

# LSTM

Pg 1

A form of RNN that addresses short term memory problem : Introduced by Hochreiter & Schmidhuber (1997)

Today, due to current COVID-19 situation, I . . . . .

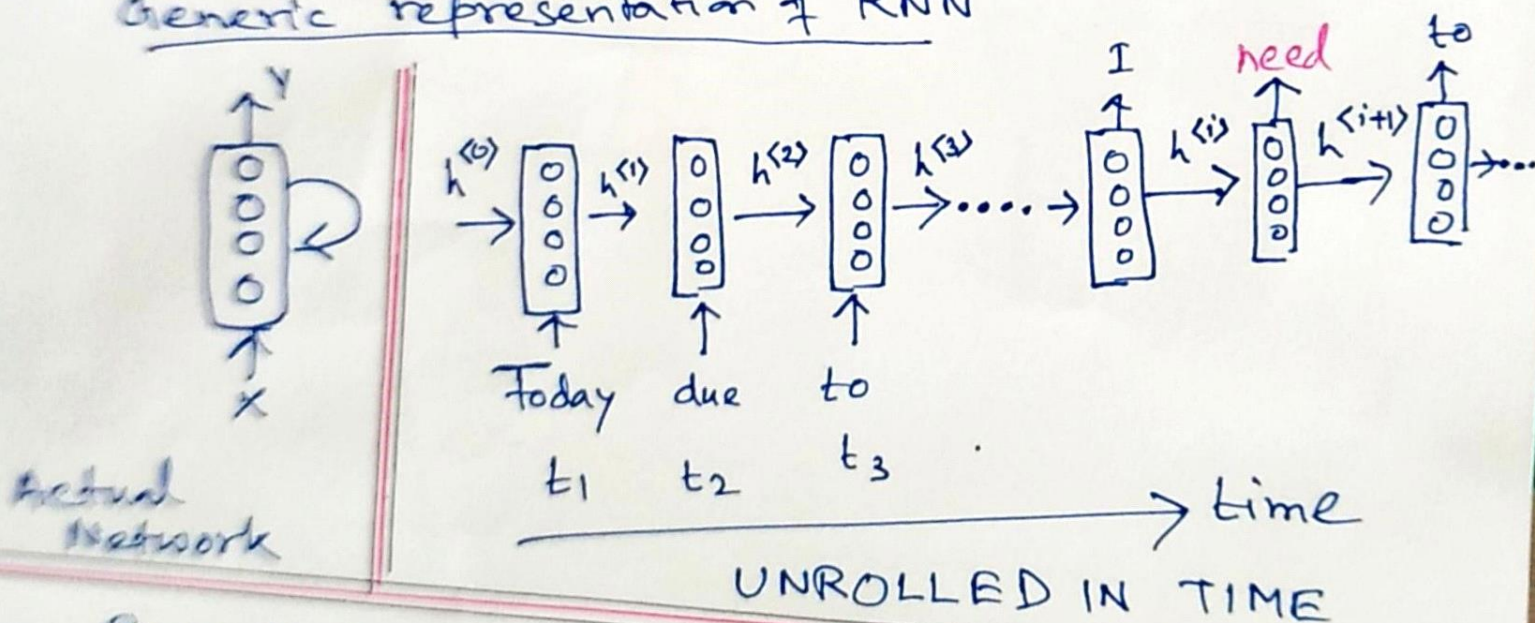
Last year, due to current COVID-19 situation, I . . . . .

Autocompletion will be significantly influenced by the beginning of the sentence.

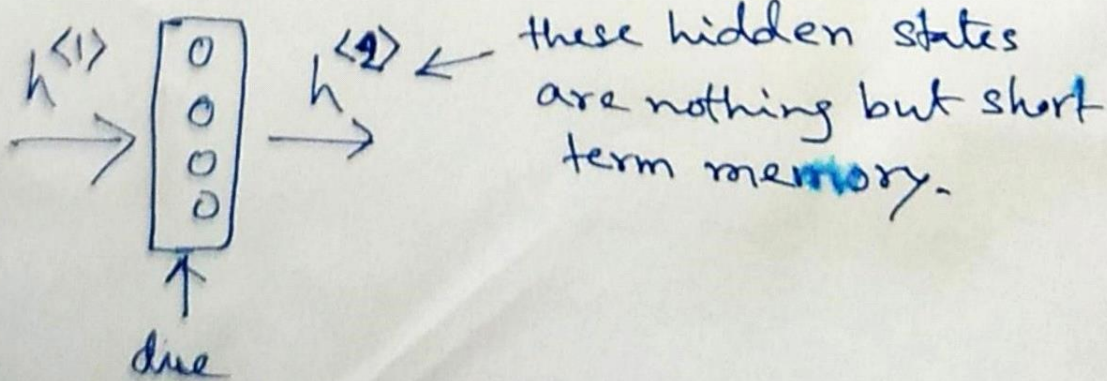
Today, due to current COVID-19 situation, I **need** to take a loan

