

1. Show your model architecture and testing accuracy.

2-layer NN, 500 hidden units, loss function use CrossEntropy.

Weight initial all random small value use normal distribution.

Bias initial all zero.

1 layer: activation function use ReLU

2 layer: activation function use Softmax

```
epoch: 0, train_loss: 0.251503, validation_loss: 0.260554, val_acc: 92.6889
epoch: 1, train_loss: 0.180763, validation_loss: 0.196883, val_acc: 94.5056
epoch: 2, train_loss: 0.140018, validation_loss: 0.161661, val_acc: 95.4778
epoch: 3, train_loss: 0.113483, validation_loss: 0.139584, val_acc: 96.0556
epoch: 4, train_loss: 0.094883, validation_loss: 0.124898, val_acc: 96.4444
epoch: 5, train_loss: 0.080989, validation_loss: 0.114401, val_acc: 96.6833
epoch: 6, train_loss: 0.070271, validation_loss: 0.106960, val_acc: 96.8889
epoch: 7, train_loss: 0.061667, validation_loss: 0.101497, val_acc: 96.9944
epoch: 8, train_loss: 0.054579, validation_loss: 0.097283, val_acc: 97.0833
epoch: 9, train_loss: 0.048798, validation_loss: 0.094185, val_acc: 97.1944
epoch: 10, train_loss: 0.043958, validation_loss: 0.091835, val_acc: 97.2278
epoch: 11, train_loss: 0.039766, validation_loss: 0.089931, val_acc: 97.2722
epoch: 12, train_loss: 0.036047, validation_loss: 0.088304, val_acc: 97.3389
epoch: 13, train_loss: 0.032726, validation_loss: 0.086993, val_acc: 97.3500
epoch: 14, train_loss: 0.029842, validation_loss: 0.085940, val_acc: 97.3722
test acc: 97.6300%
```

Testing accuracy: 97.63%

2. How do you implement feed forward and backward

propagation? A brief explanation is fine.

Forward:

each layer output is $F(W * X + B)$.

-- W is weight matrix.

-- B is bias matrix.

-- F is activation function.

-- X is input dataset

Backward:

From end to begin layer. Calculate derivative.

Main idea is $\frac{dL}{dw} = \frac{dL}{da} \times \frac{da}{dz} \times \frac{dz}{dw}$

Output layer $\frac{dL}{da} \times \frac{da}{dz}$ = model output – target. because cross-Entropy +

Softmax .

$\frac{dz}{dw}$ = hidden layer output.

So, output layer $\frac{dL}{dw} = (\text{model output} - \text{target}) * \text{hidden layer output}$

Hidden layer, use chain rule.

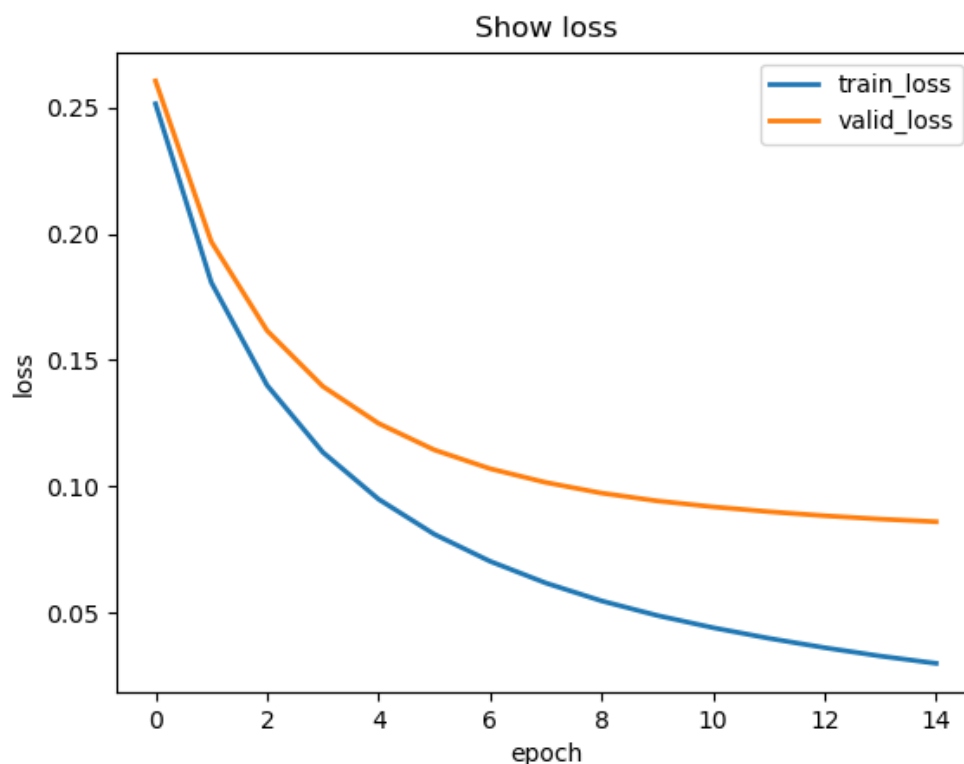
$$\frac{dL}{da} = \sum \text{weight} * (\text{output layer's } \frac{dL}{da} \times \frac{da}{dz})$$

