[i]:

    for j in range(len(inputs)):

        neuron['weights'][j] -= l_rate * neuron['delta'] * inputs[j]

    neuron['weights'][-1] -= l_rate * neuron['delta']
```

Train a network for a fixed number of epochs

```
def train_network(network, train, l_rate, n_epoch, n_outputs):

    for epoch in range(n_epoch):

        for row in train:

            outputs = forward_propagate(network, row)

            expected = [0 for i in range(n_outputs)]

            expected[row[-1]] = 1

            backward_propagate_error(network, expected)

            update_weights(network, row, l_rate)
```

Initialize a network

```
def initialize_network(n_inputs, n_hidden, n_outputs):

    network = list()

    hidden_layer = [{'weights':[random() for i in range(n_inputs + 1)]] for i in range(n_hidden)]

    network.append(hidden_layer)

    output_layer = [{'weights':[random() for i in range(n_hidden + 1)]] for i in range(n_outputs)]

    network.append(output_layer)

    return network
```

Make a prediction with a network

```
def predict(network, row):

    outputs = forward_propagate(network, row)

    return outputs.index(max(outputs))
```

Backpropagation Algorithm With Stochastic Gradient Descent

```
def back_propagation(train, test, l_rate, n_epoch, n_hidden):
```

```
    n_inputs = len(train[0]) - 1
```

```
    n_outputs = len(set([row[-1] for row in train]))
```

```
    network = initialize_network(n_inputs, n_hidden, n_outputs)
```

```
    train_network(network, train, l_rate, n_epoch, n_outputs)
```

```
    predictions = list()
```

```
    for row in test:
```

```
        prediction = predict(network, row)
```

```
        predictions.append(prediction)
```

```
    return(predictions)
```

Test Backprop on Seeds dataset

```
seed(1)
```

```
er = []
```

load and prepare data

```
filename = 'dataset.csv'
```

```
dataset = load_csv(filename)
```

```
for i in range(len(dataset[0])-1):
```

```
    str_column_to_float(dataset, i)
```

convert class column to integers

```
str_column_to_int(dataset, len(dataset[0])-1)
```

normalize input variables

```
minmax = dataset_minmax(dataset)
```

```
normalize_dataset(dataset, minmax)
```

evaluate algorithm

```
n_folds = 5
```

```
l_rate = 1
```

```
n_epoch = 500
```

```
n_hidden = 5
```

```
scores = evaluate_algorithm(dataset, back_propagation, n_folds, l_rate, n_epoch, n_hidden)
```