Last year, due to current COVID-19 situation, I **had** to take a loan

## Generic representation of RNN



Because of vanishing gradients, traditional RNN have very short time memory.

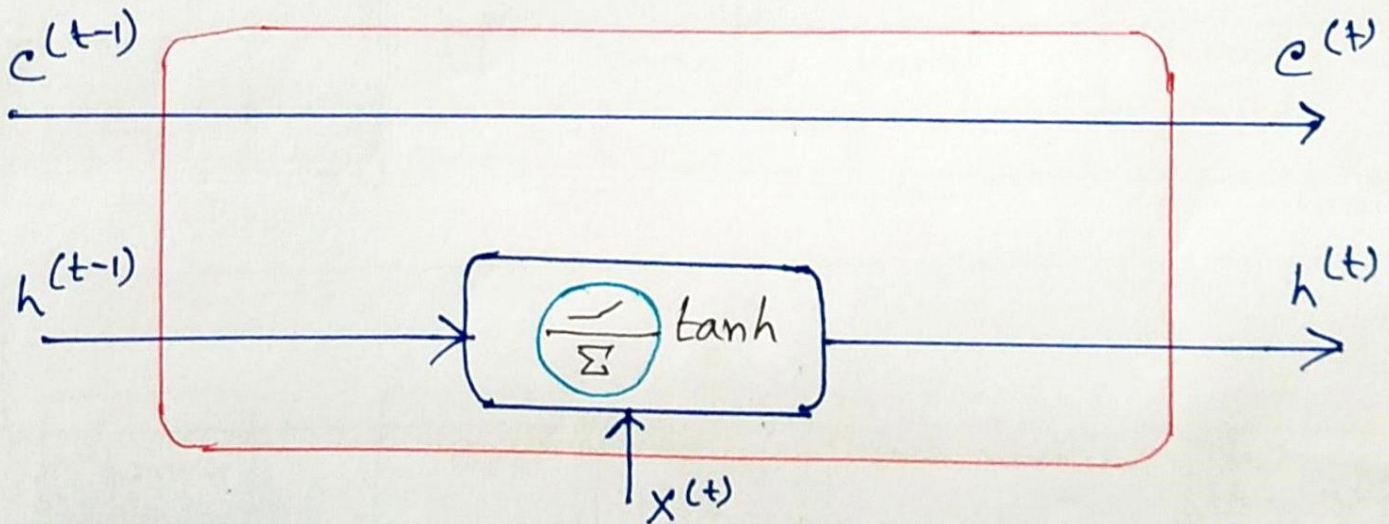
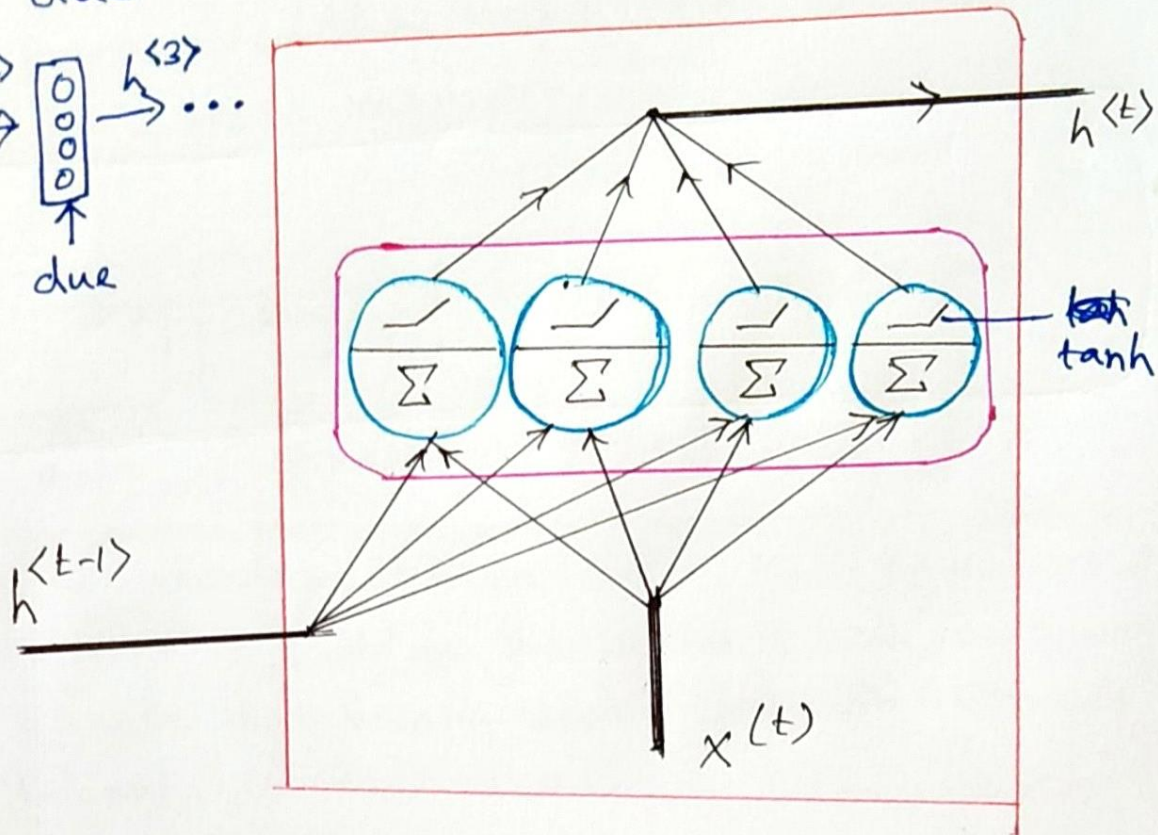
