CS 354 - Assignment 6

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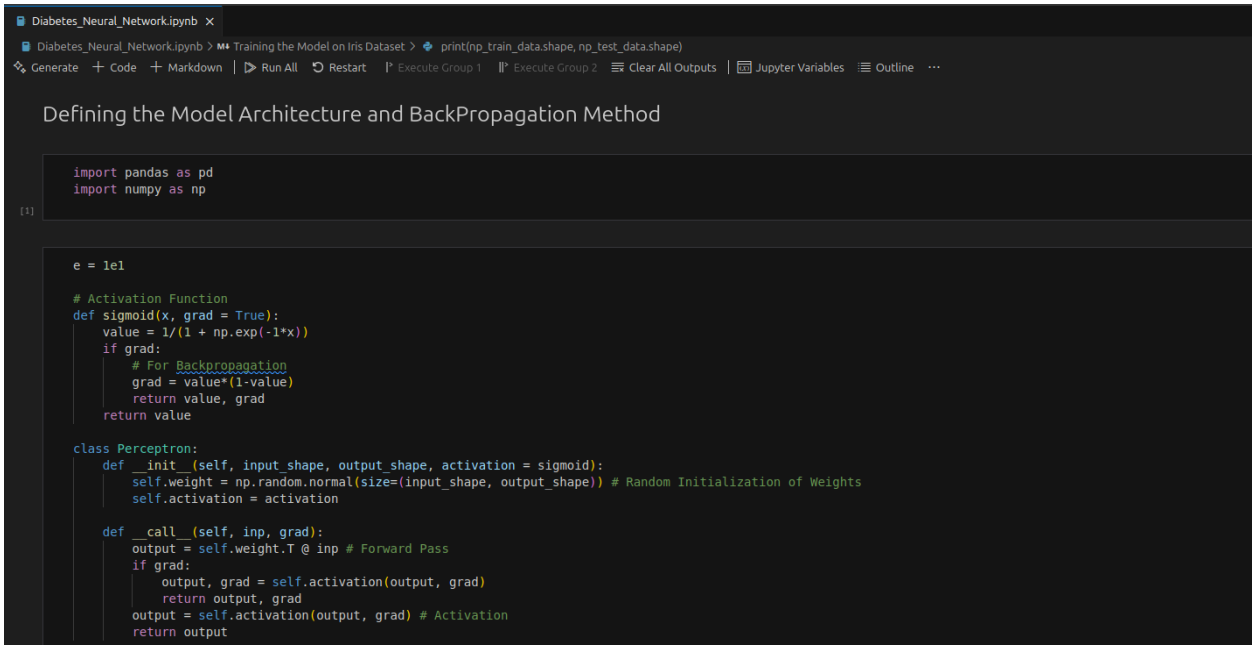
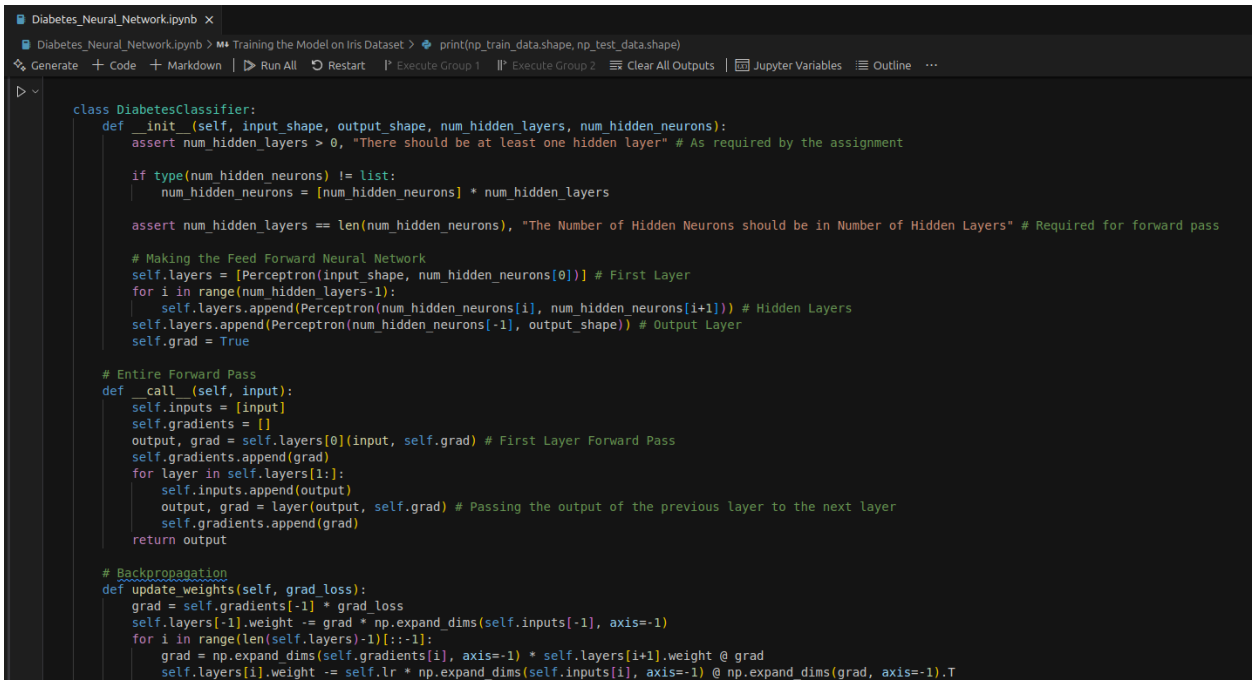
Question 1




