Logistic Regression

```
In [ ]:
        import sklearn
         import torch
         import torch.nn as nn
         import torch.nn.functional as F
        from sklearn.datasets import load_iris
        from sklearn.model_selection import train_test_split
        from sklearn.preprocessing import StandardScaler
In [ ]: | iris=load_iris()
        X=iris['data']
        y=iris['target']
In [ ]: | y.shape
Out[]: (150,)
In [ ]: ### Preprocess the Data
        scaler=StandardScaler()
        X=scaler.fit_transform(X)
        X=torch.tensor(X,dtype=torch.float32)
        y=torch.tensor(y,dtype=torch.float32)
In [ ]: X_train,X_test,y_train,y_test=train_test_split(X,y,test_size=.2,random_state=32)
In [ ]: X_train.shape
Out[ ]: torch.Size([120, 4])
In [ ]: | class LogisticRegression(nn.Module):
          def __init__(self,n_features,h1):
             super(LogisticRegression, self).__init__()
             self.layer1=nn.Linear(n_features,h1)
             self.layer2=nn.Linear(h1,3)
             self.softmax=nn.Softmax(dim=1)
          def forward(self,x):
            out=F.relu(self.layer1(x))
            out=F.relu(self.layer2(out))
             out=self.softmax(out)
             return out
```

```
In [ ]: def train(model,X,y,epocs=50,optimizer=None,loss_fn=None,bsz=20,seed=20):
          torch.manual_seed(seed)
          cost=[]
          for epoch in range(epocs):
            suffle_idx=torch.randperm(y.size(0))
            batches =torch.split(suffle_idx,bsz)
            for batch_idx in batches:
              ###1. Compute the output
              yhat=model(X[batch_idx])
              ###2. Compute the error
              loss=loss_fn(yhat,y[batch_idx].long())
              ###3. Compute the Gradient
              optimizer.zero_grad()
              loss.backward()
              ###4. Update the Parameters
              optimizer.step()
              ### Log the loss
            with torch.no_grad():
              yhat=model(X)
              curr_loss=loss_fn(yhat,y.long())
              print('Epoc:%3d ' % epoch,end="")
              print('BCE %.2f' % curr_loss)
              cost.append(curr_loss)
          return cost
```

```
In [ ]: model=LogisticRegression(n_features=X_train.size(1),h1=20)
    optimizer=torch.optim.Adam(model.parameters(),lr=.001)
    loss_fn=nn.CrossEntropyLoss()
```

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In [ ]: | cost =train(model, X_train, y_train, optimizer=optimizer, loss_fn=loss_fn)
        Epoc:
               0 BCE 1.13
        Epoc:
               1 BCE 1.12
        Epoc:
              2 BCE 1.11
        Epoc:
               3 BCE 1.10
        Epoc:
               4 BCE 1.10
        Epoc: 5 BCE 1.09
        Epoc:
               6 BCE 1.09
        Epoc: 7 BCE 1.08
        Epoc: 8 BCE 1.08
        Epoc: 9 BCE 1.07
        Epoc: 10 BCE 1.06
        Epoc: 11 BCE 1.06
        Epoc: 12 BCE 1.05
        Epoc: 13 BCE 1.05
        Epoc: 14 BCE 1.05
        Epoc: 15 BCE 1.04
        Epoc: 16 BCE 1.04
        Epoc: 17 BCE 1.03
        Epoc: 18 BCE 1.03
        Epoc: 19 BCE 1.03
        Epoc: 20 BCE 1.02
        Epoc: 21 BCE 1.02
        Epoc: 22 BCE 1.02
        Epoc: 23 BCE 1.01
        Epoc: 24 BCE 1.01
        Epoc: 25 BCE 1.01
        Epoc: 26 BCE 1.01
        Epoc: 27 BCE 1.01
        Epoc: 28 BCE 1.00
        Epoc: 29 BCE 1.00
        Epoc: 30 BCE 1.00
        Epoc: 31 BCE 1.00
        Epoc: 32 BCE 1.00
        Epoc: 33 BCE 1.00
        Epoc: 34 BCE 1.00
        Epoc: 35 BCE 1.00
        Epoc: 36 BCE 0.99
        Epoc: 37 BCE 0.99
        Epoc: 38 BCE 0.99
        Epoc: 39 BCE 0.99
        Epoc: 40 BCE 0.99
        Epoc: 41 BCE 0.99
        Epoc: 42 BCE 0.99
        Epoc: 43 BCE 0.99
        Epoc: 44 BCE 0.99
        Epoc: 45 BCE 0.99
        Epoc: 46 BCE 0.99
        Epoc: 47 BCE 0.99
        Epoc: 48 BCE 0.99
```

Epoc: 49 BCE 0.99

```
In [ ]: import matplotlib.pyplot as plt
    plt.plot(range(len(cost)),cost)
    plt.ylabel('Binary Cross Entropy')
    plt.xlabel('Epoc')
    plt.show()
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In [ ]: from sklearn.metrics import accuracy_score
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In [ ]: with torch.no_grad():
    train_pred=model(X_train)
    #train_pred
    train_acc=(torch.argmax(train_pred,dim=1)==y_train).float().mean()
    print('Training Accuracy: %.2f' %train_acc)
    test_pred=model(X_test)
    test_acc=(torch.argmax(test_pred,dim=1)==y_test).float().mean()
    print('Testing Accuracy: %.2f' %test_acc)
```

Training Accuracy: 0.65 Testing Accuracy: 0.70

Multinomial Logistic Regression

```
In [ ]: import torch
import torchvision
import torchvision.transforms as transforms
```

