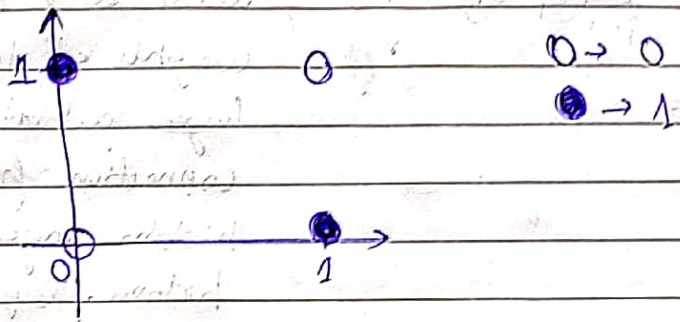


NL Assignment - 3

3. No. We cannot model the XOR truth table using any number of linear activation functions. This is because, XOR data samples are not linearly-separable:-

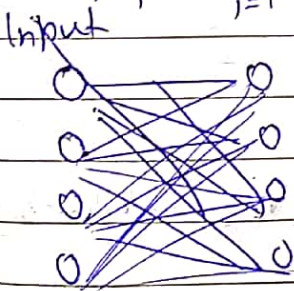


Let us take multiple layers with only linear activation function as the activation function. For the first layer:-

Output = $W^T X + b$, X is the input vector.

For the next layer (~~first hidden layer~~):-

Output = $\sum_{i=1}^n W_i (W_i^T X + b)$



Like this, we will have multiple output from the first hidden layer like CPTO:-

$$O_i = w_i^T x + b_i$$

In The second hidden layer, the output will be

$$O = \sum_{i=1}^n (V_i O_i) + b, \text{ where } v \text{ are the weights of the previous layer connection from the hidden layer 1 to hidden layer 2.}$$

$$O = V_1 (w_1^T x + b_1) + V_2 (w_2^T x + b_2) + \dots + V_n (w_n^T x + b_n) + b$$

$$= (V_1 w_1^T + V_2 w_2^T + \dots + V_n w_n^T) x + (V_1 b_1 + V_2 b_2 + \dots + V_n b_n + b)$$

Here, we can express weights as:

$$(V_1 w_1 + V_2 w_2 + \dots + V_n w_n)$$

And the new bias as: $V_1 b_1 + V_2 b_2 + \dots + V_n b_n + b$.

Hence, We still get a linear classifier for a ~~non-linear~~ data that is not linearly separable.

4. These are the components of the deep CNN:-

i) Convolution layer: This layer works on the input data to extract various features by using kernels. This kernel is chosen in such a way that it extracts the relevant feature (such as a diagonal line) from the input image. Many such kernels can be used to generate various types of input feature matrix, the ~~output~~ size of which depends on the stride.

ii) Activation layer: Generally ~~ReLU~~ ReLU activation function is used because it preserves the value (magnitude) of the feature.

iii) Pooling layer: Generally max (average pooling) are used. This uses a kernel with a ~~def~~ defined size & stride. Now, max-pooling selects the maximum value from the output, hence selecting only relevant outputs & thereby reducing the dimensions of the input. We use pooling for feature.

iv) Fully connected layer: It is like MLP (Multi layer perceptron). In this layer, every node on the left is connected to every node in the right.

Deep CNN is better than ~~feedforward~~ traditional neural network techniques because ~~it is~~ unlike in MLP, the neurons in 2 layers are not fully connected. ~~It is~~ but sparse. Also, the outputs are affected only by a small neighbourhood of the inputs, by the introduction of kernels. ~~by~~

In MLP, fully connected layers may introduce some redundancy. ~~also~~