



Watson Studio democratizes machine learning and deep learning to accelerate infusion of AI in your business to drive innovation. Watson Studio provides a suite of tools and a collaborative environment for data scientists, developers and domain experts.

(http://cocl.us/pytorch_link_top)



Deeper Neural Networks with `nn.ModuleList()`

Table of Contents

In this lab, you will create a Deeper Neural Network with `nn.ModuleList()`

- [Neural Network Module and Function for Training](#)
- [Train and Validate the Model](#)

Estimated Time Needed: **25 min**

Preparation

We'll need the following libraries

In [1]:

```
# Import the libraries we need for this lab

import matplotlib.pyplot as plt
import numpy as np
import torch
import torch.nn as nn
import torch.nn.functional as F
from matplotlib.colors import ListedColormap
from torch.utils.data import Dataset, DataLoader

torch.manual_seed(1)
```

Out[1]:

<torch._C.Generator at 0x11c408610>

Function used to plot:

In [2]:

```
# Define the function to plot the diagram

def plot_decision_regions_3class(model, data_set):
    cmap_light = ListedColormap(['#FFAAAA', '#AAFFAA', '#00AAFF'])
    cmap_bold = ListedColormap(['#FF0000', '#00FF00', '#00AAFF'])
    X = data_set.x.numpy()
    y = data_set.y.numpy()
    h = .02
    x_min, x_max = X[:, 0].min() - 0.1, X[:, 0].max() + 0.1
    y_min, y_max = X[:, 1].min() - 0.1, X[:, 1].max() + 0.1
    xx, yy = np.meshgrid(np.arange(x_min, x_max, h), np.arange(y_min, y_max, h))
    XX = torch.Tensor(np.c_[xx.ravel(), yy.ravel()])
    _, yhat = torch.max(model(XX), 1)
    yhat = yhat.numpy().reshape(xx.shape)
    plt.pcolormesh(xx, yy, yhat, cmap=cmap_light)
    plt.plot(X[y[:, 0] == 0, 0], X[y[:, 0] == 0, 1], 'ro', label = 'y=0')
    plt.plot(X[y[:, 1] == 1, 0], X[y[:, 1] == 1, 1], 'go', label = 'y=1')
    plt.plot(X[y[:, 2] == 2, 0], X[y[:, 2] == 2, 1], 'o', label = 'y=2')
    plt.title("decision region")
    plt.legend()
```

Create Dataset Class

In [3]:

```
# Create Data Class

class Data(Dataset):

    # modified from: http://cs231n.github.io/neural-networks-case-study/
    # Constructor
    def __init__(self, K=3, N=500):
        D = 2
        X = np.zeros((N * K, D)) # data matrix (each row = single example)
        y = np.zeros(N * K, dtype='uint8') # class labels
        for j in range(K):
            ix = range(N * j, N * (j + 1))
            r = np.linspace(0.0, 1, N) # radius
            t = np.linspace(j * 4, (j + 1) * 4, N) + np.random.randn(N) * 0.2 # theta
            X[ix] = np.c_[r * np.sin(t), r * np.cos(t)]
            y[ix] = j
        self.y = torch.from_numpy(y).type(torch.LongTensor)
        self.x = torch.from_numpy(X).type(torch.FloatTensor)
        self.len = y.shape[0]

    # Getter
    def __getitem__(self, index):
        return self.x[index], self.y[index]

    # Get Length
    def __len__(self):
        return self.len

    # Plot the diagram
    def plot_stuff(self):
        plt.plot(self.x[self.y[:] == 0, 0].numpy(), self.x[self.y[:] == 0, 1].numpy()
        (), 'o', label="y = 0")
        plt.plot(self.x[self.y[:] == 1, 0].numpy(), self.x[self.y[:] == 1, 1].numpy()
        (), 'ro', label="y = 1")
        plt.plot(self.x[self.y[:] == 2, 0].numpy(), self.x[self.y[:] == 2, 1].numpy()
        (), 'go', label="y = 2")
        plt.legend()
```

Neural Network Module and Function for Training

Neural Network Module using ModuleList()

In [4]:

```
# Create Net model class

class Net(nn.Module):

    # Constructor
    def __init__(self, Layers):
        super(Net, self).__init__()
        self.hidden = nn.ModuleList()
        for input_size, output_size in zip(Layers, Layers[1:]):
            self.hidden.append(nn.Linear(input_size, output_size))

    # Prediction
    def forward(self, activation):
        L = len(self.hidden)
        for (l, linear_transform) in zip(range(L), self.hidden):
            if l < L - 1:
                activation = F.relu(linear_transform(activation))
            else:
                activation = linear_transform(activation)
        return activation
```

A function used to train.

