

# **EXPERIMENT NO – 9**

## **STUDY OF DIFFERENT TYPES OF NEURAL MACHINE TRANSLATION MODEL (NMT)**

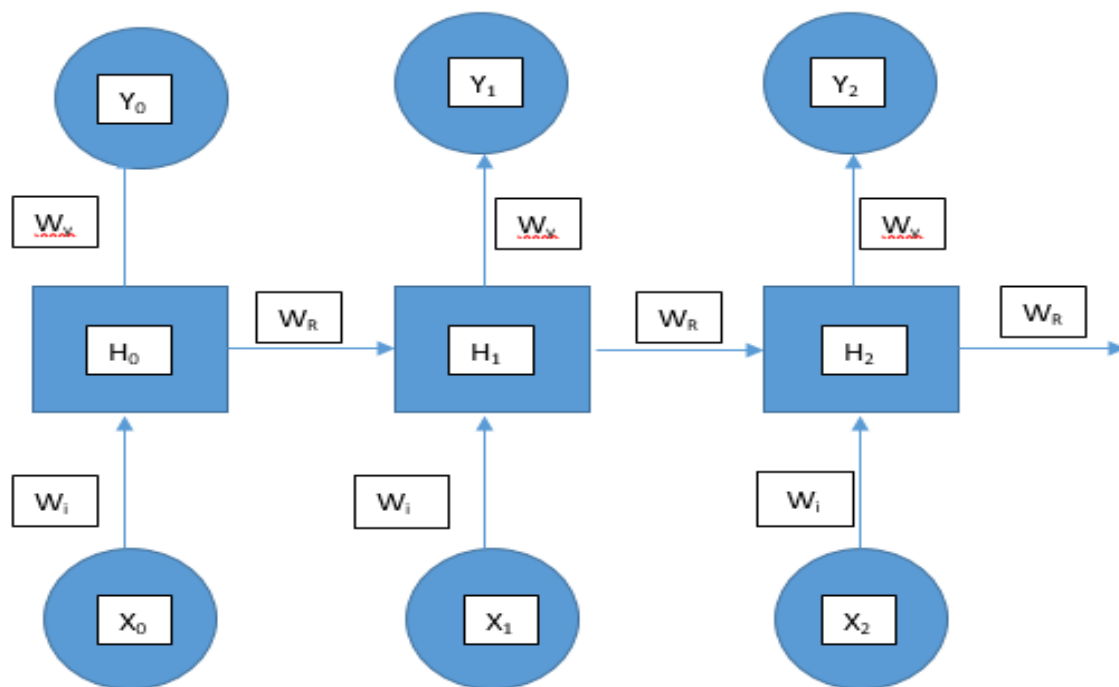
### ❖ **Basic encoder decoder model using Recurrent Neural Network (RNN):**

#### **About the model:**

1. The encoder-decoder model is a way of using recurrent neural networks for sequence-to-sequence prediction problems.
2. It was initially developed for machine translation problems, although it has proven successful at related sequence-to-sequence prediction problems such as text summarization and question answering.
3. The approach involves two recurrent neural networks, one to encode the input sequence, called the encoder, and a second to decode the encoded input sequence into the target sequence called the decoder.

#### **Need of the Recurrent Neural Network (RNN):**

1. The main problem with feed forward neural network is it only focuses on the current value/context. This value is not retained in the next time stamp.
2. Sometimes in the real world situation we have to retain the previous value and based on the previous and current input/context we have to decide the output.
3. For example, in case of sentences what will be the next word, that can be decided with the help of the context of the previous words.
4. Hence in language modelling we require long term temporal dependencies to decide the next word in the sentences.
5. RNN model helps in such type of situation. Long term temporal dependencies cannot be achieved with feed forward neural network.
6. A simple model of recurrent neural network uses tanh or sigmoid as an activation function.
7. When training a deep neural network with gradient based learning and backpropagation, we find the partial derivatives by traversing the network from the final layer to the initial layer.
8. Using the chain rule, layers that are deeper into the network go through continuous matrix multiplications in order to compute their derivatives.
9. Structure of a Recurrent Neural Network (RNN):



### ✚ Disadvantage of Recurrent Neural Network (RNN):

1. In a network of  $n$  hidden layers,  $n$  derivatives will be multiplied together.
2. If the derivatives are large then the gradient will increase exponentially as we propagate down the model until they eventually explode, and this is what we call the problem of **exploding gradient**.
3. Alternatively, if the derivatives are small then the gradient will decrease exponentially as we propagate through the model until it eventually vanishes, and this is the **vanishing gradient** problem.