$$\frac{da}{dz} = \text{ReLU derivative.}$$

$$\frac{dz}{dw} = \text{input dataset.}$$

$$\text{So, hidden layer } \frac{dL}{dw} = \frac{dL}{da} * \text{input dataset}$$

3. Plot training loss and validation loss. (loss vs. epochs figure)



4. If we use a very deep NN with a large number of neurons, will the accuracy increase? Why or why not?

May not increase accuracy only use a deep NN. Because it is same if we use a lots of hidden units. Need other additional strategies to increase

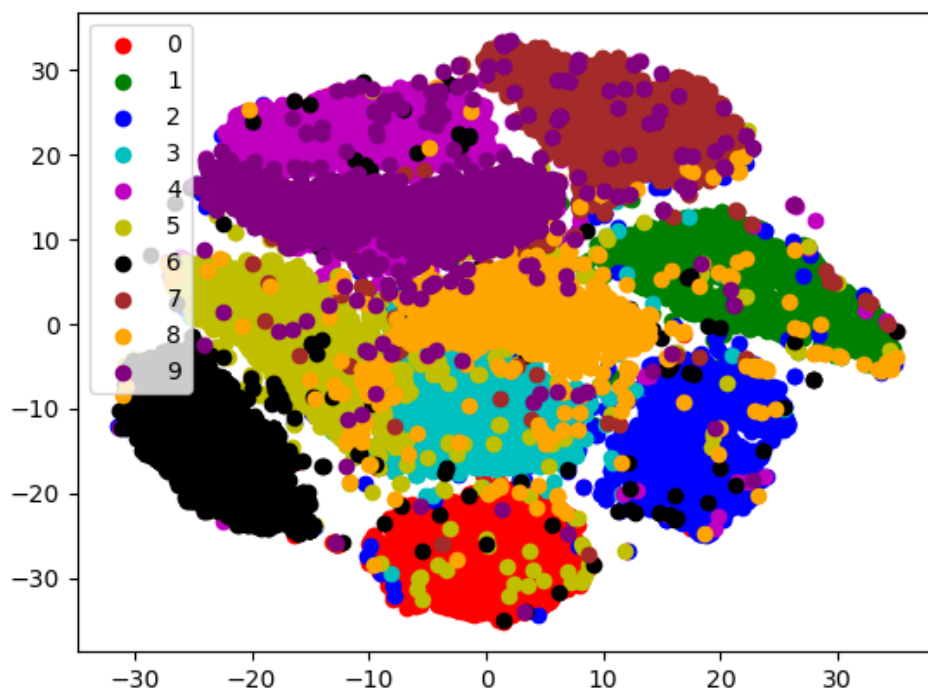
accuracy.

5. Why do we need to validate our model?

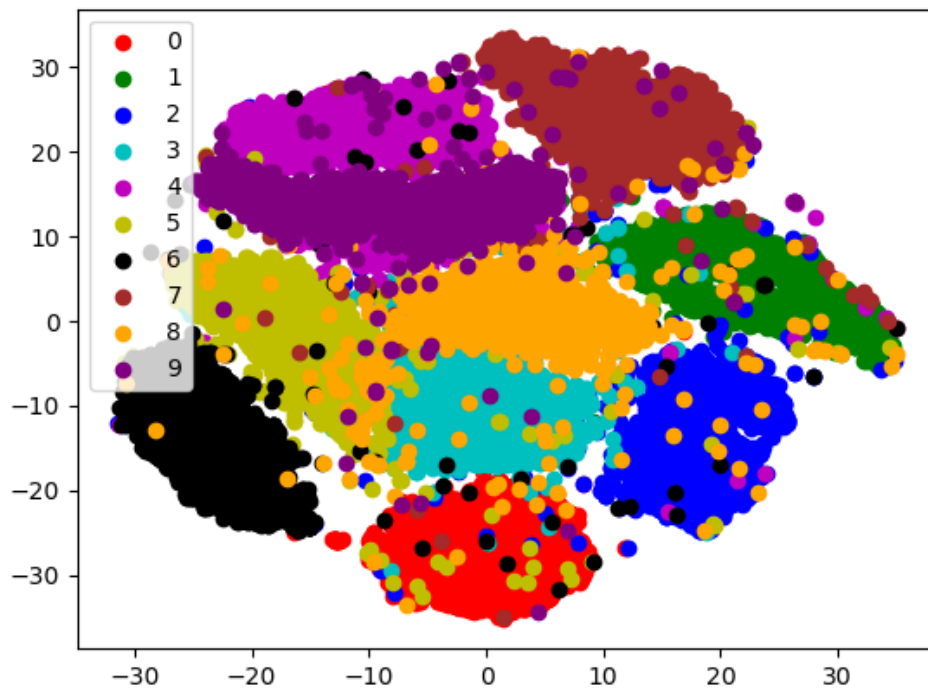
Because it is a method to make sure our model is not training bad, and use for early stopping method to avoid overfitting.

6. t-SNE results (optional, not included in 20%)

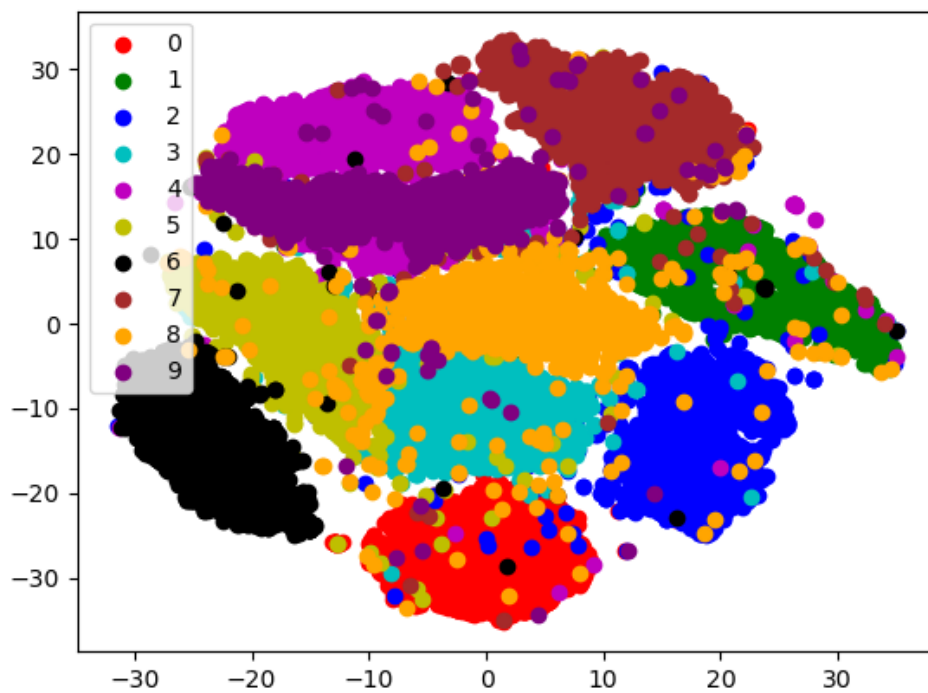
0 epoch validation data visualization



2 epoch validation data visualization



14 epoch validation data visualization



Test data

