

A hand is holding a piece of white paper with handwritten digits 0 through 9. The digits are written in a cursive style. A black rectangular box with a white border is overlaid on the center of the image, containing the title and group information. In the foreground, there is a white plastic component and a green circuit board with a blue and brown ribbon cable.

# Identify Hand Written Digits

**Group 8**

**Walia, Neeraj, Zhen Zhang**

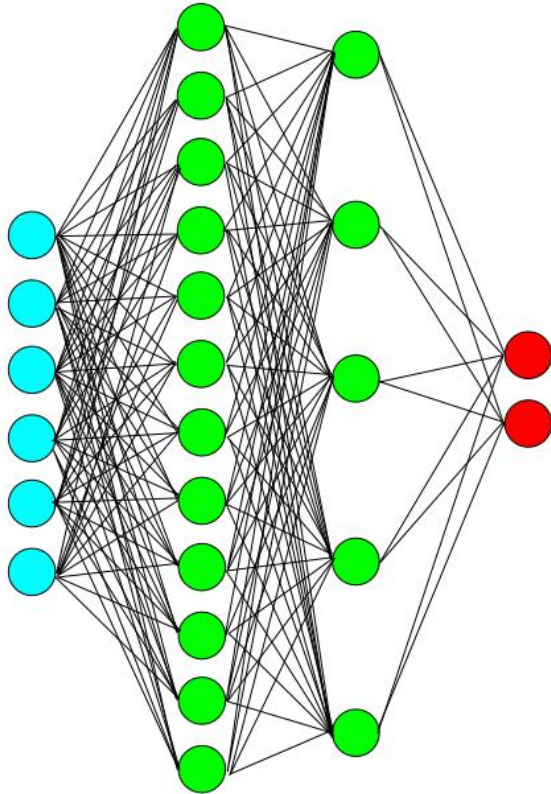
# MNIST DATASET

- **Data contains grey scaled images of hand written digits from 0 to 9.**
- **Each image size is 28 \* 28 pixels**
- **Each pixel has a number between 0 to 255 attached to it. The number identifies the darkness of the pixel.**
- **Training data set has 785 columns with first column as label which describes the hand written digit.**
- **We got the data from kaggle website.**
- **Data is already divided into training and test data sets.**
- **For training data set there are 42000 rows of input.**
- **For test data set there are 28000 rows of data.**



# APPROACH

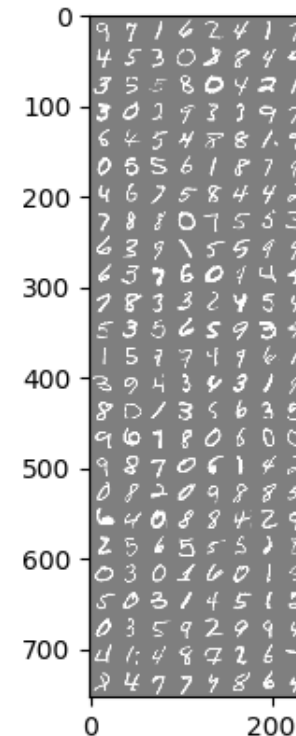
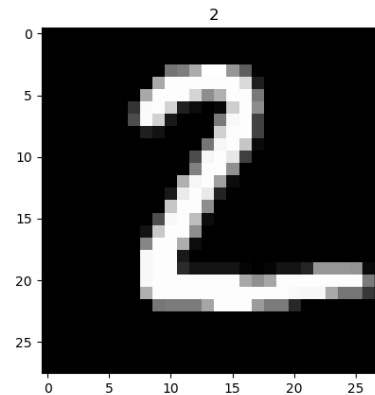
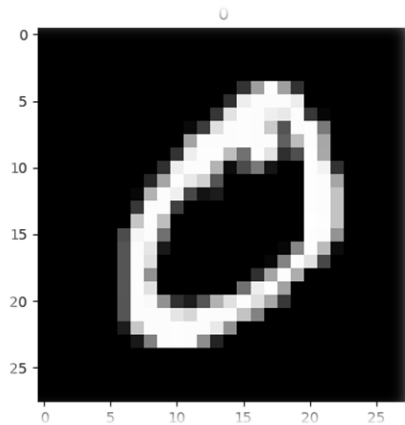
Input layer      Hidden Layers      Output Layer