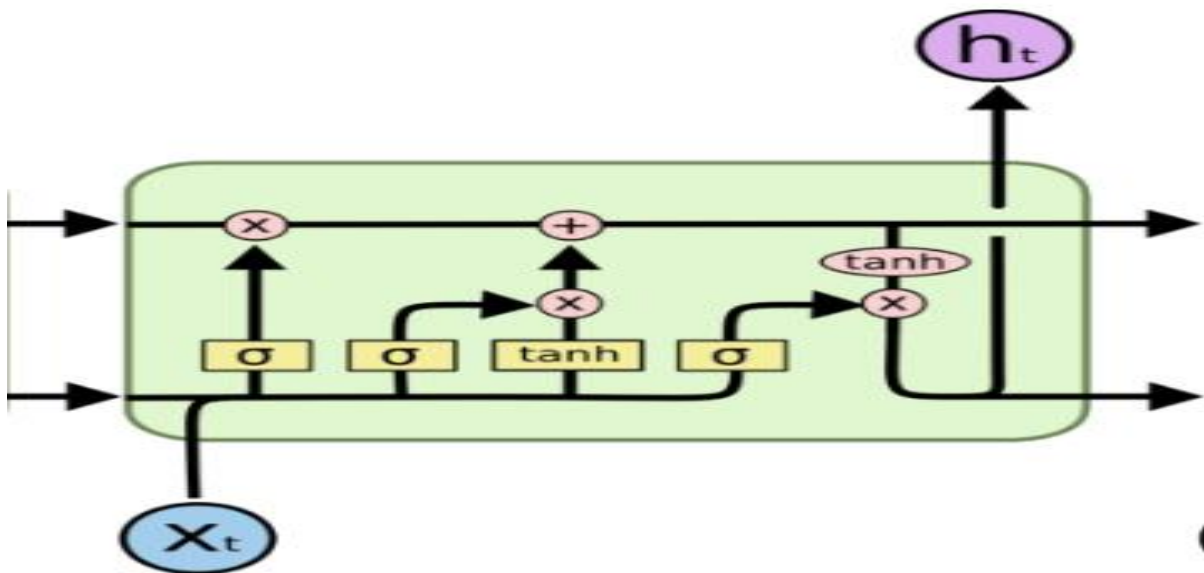
### ❖ Basic encoder decoder model using Long Short-term Memory (LSTM):

#### ✚ About the model:

1. Long Short Term Memory networks – usually just called “LSTMs” – are a special kind of RNN, capable of learning long-term dependencies.
2. LSTMs are explicitly designed to avoid the long-term dependency problem. Remembering information for long periods of time is practically their default behavior
3. All recurrent neural networks have the form of a chain of repeating modules of neural network.
4. LSTMs also have this chain like structure, but the repeating module has a different structure. Instead of having a single neural network layer, there are four, interacting in a very special way.

### Need of the Long Short-term Memory (LSTM):

1. The long term dependency problem can be solved with LSTM models.
2. LSTM cells are operated with the help of real number parameter called gates.
3. The input gate parameter helps to decide how much new input is required to change the memory state.
4. The forget gate parameter helps to decide how much previous value to retain in-memory state.
5. And the output gate parameter controls how strongly the current memory state is to pass into the next layer.
6. Structure of a Recurrent Neural Network (RNN):



### Disadvantage of Long Short-term Memory (LSTM):

1. LSTMs are prone to overfitting and it is difficult to apply the dropout algorithm to curb this issue.
2. They require a lot of resources and time to get trained and become ready for real-world applications.
3. In technical terms, they need high memory-bandwidth because of linear layers present in each cell which the system usually fails to provide for. Thus, hardware-wise, LSTMs become quite inefficient.

## ❖ **Basic encoder decoder model using Attention-based:**

### **About the model:**

1. *This mechanism helps the neural network pay more attention to certain parts of inputs (instead of the entire input) while generating the output.*
2. *Let's take an example. Say we want to translate the Spanish sentence to English. As humans, we don't pay attention to every single word in the input all the time.*
3. *We process it phrase by phrase to come up with the output. This is how we can translate this sentence as humans.*
4. *Basically, we paid more attention to certain words in the source sentence while generating different words in the output.*

### **Need of the Attention-based model:**

1. *Attention mechanisms are being increasingly used to improve the performance of Neural Machine Translation (NMT) by selectively focusing on sub-parts of the sentence during translation.*
2. *Conventional encoder-decoder architectures for machine translation encoded every source sentence into a fixed-length vector, irrespective of its length, from which the decoder would then generate a translation.*
3. *This made it difficult for the neural network to cope with long sentences, essentially resulting in a performance bottleneck.*
4. *These are two simple and effective classes of attentional mechanism: a global approach which always attends to all source words and a local one that only looks at a subset of source words at a time*
5. *In local level attention, we use only a subset of entire sentence and based on that subset the context is figured.*
6. *In global level attention, we use the entire sentence to understand and figure sentence context*

### **Disadvantage of Attention-based model:**

1. *The only disadvantage of the Attention mechanism is that it is a very time consuming and hard to parallelize system.*
2. *The main disadvantage of the attention mechanism is that it adds more weight parameters to the model*
3. *Which can increase training time especially if the input data for the model are long sequences.*

## ❖ Basic encoder decoder model using the Bahdanau Attention Mechanism:

### ✚ About the model:

1. The most important distinguishing feature of this approach from the basic encoder-decoder is that it does not attempt to encode a whole input sentence into a single fixed-length vector.
2. Instead, it encodes the input sentence into a sequence of vectors and chooses a subset of these vectors adaptively while decoding the translation.
3. This improves the translation performance of the basic encoder-decoder model.

### ✚ Need of the Bahdanau Attention Mechanism model:

1. Conventional encoder-decoder architectures for machine translation encoded every source sentence into a fixed-length vector, irrespective of its length, from which the decoder would then generate a translation.
2. This made it difficult for the neural network to cope with long sentences, essentially resulting in a performance bottleneck.
3. The Bahdanau attention was proposed to address the performance bottleneck of conventional encoder-decoder architectures, achieving significant improvements over the conventional approach.
4. Each time the proposed model generates a word in a translation, it (soft-)searches for a set of positions in a source sentence where the most relevant information is concentrated.
5. The model then predicts a target word based on the context vectors associated with these source positions and all the previous generated target words.
6. Structure of the Bahdanau Attention Mechanism model:

