

## Python file “Assignment\_5\_student\_id” information.

### python functions:

Input\_read. Read data and preprocess data.

Noise\_addition. Adding noise to the data

Plot1. Plots the learning curves

Plot2. Plots the original image, noisy image and reconstructed image

Plot 3. Reconstructs images by varying the value of the latent variable.

Architecture\_1\_Variation\_AE. first VAE model discussed in question 2-3

Architecture\_2\_Variation\_AE. Second VAE model discussed in question 2-3

Architecture\_2\_Variation\_AE. third VAE model discussed in question 2-3

Architecture\_1\_VAE. Its an autoencoder model, coded just for cross checking the models.

### Q1. Preprocessing dataset.

As mentioned in the question we have to use balance dataset of emnist. So, I used that data in this whole report. The following step by step process is carried out in “input\_data” function of “data\_prepro.py” file.

- First to download the dataset, the emnist library was installed using “pip install emnist” in command window.
- I imported the emnist “balanced” training and test samples.
- The training samples are 112800, while the test samples are 18800. All the images have shape (28, 28)
- After analyzing the dataset, I found that that numerical data is labelled from 0-9. The capital alphabets are labelled from 10 to 35 and onward from 35, the labels are given to small alphabets and special characters.
- As we have to use capital alphabets, so I extracted capital alphabets are those data sets.
- The training samples of alphabets only balance dataset is now 62400 while that of testing set are 10400.
- After obtaining our data, then I analyzed the image pixels. I found that pixel value ranges from 0-255.
- So, the final thing which I did was to normalizing each data by dividing it by 255, and changed the datatype to float values.

The data\_prepro.py also contain the following two function.

noise\_addition: It generate noisy copies of our dataset. In that function we only add gaussian noise to image and the clip it between 0 and 1.

**Note:** Question 2 and question 3 are answered simultaneously here.

**Q2. and Q3.** The VAE Model and its learning curve plots.

I trained different VAE architectures. Three of them are below.

- Architecture 1:

The architecture is shown in fig.1. This encoder has two convolutional and two maxpooling layers in the encoder. The latent dimension were 2 and the batch\_size was 100. The decoder has only dense layers.

The Train loss, validation loss is shown in figure 2. While train accuracy and validation accuracy are shown in figure 3.

Encoder				Decoder		
Layer (type)	Output Shape	Param #	Connected to	Layer (type)	Output Shape	Param #
input_19 (InputLayer)	(None, 28, 28, 1)	0		sampling (InputLayer)	(None, 2)	0
conv2d_34 (Conv2D)	(None, 28, 28, 16)	160	input_19[0][0]	decoding (Dense)	(None, 512)	1536
max_pooling2d_33 (MaxPooling2D)	(None, 14, 14, 16)	0	conv2d_34[0][0]	decoding1 (Dense)	(None, 256)	131328
conv2d_35 (Conv2D)	(None, 14, 14, 8)	1160	max_pooling2d_33[0][0]	flat_decoded (Dense)	(None, 784)	201488
max_pooling2d_34 (MaxPooling2D)	(None, 7, 7, 8)	0	conv2d_35[0][0]	out_decoded (Reshape)	(None, 28, 28, 1)	0
flatten (Flatten)	(None, 392)	0	max_pooling2d_34[0][0]			
mean (Dense)	(None, 2)	786	flatten[0][0]			
var (Dense)	(None, 2)	786	flatten[0][0]			
sampling (Lambda)	(None, 2)	0	mean[0][0] var[0][0]			
Total params: 2,892 Trainable params: 2,892 Non-trainable params: 0				Total params: 334,352 Trainable params: 334,352 Non-trainable params: 0		

VAE		
Layer (type)	Output Shape	Param #
input_19 (InputLayer)	(None, 28, 28, 1)	0
encoder (Model)	[ (None, 2), (None, 2) ]	2892
decoder (Model)	(None, 28, 28, 1)	334352
Total params: 337,244 Trainable params: 337,244 Non-trainable params: 0		

Figure 1. VAE Architecture 1

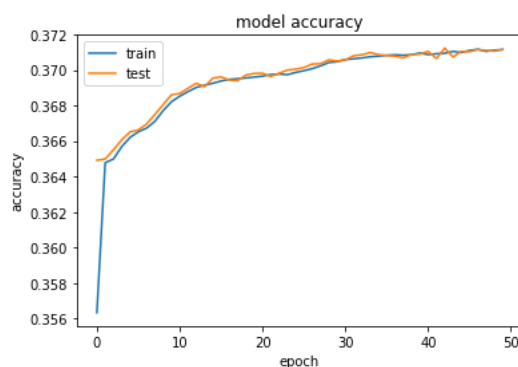


Figure 2. Architecture 1 accuracy.