```
In [ ]: from matplotlib.transforms import Transform
        transform=transforms.ToTensor()
        ### Data Loading
        traindata=torchvision.datasets.MNIST(root='./data',train=True,download=True,transform=tr
        trainloader=torch.utils.data.DataLoader(traindata,batch_size=32,shuffle=False)
        testdata=torchvision.datasets.MNIST(root='./data',train=False,download=True,transform=tr
        ansform)
        testloader=torch.utils.data.DataLoader(testdata,batch_size=32,shuffle=False)
        Downloading http://yann.lecun.com/exdb/mnist/train-images-idx3-ubyte.gz
        Downloading http://yann.lecun.com/exdb/mnist/train-images-idx3-ubyte.gz to ./data/MNIS
        T/raw/train-images-idx3-ubyte.gz
        Extracting ./data/MNIST/raw/train-images-idx3-ubyte.gz to ./data/MNIST/raw
        Downloading http://yann.lecun.com/exdb/mnist/train-labels-idx1-ubyte.gz
        Downloading http://yann.lecun.com/exdb/mnist/train-labels-idx1-ubyte.gz to ./data/MNIS
        T/raw/train-labels-idx1-ubyte.gz
        Extracting ./data/MNIST/raw/train-labels-idx1-ubyte.gz to ./data/MNIST/raw
        Downloading http://yann.lecun.com/exdb/mnist/t10k-images-idx3-ubyte.gz
        Downloading http://yann.lecun.com/exdb/mnist/t10k-images-idx3-ubyte.gz to ./data/MNIST/
        raw/t10k-images-idx3-ubyte.gz
        Extracting ./data/MNIST/raw/t10k-images-idx3-ubyte.gz to ./data/MNIST/raw
        Downloading http://yann.lecun.com/exdb/mnist/t10k-labels-idx1-ubyte.gz
        Downloading http://yann.lecun.com/exdb/mnist/t10k-labels-idx1-ubyte.gz to ./data/MNIST/
        raw/t10k-labels-idx1-ubyte.gz
        Extracting ./data/MNIST/raw/t10k-labels-idx1-ubyte.gz to ./data/MNIST/raw
```

In []: import matplotlib.pyplot as plt
import numpy as np
def imshow(img):
 img=img.numpy()

plt.imshow(img)
plt.show()

img=np.transpose(img,(1,2,0))

```
In [ ]:
        dataiter=iter(trainloader)
        images,labels=dataiter.next()
         imshow(torchvision.utils.make_grid(images))
         print(labels)
          40
          60
         100
         120
        tensor([5, 0, 4, 1, 9, 2, 1, 3, 1, 4, 3, 5, 3, 6, 1, 7, 2, 8, 6, 9, 4, 0, 9, 1,
                 1, 2, 4, 3, 2, 7, 3, 8])
In [ ]: | images.view(images.size(0),-1).shape
Out[ ]: torch.Size([32, 784])
In [ ]: | print(images[0].shape)
        torch.Size([1, 28, 28])
In [ ]:
        import torch.nn as nn
         import torch.nn.functional as F
        class LogisticNet(nn.Module):
          def __init__(self):
            super().__init__()
             self.layer1=nn.Linear(1*28*28,100)
             self.layer2=nn.Linear(100,30)
             self.layer3=nn.Linear(30,10)
          def forward(self,x):
            x=F.relu(self.layer1(x))
            x=F.relu(self.layer2(x))
            x=self.layer3(x)
             return x
        model=LogisticNet()
In [ ]: | model
Out[ ]: LogisticNet(
          (layer1): Linear(in_features=784, out_features=100, bias=True)
          (layer2): Linear(in_features=100, out_features=30, bias=True)
          (layer3): Linear(in_features=30, out_features=10, bias=True)
In [ ]: | import torch.optim as optim
```

```
In [ ]: | def train_mnst(model,train_loader,test_loader,EPOCHS=100,lr=.001,seed=32):
          torch.manual seed(seed)
          device = torch.device('cuda' if torch.cuda.is_available() else 'cpu')
          model.to(device)
          optimizer=optim.SGD(model.parameters(),lr=lr)
          loss_function=nn.CrossEntropyLoss()
          for epoch in range(EPOCHS):
             curr_loss=0.0
            currect=0.0
            t_currect=0.0
             for i,data in enumerate(train_loader,0):
               ## Get a Batch of Data
               input, target=data[0].to(device), data[1].to(device)
               ###1. Compute the output
               ###Flatten the input
               input=input.view(input.size(0),-1)
               yhat=model(input)
               ###2. Compute the Loss
               loss=loss_function(yhat, target)
              ###3. Compute the gradient
               optimizer.zero_grad()
               loss.backward()
              ###4. Update the parameters
              optimizer.step()
               ### Log the record
              with torch.no_grad():
                 yhat=model(input)
                 err=loss_function(yhat, target)
                 curr loss+=err
                 b_currect=((torch.argmax(yhat,dim=1)==target).sum()).float()
                 currect+=b currect
             ### Calculate the accuracy
            with torch.no_grad():
               for data in test_loader:
                   ims,lbs=data[0].to(device),data[1].to(device)
                   pred=model(ims.view(ims.size(0),-1))
                   t_currect+=(torch.argmax(pred,dim=1)==lbs).sum()
               print('Current Epoch: %3d' %(epoch+1),end="")
               print(' | Loss: %5.2f' %curr loss,end="")
               accuracy=currect.sum()/len(traindata)*100
               test_accuracy=t_currect.sum()/len(testdata)*100
               #print(currect)
               print(' | Accuracy: %.2f' %accuracy,end="")
               print(' | Test Accuracy: %.2f' %test_accuracy )
          return cost
```