```
import torch
import torch.optim as optim
from torch.nn.functional import normalize
from torch import nn as nn
import matplotlib.pyplot as plt
%matplotlib inline
import numpy as np
import os
import pandas as pd
from sklearn.preprocessing import MinMaxScaler, StandardScaler
from sklearn.model_selection import train_test_split
import time
from torchvision import datasets, transforms
import torch.nn.functional as F
os.environ['KMP_DUPLICATE_LIB_OK'] = 'True'
```

## **Final Project**

```
In [31]: | data_path = r"C:\Users\ccm51\Documents\ECGR_4105\Final Project\dataset csv file.csv"
            league_numpy = np.loadtxt(data_path, dtype = np.float32, delimiter = ",", skiprows=1)
            print(league_numpy.shape)
            (4028, 11)
data = League[:, :-1]
            torch.reshape(data, (-1, 10, 1, 1))
            data, data.shape, data.dtype
   Out[16]: (tensor([[ 95., 79., 31., ..., 17., 27., 3.],
                     [118., 79., 65., ..., 17., 47., 66.],
[ 68., 82., 115., ..., 65., 100., 85.],
                     [103., 56., 83., ..., 112., 15., 74.],
                     [ 16., 28., 88., ..., 148., 7., 87.], [122., 63., 88., ..., 83., 27., 85.]]),
             torch.Size([4028, 10]),
             torch.float32)
In [74]: | target = League[:,-1].long()
            target = torch.unsqueeze(target, dim=-1)
            target, target.shape, target.dtype
   Out[74]: (tensor([[1],
                     [0],
                     [0],
                     [1],
                     [0],
                     [0]]),
             torch.Size([4028, 1]),
             torch.int64)
data_std = torch.std(data, dim=1).unsqueeze(-1)
            data_var = torch.var(data, dim=1).unsqueeze(-1)
            data_normalized = (data - data_mean) / torch.sqrt(data_var)
            data_normalized = torch.cat((data_normalized, target), dim=1)
            train, test = train_test_split(data_normalized, train_size = 0.9, test_size = 0.1)
            print(train.shape)
            torch.Size([3625, 11])
```

```
model = nn.Sequential(nn.Linear(10, 512),
                                nn.Tanh(),
                                nn.Linear(512, 256),
                                nn.Tanh(),
                                nn.Linear(256, 64),
                                nn.Tanh(),
                                nn.Linear(64, 2),
                                nn.LogSoftmax(dim=1))
             loss_fn = nn.NLLLoss()
             learning_rate = 1e-2
             optimizer = optim.SGD(model.parameters(), lr = learning_rate)
             n_{epochs} = 500
for epoch in range(1, n_epochs + 1):
                tic = time.perf_counter()
                loss_train = 0
                for imgs in train_loader:
                    labels = imgs[:,-1].long()
                    imgs = imgs[:,:-1]
                    batch_size = imgs.shape[0]
                    outputs = model(imgs.view(batch_size, -1))
                    loss = loss_fn(outputs, labels)
                    optimizer.zero_grad()
                    loss.backward()
                    optimizer.step()
                    loss_train += loss.item()
                toc = time.perf counter()
                print(f"Epoch {epoch}, Duration = {round(toc - tic, 3)} seconds, Loss: {round(loss_train / len(train_loader), 5)}")
             main_toc = time.perf_counter()
             print(f"Total Training Time = {round(main_toc - main_tic, 3)} seconds")
             print(f"Average Training Time per Epoch = {round((main_toc - main_tic) / n_epochs , 3)} seconds")
             epocn 484, υματίοη = 3.975 seconds, Loss: 0.24409
             Epoch 485, Duration = 4.121 seconds, Loss: 0.24311
             Epoch 486, Duration = 4.229 seconds, Loss: 0.24251
             Epoch 487, Duration = 3.909 seconds, Loss: 0.24329
             Epoch 488, Duration = 3.96 seconds, Loss: 0.24284
             Epoch 489, Duration = 4.348 seconds, Loss: 0.24206
             Epoch 490, Duration = 4.205 seconds, Loss: 0.24256
             Epoch 491, Duration = 3.965 seconds, Loss: 0.24239
             Epoch 492, Duration = 4.65 seconds, Loss: 0.24258
             Epoch 493, Duration = 4.435 seconds, Loss: 0.24158
             Epoch 494, Duration = 4.784 seconds, Loss: 0.24179
             Epoch 495, Duration = 4.25 seconds, Loss: 0.24228
             Epoch 496, Duration = 4.688 seconds, Loss: 0.24161
             Epoch 497, Duration = 4.671 seconds, Loss: 0.24065
             Epoch 498, Duration = 4.131 seconds, Loss: 0.24127
             Epoch 499, Duration = 4.192 seconds, Loss: 0.24105
             Epoch 500, Duration = 5.14 seconds, Loss: 0.24129
             Total Training Time = 2310.457 seconds
             Average Training Time per Epoch = 4.621 seconds
correct = 0
             total = 0
             val tic = time.perf counter()
             with torch.no_grad():
                for imgs in val_loader:
                    labels = imgs[:,-1].long()
                    imgs = imgs[:,:-1]
                    batch_size = imgs.shape[0]
                    outputs = model(imgs.view(batch_size, -1))
                    _, predicted = torch.max(outputs, dim = 1)
                    total += labels.shape[0]
                    correct += int((predicted == labels).sum())
             val_toc = time.perf_counter()
             print(f"Accuracy: {round(correct/total, 3)}, Duration = {round(val_toc - val_tic, 3)} seconds")
             Accuracy: 0.677, Duration = 0.007 seconds
```

Out[124]: tensor([[-4.4187, -0.0121]], grad\_fn=<LogSoftmaxBackward0>)