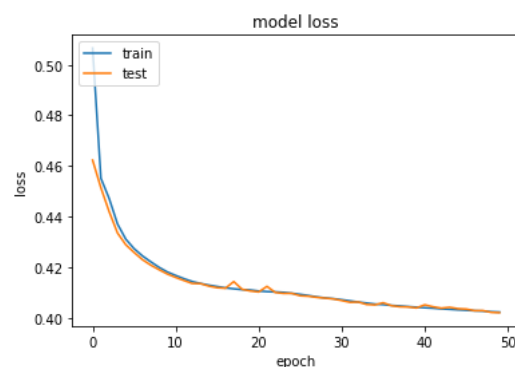


Figure 3. Architecture 1 loss

- Architecture 2:

The architecture is shown in fig.4. This encoder has only one dense layer. The decoder also has a single dense layer. The latent dimension were 2 and the batch\_size was 100.

The Train loss, validation loss is shown in figure 5. While train accuracy and validation accuracy are shown in figure 6. The loss in model is decreasing but accuracy is highly fluctuating.

Encoder			
Layer (type)	Output Shape	Param #	Connected to
input_encoder (InputLayer)	(None, 28, 28, 1)	0	
flat_incoder (Flatten)	(None, 784)	0	input_encoder[0][0]
encoding_dense (Dense)	(None, 512)	401920	flat_incoder[0][0]
mean (Dense)	(None, 2)	1026	encoding_dense[0][0]
dense_var (Dense)	(None, 2)	1026	encoding_dense[0][0]
sampling (Lambda)	(None, 2)	0	mean[0][0] dense_var[0][0]
Total params: 403,972 Trainable params: 403,972 Non-trainable params: 0			
Decoder			
Layer (type)	Output Shape	Param #	
input_decoder (InputLayer)	(None, 2)	0	
decoding_Dense (Dense)	(None, 512)	1536	
flat_decoder (Dense)	(None, 784)	402192	
output_decoder (Reshape)	(None, 28, 28, 1)	0	
Total params: 403,728 Trainable params: 403,728 Non-trainable params: 0			
VAE			
Layer (type)	Output Shape	Param #	
input_encoder (InputLayer)	(None, 28, 28, 1)	0	
encoder (Model)	[(None, 2), (None, 2)]	403972	
decoder (Model)	(None, 28, 28, 1)	403728	
Total params: 807,700 Trainable params: 807,700 Non-trainable params: 0			

Figure 4. Architecture 2

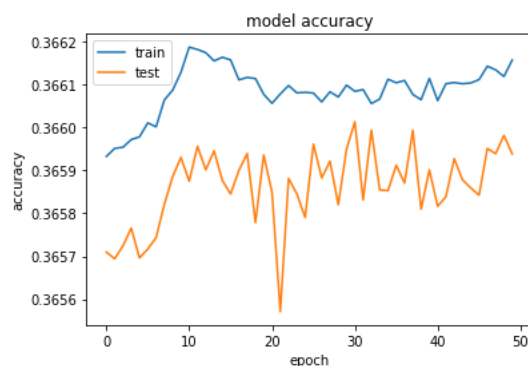


Figure 5. Architecture 2 accuracy

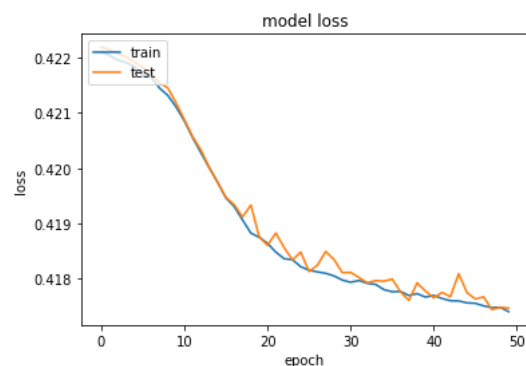


Figure 6. Architecture 2 loss

As I didn't get the required result so I increased the depth of the model and included some convolutional layers in both encoder and decoder. The architecture 3 shows that model. This model gave us some better results.

- Architecture 3:

The architecture is shown in fig.7. This encoder has three convolutional and three maxpooling layers in the encoder. The latent dimension were 2 and the batch\_size was 100. The decoder also has three convolutional layers and also upsampling has been done.

- The Train loss, validation loss is shown in figure 8. While train accuracy and validation accuracy are shown in figure 9.

Encoder				Decoder		
Layer (type)	Output Shape	Param #	Connected to	Layer (type)	Output Shape	Param #
input_encoder (InputLayer)	(None, 28, 28, 1)	0		decoding_input (InputLayer)	(None, 2)	0
encoder_conv1 (Conv2D)	(None, 28, 28, 16)	160	input_encoder[0][0]	decoder_dense1 (Dense)	(None, 128)	384
encoder_maxpool1 (MaxPooling2D)	(None, 14, 14, 16)	0	encoder_conv1[0][0]	decoder_reshape (Reshape)	(None, 4, 4, 8)	0
encoder_conv2 (Conv2D)	(None, 14, 14, 8)	1160	encoder_maxpool1[0][0]	decoder_conv1 (Conv2D)	(None, 4, 4, 8)	584
encoder_maxpool2 (MaxPooling2D)	(None, 7, 7, 8)	0	encoder_conv2[0][0]	decoder_upsampling1 (UpSamp)	(None, 8, 8, 8)	0
encoder_conv3 (Conv2D)	(None, 7, 7, 8)	584	encoder_maxpool2[0][0]	decoder_conv2 (Conv2D)	(None, 8, 8, 8)	584
encoder_maxpool3 (MaxPooling2D)	(None, 4, 4, 8)	0	encoder_conv3[0][0]	decoder_upsampling2 (UpSamp)	(None, 16, 16, 8)	0
flatten_5 (Flatten)	(None, 128)	0	encoder_maxpool3[0][0]	decoder_conv3 (Conv2D)	(None, 14, 14, 16)	1168
mean (Dense)	(None, 2)	258	flatten_5[0][0]	decoder_upsampling3 (UpSamp)	(None, 28, 28, 16)	0
sampling (Dense)	(None, 2)	258	flatten_5[0][0]	decoder_conv4 (Conv2D)	(None, 28, 28, 1)	145
z (Lambda)	(None, 2)	0	mean[0][0] sampling[0][0]			
Total params: 2,420 Trainable params: 2,420 Non-trainable params: 0				Total params: 2,865 Trainable params: 2,865 Non-trainable params: 0		

#### VAE

Layer (type)	Output Shape	Param #
input_encoder (InputLayer)	(None, 28, 28, 1)	0
encoder (Model)	[(None, 2), (None, 2), (N 2420	
decoder (Model)	(None, 28, 28, 1)	2865

Total params: 5,285  
Trainable params: 5,285  
Non-trainable params: 0

Figure 7. Architecture 3

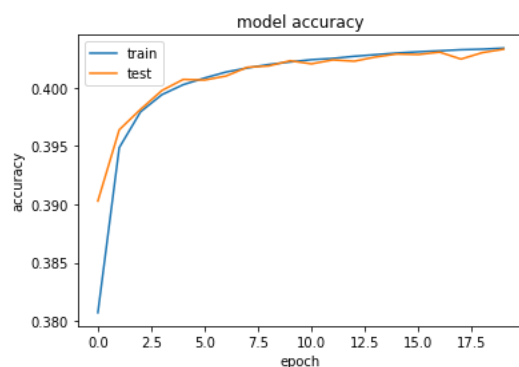


Figure 2. Architecture 3 accuracy

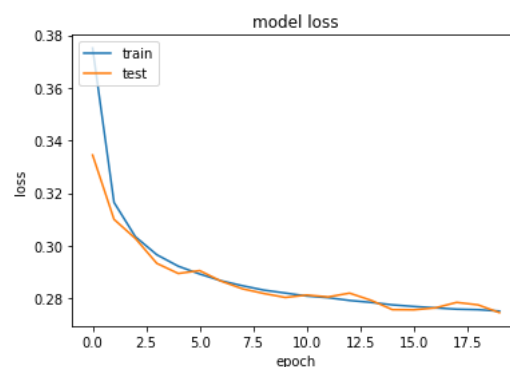
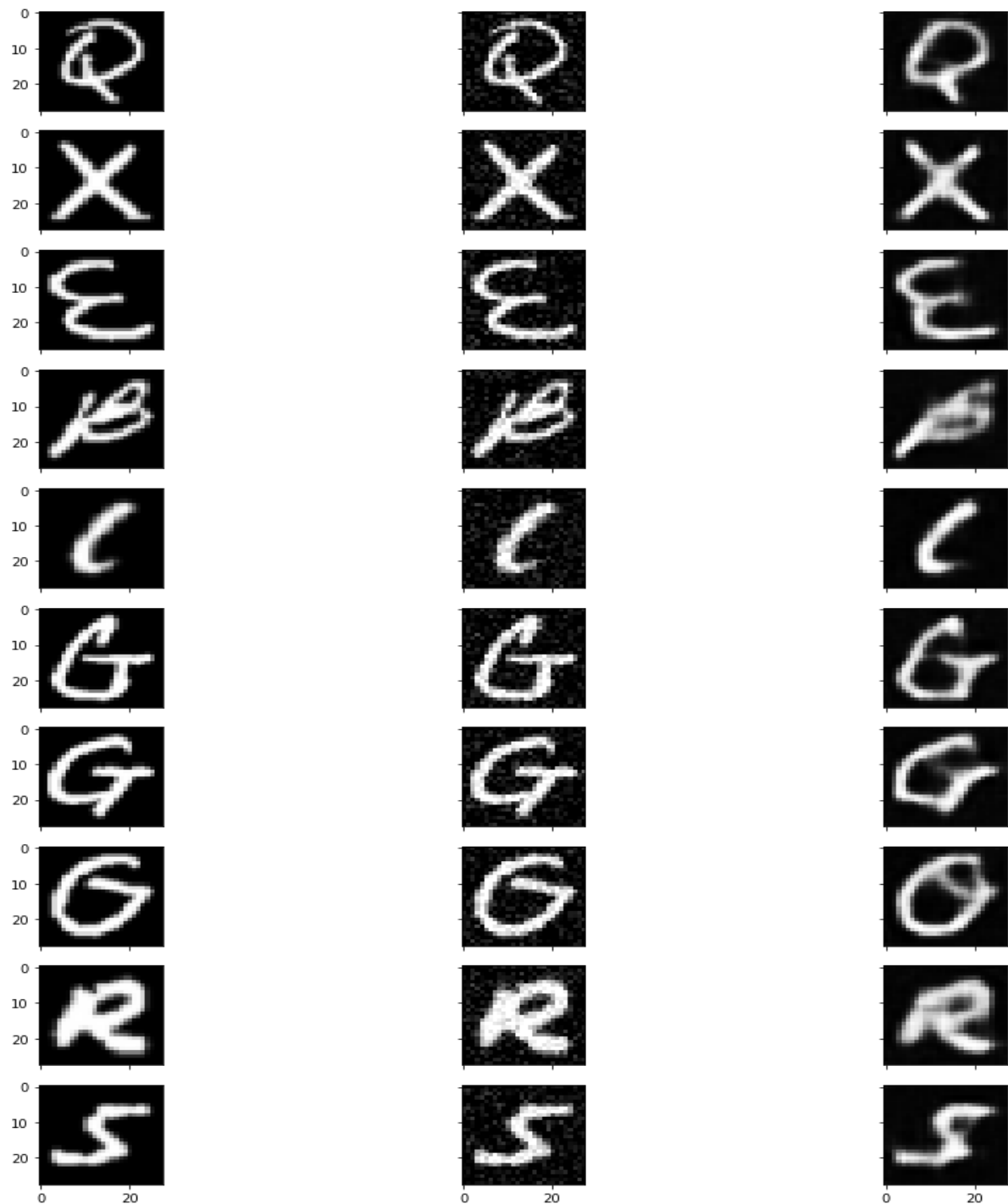


Figure 3. Architecture 3 loss

**Q4.** The column1 shows original images, column 2 shows noised images, column 3 shows regenerated images





Question 5. Figure 1 and figure 2 both are from different architectures, The figure 1 is so far best result with us