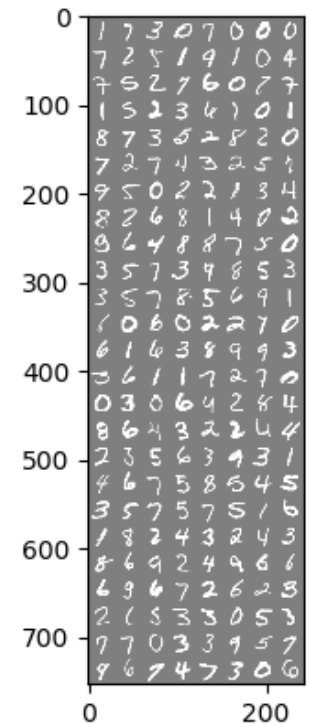
- We built a multi – layer network architecture for this problem. And we tested the number of layers in the network to achieve best accuracy with least complex model.
  - Along with layers, we have changed other variables to test the accuracy like batch size, epoch etc.
- We have designed a Multilayered Perceptron (MLP) and Convolution network for this project and compared their accuracy.
- Pytorch framework is used for this project.

# DATA OVERVIEW

```
training image:torch.Size([60000, 28, 28])
training label:torch.Size([60000])
testing image:torch.Size([10000, 28, 28])
testing label:torch.Size([10000])
```



Randomly Training Samples



Randomly Testing Samples

# MULTILAYERED PERCEPTRON

- We have finalized a two-layer network architecture with the following parameters. We chose two layered network because adding more layers was not improving the accuracy of the results.
  - Epoch\_size = 50
  - Image\_dimen = 1
  - Image\_size = 28
  - Hidden\_size = 200
  - Batch\_size = 200
  - Class\_num = 10

**We have used `CrossEntropyLoss()` criterion as this is a classification problem. This criterion helps in training data with number of classes.**

**We also test with `nn.NLLLoss()` criterion but did not found any difference in accuracy or performance.**

# MULTILAYERED PERCEPTRON

**optim.SGD()** is used to optimize the algorithm. Following are the options selected for this optimizer:

**Learning rate  
= 0.1**

**Momentum =  
0.9**



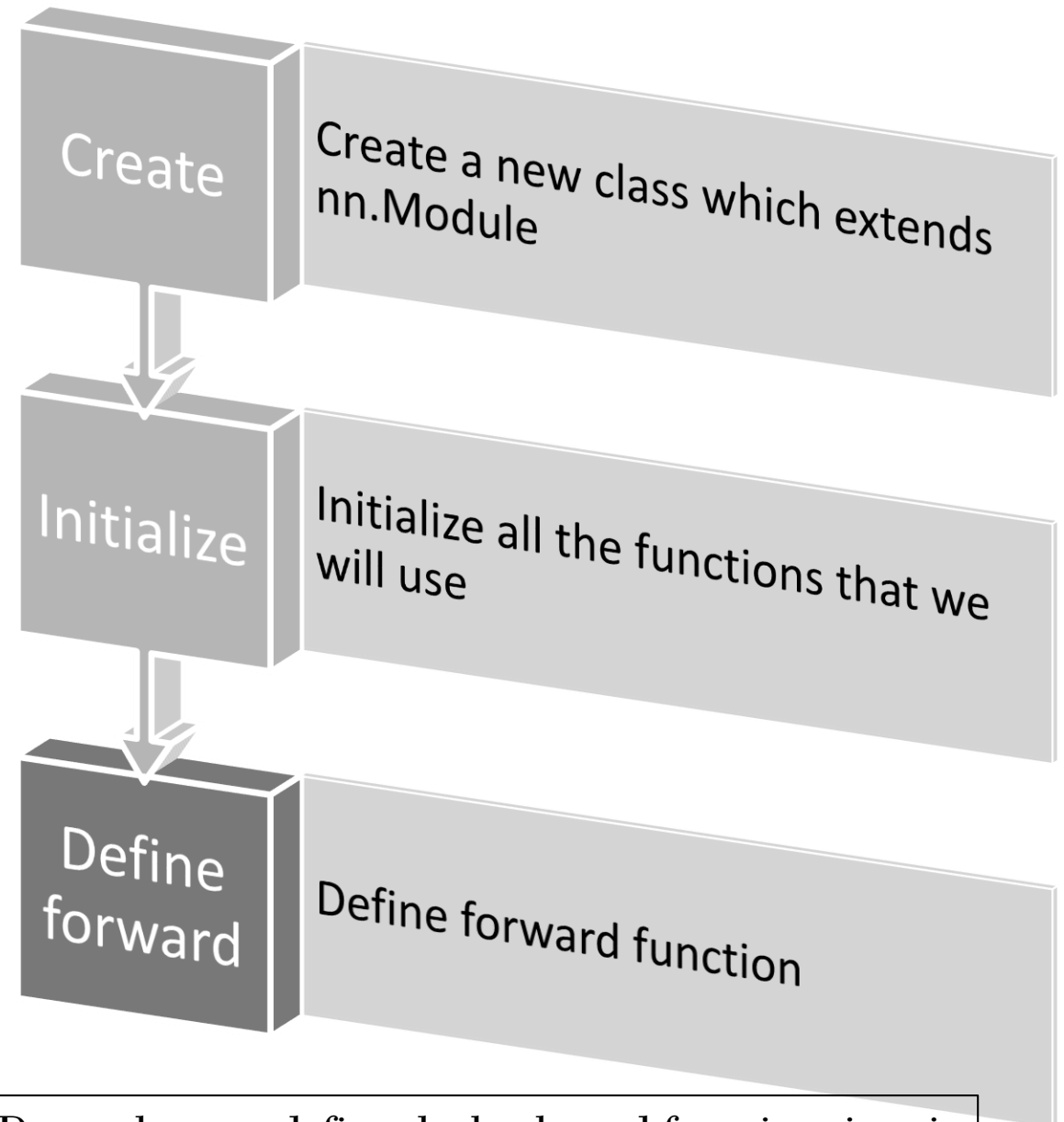
**Model got trained using the above mentioned parameters on training data set and its accuracy was test on test dataset.**