In [5]:

```
# Define the function for training the model

def train(data_set, model, criterion, train_loader, optimizer, epochs=100):
    LOSS = []
    ACC = []
    for epoch in range(epochs):
        for x, y in train_loader:
            optimizer.zero_grad()
            yhat = model(x)
            loss = criterion(yhat, y)
            optimizer.zero_grad()
            loss.backward()
            optimizer.step()
            LOSS.append(loss.item())
        ACC.append(accuracy(model, data_set))

    fig, ax1 = plt.subplots()
    color = 'tab:red'
    ax1.plot(LOSS, color = color)
    ax1.set_xlabel('Iteration', color = color)
    ax1.set_ylabel('total loss', color = color)
    ax1.tick_params(axis = 'y', color = color)

    ax2 = ax1.twinx()
    color = 'tab:blue'
    ax2.set_ylabel('accuracy', color = color) # we already handled the x-label with
h ax1
    ax2.plot(ACC, color = color)
    ax2.tick_params(axis = 'y', color = color)
    fig.tight_layout() # otherwise the right y-label is slightly clipped

    plt.show()
    return LOSS
```

A function used to calculate accuracy

In [6]:

```
# The function to calculate the accuracy

def accuracy(model, data_set):
    _, yhat = torch.max(model(data_set.x), 1)
    return (yhat == data_set.y).numpy().mean()
```

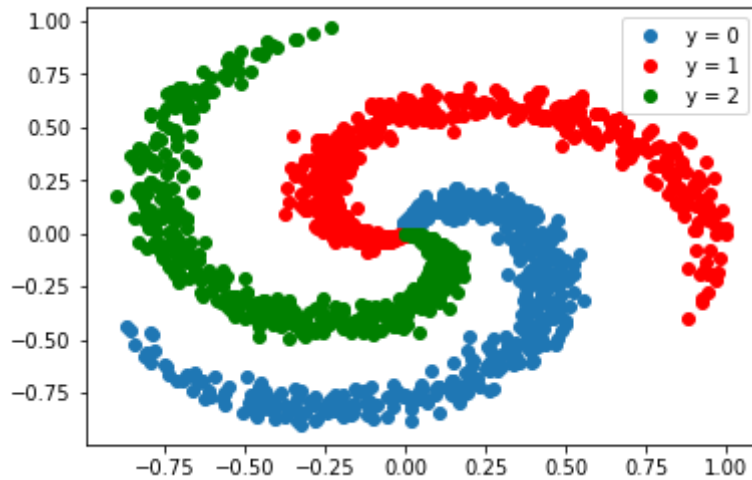
Train and Validate the Model

Create a dataset object:

In [7]:

```
# Create a Dataset object
```

```
data_set = Data()  
data_set.plot_stuff()  
data_set.y = data_set.y.view(-1)
```



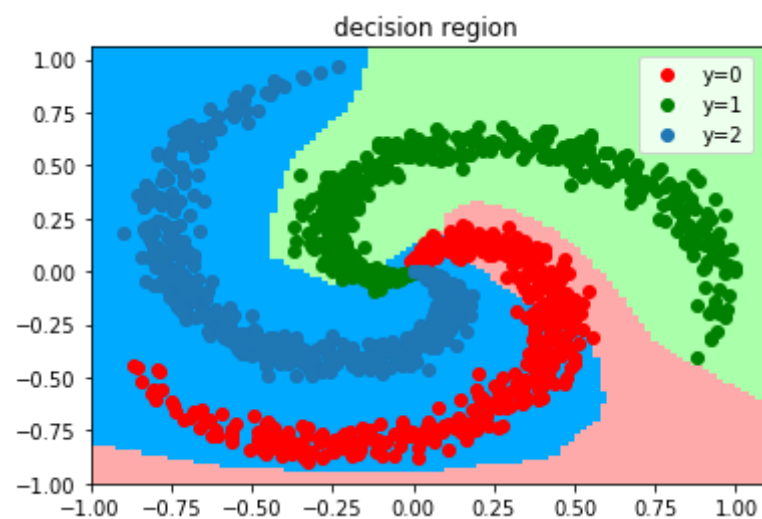
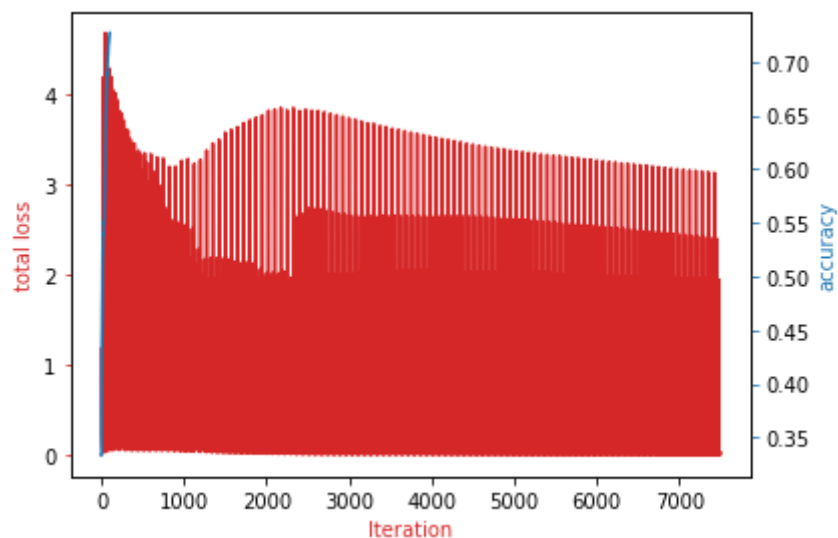
Create a network to classify three classes with 1 hidden layer with 50 neurons