Diabetes_Neural_Network.ipynb
Diabetes_Neural_Network.ipynb > Training the Model on Iris Dataset > print(np_train_data.shape, np_test_data.shape)
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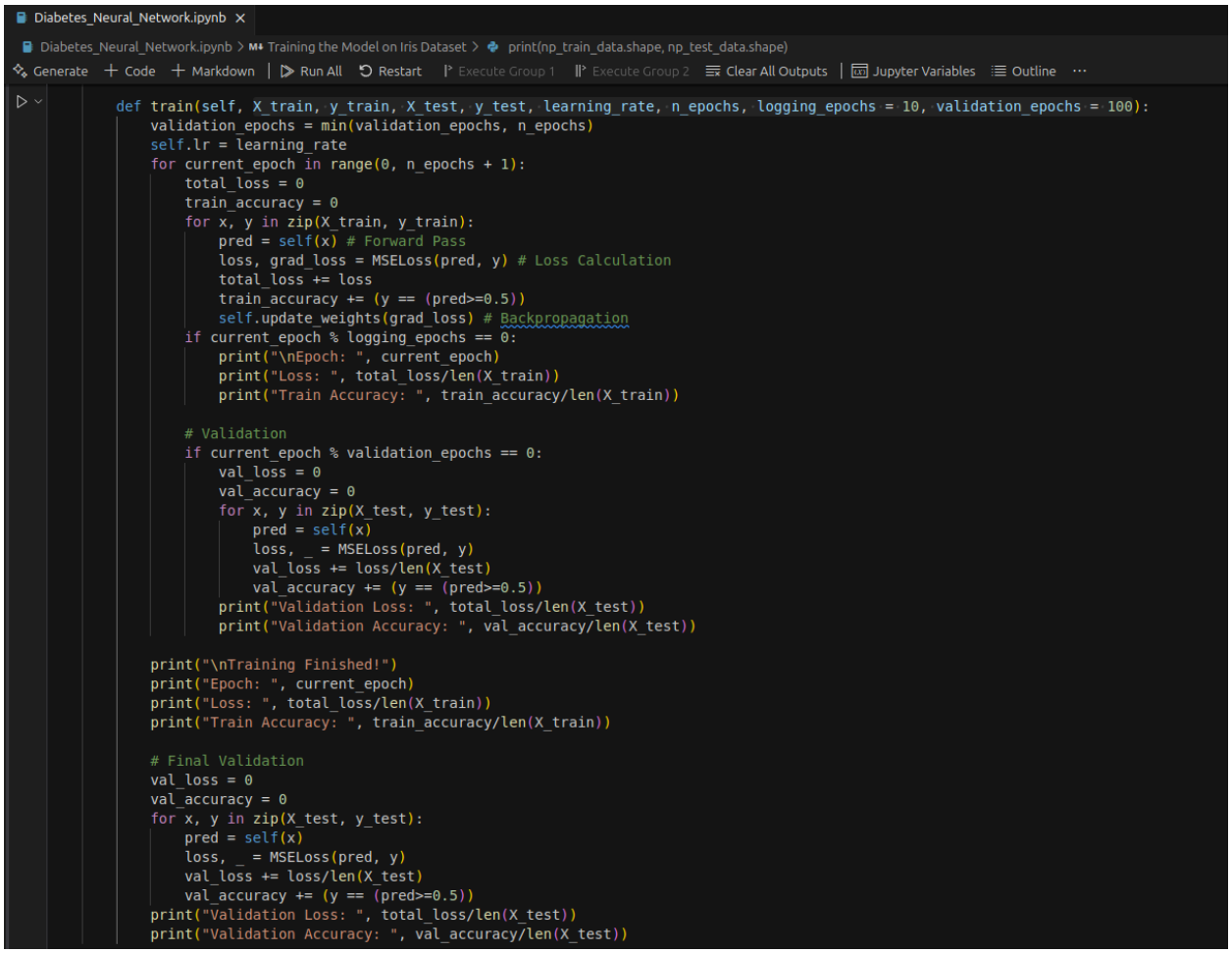
Outline
class DiabetesClassifier:
def __init__(self, input_shape, output_shape, num_hidden_layers, num_hidden_neurons):
assert num_hidden_layers > 0, "There should be at least one hidden layer" # As required by the assignment
if type(num_hidden_neurons) != list:
num_hidden_neurons = [num_hidden_neurons] * num_hidden_layers
assert num_hidden_layers == len(num_hidden_neurons), "The Number of Hidden Neurons should be in Number of Hidden Layers" # Required for forward pass
# Making the Feed Forward Neural Network
self.layers = [Perceptron(input_shape, num_hidden_neurons[0])] # First Layer
for i in range(num_hidden_layers-1):
self.layers.append(Perceptron(num_hidden_neurons[i], num_hidden_neurons[i+1])) # Hidden Layers
self.layers.append(Perceptron(num_hidden_neurons[-1], output_shape)) # Output Layer
self.grad = True
# Entire Forward Pass
def __call__(self, input):
self.inputs = [input]
self.gradients = []
output, grad = self.layers[0](input, self.grad) # First Layer Forward Pass
self.gradients.append(grad)
for layer in self.layers[1:]:
self.inputs.append(output)
output, grad = layer(output, self.grad) # Passing the output of the previous layer to the next layer
self.gradients.append(grad)
return output
# Backpropagation
def update_weights(self, grad_loss):
grad = self.gradients[-1] * grad_loss
self.layers[-1].weight -= grad * np.expand_dims(self.inputs[-1], axis=-1)
for i in range(len(self.layers)-1)[::-1]:
grad = np.expand_dims(self.gradients[i], axis=-1) * self.layers[i+1].weight @ grad
self.layers[i].weight -= self.lr * np.expand_dims(self.inputs[i], axis=-1) @ np.expand_dims(grad, axis=-1).T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Outline
def train(self, X_train, y_train, X_test, y_test, learning_rate, n_epochs, logging_epochs = 10, validation_epochs = 100):
validation_epochs = min(validation_epochs, n_epochs)
self.lr = learning_rate
for current_epoch in range(0, n_epochs + 1):
total_loss = 0
train_accuracy = 0
for x, y in zip(X_train, y_train):
pred = self(x) # Forward Pass
loss, grad_loss = MSELoss(pred, y) # Loss Calculation
total_loss += loss
train_accuracy += (y == (pred>=0.5))
self.update_weights(grad_loss) # Backpropagation
if current_epoch % logging_epochs == 0:
print("\nEpoch: ", current_epoch)
print("Loss: ", total_loss/len(X_train))
print("Train Accuracy: ", train_accuracy/len(X_train))
# Validation
if current_epoch % validation_epochs == 0:
val_loss = 0
val_accuracy = 0
for x, y in zip(X_test, y_test):
pred = self(x)
loss, _ = MSELoss(pred, y)
val_loss += loss/len(X_test)
val_accuracy += (y == (pred>=0.5))
print("Validation Loss: ", total_loss/len(X_test))
print("Validation Accuracy: ", val_accuracy/len(X_test))
print("\nTraining Finished!")
print("Epoch: ", current_epoch)
print("Loss: ", total_loss/len(X_train))
print("Train Accuracy: ", train_accuracy/len(X_train))
# Final Validation
val_loss = 0
val_accuracy = 0
for x, y in zip(X_test, y_test):
pred = self(x)
loss, _ = MSELoss(pred, y)
val_loss += loss/len(X_test)
val_accuracy += (y == (pred>=0.5))
print("Validation Loss: ", total_loss/len(X_test))
print("Validation Accuracy: ", val_accuracy/len(X_test))

