```
In [2]:
          1 import numpy as np
           2 import pandas as pd
           3 pd.options.display.max columns=100
          4 from sklearn.model_selection import train_test_split, cross_val_score, cross
          5 import sklearn.metrics
          6 from sklearn.preprocessing import StandardScaler as SSc
          7
             import torch
          8 from torch import nn, optim
          9 from torch.autograd import Variable
         10 import torch.nn.functional as F
         11 from torch.utils.data import TensorDataset, DataLoader
             import matplotlib.pyplot as plt
         12
         13 %matplotlib inline
         14
         15 #set width of window to preference
         16 from IPython.core.display import display, HTML
         17 | display(HTML("<style>.container { width:90% !important; }</style>"))
In [3]:
          1 data = pd.read csv("Data-Prepped.csv",index col=0)
          2 | data = data.astype(np.float32)
           3 data.head()
Out[3]:
                                        Diamond Master GrandMaster LeagueIndex Age HoursPerWe
            Bronze Silver Gold Platinum
         0
               0.0
                      0.0
                           0.0
                                    0.0
                                             1.0
                                                    0.0
                                                                0.0
                                                                            5.0 27.0
         1
               0.0
                      0.0
                           0.0
                                    0.0
                                             1.0
                                                    0.0
                                                                0.0
                                                                            5.0 23.0
                                                                                              1
         2
                                                    0.0
                                                                0.0
               0.0
                      0.0
                           0.0
                                    1.0
                                             0.0
                                                                            4.0 30.0
                                                                                              1
          3
               0.0
                      0.0
                           1.0
                                    0.0
                                             0.0
                                                    0.0
                                                                0.0
                                                                            3.0 19.0
                                                                                              2
               0.0
                      0.0
                           1.0
                                    0.0
                                             0.0
                                                    0.0
                                                                0.0
                                                                            3.0 32.0
                                                                                              1
          1 X = data.iloc[:,data.columns != 'APM']
In [5]:
           2 Y = data.iloc[:,data.columns == 'APM']
           3 #transform input data (normalize scaling)
          4 \operatorname{ssc} = \operatorname{SSc}()
           5 Xft = ssc.fit transform(X)
          6 X = pd.DataFrame(Xft)
          7
             print("Xtr(Xtrain),Xtst(Xtest),Ytr(Ytrain),Ytst(Ytest) shapes: ")
          8 Xtr,Xtst,Ytr,Ytst = train_test_split(X,Y.values.ravel(),test_size=0.2,random
             print(Xtr.shape, Xtst.shape, Ytr.shape, Ytst.shape)
         10 Ytr = pd.DataFrame(Ytr)
```

Xtr(Xtrain),Xtst(Xtest),Ytr(Ytrain),Ytst(Ytest) shapes:

(2670, 25) (668, 25) (2670,) (668,)

```
In [7]:
          1
          2
             class Swish(torch.autograd.Function):
          3
                 @staticmethod
          4
                 def forward(ctx, i):
          5
                     result = i * torch.sigmoid(i)
          6
                     ctx.save_for_backward(i)
          7
                     return result
          8
          9
                 @staticmethod
                 def backward(ctx, grad_output):
         10
                     i = ctx.saved_variables[0]
         11
         12
                     sigmoid_i = torch.sigmoid(i)
         13
                     return grad_output * (sigmoid_i * (1 + i * (1 - sigmoid_i)))
         14
         15
             class swish(nn.Module):
                 def forward(self, input_tensor):
         16
         17
                     return Swish.apply(input_tensor)
             1.1.1
         18
         19
         20
         21
             class resultNet(nn.Module):
         22
                 def __init__(self, X_sz, Y_sz, a=0, b=0, c=0, d=0):
         23
                     super(resultNet, self).__init__()
         24
         25
                     self.inputSize = len(X.columns)
         26
                     self.outputSize = len(Y.columns)
         27
                     self.hidden0Size = a
         28
                     self.hidden1Size = b
         29
                     self.hidden2Size = c
         30
                     self.hidden3Size = d
         31
                     self.activation = F.selu
         32
                     self.outactivation = F.selu
         33
                     self.outsquish = torch.sigmoid
         34
         35
                     #Connect network
                     self.dpth = 0
         36
         37
                     if (self.hidden0Size != 0):
         38
                          self.c1 = nn.Linear(self.inputSize,self.hidden0Size)
         39
                          self.dpth += 1
                          print("adding layer 1")
         40
         41
                          if (self.hidden1Size != 0):
                              self.c2 = nn.Linear(self.hidden0Size,self.hidden1Size)
         42
         43
                              self.dpth += 1
                              print("adding layer 2")
         44
         45
                              if (self.hidden2Size != 0):
                                  self.c3 = nn.Linear(self.hidden1Size, self.hidden2Size)
         46
         47
                                  self.dpth += 1
         48
                                  print("adding layer 3")
         49
                                  if (self.hidden3Size != 0):
         50
                                      self.c4 = nn.Linear(self.hidden2Size,self.hidden3Siz
         51
                                      self.dpth += 1
         52
                                      print("adding layer 4")
         53
                                      self.c5 = nn.Linear(self.hidden3Size,self.outputSize
         54
                                  else:
         55
                                      self.c4 = nn.Linear(self.hidden2Size,self.outputSize
         56
                              else:
```

```
57
                        self.c3 = nn.Linear(self.hidden1Size,self.outputSize)
58
                else:
59
                    self.c2 = nn.Linear(self.hidden0Size,self.outputSize)
60
            else:
61
                self.c1 = nn.Linear(self.inputSize,self.outputSize)
62
63
        def forward(self, x):
64
65
            if (self.dpth == 0):
66
                out = self.outsquish(self.outactivation(self.c1(x)))
67
                #print("fwd dpth 0")
68
            elif (self.dpth == 1):
69
                x = self.activation(self.c1(x))
                out = self.outsquish(self.outactivation(self.c2(x)))
70
71
                #print("fwd dpth 1")
            elif (self.dpth == 2):
72
73
                x = self.activation(self.c1(x))
74
                x = self.activation(self.c2(x))
75
                out = self.outsquish(self.outactivation(self.c3(x)))
                #print("fwd dpth 2")
76
77
            elif (self.dpth == 3):
78
                x = self.activation(self.c1(x))
79
                x = self.activation(self.c2(x))
80
                x = self.activation(self.c3(x))
81
                out = self.outsquish(self.outactivation(self.c4(x)))
82
                #print("fwd dpth 3")
83
            elif (self.dpth == 4):
84
                x = self.activation(self.c1(x))
85
                x = self.activation(self.c2(x))
86
                x = self.activation(self.c3(x))
87
                x = self.activation(self.c4(x))
88
                out = self.outsquish(self.outactivation(self.c5(x)))
89
                #print("fwd dpth 4")
90
            return out
91
92 | h size = 12
93 | testNet = resultNet(len(Xtr.columns),len(Ytr.columns),25)
    for p in testNet.parameters():
94
        print(p)
95
adding layer 1
Parameter containing:
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         -1.0560e-01, 1.0589e-01, 1.7890e-01, -4.0324e-02, -1.3308e-01,
         2.0105e-02, -7.0236e-03, -8.2950e-02, -4.0359e-02, -1.9940e-02],
        [-1.6024e-02, 1.4258e-01, -1.5780e-01, -1.0095e-01, 9.8480e-02,
         1.4795e-01, -2.2709e-02, 1.9031e-01, 4.1711e-03,
                                                            1.9988e-01,
                     1.8523e-02,
         -1.5515e-01,
                                  9.3929e-02, 8.4581e-02,
                                                            6.8450e-03,
         -9.4609e-03, 1.6125e-01, 1.7621e-01, 3.9856e-02,
                                                            1.7383e-01,
         -1.4914e-01,
                      1.5930e-01, 9.3127e-02, 1.5983e-01,
                                                            1.2004e-01]],
       requires_grad=True)
Parameter containing:
tensor([-0.1468, 0.0856, 0.1191, 0.1105, -0.0675, -0.0679, 0.0974, -0.190
        -0.1052, -0.1337, -0.1347, -0.0526, 0.0032, 0.1431, -0.1516,
```

2,

```
8,
                -0.1948, 0.0931, 0.0579, 0.0799, -0.1024, 0.0350, 0.1849, -0.148
        1,
                 0.0133], requires grad=True)
        Parameter containing:
        tensor([[ 0.0362, -0.1253, -0.1335, -0.1459, 0.1180, -0.1118, 0.1488,
        61,
                 -0.0589, -0.0037, 0.1676, 0.0412, 0.1529, 0.0831, -0.0608, -0.03
        04,
                  0.0924, 0.1343, 0.1946, -0.0999, 0.1608, 0.0416, 0.1986, 0.10
        57,
                 -0.0635]], requires_grad=True)
        Parameter containing:
        tensor([0.1583], requires_grad=True)
In [8]:
            def test(model, lss fn, tst dload):
          1
          2
                 scores = []
          3
                with torch.no_grad():
          4
                     model.eval()
          5
                     for (x_btch, y_btch) in tst_dload:
          6
                         out_btch = model(x_btch.float())
          7
                         lss = lss_fn(out_btch.float()[:,0], y_btch.long())
          8
                         scores.append(lss.item())
          9
                     model.train()
                 return np.array(scores).mean()
         10
         11
         12 test(testNet, nn.MSELoss(), tst_dload)
Out[8]: 14623.01309742647
In [9]:
            rnet = resultNet(Xtr,Ytr)
          2
            print(rnet)
          3
          4 | learn rate = 1
          5
            inertia = .8
          6
          7
            criterion = nn.MSELoss()
          8
            optimizer = optim.SGD(rnet.parameters(), lr=learn_rate, momentum = inertia)
          9
         10
         11
            gpu_rdy = torch.cuda.is_available()
         12 if gpu_rdy:
         13
                 print("Using GPU")
         14
         15 else:
                 print("Using CPU")
         16
        resultNet(
          (c1): Linear(in_features=25, out_features=1, bias=True)
        Using GPU
```

```
In [11]:
              #you can keep iterating this block to continue training the network
           1
           2
           3 if gpu_rdy:
                  print("On GPU")
           4
           5 print("\nDisplayed score is MSE on 289 test data points while model is train
             rnet.train() #just in case
           6
           7
           8
             print("\nUntrained score:
                                                    {}\n".format(OGscr))
           9
          10 lr_= lambda epoch: (0.95 ** epoch)/10
          11 | scheduler = optim.lr_scheduler.LambdaLR(optimizer, lr_lambda=lr_)
          12 for epoch in range(n_epochs):
          13
          14
          15
                  if idx + btch_sz >= tr_shp:
          16
                      idx = 0
          17
                  else:
          18
                      idx += btch_sz
          19
          20
                  x_tr = Variable(x_tr[idx:(idx+btch_sz)].clone())
          21
          22
          23
                  for i, (x_btch, y_btch) in enumerate(tr_dload):
          24
          25
                      if gpu_rdy:
                          rnet.to(device)
          26
          27
                          x btch = x btch.cuda()
          28
                          y_btch = y_btch.cuda()
          29
          30
                      optimizer.zero_grad()
          31
                      out_btch = rnet(x_btch.float())
          32
          33
                      out_lss = criterion(out_btch, y_btch)
          34
          35
                      out_lss.backward()
                      optimizer.step()
          36
          37
                  t epochs += 1
          38
                  rnet.to('cpu')
          39
                  scr = test(rnet, criterion, tst dload)
          40
                  if (epoch %5) == 0:
          41
                      print("epoch {:06d} test data score: {}".format(t_epochs,scr))
          42
                  t epochs += 1
```