```
# printing the final accuracy and score
```

```
print('Scores: %s' % scores)
```

```
print()
```

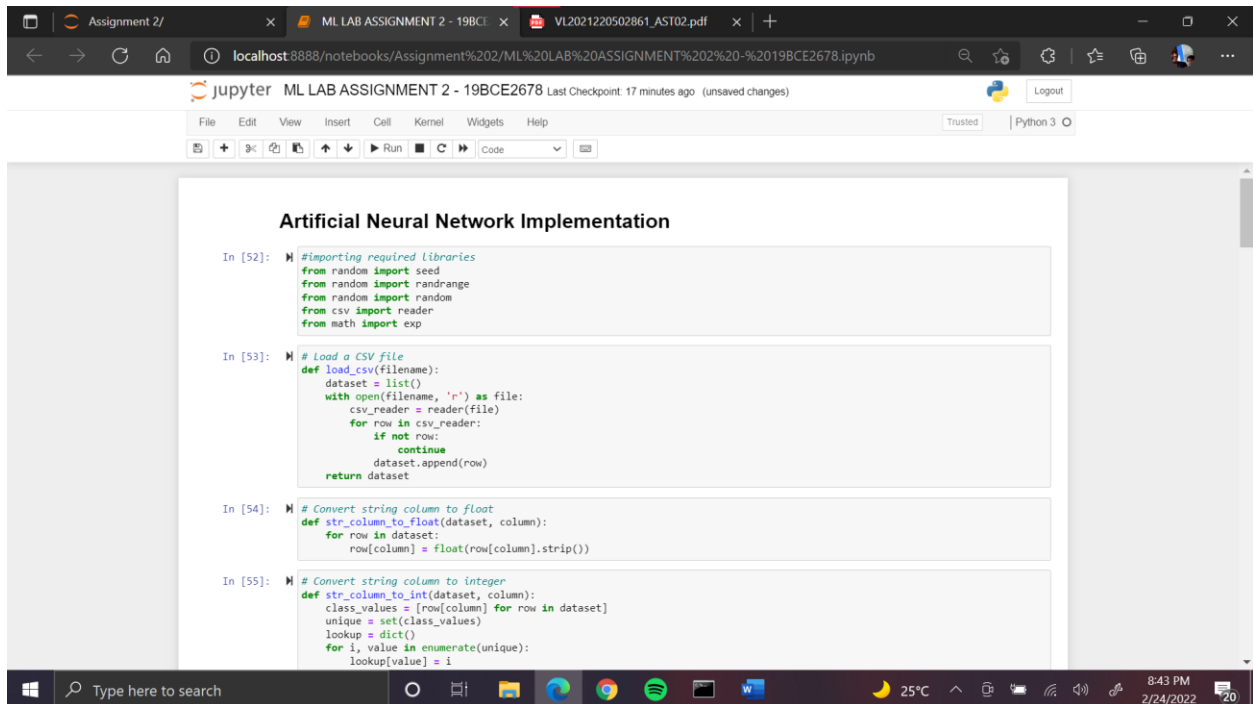
```
print('Mean Accuracy: %.3f%%' % (sum(scores)/float(len(scores))))
```

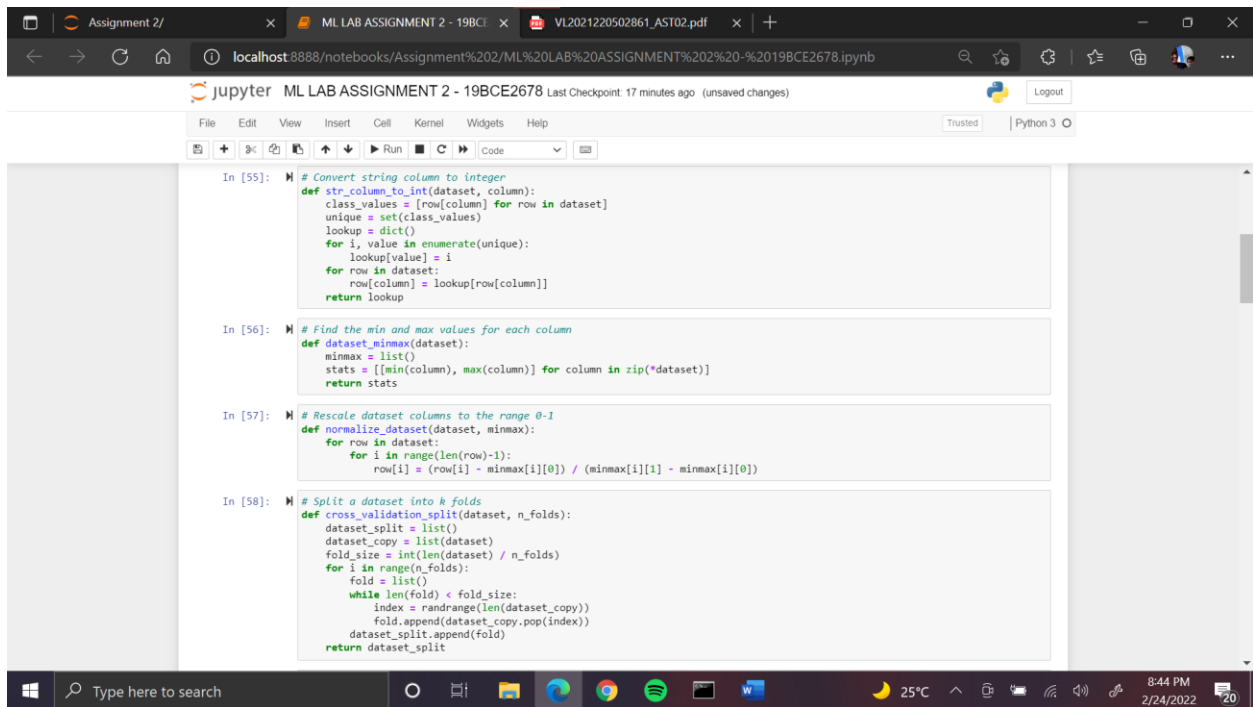
```
# printing error values of each epoch
```