Diabetes_Neural_Network.ipynb

Diabetes_Neural_Network.ipynb > Training the Model on Iris Dataset > print(np_train_data.shape, np_test_data.shape)

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```
# Loss Function
def MSELoss(prediction, truth):
    loss = 0.5 * (prediction - truth) ** 2
    grad = prediction - truth
    return loss, grad
```

[2]

Training the Model on Iris Dataset

```
train_data = pd.read_csv("diabetes_train.csv")
print(train_data.columns)

test_data = pd.read_excel("diabetes_test.xlsx")
print(test_data.columns)
```

[4]

```
Index(['Pregnancies', 'Glucose', 'BloodPressure', 'SkinThickness', 'Insulin',
       'BMI', 'DiabetesPedigreeFunction', 'Age', 'Outcome'],
      dtype='object')
Index(['Pregnancies', 'Glucose', 'BloodPressure', 'SkinThickness', 'Insulin',
       'BMI', 'DiabetesPedigreeFunction', 'Age', 'Outcome'],
      dtype='object')
```

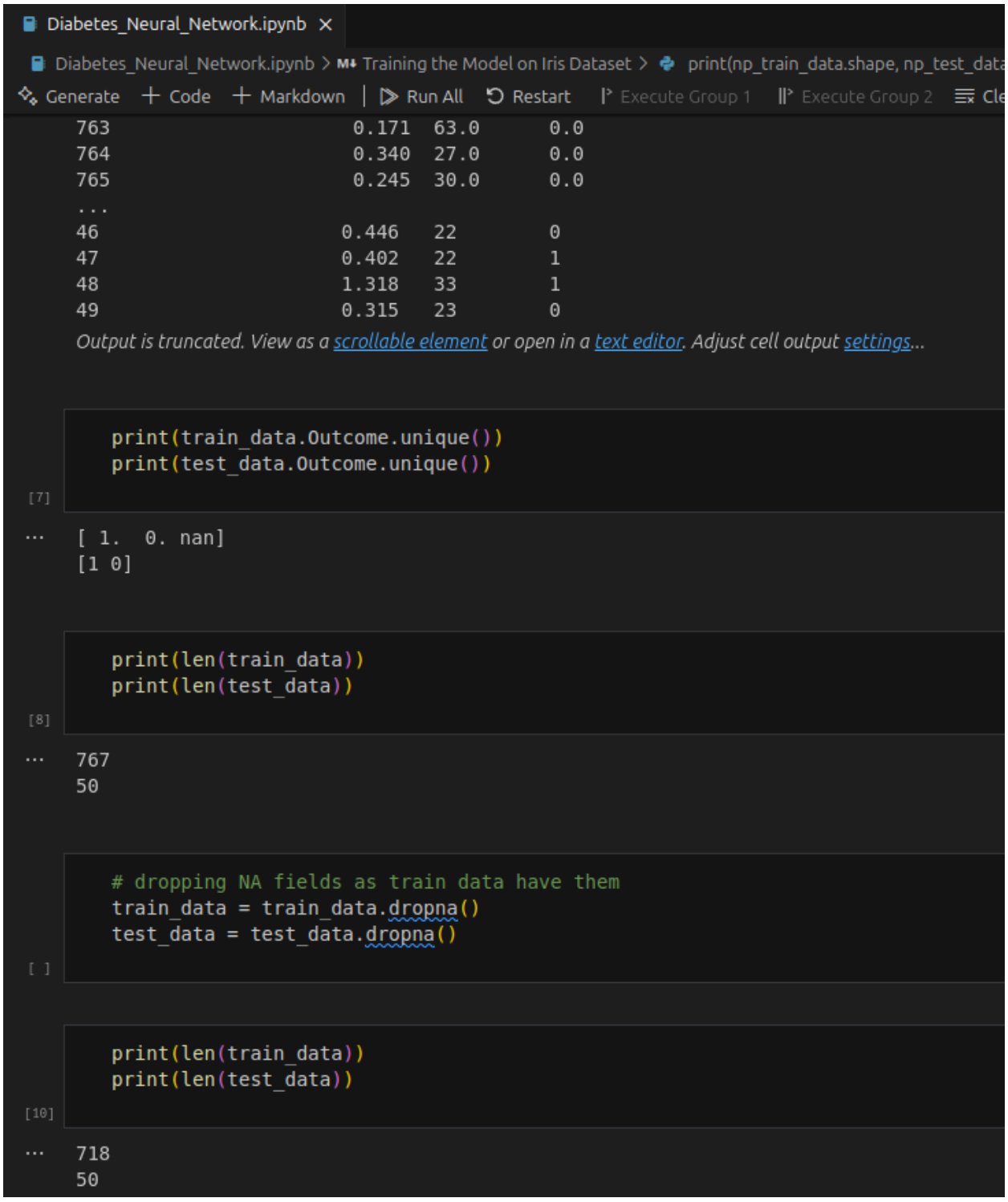
```
print(train_data.head())
print(train_data.tail())

print(test_data.head())
print(test_data.tail())
```

[5]

