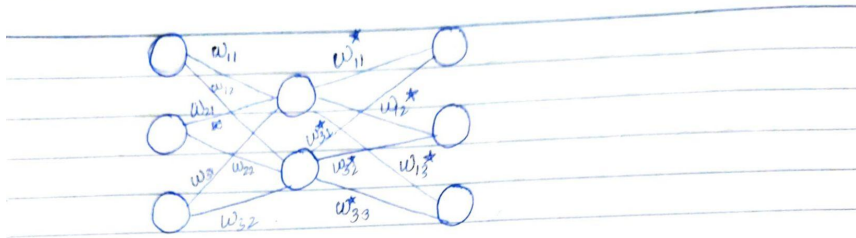


The input layer will have three nodes, the hidden layer will have 2 nodes and the output layer will have 3 nodes (as autoencoder)

I have used matrices to keep track of Weights and biases,
The derivation for backpropagation is done down below:

Ques 2 Part C.



$$W_1 = \begin{bmatrix} w_{11} & w_{12} \\ w_{21} & w_{22} \\ w_{31} & w_{32} \end{bmatrix}$$

Input in hidden layer

$$w_1^T x + b_1$$

$$\begin{bmatrix} w_{11} & w_{21} & w_{31} \\ w_{12} & w_{22} & w_{32} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} + \begin{bmatrix} b_1 \\ b_2 \end{bmatrix}$$

$$I_1 = \begin{bmatrix} w_{11}x_1 + w_{21}x_2 + w_{31}x_3 + b_1 \\ w_{12}x_1 + w_{22}x_2 + w_{32}x_3 + b_2 \end{bmatrix} \left. \begin{array}{l} \text{Input} \\ \text{to hidden} \\ \text{layer} \end{array} \right\}$$

α

$$\frac{\partial E}{\partial W_1} = \frac{\partial L}{\partial O_2} \frac{\partial O_2}{\partial I_2} \frac{\partial I_2}{\partial O_1} \frac{\partial O_1}{\partial I_1} \frac{\partial I_1}{\partial W_1}$$

found in previous step.

$$I_2 = W_2 O_1 + B_2$$

$$\frac{\partial I_2}{\partial O_1} = W_2$$

\rightarrow weight matrix of second layer.

$$\frac{\partial O_1}{\partial I_1} = \sigma'(I_1)$$

$$O_1 = \sigma(I_1)$$

$$\frac{\partial I_1}{\partial W_1} = x$$

\rightarrow input

$$W_1 x + B_1$$

\downarrow

weight matrix of first layer

$$\frac{\partial E}{\partial B_2} = \frac{\partial L}{\partial O_2} \frac{\partial O_2}{\partial I_2} \frac{\partial I_2}{\partial B_2}$$

Bias
b/w
hidden

calculated

\rightarrow

\rightarrow identity matrix of B_2 dimension having all 1.

& output layer.

$$\frac{\partial E}{\partial B_1} = \alpha \times \frac{\partial I_1}{\partial B_1}$$

\rightarrow matrix of B_1 dimension with all 1.

Output of hidden layer

$$O_1 = \sigma(I_1) = \begin{bmatrix} \sigma(W_{11}x_1 + W_{21}x_2 + W_{31}x_3 + b_1) \\ \sigma(W_{12}x_1 + W_{22}x_2 + W_{32}x_3 + b_2) \end{bmatrix}$$

$$\hookrightarrow \begin{bmatrix} O_1 \\ O_2 \end{bmatrix}$$

Input to Output layer $\rightarrow W^*{}^T O_1 + B_2 = I_2$

$$O_2 = \text{Output of Output layer} = \sigma((W^*)^T O_1 + B_2)$$

$\hookrightarrow 3 \times 2 \times 1$

$$W^* = \begin{bmatrix} W_{11}^* & W_{12}^* & W_{13}^* \\ W_{21}^* & W_{22}^* & W_{23}^* \end{bmatrix} \quad B_2 = \begin{bmatrix} b_1^* \\ b_2^* \\ b_3^* \end{bmatrix}$$