```
for i in range(len(er)):
```

```
    print(er[i])
```

SCREENSHOTS:





Assignment 2/ ML LAB ASSIGNMENT 2 - 19BCE2678 VL2021220502861_AST02.pdf

localhost:8888/notebooks/Assignment%202/ML%20LAB%20ASSIGNMENT%202-%2019BCE2678.ipynb

Jupyter ML LAB ASSIGNMENT 2 - 19BCE2678 Last Checkpoint: 17 minutes ago (unsaved changes)

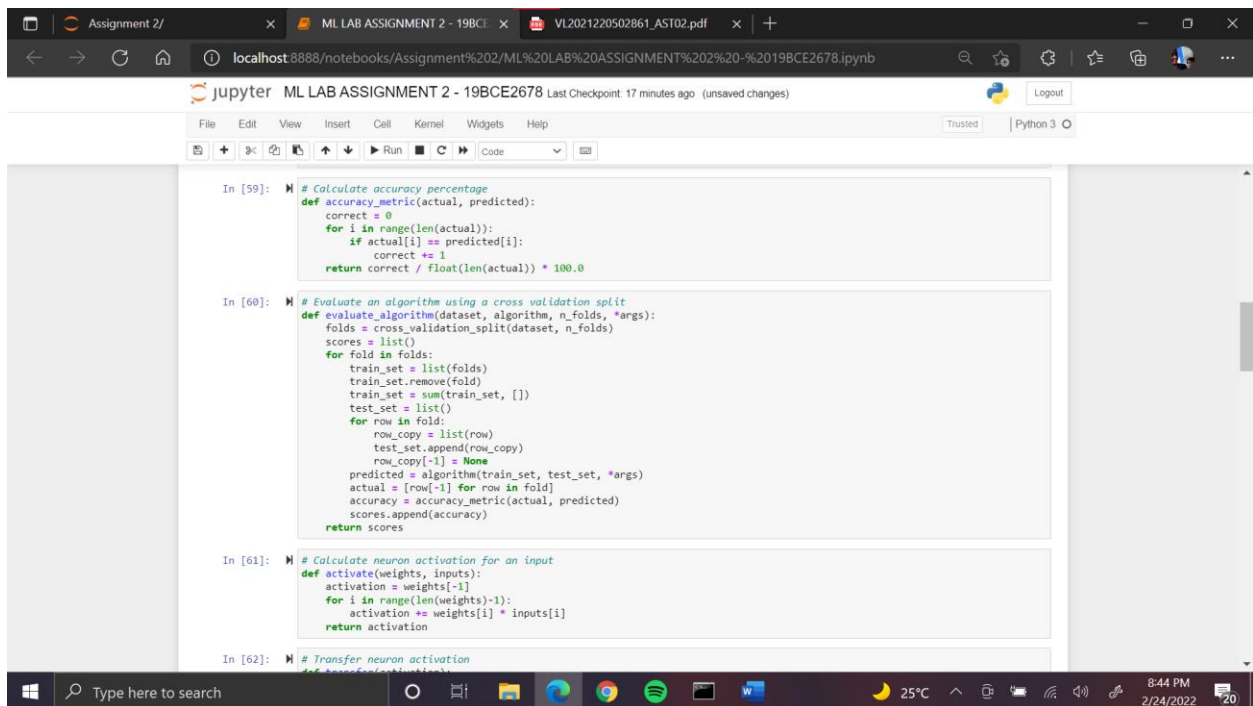
Python 3

```
In [55]: # Convert string column to integer
def str_column_to_int(dataset, column):
    class_values = [row[column] for row in dataset]
    unique = set(class_values)
    lookup = dict()
    for i, value in enumerate(unique):
        lookup[value] = i
    for row in dataset:
        row[column] = lookup[row[column]]
    return lookup

In [56]: # Find the min and max values for each column
def dataset_minmax(dataset):
    minmax = list()
    stats = [[min(column), max(column)] for column in zip(*dataset)]
    return stats

In [57]: # Rescale dataset columns to the range 0-1
def normalize_dataset(dataset, minmax):
    for row in dataset:
        for i in range(len(row)-1):
            row[i] = (row[i] - minmax[i][0]) / (minmax[i][1] - minmax[i][0])