**Accuracy = 98%**



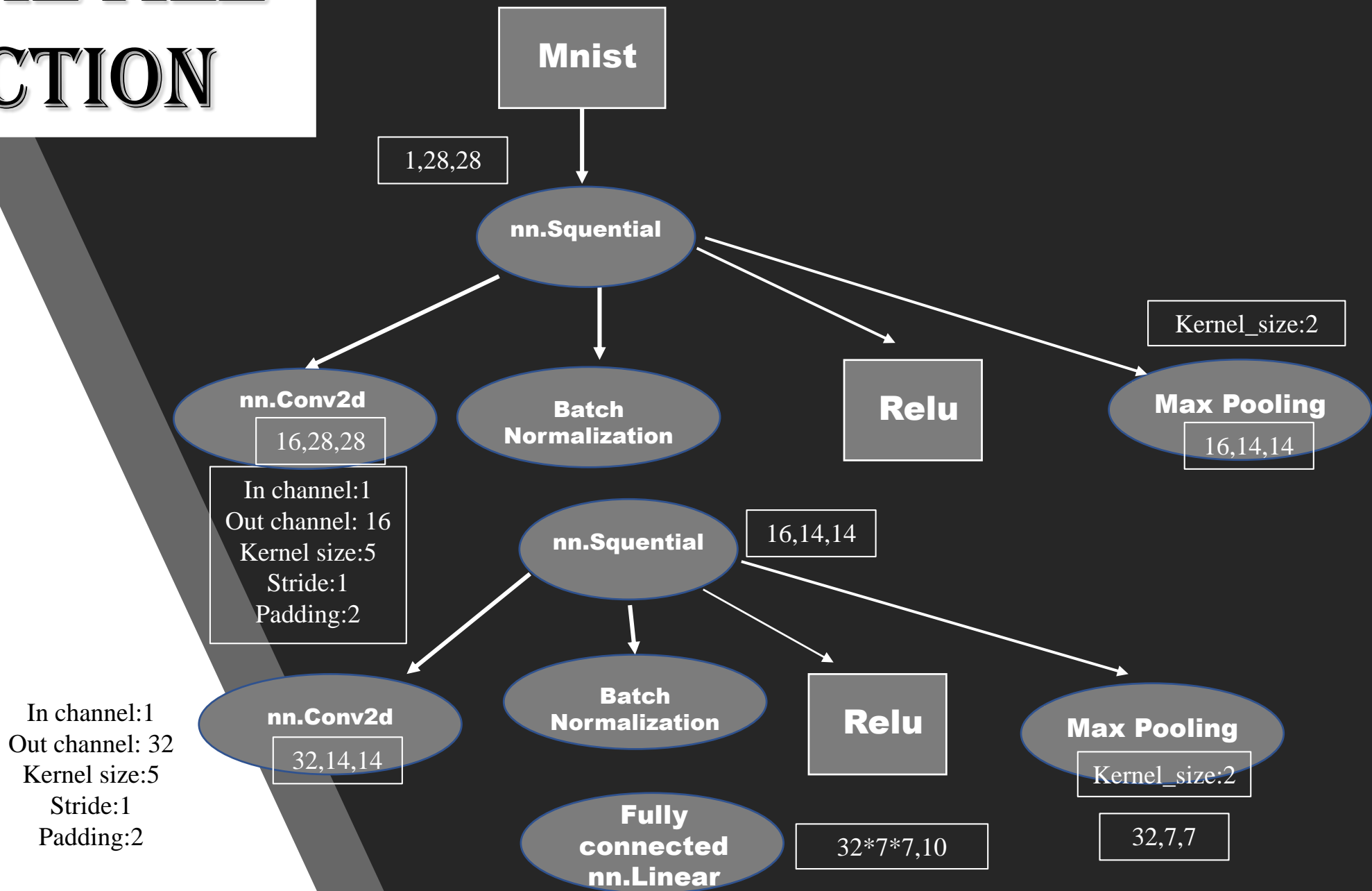
# DEFINE A CONVOLUTION NEURAL NETWORK

BATCH\_SIZE = 200  
LEARNING\_RATE = 0.1  
EPOCH\_SIZE = 50  
