

**CAP 5415 Fall 2023**

**Programming Assignment 3**

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## **Objective**

1. To implement an autoencoder and a convolutional autoencoder and compare the results along with the number of model parameters
2. To implement k-nearest neighbor algorithms with different numbers of neighbors and compare the accuracies

## **Autoencoder and Convolutional Autoencoder**

### **Code output**

The output of the python file 'autoencoder.py' is given below

running on: cpu

Autoencoder

number of autoencoder params: 468368

epoch: 1/20, train\_loss: 0.0469

epoch: 2/20, train\_loss: 0.0210

epoch: 3/20, train\_loss: 0.0167

epoch: 4/20, train\_loss: 0.0149

epoch: 5/20, train\_loss: 0.0139

epoch: 6/20, train\_loss: 0.0132

epoch: 7/20, train\_loss: 0.0125

epoch: 8/20, train\_loss: 0.0119

epoch: 9/20, train\_loss: 0.0116

epoch: 10/20, train\_loss: 0.0113

epoch: 11/20, train\_loss: 0.0111

epoch: 12/20, train\_loss: 0.0109

epoch: 13/20, train\_loss: 0.0107

epoch: 14/20, train\_loss: 0.0105

epoch: 15/20, train\_loss: 0.0104

epoch: 16/20, train\_loss: 0.0103

epoch: 17/20, train\_loss: 0.0102

epoch: 18/20, train\_loss: 0.0101

epoch: 19/20, train\_loss: 0.0100

epoch: 20/20, train\_loss: 0.0099

### CNN Autoencoder

number of params of CNN autoencoder: 2349

epoch: 1/20, train\_loss: 0.0891

epoch: 2/20, train\_loss: 0.0339

epoch: 3/20, train\_loss: 0.0228

epoch: 4/20, train\_loss: 0.0175

epoch: 5/20, train\_loss: 0.0145

epoch: 6/20, train\_loss: 0.0127

epoch: 7/20, train\_loss: 0.0115

epoch: 8/20, train\_loss: 0.0107

epoch: 9/20, train\_loss: 0.0100

epoch: 10/20, train\_loss: 0.0096

epoch: 11/20, train\_loss: 0.0092

epoch: 12/20, train\_loss: 0.0089

epoch: 13/20, train\_loss: 0.0086

epoch: 14/20, train\_loss: 0.0084

epoch: 15/20, train\_loss: 0.0082

epoch: 16/20, train\_loss: 0.0081

epoch: 17/20, train\_loss: 0.0079

epoch: 18/20, train\_loss: 0.0078

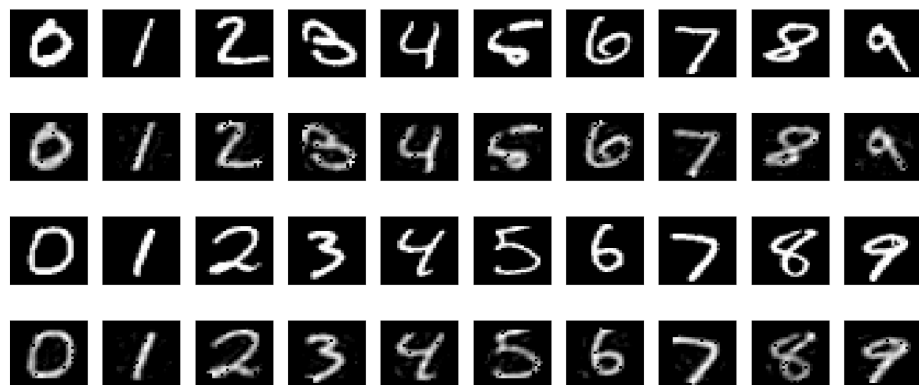
epoch: 19/20, train\_loss: 0.0077

epoch: 20/20, train\_loss: 0.0076

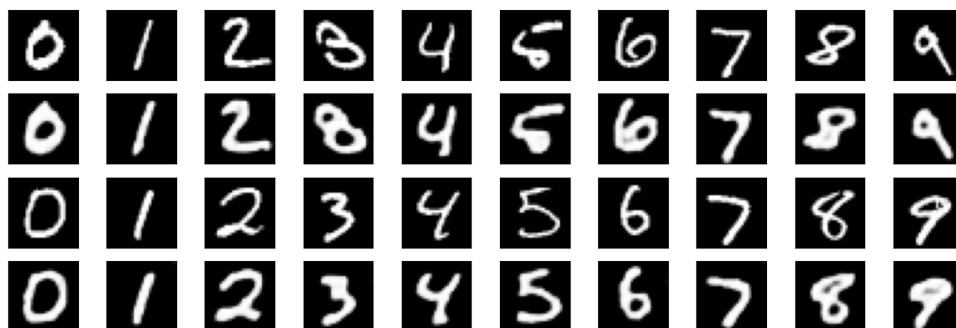
So the training loss goes down in both cases as training goes on. Autoencoder has 468k+ parameters and the CNN one has only 2349 parameters, and still the CNN autoencoder has lower training loss. This shows that in terms of our used evaluation metric MSE, CNN performs better

### Representation and Reconstruction

This shows Autoencoder's performance on two sets of original and reconstructed images for each digit.



This shows the CNN autoencoder's performance on two sets of original and reconstructed images for each digit.



