## **Fashion-MNIST Project**

## **Preparation**

```
In [5]: # PyTorch Modules
    from torch.utils.data import Dataset, DataLoader
    from torchvision import transforms
    import torch.nn as nn
    import torchvision.transforms as transforms
    import torchvision.datasets as dsets
    torch.manual_seed(0)

In [2]: # Other non-PyTorch Modules
    from matplotlib.pyplot import imshow
    import matplotlib.pylab as plt
    from PIL import Image

In [3]: def show_data(data_sample):
        plt.imshow(data_sample[0].numpy().reshape(IMAGE_SIZE, IMAGE_SIZE), cmap='gray')
        plt.title('y = '+ str(data_sample[1].item()))
```

## 1. Create a Dataset Class ¶

In this section, you will load a Dataset object and transform the dataset. U

Resize to resize the image.

ToTensor to concert the image to a tensor.

Use the compose function of compse the

Create two dataset objects for the Fashion MNIST dataset. One for training data called dataset\_train and one for validation data dataset\_val .

```
In [5]:
    dataset_train=dsets.FashionMNIST(root= '.fashion/data', train=True, transform=composed, download=True)
    dataset_val=dsets.FashionMNIST(root= '.fashion/data', train=False, transform=composed, download=True)
```

```
Downloading http://fashion-mnist.s3-website.eu-central-1.amazonaws.com/train-images-idx3-ubyte.gz (http://fashion-mnist.s3-website.eu-central-1.amazonaws.com/train-images-idx3-ubyte.gz)

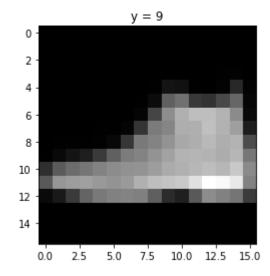
Downloading http://fashion-mnist.s3-website.eu-central-1.amazonaws.com/train-labels-idx1-ubyte.gz (http://fashion-mnist.s3-website.eu-central-1.amazonaws.com/train-labels-idx1-ubyte.gz)

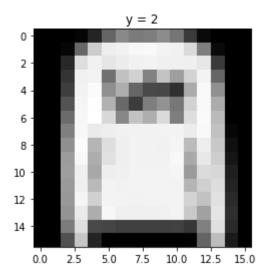
Downloading http://fashion-mnist.s3-website.eu-central-1.amazonaws.com/t10k-images-idx3-ubyte.gz (http://fashion-mnist.s3-website.eu-central-1.amazonaws.com/t10k-images-idx3-ubyte.gz)

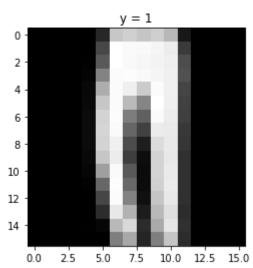
Downloading http://fashion-mnist.s3-website.eu-central-1.amazonaws.com/t10k-labels-idx1-ubyte.gz (http://fashion-mnist.s3-website.eu-central-1.amazonaws.com/t10k-labels-idx1-ubyte.gz)

Processing...

Done!
```







## # 2. Create a Convolutional Neural Network class

Constructor using Batch Norm

```
In [7]: class CNN batch(nn.Module):
            # Contructor
            def init (self, out 1=16, out 2=32, number of classes=10):
                super(CNN batch, self). init ()
                self.cnn1 = nn.Conv2d(in channels=1, out channels=out 1, kernel size=5, padding=2)
                self.conv1 bn = nn.BatchNorm2d(out 1)
                self.maxpool1=nn.MaxPool2d(kernel size=2)
                self.cnn2 = nn.Conv2d(in channels=out 1, out channels=out 2, kernel size=5, stride=1, padding=2)
                self.conv2 bn = nn.BatchNorm2d(out 2)
                self.maxpool2=nn.MaxPool2d(kernel size=2)
                self.fc1 = nn.Linear(out 2 * 4 * 4, number of classes)
                self.bn fc1 = nn.BatchNorm1d(10)
            # Prediction
            def forward(self, x):
                x = self.cnn1(x)
                x=self.conv1 bn(x)
                x = torch.relu(x)
                x = self.maxpool1(x)
                x = self.cnn2(x)
                x=self.conv2 bn(x)
                x = torch.relu(x)
                x = self.maxpool2(x)
                x = x.view(x.size(0), -1)
                x = self.fc1(x)
                x=self.bn fc1(x)
                return x
```