HIDDEN\_LAYERS=100



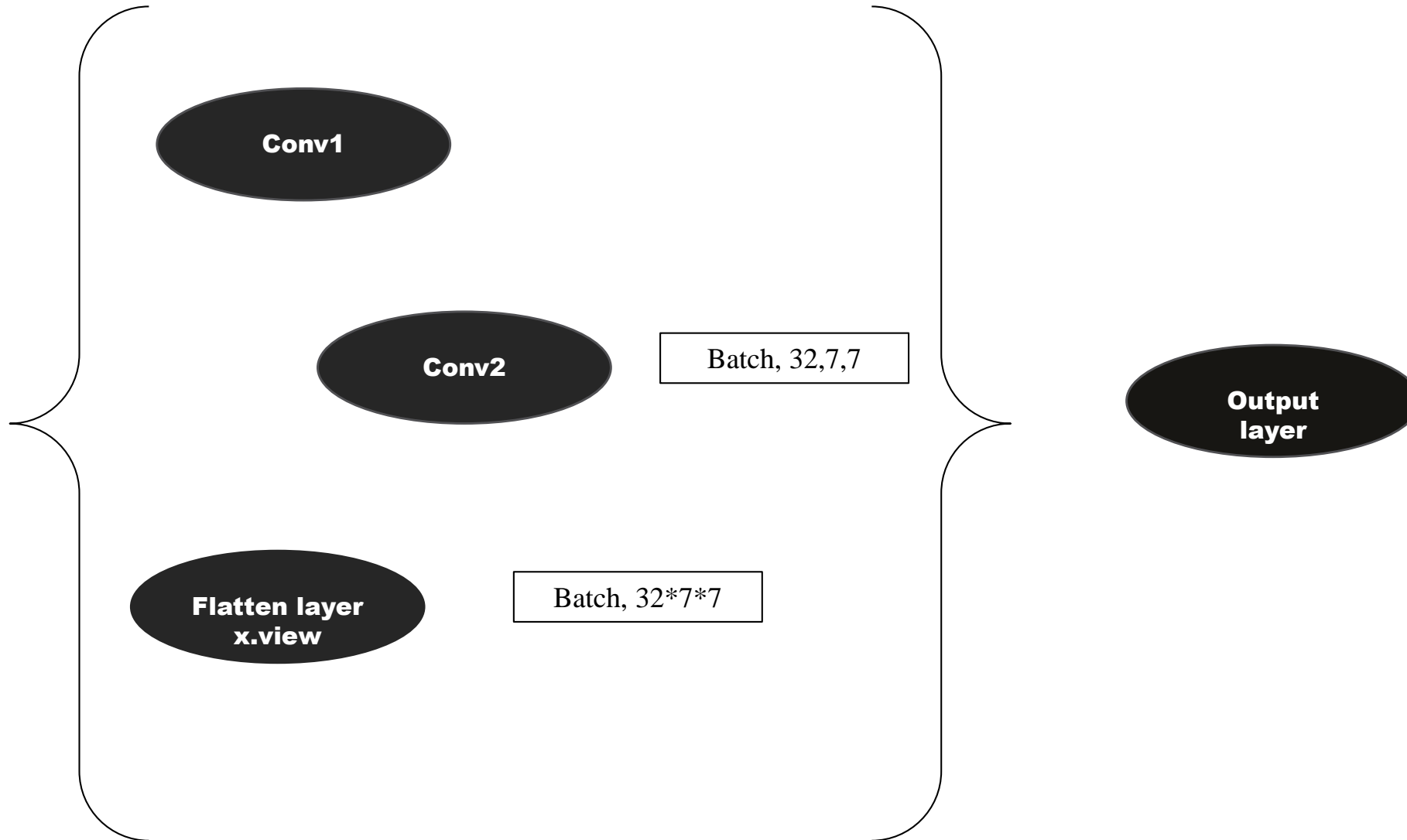
Do not have to define the backward function since it automatically determined by the autograd package.