In both cases, rows 1 and 3 are the original images and rows 2 and 4 show the reconstructed ones. From these, we can say that the autoencoder reconstructions contain grains similar to salt and pepper noise. On the other hand, although the CNN version has smoother images, it doesn't have such grain.

### K-nearest neighbors

***(I know the question said to test with `n_neighbors` with 3 5 and 7 but I checked with other values as well to understand the model behaviour better so I am showing those results here)***

#### **Code Output**

KNN result for `n_neighbours = 2`  
Train accuracy: 0.9923195084485407  
Test accuracy: 0.9818181818181818

KNN result for `n_neighbours = 3`  
train accuracy: 99.23195084485407  
Test accuracy: 98.38383838383838

KNN result for `n_neighbours = 4`  
train accuracy: 99.15514592933948  
Test accuracy: 98.18181818181819

KNN result for `n_neighbours = 5`  
train accuracy: 98.84792626728111

Test accuracy: 98.18181818181819

KNN result for n\_neighbours = 6  
train accuracy: 98.77112135176651  
Test accuracy: 97.97979797979798

KNN result for n\_neighbours = 7  
train accuracy: 98.84792626728111  
Test accuracy: 98.18181818181819

KNN result for n\_neighbours = 8  
train accuracy: 98.61751152073732  
Test accuracy: 98.18181818181819

KNN result for n\_neighbours = 9  
train accuracy: 98.54070660522274  
Test accuracy: 97.97979797979798

KNN result for n\_neighbours = 10  
train accuracy: 98.46390168970814  
Test accuracy: 97.77777777777777

KNN result for n\_neighbours = 11  
train accuracy: 98.38709677419355  
Test accuracy: 98.18181818181819

KNN result for n\_neighbours = 12  
train accuracy: 98.31029185867895  
Test accuracy: 97.77777777777777

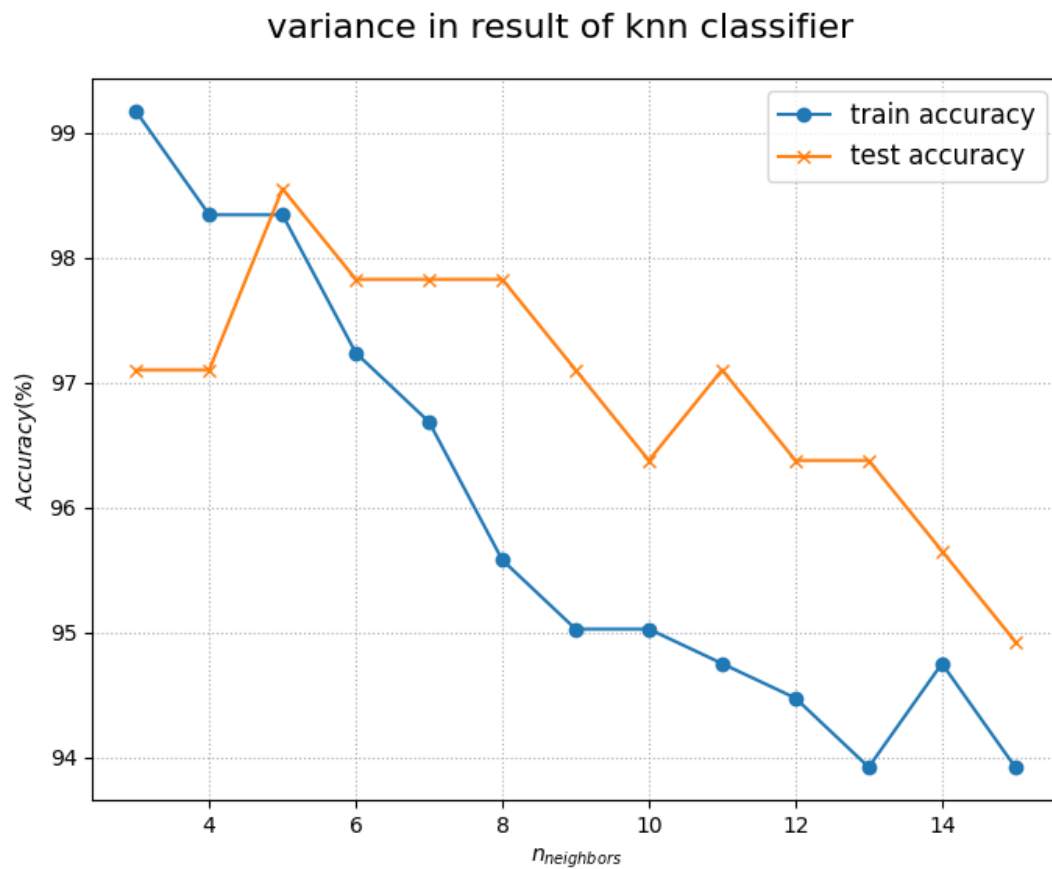
KNN result for n\_neighbours = 13  
train accuracy: 98.15668202764977  
Test accuracy: 98.18181818181819

KNN result for n\_neighbours = 14  
train accuracy: 98.15668202764977  
Test accuracy: 97.77777777777777

KNN result for n\_neighbours = 15  
train accuracy: 97.6958525345622  
Test accuracy: 98.38383838383838

So the best training accuracy is achieved with 2 neighbors only and the best is achieved with 5 neighbors. Going beyond that, the performance always gets worse.

## Accuracy vs number of neighbors plot



So the number of neighbors affects both the training and the testing performance.