

Long Short-term memory (LSTM)

* LSTM is one of the variant of RNN.

Why LSTM is required ?

Some of the usecases are

- a) Sometimes model needs to remember context and sometimes model needs to remember previous word.
- b) Sometimes sentences was too long and supposed to remember everything.
- c) Sometimes need to remember previous small context not all sentences.

Examples:

- a) How are _____ ? **you** → short term memory
* 'you' can have high probability, because in most of the scenarios we use it.

* Why short term ?

Because 'you' is predicted from sequence 'How are' which is short context. Model didn't predict 'you' from the overall context in this page.

b) I am going to learn _____ **LSTM**
↓
Long term memory

* It can be anything like NLP, DL, ML or LSTM.

* Why LSTM?

Because from the starting, we are learning about LSTM. So, it will have high probability.

* Why Long term?

→ Because 'LSTM' is predicted from long context (Mostly all sentences from start).

→ It didn't predict from short context
i.e. 'I am going to learn'

* So, sometimes we need short context, sometimes we need long context, sometimes we need very long memory, sometimes we need to change the context.

c) Changing context

After learning this, I am going to do

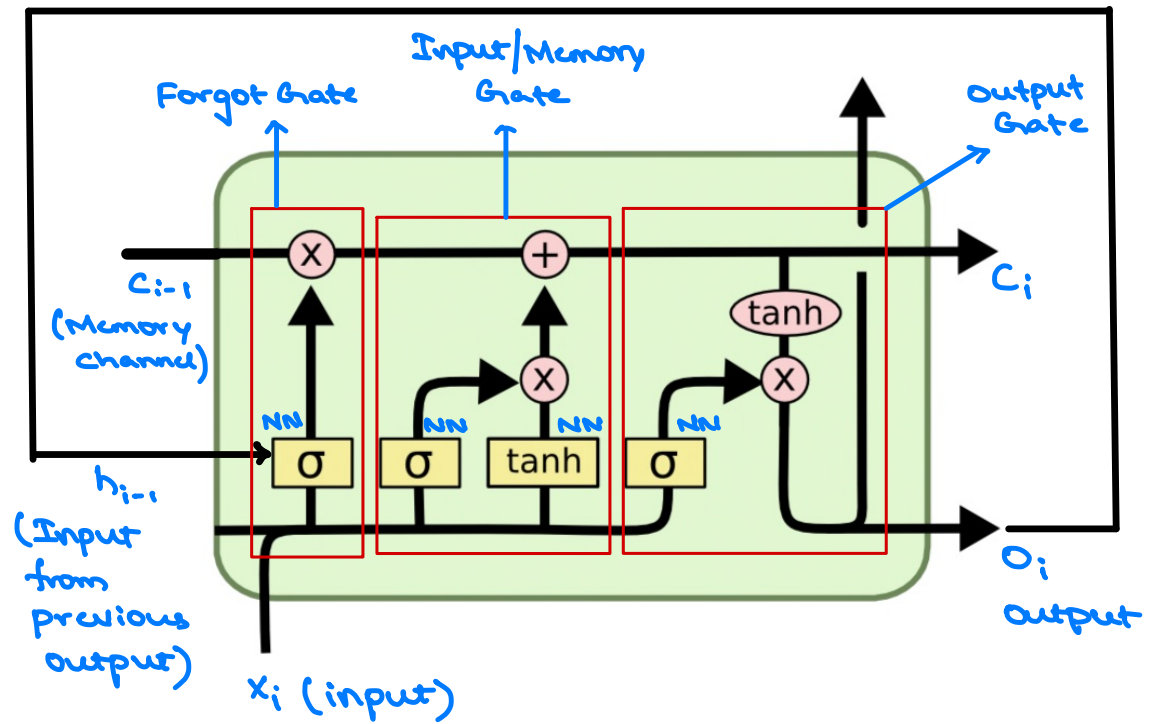
_____ assignment
cooking
rest
project

Here we don't need to remember previous context.

So, we need a model that can satisfy all previous criteria. This is why we need LSTM.