# INITIAL ALL FUNCTION





# DEFINE FORWARD FUNCTION



# CONVOLUTION NEURAL NETWORK

---

## Net Architecture

```
CNN(  
  (layer1): Sequential(  
    (0): Conv2d(1, 16, kernel_size=(5, 5), stride=(1, 1), padding=(2, 2))  
    (1): BatchNorm2d(16, eps=1e-05, momentum=0.1, affine=True)  
    (2): ReLU()  
    (3): MaxPool2d(kernel_size=(2, 2), stride=(2, 2), dilation=(1, 1), ceil_mode=False)  
  )
```

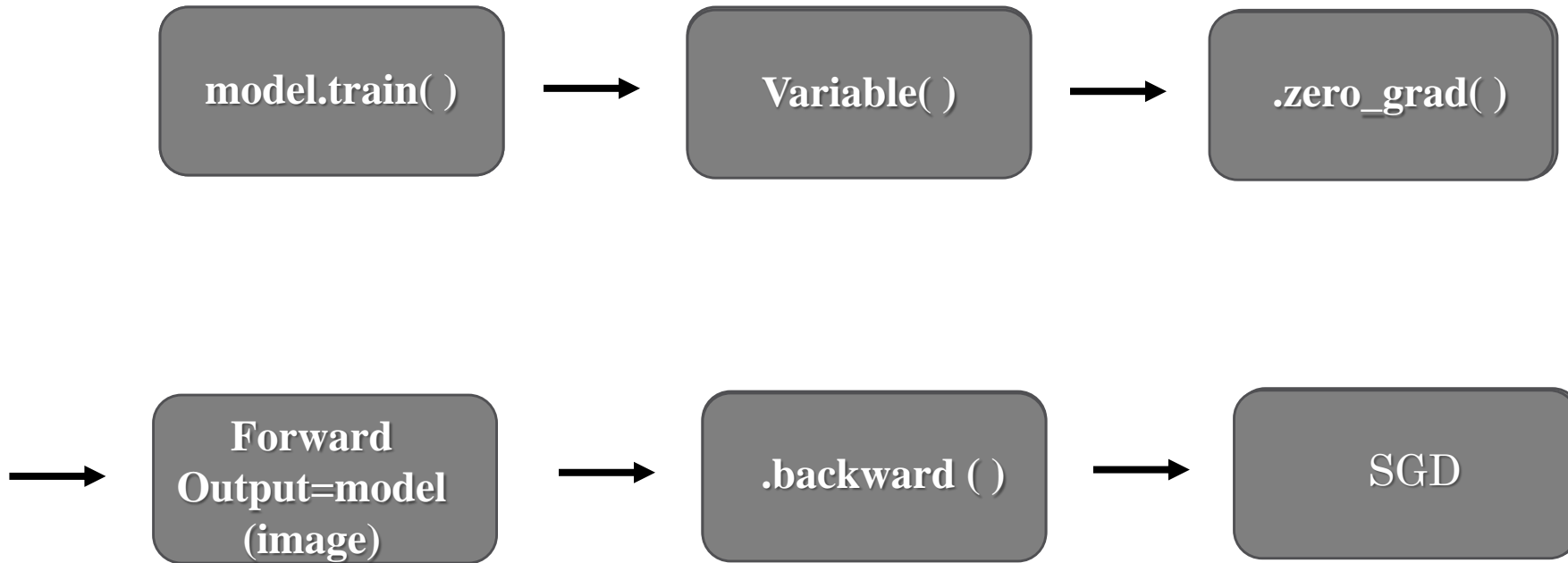
---

---

```
(layer2): Sequential(  
  (0): Conv2d(16, 32, kernel_size=(5, 5), stride=(1, 1), padding=(2, 2))  
  (1): BatchNorm2d(32, eps=1e-05, momentum=0.1, affine=True)  
  (2): ReLU()  
  (3): MaxPool2d(kernel_size=(2, 2), stride=(2, 2), dilation=(1, 1), ceil_mode=False)  
)
```