On_GPU

Displayed score is MSE on 289 test data points while model is trained on 1155 training data points

```
Untrained score: 14624.968778722427

epoch 000001 test data score: 14514.310403262867

epoch 000011 test data score: 14514.310345818014

epoch 000021 test data score: 14514.310288373163

epoch 000031 test data score: 14514.310288373163

epoch 000041 test data score: 14514.310288373163
```

epoch 000051 test data score: 14514.310259650736

```
epoch 000061 test data score: 14514.310259650736
epoch 000071 test data score: 14514.310259650736
epoch 000081 test data score: 14514.310259650736
epoch 000091 test data score: 14514.310259650736
epoch 000101 test data score: 14514.310259650736
epoch 000111 test data score: 14514.310259650736
epoch 000121 test data score: 14514.31023092831
epoch 000131 test data score: 14514.31023092831
epoch 000141 test data score: 14514.31023092831
epoch 000151 test data score: 14514.31023092831
epoch 000161 test data score: 14514.31023092831
epoch 000171 test data score: 14514.31023092831
epoch 000181 test data score: 14514.31023092831
epoch 000191 test data score: 14514.31023092831
epoch 000201 test data score: 14514.31023092831
epoch 000211 test data score: 14514.31023092831
epoch 000221 test data score: 14514.31023092831
epoch 000231 test data score: 14514.31023092831
epoch 000241 test data score: 14514.31023092831
epoch 000251 test data score: 14514.310202205883
epoch 000261 test data score: 14514.310202205883
epoch 000271 test data score: 14514.310202205883
epoch 000281 test data score: 14514.310202205883
epoch 000291 test data score: 14514.310202205883
epoch 000301 test data score: 14514.310202205883
epoch 000311 test data score: 14514.310202205883
epoch 000321 test data score: 14514.310202205883
epoch 000331 test data score: 14514.310202205883
epoch 000341 test data score: 14514.310202205883
epoch 000351 test data score: 14514.310202205883
epoch 000361 test data score: 14514.310202205883
epoch 000371 test data score: 14514.310202205883
epoch 000381 test data score: 14514.310202205883
epoch 000391 test data score: 14514.310202205883
epoch 000401 test data score: 14514.310202205883
```

```
In [12]: 1 tr_scr = test(rnet, criterion, tr_dload)
2 print("Score on training data for comparison: {}".format(tr_scr))
3
```

Score on training data for comparison: 15276.030535797574

C:\Users\Triplea657\anaconda3\envs\MSCS335\lib\site-packages\torch\nn\modules\l oss.py:528: UserWarning: Using a target size (torch.Size([20, 1])) that is different to the input size (torch.Size([20])). This will likely lead to incorrect results due to broadcasting. Please ensure they have the same size.

return F.mse_loss(input, target, reduction=self.reduction)

C:\Users\Triplea657\anaconda3\envs\MSCS335\lib\site-packages\torch\nn\modules\l oss.py:528: UserWarning: Using a target size (torch.Size([10, 1])) that is diff erent to the input size (torch.Size([10])). This will likely lead to incorrect results due to broadcasting. Please ensure they have the same size.

return F.mse_loss(input, target, reduction=self.reduction)

computer sit and crunch while I worked on other things, I would've been able to test more.