In [58]: # Split a dataset into k folds
def cross_validation_split(dataset, n_folds):
    dataset_split = list()
    dataset_copy = list(dataset)
    fold_size = int(len(dataset) / n_folds)
    for i in range(n_folds):
        fold = list()
        while len(fold) < fold_size:
            index = randrange(len(dataset_copy))
            fold.append(dataset_copy.pop(index))
        dataset_split.append(fold)
    return dataset_split
```



Assignment 2/ ML LAB ASSIGNMENT 2 - 19BCE2678 VL2021220502861_AST02.pdf

localhost:8888/notebooks/Assignment%202/ML%20LAB%20ASSIGNMENT%202-%2019BCE2678.ipynb

Jupyter ML LAB ASSIGNMENT 2 - 19BCE2678 Last Checkpoint: 17 minutes ago (unsaved changes)

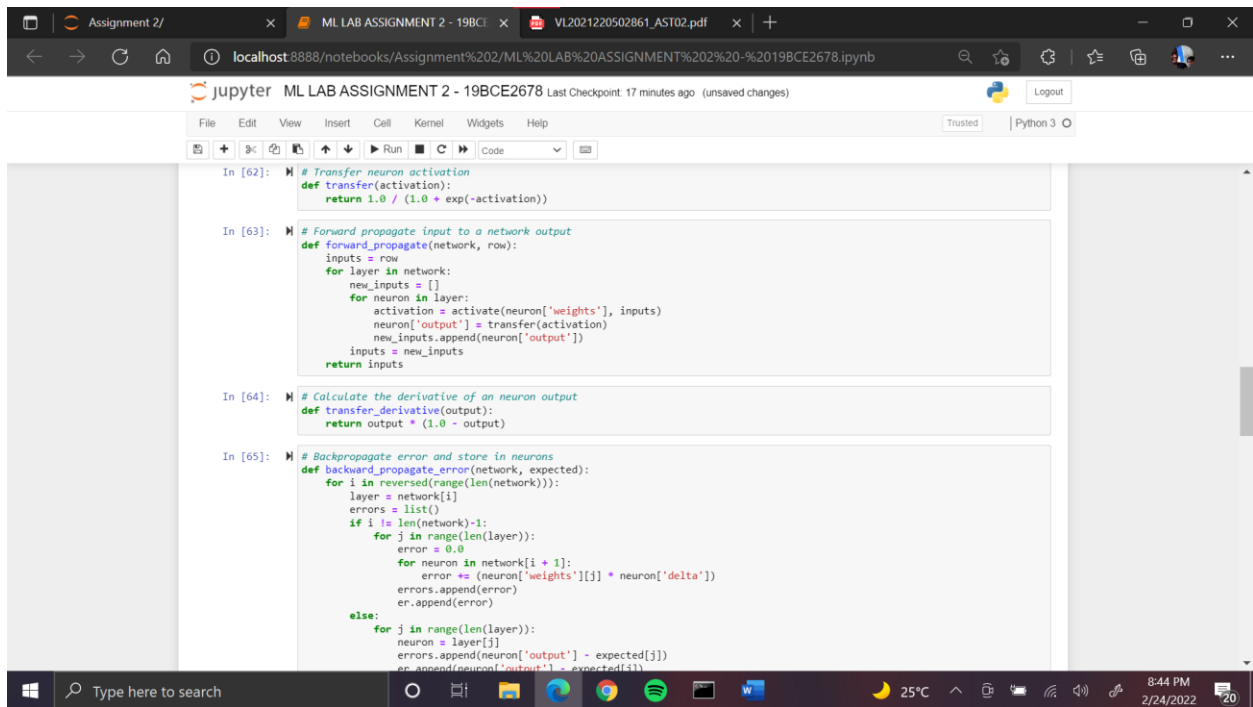
Python 3

```
In [59]: # Calculate accuracy percentage
def accuracy_metric(actual, predicted):
    correct = 0
    for i in range(len(actual)):
        if actual[i] == predicted[i]:
            correct += 1
    return correct / float(len(actual)) * 100.0

In [60]: # Evaluate an algorithm using a cross validation split
def evaluate_algorithm(dataset, algorithm, n_folds, *args):
    folds = cross_validation_split(dataset, n_folds)
    scores = list()
    for fold in folds:
        train_set = list(folds)
        train_set.remove(fold)
        train_set = sum(train_set, [])
        test_set = list()
        for row in fold:
            row_copy = list(row)
            test_set.append(row_copy)
            row_copy[-1] = None
        predicted = algorithm(train_set, test_set, *args)
        actual = [row[-1] for row in fold]
        accuracy = accuracy_metric(actual, predicted)
        scores.append(accuracy)
    return scores

In [61]: # Calculate neuron activation for an input
def activate(weights, inputs):
    activation = weights[-1]
    for i in range(len(weights)-1):
        activation += weights[i] * inputs[i]
    return activation

In [62]: # Transfer neuron activation
```

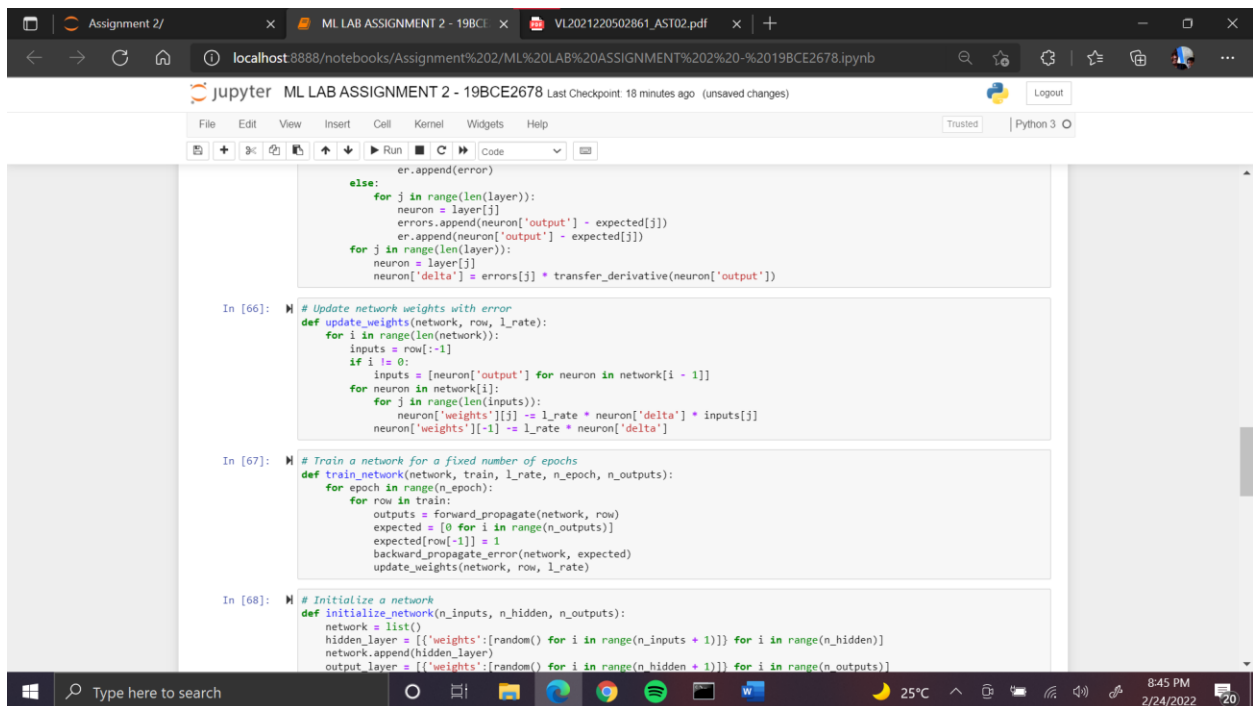


```
In [62]: # Transfer neuron activation
def transfer(activation):
    return 1.0 / (1.0 + exp(-activation))

In [63]: # Forward propagate input to a network output
def forward_propagate(network, row):
    inputs = row
    for layer in network:
        new_inputs = []
        for neuron in layer:
            activation = activate(neuron['weights'], inputs)
            neuron['output'] = transfer(activation)
            new_inputs.append(neuron['output'])
        inputs = new_inputs
    return inputs

In [64]: # Calculate the derivative of an neuron output
def transfer_derivative(output):
    return output * (1.0 - output)

In [65]: # Backpropagate error and store in neurons
def backward_propagate_error(network, expected):
    for i in reversed(range(len(network))):
        layer = network[i]
        errors = list()
        if i != len(network)-1:
            for j in range(len(layer)):
                error = 0.0
                for neuron in network[i + 1]:
                    error += (neuron['weights'][j] * neuron['delta'])
                errors.append(error)
            er.append(error)
        else:
            for j in range(len(layer)):
                neuron = layer[j]
                errors.append(neuron['output'] - expected[j])
            er.append(neuron['output'] - expected[j])
```

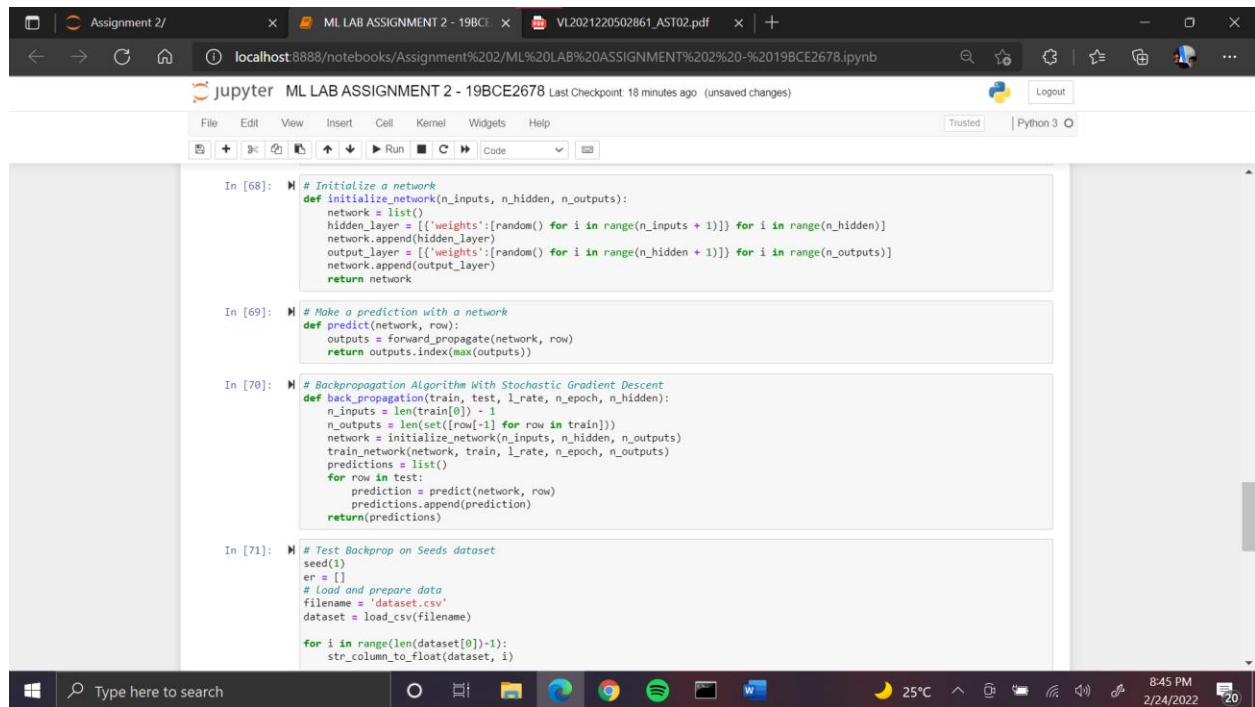


```
er.append(error)
else:
    for j in range(len(layer)):
        neuron = layer[j]
        errors.append(neuron['output'] - expected[j])
        er.append(neuron['output'] - expected[j])
    for j in range(len(layer)):
        neuron = layer[j]
        neuron['delta'] = errors[j] * transfer_derivative(neuron['output'])

In [66]: # Update network weights with error
def update_weights(network, row, l_rate):
    for i in range(len(network)):
        inputs = row[:-1]
        if i != 0:
            inputs = [neuron['output'] for neuron in network[i - 1]]
        for neuron in network[i]:
            for j in range(len(inputs)):
                neuron['weights'][j] += l_rate * neuron['delta'] * inputs[j]
            neuron['weights'][-1] += l_rate * neuron['delta']

In [67]: # Train a network for a fixed number of epochs
def train_network(network, train, l_rate, n_epoch, n_outputs):
    for epoch in range(n_epoch):
        for row in train:
            outputs = forward_propagate(network, row)
            expected = [0 for i in range(n_outputs)]
            expected[row[-1]] = 1
            backward_propagate_error(network, expected)
            update_weights(network, row, l_rate)

In [68]: # Initialize a network
def initialize_network(n_inputs, n_hidden, n_outputs):
    network = list()
    hidden_layer = [{'weights': [random() for i in range(n_inputs + 1)]} for i in range(n_hidden)]
    network.append(hidden_layer)
    output_layer = [{'weights': [random() for i in range(n_hidden + 1)]} for i in range(n_outputs)]
```



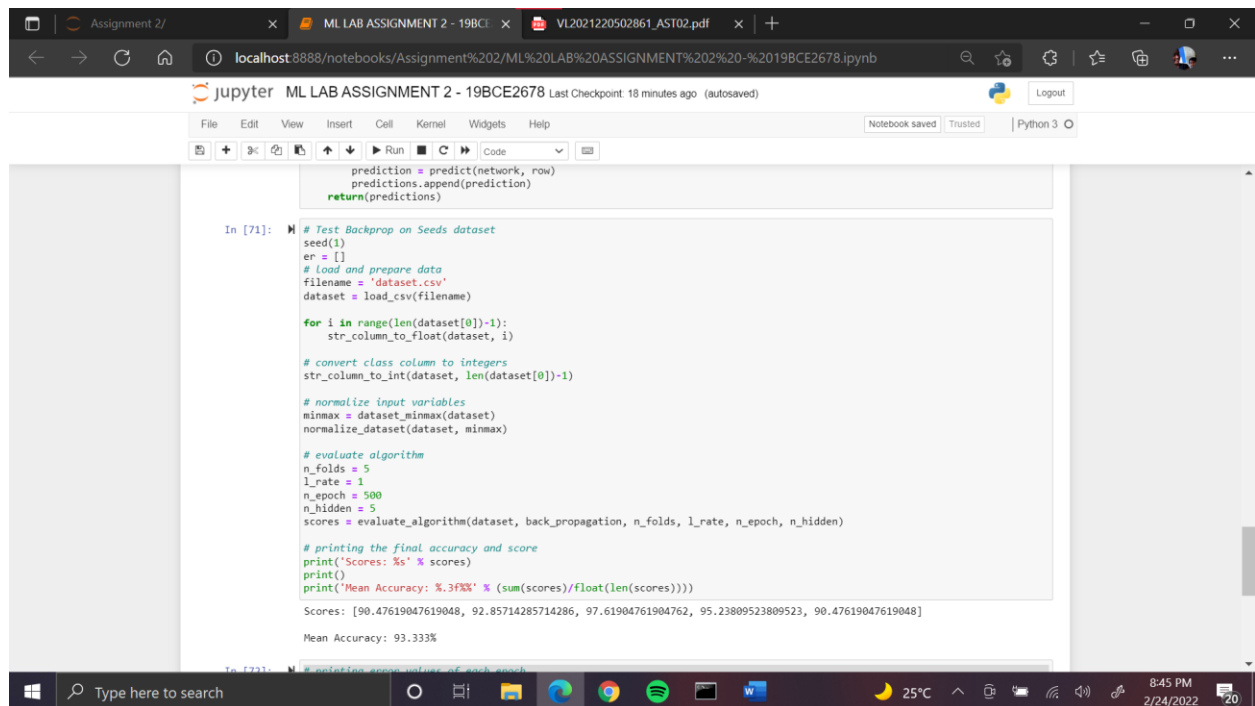
```
In [68]: # Initialize a network
def initialize_network(n_inputs, n_hidden, n_outputs):
    network = list()
    hidden_layer = [{'weights':[random() for i in range(n_inputs + 1)] for i in range(n_hidden)]
    network.append(hidden_layer)
    output_layer = [{'weights':[random() for i in range(n_hidden + 1)] for i in range(n_outputs)]
    network.append(output_layer)
    return network

In [69]: # Make a prediction with a network
def predict(network, row):
    outputs = forward_propagate(network, row)
    return outputs.index(max(outputs))

In [70]: # Backpropagation Algorithm With Stochastic Gradient Descent
def back_propagation(train, test, l_rate, n_epoch, n_hidden):
    n_inputs = len(train[0]) - 1
    n_outputs = len(set([row[-1] for row in train]))
    network = initialize_network(n_inputs, n_hidden, n_outputs)
    train_network(network, train, l_rate, n_epoch, n_outputs)
    predictions = list()
    for row in test:
        prediction = predict(network, row)
        predictions.append(prediction)
    return(predictions)

In [71]: # Test Backprop on Seeds dataset
seed(1)
er = []
# Load and prepare data
filename = 'dataset.csv'
dataset = load_csv(filename)

for i in range(len(dataset[0])-1):
    str_column_to_float(dataset, i)
```



```
prediction = predict(network, row)
predictions.append(prediction)
return(predictions)

In [71]: # Test Backprop on Seeds dataset
seed(1)
er = []
# Load and prepare data
filename = 'dataset.csv'
dataset = load_csv(filename)

for i in range(len(dataset[0])-1):
    str_column_to_float(dataset, i)

# convert class column to integers
str_column_to_int(dataset, len(dataset[0])-1)

# normalize input variables
minmax = dataset_minmax(dataset)
normalize_dataset(dataset, minmax)

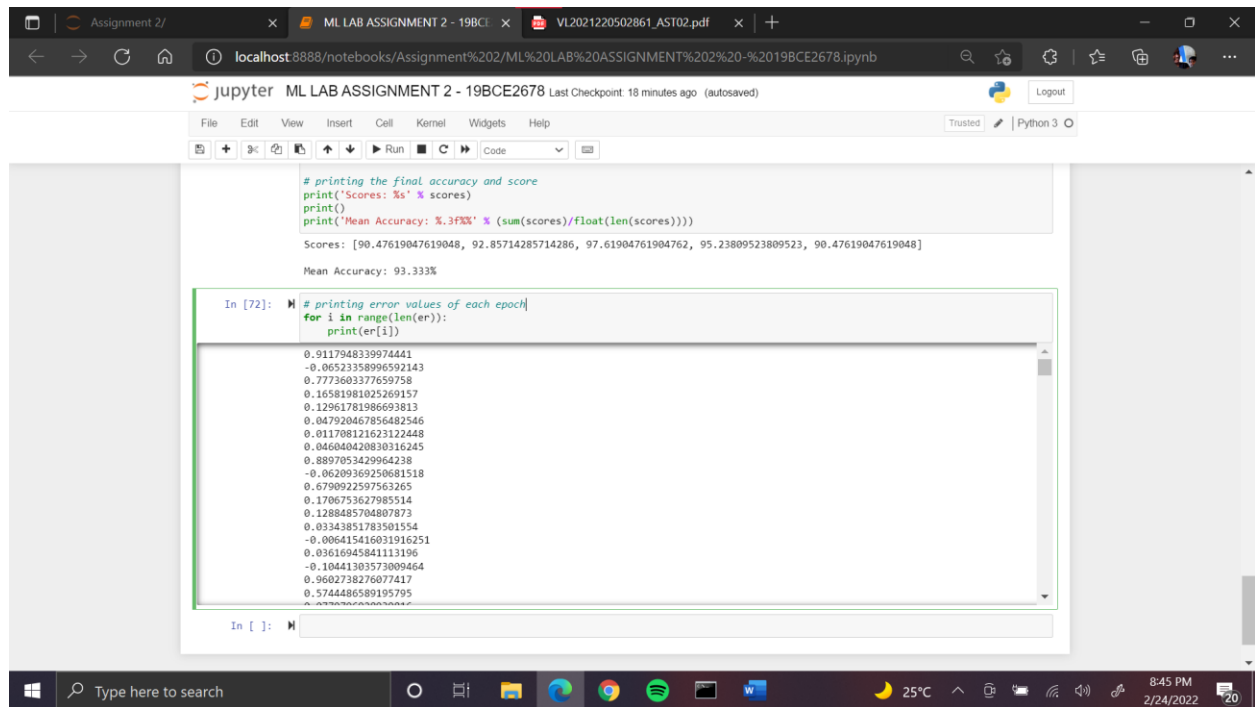
# evaluate algorithm
n_folds = 5
l_rate = 1
n_epoch = 500
n_hidden = 5
scores = evaluate_algorithm(dataset, back_propagation, n_folds, l_rate, n_epoch, n_hidden)

# printing the final accuracy and score
print('Scores: %s' % scores)
print()
print('Mean Accuracy: %.3f%%' % (sum(scores)/float(len(scores))))

Scores: [90.47619047619048, 92.85714285714286, 97.61904761904762, 95.23809523809523, 90.47619047619048]

Mean Accuracy: 93.333%

In [72]: # Evaluate algorithm with a random seed
```



The screenshot displays a Jupyter Notebook window titled "ML LAB ASSIGNMENT 2 - 19BCE2678". The browser address bar shows the URL "localhost:8888/notebooks/Assignment%202/ML%20LAB%20ASSIGNMENT%202-%2019BCE2678.ipynb". The notebook interface includes a menu bar (File, Edit, View, Insert, Cell, Kernel, Widgets, Help) and a toolbar with icons for file operations, cell execution, and code editing. The code is written in Python and is organized into two cells. The first cell contains code to print the final accuracy and score, and the second cell contains code to print error values for each epoch. The output of the first cell shows a list of scores and a mean accuracy of 93.33%. The output of the second cell shows a list of error values.

```
# printing the final accuracy and score
print('Scores: %s' % scores)
print()
print('Mean Accuracy: %.3f%%' % (sum(scores)/float(len(scores))))

Scores: [90.47619047619048, 92.85714285714286, 97.61904761904762, 95.23809523809523, 90.47619047619048]

Mean Accuracy: 93.33%

In [72]: # printing error values of each epoch
for i in range(len(er)):
    print(er[i])

0.9117948339974441
-0.06523358996592143
0.7773603377659758
0.16581981025269157
0.12961781986693813
0.047920467856482546
0.011708121623122448
0.046040420830316245
0.8897053429964238
-0.06209369250681518
0.6790922597563265
0.1706753627985514
0.1288485704807873
0.03343851783501554
-0.006415416031916251
0.03616945841113196
-0.10441303573009464
0.9602738276077417
0.5744486589195795
```