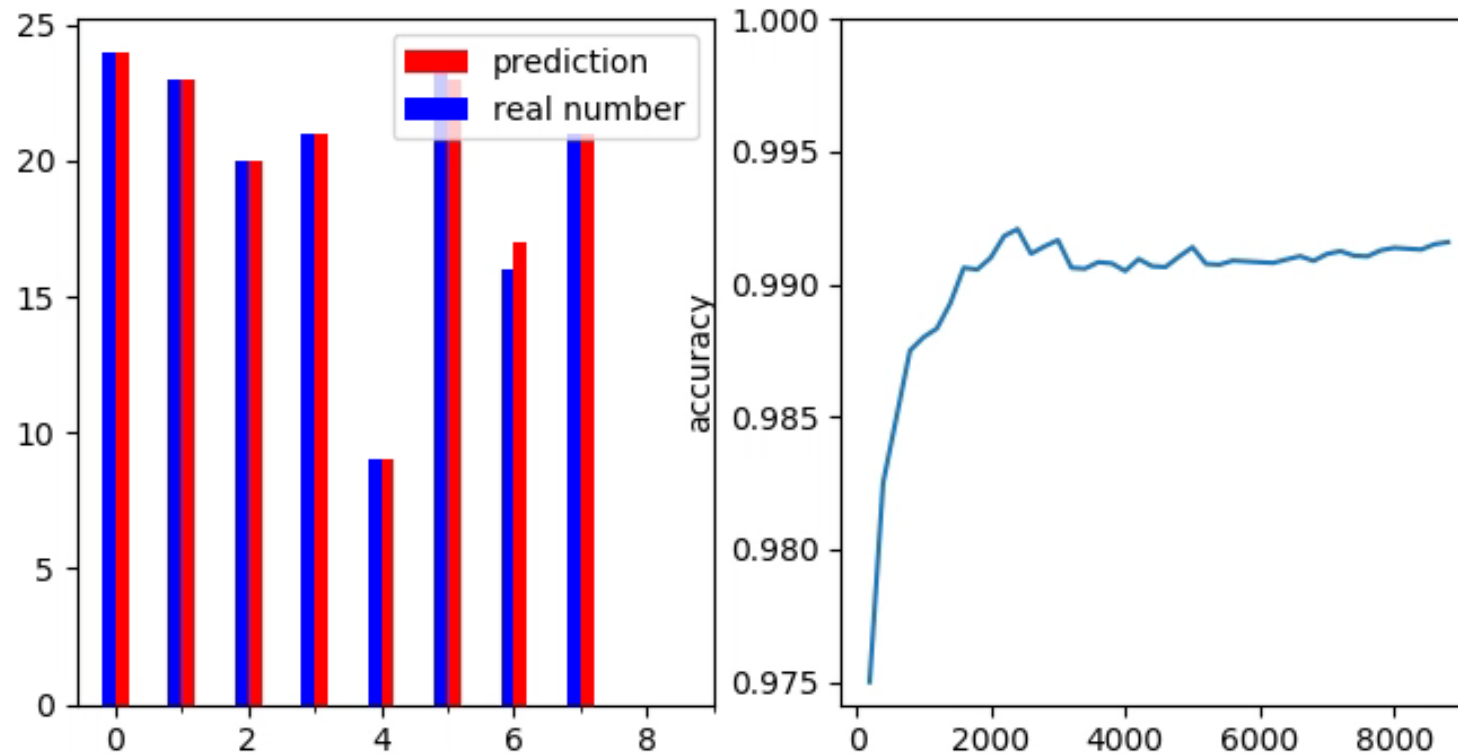
# TRAINING NETWORK

Optimizer: Stochastic Gradient Descent  
Loss: Cross entropy

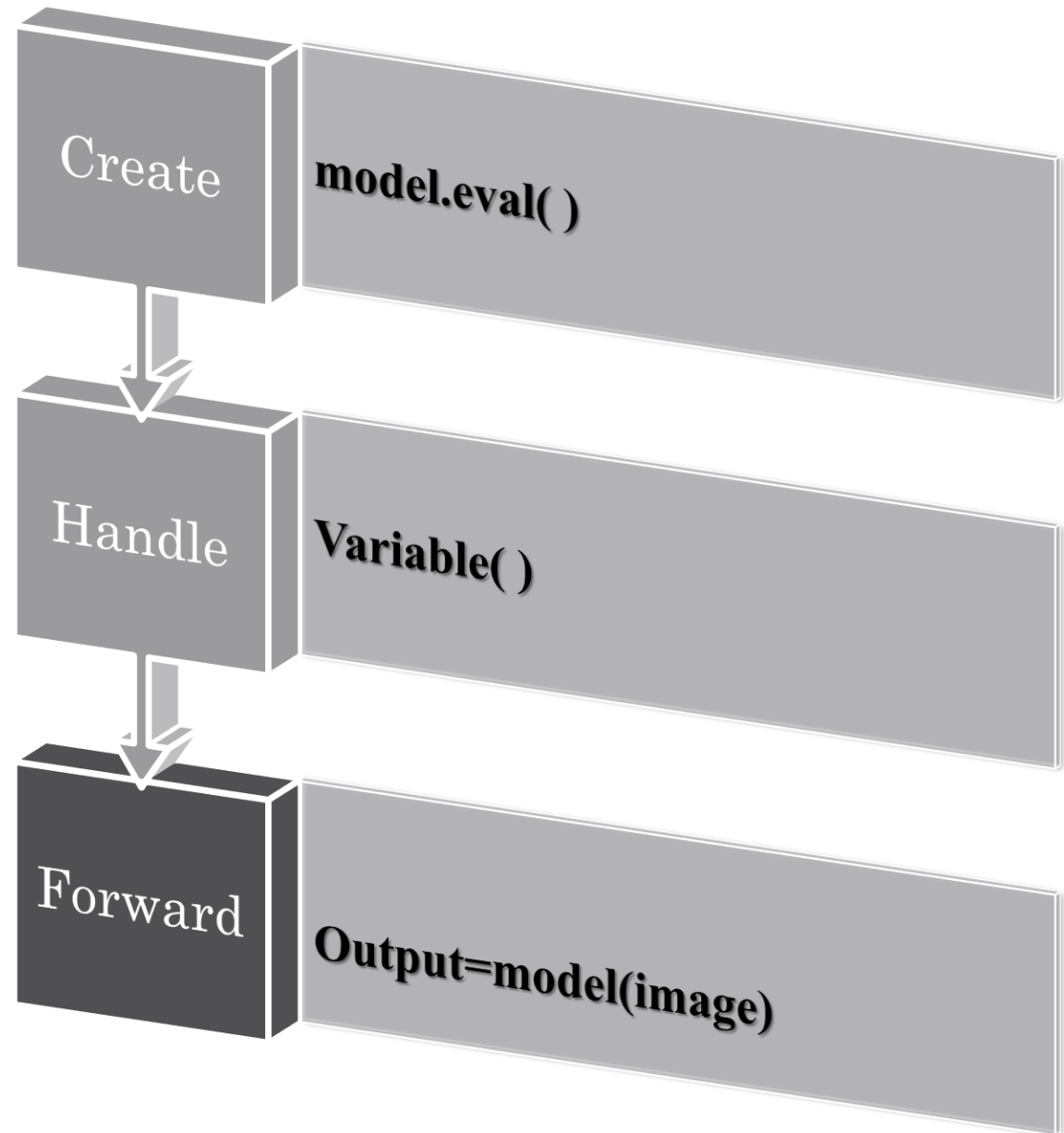


- `Torch.optim`
- $\text{weight} = \text{weight} - \text{learning\_rate} * \text{gradient}$