```
   Pregnancies  Glucose  BloodPressure  SkinThickness  Insulin   BMI  \
0          6.0    148.0           72.0           35.0      0.0  33.6
1          1.0     85.0           66.0           29.0      0.0  26.6
2          8.0    183.0           64.0            0.0      0.0  23.3
3          1.0     89.0           66.0           23.0     94.0  28.1
4          0.0    137.0           40.0           35.0    168.0  43.1

   DiabetesPedigreeFunction   Age  Outcome
0                     0.627  50.0      1.0
1                     0.351  31.0      0.0
```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763 0.171 63.0 0.0
764 0.340 27.0 0.0
765 0.245 30.0 0.0
...
46 0.446 22 0
47 0.402 22 1
48 1.318 33 1
49 0.315 23 0
Output is truncated. View as a scrollable element or open in a text editor. Adjust cell output settings...
print(train_data.Outcome.unique())
print(test_data.Outcome.unique())
[7]
[1. 0. nan]
[1 0]
print(len(train_data))
print(len(test_data))
[8]
767
50
# dropping NA fields as train data have them
train_data = train_data.dropna()
test_data = test_data.dropna()
[ ]
print(len(train_data))
print(len(test_data))
[10]
718
50

Diabetes_Neural_Network.ipynb

Diabetes_Neural_Network.ipynb > Training the Model on Iris Dataset > print(np_train_data.shape, np_test_data.shape)

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[11]
```
train_data = train_data.sample(frac = 1)
```

[12]
```
print(train_data.head())
print(train_data.tail())
```

```
     Pregnancies  Glucose  BloodPressure  SkinThickness  Insulin   BMI  \
526          1.0     97.0           64.0           19.0     82.0  18.2
340          1.0    130.0           70.0           13.0    105.0  25.9
732          2.0    174.0           88.0           37.0    120.0  44.5
402          5.0    136.0           84.0           41.0     88.0  35.0
412          1.0    143.0           84.0           23.0    310.0  42.4

     DiabetesPedigreeFunction   Age  Outcome
526                     0.299  21.0      0.0
340                     0.472  22.0      0.0
732                     0.646  24.0      1.0
402                     0.286  35.0      1.0
412                     1.076  22.0      0.0
     Pregnancies  Glucose  BloodPressure  SkinThickness  Insulin   BMI  \
688          1.0    140.0           74.0           26.0    180.0  24.1
631          0.0    102.0           78.0           40.0     90.0  34.5
756          7.0    137.0           90.0           41.0      0.0  32.0
35           4.0    103.0           60.0           33.0    192.0  24.0
336          0.0    117.0            0.0            0.0      0.0  33.8

     DiabetesPedigreeFunction   Age  Outcome
688                     0.828  23.0      0.0
631                     0.238  24.0      0.0
756                     0.391  39.0      0.0
35                      0.966  33.0      0.0
336                     0.932  44.0      0.0
```

[13]
```
np_train_data = train_data.to_numpy()
np_test_data = test_data.to_numpy()
```

[14]
```
X_train, y_train = np_train_data[:,:-1], np_train_data[:,-1]
X_test, y_test = np_test_data[:,:-1], np_test_data[:,-1]
```

Diabetes_Neural_Network.ipynb

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[18]
```
print(np_train_data.shape, np_test_data.shape)
print(X_train.shape, y_train.shape, X_test.shape, y_test.shape)
```

```
(718, 9) (50, 9)
(718, 8) (718,) (50, 8) (50,)
```

[19]
```
# IrisClassifier : input_shape, output_shape, num_hidden_layers, num_hidden_neurons
model = DiabetesClassifier(8, 1, num_hidden_layers = 1, num_hidden_neurons = [10])
# train method() : X_train, y_train, X_test, y_test, learning_rate, n_epochs, logging_epochs = 10, validation_epochs = 100
model.train(X_train, y_train, X_test, y_test, 0.0001 , 30 , 5, 10)
```

```
/tmp/ipykernel_9327/1659400037.py:5: RuntimeWarning: overflow encountered in exp
  value = 1/(1 + np.exp(-1*x))

Epoch:  0
Loss:  [0.12353083]
Train Accuracy:  [0.63509749]
Validation Loss:  [1.77390278]
Validation Accuracy:  [0.52]

Epoch:  5
Loss:  [0.12046696]
Train Accuracy:  [0.63231198]

Epoch:  10
Loss:  [0.12022412]
Train Accuracy:  [0.62952646]
Validation Loss:  [1.72641837]
Validation Accuracy:  [0.56]

Epoch:  15
Loss:  [0.1200828]
Train Accuracy:  [0.6281337]