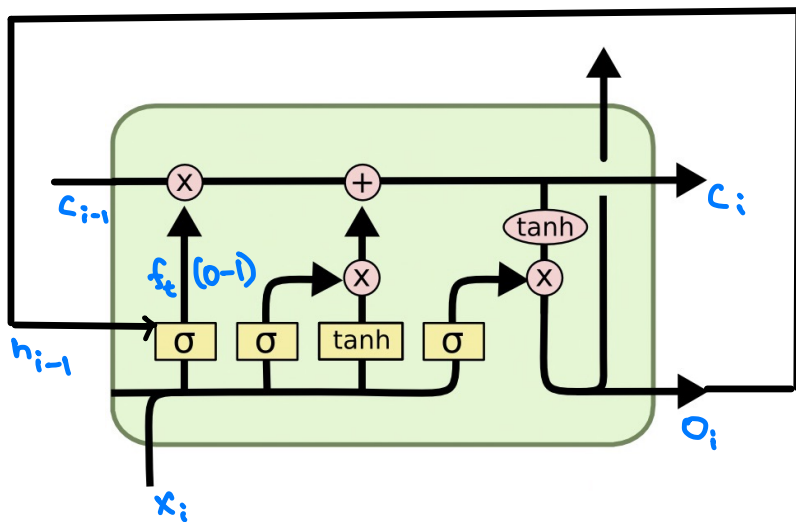
Next →

Architecture



Forget Gate:

- * At $t=0$, let us assume $x_i = H$. Now convert 'H' into vector and send to NN which is using sigmoid activation function.
- * As $t=0$, h_{i-1} is empty. No previous output.
- * This NN always gives output b/w 0 to 1.
 - close to 0 \rightarrow NN learnt that forgot about previous context
 - close to 1 \rightarrow NN learnt that remember previous context

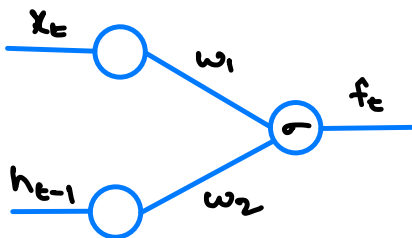


* At $t=0$, Memory channel (c_{i-1}) is 0. So, the output of forget gate at $t=0$, will be zero always because $0 * (0-1) = 0$.

* Basically forget gate tells you, how much I need to forget from the previous context.

Equation

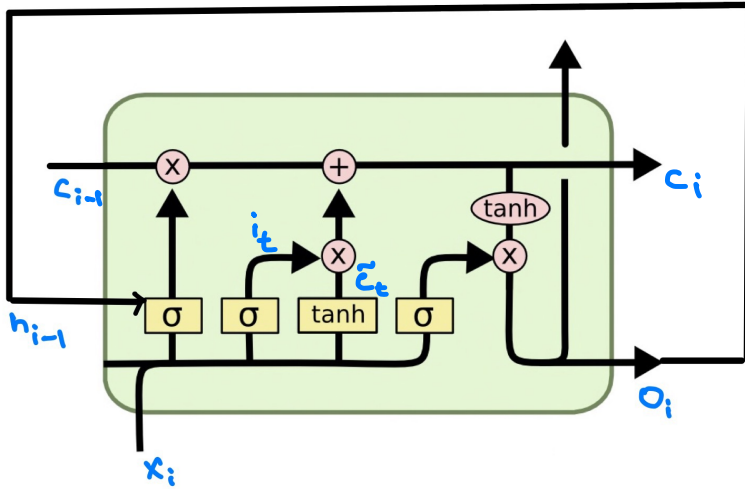
$$f_t = \sigma(W_f \cdot [h_{t-1}, x_t] + b_f)$$



$$w_f = [w_1, w_2]$$

Input / Memory Gate:

- * In forget gate, we learnt how to forget previous word / context.
- * In Input / Memory gate, we learn how to memorize current word.