In [8]:

```
# Train the model with 1 hidden layer with 50 neurons

Layers = [2, 50, 3]
model = Net(Layers)
learning_rate = 0.10
optimizer = torch.optim.SGD(model.parameters(), lr=learning_rate)
train_loader = DataLoader(dataset=data_set, batch_size=20)
criterion = nn.CrossEntropyLoss()
LOSS = train(data_set, model, criterion, train_loader, optimizer, epochs=100)

plot_decision_regions_3class(model, data_set)
```



Create a network to classify three classes with 2 hidden layers with 20 neurons in total

In [15]:

```
Net([3,3,4,3]).parameters
```

Out[15]:

```
<bound method Module.parameters of Net(
  (hidden): ModuleList(
    (0): Linear(in_features=3, out_features=3, bias=True)
    (1): Linear(in_features=3, out_features=4, bias=True)
    (2): Linear(in_features=4, out_features=3, bias=True)
  )
)>
```

In []:

```
# Train the model with 2 hidden layers with 20 neurons

Layers = [2, 10, 10, 3]
model = Net(Layers)
learning_rate = 0.01
optimizer = torch.optim.SGD(model.parameters(), lr=learning_rate)
train_loader = DataLoader(dataset=data_set, batch_size=20)
criterion = nn.CrossEntropyLoss()
LOSS = train(data_set, model, criterion, train_loader, optimizer, epochs=1000)

plot_decision_regions_3class(model, data_set)
```

Practice

Create a network with three hidden layers each with ten neurons, then train the network using the same process as above

In []:

```
# Practice: Create a network with three hidden layers each with ten neurons.
Layers = [2, 10, 10, 10, 3]
model = Net(Layers)
learning_rate = 0.01
optimizer = torch.optim.SGD(model.parameters(), lr = learning_rate)
train_loader = DataLoader(dataset = data_set, batch_size = 20)
criterion = nn.CrossEntropyLoss()
LOSS = train(data_set, model, criterion, train_loader, optimizer, epochs = 1000)
plot_decision_regions_3class(model, data_set)
```

Double-click **here** for the solution.

Get IBM Watson Studio free of charge!

Build and train AI & machine learning models, prepare and analyze data – all in a flexible, hybrid cloud environment. Get IBM Watson Studio Lite Plan free of charge.



Learn

Get started or get better with built-in learning.



Create

Use the best of open source tooling with IBM innovation.



Collaborate

Work smarter using community, work faster with your team.

[Sign Up For a Free Trial](#)

(http://cocl.us/pytorch_link_bottom)

(http://cocl.us/pytorch_link_bottom)

(http://cocl.us/pytorch_link_bottom)

(http://cocl.us/pytorch_link_bottom)

(http://cocl.us/pytorch_link_bottom)

(http://cocl.us/pytorch_link_bottom)

(http://cocl.us/pytorch_link_bottom)

About the Authors:

(http://cocl.us/pytorch_link_bottom)

(http://cocl.us/pytorch_link_bottom) Joseph Santarcangelo (<https://www.linkedin.com/in/joseph-s-50398b136/>) has a PhD in Electrical Engineering, his research focused on using machine learning, signal processing, and computer vision to determine how videos impact human cognition. Joseph has been working for IBM since he completed his PhD.

Other contributors: [Michelle Carey](https://www.linkedin.com/in/michelleccarey/) (<https://www.linkedin.com/in/michelleccarey/>), [Mavis Zhou](https://www.linkedin.com/in/jiahui-mavis-zhou-a4537814a/) (www.linkedin.com/in/jiahui-mavis-zhou-a4537814a)

Copyright © 2018 cognitiveclass.ai (cognitiveclass.ai?utm_source=bducopyrightlink&utm_medium=dswb&utm_campaign=bdu). This notebook and its source code are released under the terms of the [MIT License](https://bigdatauniversity.com/mit-license/) (<https://bigdatauniversity.com/mit-license/>).