Epoch:  20
Loss:  [0.11997278]
Train Accuracy:  [0.6281337]
Validation Loss:  [1.72280912]
Validation Accuracy:  [0.56]
```

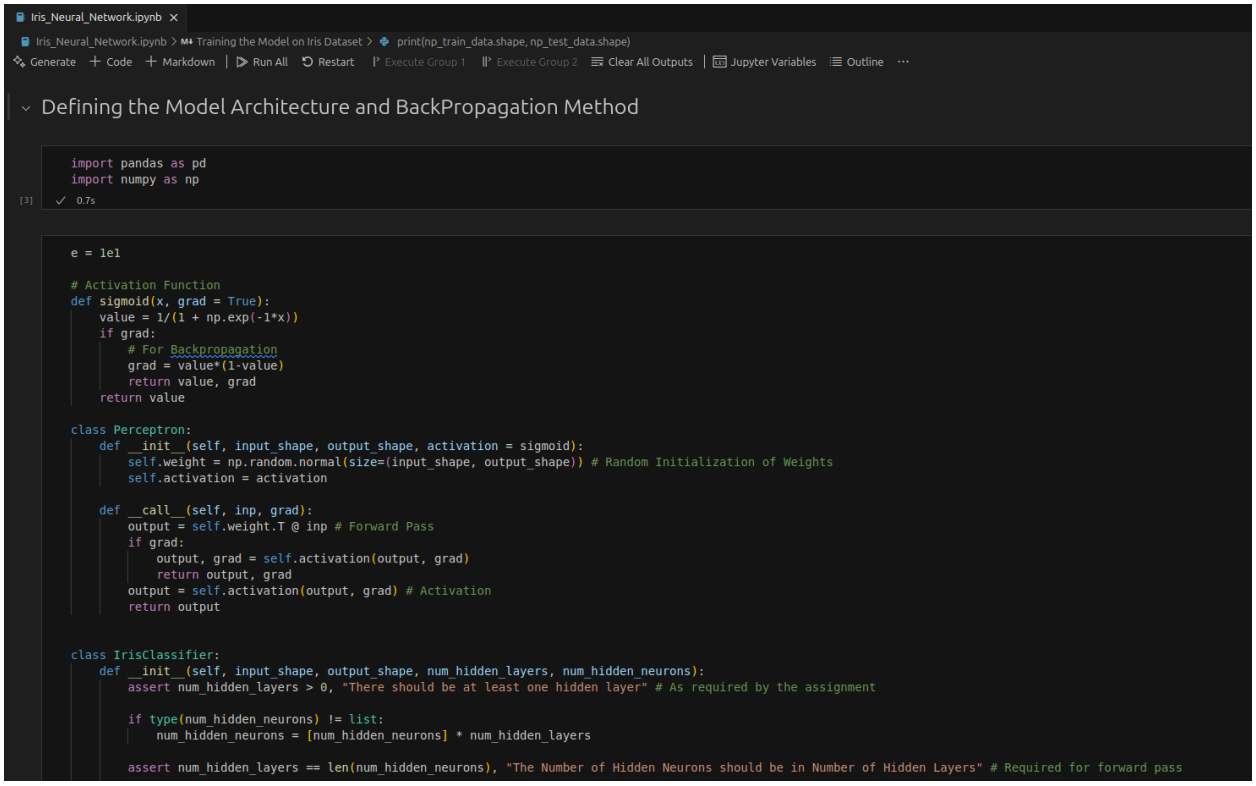
```
Epoch:  25
Loss:  [0.11983622]
Train Accuracy:  [0.63091922]

Epoch:  30
Loss:  [0.11957124]
Train Accuracy:  [0.63509749]
Validation Loss:  [1.71704308]
Validation Accuracy:  [0.56]

Training Finished!
Epoch:  30
Loss:  [0.11957124]
Train Accuracy:  [0.63509749]
Validation Loss:  [1.71704308]
Validation Accuracy:  [0.56]
```

```
[22]
# IrisClassifier : input_shape, output_shape, num_hidden_layers, num_hidden_neurons
model = DiabetesClassifier(8, 1, num_hidden_layers = 3, num_hidden_neurons = [15 , 10 , 5])
# train method() : X_train, y_train, X_test, y_test, learning_rate, n_epochs, logging_epochs = 10, validation_epochs = 100
model.train(X_train, y_train, X_test, y_test, 0.0001 , 30 , 5, 10)
```

```
/tmp/ipykernel_9327/1659400037.py:5: RuntimeWarning: overflow encountered in exp
  value = 1/(1 + np.exp(-1*x))

Epoch:  0
Loss:  [0.11800861]
Train Accuracy:  [0.62952646]
Validation Loss:  [1.69460366]
Validation Accuracy:  [0.52]

Epoch:  5
Loss:  [0.11379048]
Train Accuracy:  [0.64763231]

Epoch:  10
Loss:  [0.11288279]
Train Accuracy:  [0.64345404]
Validation Loss:  [1.62099693]
Validation Accuracy:  [0.52]

Epoch:  15
Loss:  [0.11223748]
Train Accuracy:  [0.6448468]

Epoch:  20
Loss:  [0.11165935]
Train Accuracy:  [0.65320334]
Validation Loss:  [1.60342821]
Validation Accuracy:  [0.52]
```

```
Epoch:  25
Loss:  [0.11115554]
Train Accuracy:  [0.65877437]

Epoch:  30
Loss:  [0.11071066]
Train Accuracy:  [0.66295265]
Validation Loss:  [1.58980514]
Validation Accuracy:  [0.52]