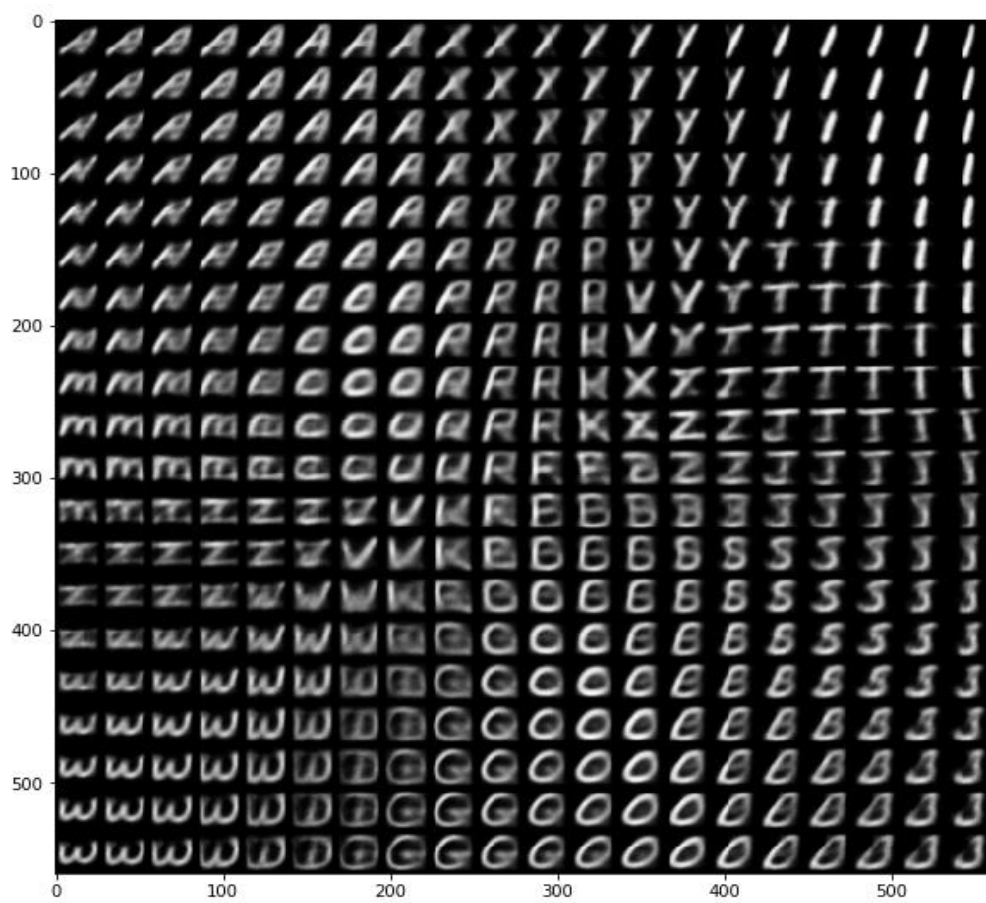


Figure 1

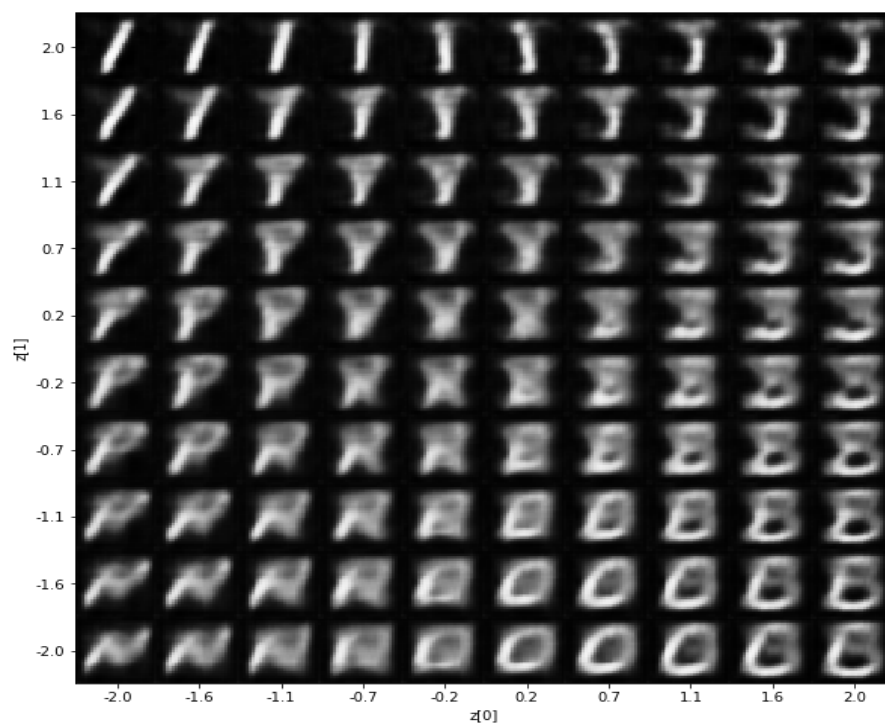


Figure 2

Question 6.