At $t=0$
 $x_i = 'H'$
 $h_{i-1} = 0$

- * Sigmoid NN \rightarrow Decides whether current input we are supposed to add into memory (or) not. [Important (or) not]
- * tanh NN \rightarrow If I have to add current input to the memory, then how much need to be added is decided by this NN. (Regulates NN)
- * For example, $i_t = 1$, $\tilde{c}_t = 0.09$ then output of memory channel
$$c_{i-1} = (1 * 0.09) + 0 = 0.09 \Rightarrow \text{vector 'H' memory}$$

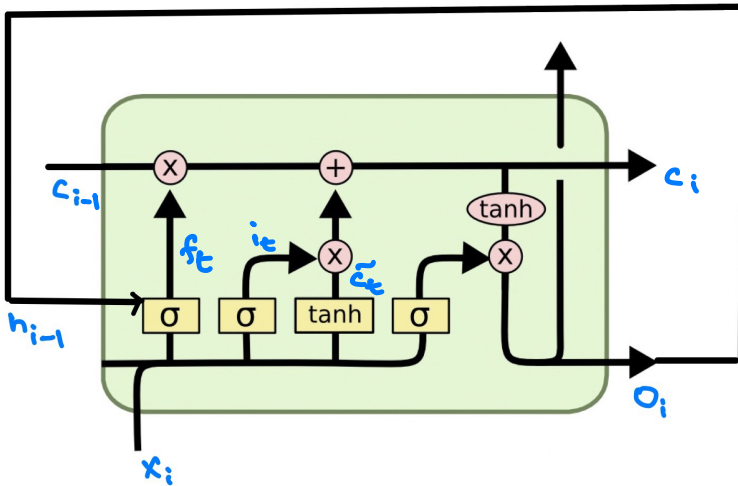
* Input/Memory gate learns new things (or) context and adds the context to the previous one.

Equations

$$i_t = \sigma(W_i \cdot [h_{t-1}, x_t] + b_i)$$

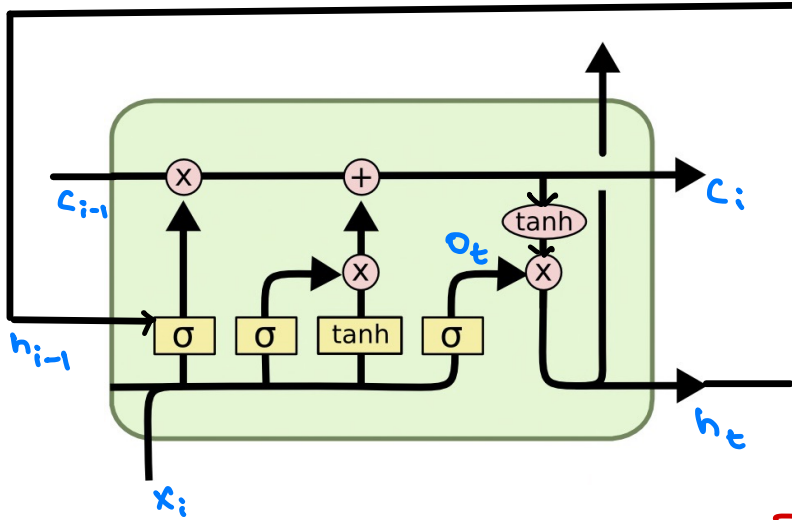
$$\tilde{c}_t = \tanh(W_c \cdot [h_{t-1}, x_t] + b_c)$$

Memory Channel



$$c_t = f_t * c_{t-1} + i_t * \tilde{c}_t$$

Output Gate



Equations:

$$o_t = \sigma(W_o \cdot [h_{t-1}, x_t] + b_o)$$

$$h_t = o_t * \tanh(c_t)$$

Ex:

$$c_t = 0.09$$

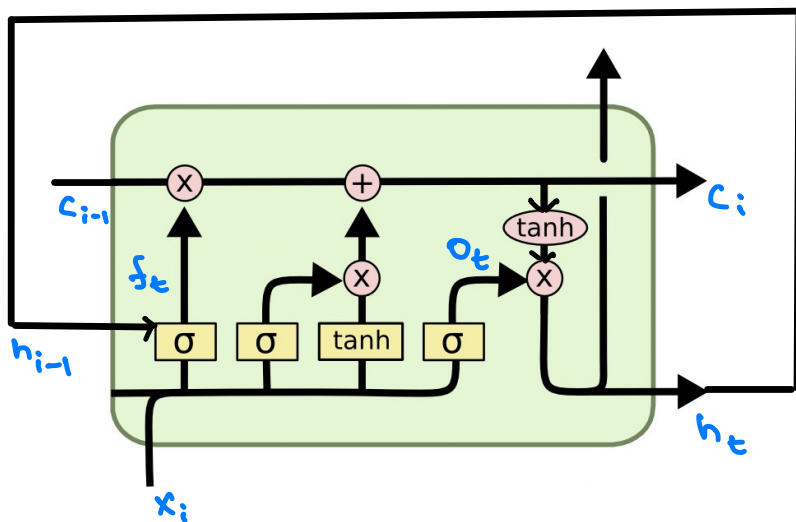
$$\tanh(c_t) = 0.089$$

$$o_t = 1$$

$$h_t = 0.089$$

- * Output gate takes memory channel (Which captures context till the current word) and output of sigmoid NN as input.
- * tanh converts memory channel output into $[-1, 1]$ range.
- * The output of output gate is sent to input of next word to memorize the previous context.

Let us assume word is 'HELLO'



At $t=0$, $x_i = H$, $c_i = 0.09$, $h_t = 0.089$

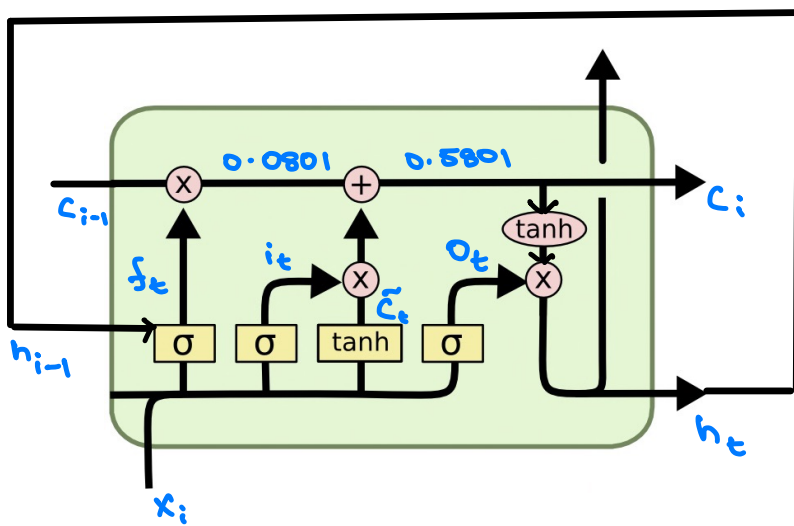
At $t=1$, $x_i = E$

* In the forget gate, sigmoid NN takes $h_t = 'H'$ and $x_i = 'E'$ as input and checks for a pattern. If our data doesn't have 'HE' pattern it gives 0 as output. Similarly if our data contains pattern 'HE', it gives close to '1' as output.

* For example, if the pattern is 'XH' it gives output close to 0, because it's hard to find words with 'XH' pattern.

$$f_t = 0.89, c_{i-1} = 0.09$$

$$\text{output} = 0.89 * 0.09 = 0.0801$$



* In the input / Memory gate, we are learning the pattern 'HE' and adding it to the memory channel.

$$\text{output} = (H + HE) \text{ memory}$$

$$= 0.0801 + (i_t * \tilde{c}_t)$$

$$= 0.0801 + (1 * 0.5)$$

$$= 0.5801$$

* In the output gate

$$h_t = o_t + \tanh(0.5801)$$

$$= 1 + 0.5227 = 1.5227$$

Example

- * Assume we passed entire wikipedia data to the model. Input is 'HEZ'.
- * LSTM tries to remember the previous context until input is letter 'E'.
- * When 'z' is given as input, forget gate gives output close to zero, because words with pattern 'HEZ' are hard to find.
- * So, basically it forgets the previous context when 'z' is given as input.

Key points

- * Role and responsibilities of every sigmoid NN is different. We can't use same sigmoid NN in entire architecture.

RNN Architecture