Training Finished!
Epoch:  30
Loss:  [0.11071066]
Train Accuracy:  [0.66295265]
Validation Loss:  [1.58980514]
Validation Accuracy:  [0.52]
```

Question 2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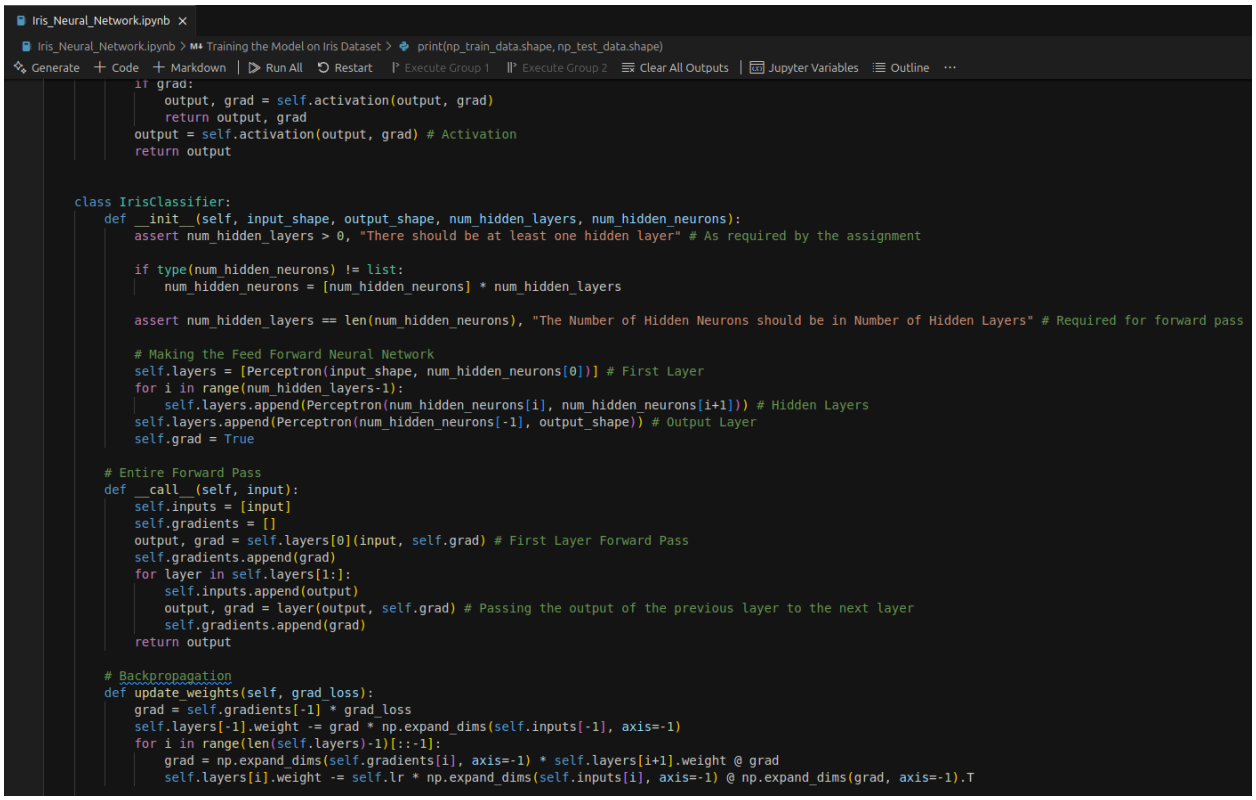
Outline
Defining the Model Architecture and BackPropagation Method
import pandas as pd
import numpy as np
[3]
0.7s
e = 1e1
# Activation Function
def sigmoid(x, grad = True):
value = 1/(1 + np.exp(-1*x))
if grad:
# For Backpropagation
grad = value*(1-value)
return value, grad
return value
class Perceptron:
def __init__(self, input_shape, output_shape, activation = sigmoid):
self.weight = np.random.normal(size=(input_shape, output_shape)) # Random Initialization of Weights
self.activation = activation
def __call__(self, inp, grad):
output = self.weight.T @ inp # Forward Pass
if grad:
output, grad = self.activation(output, grad)
return output, grad
output = self.activation(output, grad) # Activation
return output
class IrisClassifier:
def __init__(self, input_shape, output_shape, num_hidden_layers, num_hidden_neurons):
assert num_hidden_layers > 0, "There should be at least one hidden layer" # As required by the assignment
if type(num_hidden_neurons) != list:
num_hidden_neurons = [num_hidden_neurons] * num_hidden_layers
assert num_hidden_layers == len(num_hidden_neurons), "The Number of Hidden Neurons should be in Number of Hidden Layers" # Required for forward pass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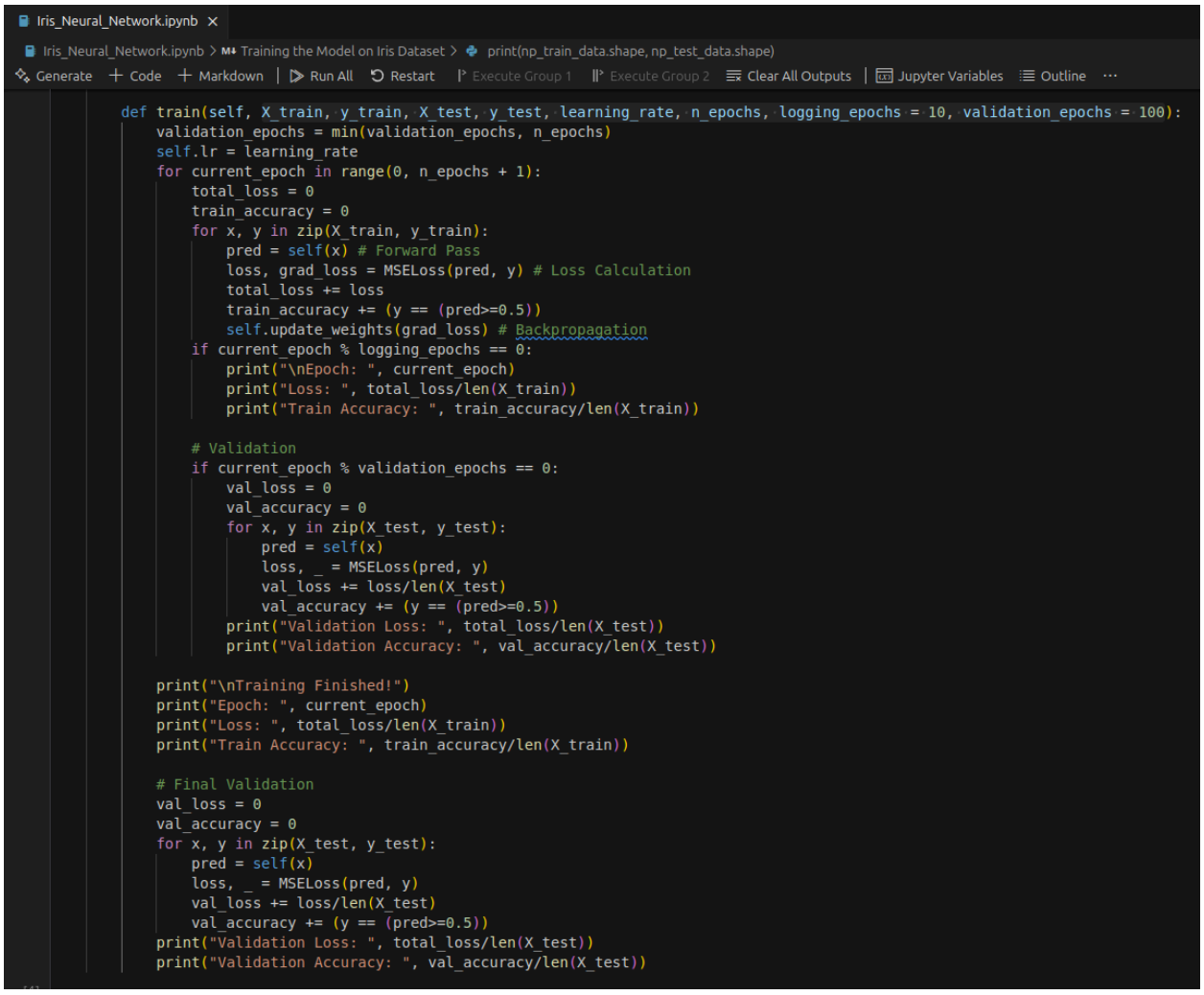
Outline
if grad:
output, grad = self.activation(output, grad)
return output, grad
output = self.activation(output, grad) # Activation
return output
class IrisClassifier:
def __init__(self, input_shape, output_shape, num_hidden_layers, num_hidden_neurons):
assert num_hidden_layers > 0, "There should be at least one hidden layer" # As required by the assignment
if type(num_hidden_neurons) != list:
num_hidden_neurons = [num_hidden_neurons] * num_hidden_layers
assert num_hidden_layers == len(num_hidden_neurons), "The Number of Hidden Neurons should be in Number of Hidden Layers" # Required for forward pass
# Making the Feed Forward Neural Network
self.layers = [Perceptron(input_shape, num_hidden_neurons[0])] # First Layer
for i in range(num_hidden_layers-1):
self.layers.append(Perceptron(num_hidden_neurons[i], num_hidden_neurons[i+1])) # Hidden Layers
self.layers.append(Perceptron(num_hidden_neurons[-1], output_shape)) # Output Layer
self.grad = True
# Entire Forward Pass
def __call__(self, input):
self.inputs = [input]
self.gradients = []
output, grad = self.layers[0](input, self.grad) # First Layer Forward Pass
self.gradients.append(grad)
for layer in self.layers[1:]:
self.inputs.append(output)
output, grad = layer(output, self.grad) # Passing the output of the previous layer to the next layer
self.gradients.append(grad)
return output
# Backpropagation
def update_weights(self, grad_loss):
grad = self.gradients[-1] * grad_loss
self.layers[-1].weight -= grad * np.expand_dims(self.inputs[-1], axis=-1)
for i in range(len(self.layers)-1)[::-1]:
grad = np.expand_dims(self.gradients[i], axis=-1) * self.layers[i+1].weight @ grad
self.layers[i].weight -= self.lr * np.expand_dims(self.inputs[i], axis=-1) @ np.expand_dims(grad, axis=-1).T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Outline
def train(self, X_train, y_train, X_test, y_test, learning_rate, n_epochs, logging_epochs = 10, validation_epochs = 100):
validation_epochs = min(validation_epochs, n_epochs)
self.lr = learning_rate
for current_epoch in range(0, n_epochs + 1):
total_loss = 0
train_accuracy = 0
for x, y in zip(X_train, y_train):
pred = self(x) # Forward Pass
loss, grad_loss = MSELoss(pred, y) # Loss Calculation
total_loss += loss
train_accuracy += (y == (pred>=0.5))
self.update_weights(grad_loss) # Backpropagation
if current_epoch % logging_epochs == 0:
print("\nEpoch: ", current_epoch)
print("Loss: ", total_loss/len(X_train))
print("Train Accuracy: ", train_accuracy/len(X_train))
# Validation
if current_epoch % validation_epochs == 0:
val_loss = 0
val_accuracy = 0
for x, y in zip(X_test, y_test):
pred = self(x)
loss, _ = MSELoss(pred, y)
val_loss += loss/len(X_test)
val_accuracy += (y == (pred>=0.5))
print("Validation Loss: ", total_loss/len(X_test))
print("Validation Accuracy: ", val_accuracy/len(X_test))
print("\nTraining Finished!")
print("Epoch: ", current_epoch)
print("Loss: ", total_loss/len(X_train))
print("Train Accuracy: ", train_accuracy/len(X_train))
# Final Validation
val_loss = 0
val_accuracy = 0
for x, y in zip(X_test, y_test):
pred = self(x)
loss, _ = MSELoss(pred, y)
val_loss += loss/len(X_test)
val_accuracy += (y == (pred>=0.5))
print("Validation Loss: ", total_loss/len(X_test))
print("Validation Accuracy: ", val_accuracy/len(X_test))

Iris_Neural_Network.ipynb

Iris_Neural_Network.ipynb > Training the Model on Iris Dataset > print(np_train_data.shape, np_test_data.shape)

Generate | Code | Markdown | Run All | Restart | Execute Group 1 | Execute Group 2 | Clear All Outputs | Jupy

```
# Loss Function
def MSELoss(prediction, truth):
    loss = 0.5 * (prediction - truth) ** 2
    grad = prediction - truth
    return loss, grad
```

[4] ✓ 0.0s

Training the Model on Iris Dataset

```
train_data = pd.read_excel("iris_train.xlsx")
print(train_data.columns)

test_data = pd.read_excel("iris_test.xlsx")
print(test_data.columns)
```

[25] ✓ 0.0s

```
Index(['sepal_length', 'sepal_width', 'petal_length', 'petal_width',
       'species'],
      dtype='object')
Index(['sepal_length', 'sepal_width', 'petal_length', 'petal_width',
       'species'],
      dtype='object')
```

```
print(train_data.head())
print(train_data.tail())

print(test_data.head())
print(test_data.tail())
```

[26] ✓ 0.0s

```
   sepal_length  sepal_width  petal_length  petal_width species
0           5.4          3.9           1.7          0.4  setosa
1           4.6          3.4           1.4          0.3  setosa
2           5.0          3.4           1.5          0.2  setosa
3           4.4          2.9           1.4          0.2  setosa
4           4.9          3.1           1.5          0.1  setosa
    sepal_length  sepal_width  petal_length  petal_width     species
75           5.7          3.0           4.2          1.2  versicolor
76           5.7          2.9           4.2          1.3  versicolor
77           6.2          2.9           4.3          1.3  versicolor
```

Iris_Neural_Network.ipynb

Iris_Neural_Network.ipynb > Training the Model on Iris Dataset > print(np_train_data.shape, np_test_data.shape)

Generate | Code | Markdown | Run All | Restart | Execute Group 1 | Execute Group 2 | Clear All Outputs | Jupyter

```
print(train_data.species.unique())
print(test_data.species.unique())
```

[28] ✓ 0.0s

```
['setosa' 'versicolor']
['setosa' 'versicolor']
```

```
def convert(x):
    values = ["setosa", "versicolor"]
    return values.index(x)

train_data["species"] = train_data["species"].apply(convert)
test_data["species"] = test_data["species"].apply(convert)
```

[30] ✓ 0.0s

```
print(train_data.head())
print(test_data.tail())
```

[19] ✓ 0.0s

```
   sepal_length  sepal_width  petal_length  petal_width  species
0           5.4          3.9           1.7          0.4        0
1           4.6          3.4           1.4          0.3        0
2           5.0          3.4           1.5          0.2        0
3           4.4          2.9           1.4          0.2        0
4           4.9          3.1           1.5          0.1        0
    sepal_length  sepal_width  petal_length  petal_width  species
75           5.7          3.0           4.2          1.2        1
76           5.7          2.9           4.2          1.3        1
77           6.2          2.9           4.3          1.3        1
78           5.1          2.5           3.0          1.1        1
79           5.7          2.8           4.1          1.3        1
```

```
len(train_data), len(test_data)
```

[31] ✓ 0.0s

```
(80, 20)
```

Iris_Neural_Network.ipynb

Iris_Neural_Network.ipynb > Training the Model on Iris Dataset > print(np_train_data.shape, np_test_data.shape)

Generate | Code | Markdown | Run All | Restart | Execute Group 1 | Execute Group 2 | Clear All Outputs | Jupyter Va

```
# Shuffle the Dataset
train_data = train_data.sample(frac = 1)
```

[32] ✓ 0.0s

```
print(train_data.head())
print(train_data.tail())
```

[33] ✓ 0.0s

```
     sepal_length  sepal_width  petal_length  petal_width  species
7             4.8          3.0           1.4          0.1        0
66            6.7          3.1           4.7          1.5        1
34            5.1          3.8           1.9          0.4        0
5             5.4          3.7           1.5          0.2        0
2             5.0          3.4           1.5          0.2        0
    sepal_length  sepal_width  petal_length  petal_width  species
22            5.2          3.5           1.5          0.2        0
74            5.6          2.7           4.2          1.3        1
21            5.0          3.4           1.6          0.4        0
9             5.8          4.0           1.2          0.2        0
12            5.1          3.5           1.4          0.3        0
```

```
np_train_data = train_data.to_numpy()
np_test_data = test_data.to_numpy()
```

[34] ✓ 0.0s

```
X_train, y_train = np_train_data[:,:-1], np_train_data[:,-1]
X_test, y_test = np_test_data[:,:-1], np_test_data[:,-1]
```

[35] ✓ 0.0s

```
print(np_train_data.shape, np_test_data.shape)
print(X_train.shape, y_train.shape, X_test.shape, y_test.shape)
```

[36] ✓ 0.0s

```
((80, 5), (20, 5))
```

Iris_Neural_Network.ipynb

Iris_Neural_Network.ipynb > Training the Model on Iris Dataset > print(np_train_data.shape, np_test_data.shape)

Generate | Code | Markdown | Run All | Restart | Execute Group 1 | Execute Group 2 | Clear All Outputs | Jupyter Variables | Outline | ...

```
# IrisClassifier : input_shape, output_shape, num_hidden_layers, num_hidden_neurons
model = IrisClassifier(4, 1, num_hidden_layers = 1, num_hidden_neurons = [15])
# train method() : X_train, y_train, X_test, y_test, learning_rate, n_epochs, logging_epochs = 10, validation_epochs = 100
model.train(X_train, y_train, X_test, y_test, 0.0001 , 35 , 5, 10)
```

[37] ✓ 0.2s

```
...
Epoch:  0
Loss:  [0.06153589]
Train Accuracy:  [0.875]
Validation Loss:  [0.24614358]
Validation Accuracy:  [1.]

Epoch:  5
Loss:  [0.00326697]
Train Accuracy:  [1.]

Epoch:  10
Loss:  [0.00164252]
Train Accuracy:  [1.]
Validation Loss:  [0.00657008]
Validation Accuracy:  [1.]

Epoch:  15
Loss:  [0.00109697]
Train Accuracy:  [1.]

Epoch:  20
Loss:  [0.00082412]
Train Accuracy:  [1.]
Validation Loss:  [0.00329646]
Validation Accuracy:  [1.]

Epoch:  25
Loss:  [0.00066044]
Train Accuracy:  [1.]
```

```
Epoch:  15
Loss:  [0.00109697]
Train Accuracy:  [1.]

Epoch:  20
Loss:  [0.00082412]
Train Accuracy:  [1.]
Validation Loss:  [0.00329646]
Validation Accuracy:  [1.]

Epoch:  25
Loss:  [0.00066044]
Train Accuracy:  [1.]

Epoch:  30
Loss:  [0.00055132]
Train Accuracy:  [1.]
Validation Loss:  [0.00220529]
Validation Accuracy:  [1.]

Epoch:  35
Loss:  [0.00047337]
Train Accuracy:  [1.]

Training Finished!
Epoch:  35
Loss:  [0.00047337]
Train Accuracy:  [1.]
Validation Loss:  [0.0018935]
Validation Accuracy:  [1.]
```

For code , refer GitHub

https://github.com/arnavjain2710/Computational-Intelligence-Lab-CS354N/tree/main/LAB%206