Upon backpropagation $W_2 = W^*$

$$E = (\hat{x} - x)^2$$

$$O_2 = \sigma(I_2)$$

$$\frac{\partial E}{\partial W_2} = \frac{\partial E}{\partial O_2} \frac{\partial O_2}{\partial I_2} \frac{\partial I_2}{\partial W_2}$$

$$\hookrightarrow 2(\hat{x} - x)$$

$$\frac{\partial O_2}{\partial I_2} = \sigma'(I_2)$$

$$\frac{\partial I_2}{\partial W_2} = O_1$$

Update Rule

$$W_1\text{-new} = W_1 - \text{learning-rate} \frac{\partial E}{\partial W_1}$$

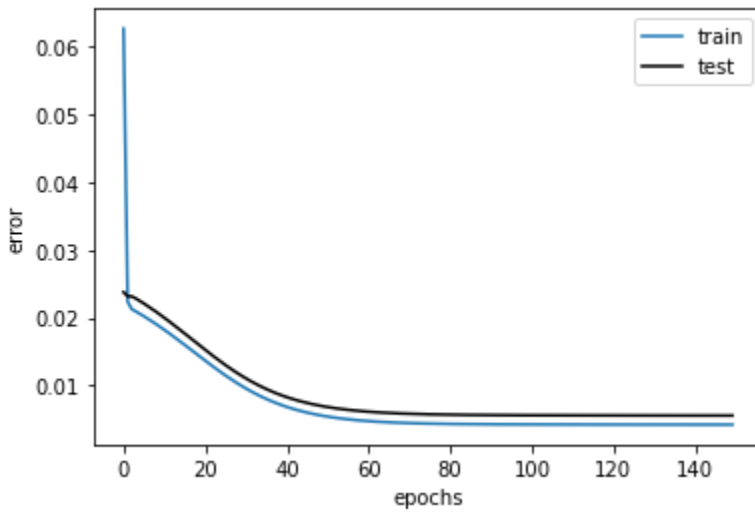
$$W_2\text{-new} = W_2 - \text{learning-rate} \frac{\partial E}{\partial W_2}$$

$$B_1\text{-new} = B_1 - \text{learning-rate} \frac{\partial E}{\partial B_1}$$

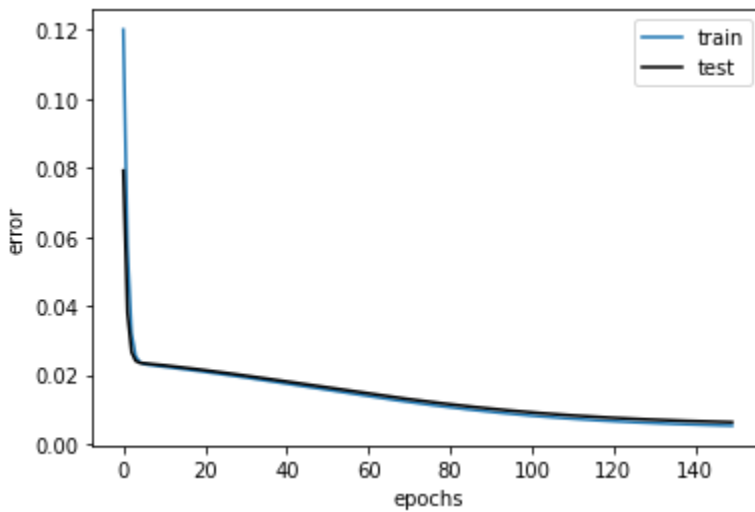
$$B_2\text{-new} = B_2 - \text{learning-rate} \frac{\partial E}{\partial B_2}$$

Question 3)

Backpropagation which is implemented from scratch



For the autograd part:



As can be seen from both graphs that error decreases as epochs increase.