Constructor for regular Convolutional Neural Network

```
In [8]: | class CNN(nn.Module):
            # Contructor
            def __init__(self, out_1=16, out_2=32,number_of_classes=10):
                super(CNN, self). init ()
                self.cnn1 = nn.Conv2d(in channels=1, out channels=out 1, kernel size=5, padding=2)
                self.maxpool1=nn.MaxPool2d(kernel size=2)
                self.cnn2 = nn.Conv2d(in channels=out 1, out channels=out 2, kernel size=5, stride=1, padding=2)
                self.maxpool2=nn.MaxPool2d(kernel size=2)
                self.fc1 = nn.Linear(out_2 * 4 * 4, number of classes)
            # Prediction
            def forward(self, x):
                x = self.cnn1(x)
                x = torch.relu(x)
                x = self.maxpool1(x)
                x = self.cnn2(x)
                x = torch.relu(x)
                x = self.maxpool2(x)
                x = x.view(x.size(0), -1)
                x = self.fc1(x)
                return x
```

train loader and validation loader

```
In [9]: train_loader = torch.utils.data.DataLoader(dataset=dataset_train, batch_size=100 )
  test_loader = torch.utils.data.DataLoader(dataset=dataset_val, batch_size=100 )
```

Convolutional Neural Network object

```
In [12]: #model = CNN(out_1=16, out_2=32,number_of_classes=10)
model =CNN_batch(out_1=16, out_2=32,number_of_classes=10)
```

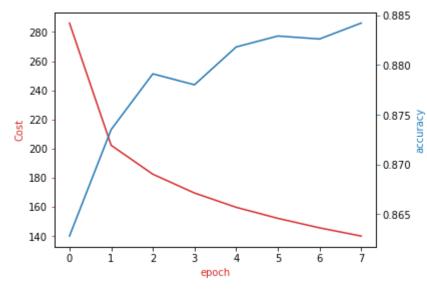
Code used to train the model

```
In [13]: | import time
         start_time = time.time()
          cost list=[]
         accuracy list=[]
         N test=len(dataset val)
         learning_rate =0.1
         optimizer = torch.optim.SGD(model.parameters(), lr = learning rate, momentum=0.3)
         criterion = nn.CrossEntropyLoss()
         n epochs=8
         for epoch in range(n_epochs):
             cost=0
             model.train()
             for x, y in train_loader:
                 optimizer.zero_grad()
                 z = model(x)
                 loss = criterion(z, y)
                 loss.backward()
                 optimizer.step()
                 cost+=loss.item()
              correct=0
             #perform a prediction on the validation data
             model.eval()
             for x_test, y_test in test_loader:
                 z = model(x test)
                 _, yhat = torch.max(z.data, 1)
                 correct += (yhat == y_test).sum().item()
             accuracy = correct / N_test
             accuracy_list.append(accuracy)
             cost list.append(cost)
```

To plot the Cost and accuracy for each epoch for the training and testing data, respectively.

```
In [14]:
    fig, ax1 = plt.subplots()
    color = 'tab:red'
    ax1.plot(cost_list, color=color)
    ax1.set_xlabel('epoch', color=color)
    ax1.set_ylabel('Cost', color=color)
    ax1.tick_params(axis='y', color=color)

ax2 = ax1.twinx()
    color = 'tab:blue'
    ax2.set_ylabel('accuracy', color=color)
    ax2.set_xlabel('epoch', color=color)
    ax2.plot( accuracy_list, color=color)
    ax2.tick_params(axis='y', color=color)
    fig.tight_layout()
```



dataset: <a href="https://github.com/zalandoresearch/fashion-mnist">https://github.com/zalandoresearch/fashion-mnist</a> (<a href="https://github.com/zalandoresearch/fashion-mnist</a> (<a href="https://github.com/zalandoresearch/fashion-m