# Training Results



# TESTING OUR TRAINED NETWORK



# TESTING RESULTS

Test Accuracy of the model on the 10000 test images: 98 %

Training Epoch 9

[9, 50] loss: 0.009

[9, 100] loss: 0.010

[9, 150] loss: 0.008

[9, 200] loss: 0.010

[9, 250] loss: 0.018

[9, 300] loss: 0.012

Training Completed

Test Accuracy of the model on the 10000 test images: 99 %

Training Epoch 10

[10, 50] loss: 0.007

[10, 100] loss: 0.009

[10, 150] loss: 0.009

[10, 200] loss: 0.008

[10, 250] loss: 0.011

[10, 300] loss: 0.009

Training Completed

Test Accuracy of the model on the 10000 test images: 99 %

Training Epoch 11

[11, 50] loss: 0.008

[11, 100] loss: 0.004

[11, 150] loss: 0.006

[11, 200] loss: 0.008

[11, 250] loss: 0.009

[11, 300] loss: 0.010

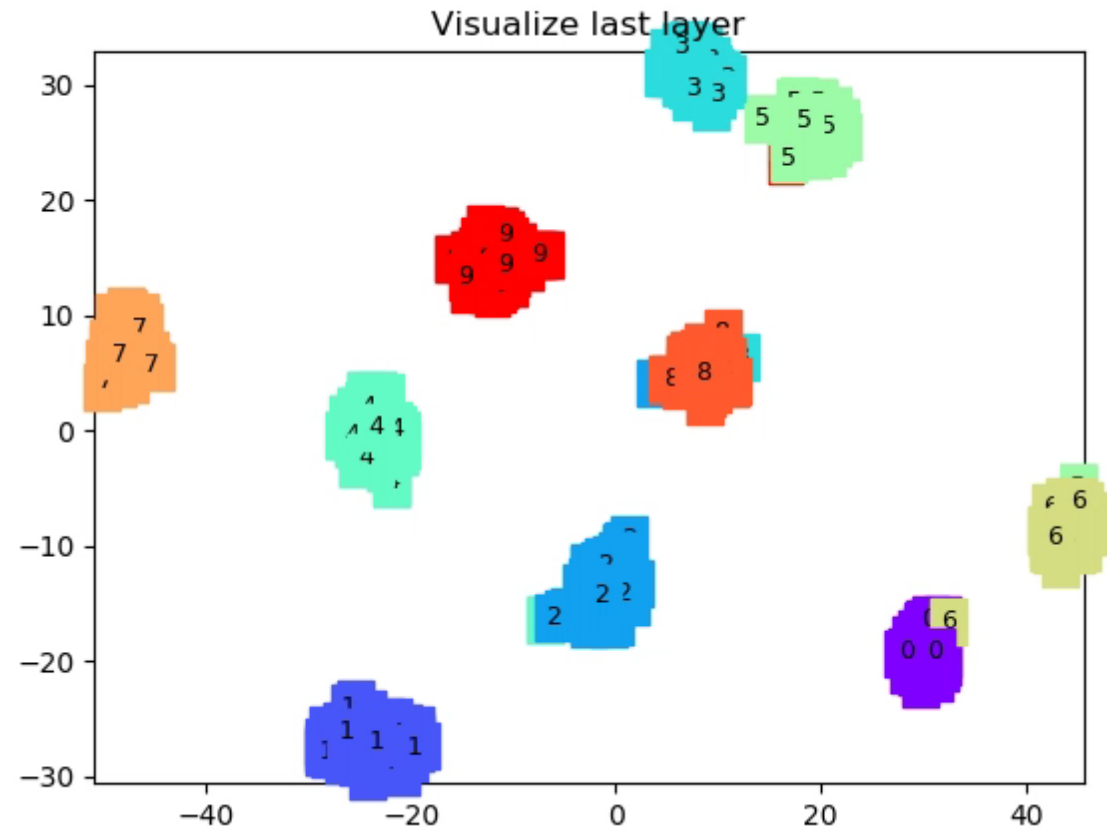
Training Completed

**Average Accuracy of the model on the 10000 test images is 99%**

```
Accuracy for number 0 is 0.9948979591836735
Accuracy for number 1 is 0.9982378854625551
Accuracy for number 2 is 0.9932170542635659
Accuracy for number 3 is 0.9920792079207921
Accuracy for number 4 is 0.9938900203665988
Accuracy for number 5 is 0.9932735426008968
Accuracy for number 6 is 0.9906054279749478
Accuracy for number 7 is 0.9873540856031129
Accuracy for number 8 is 0.9917864476386037
Accuracy for number 9 is 0.9861248761149654
```



# VISUALIZATION



- Almost every digit goes to their own group

# CONCLUSION

- We got accuracy of MLP network is rated as 98% and of the convolution network at 99%.
- This shows that both networks are highly accurate in reading the hand written digit and with further tweaking they can perform at almost 100% accuracy.
- So according to our testing both network are suitable for this project of reading hand written digits



**End....**

---

**Thank You**