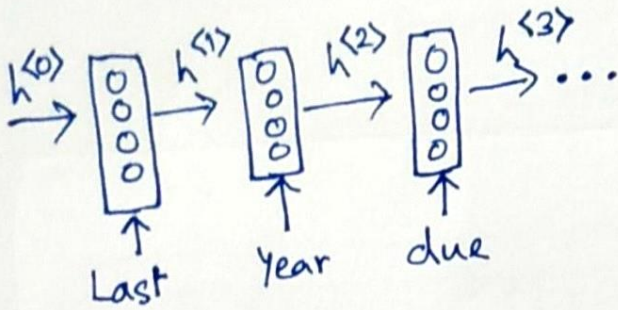
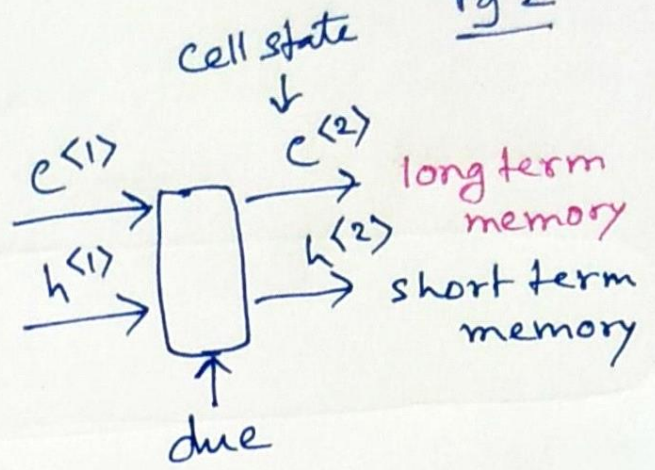
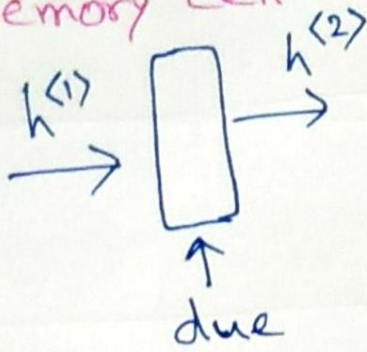




# LSTM

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Memory cell



Lara eats samosa everyday, it should not be hard to guess that her favourite cuisine is .....

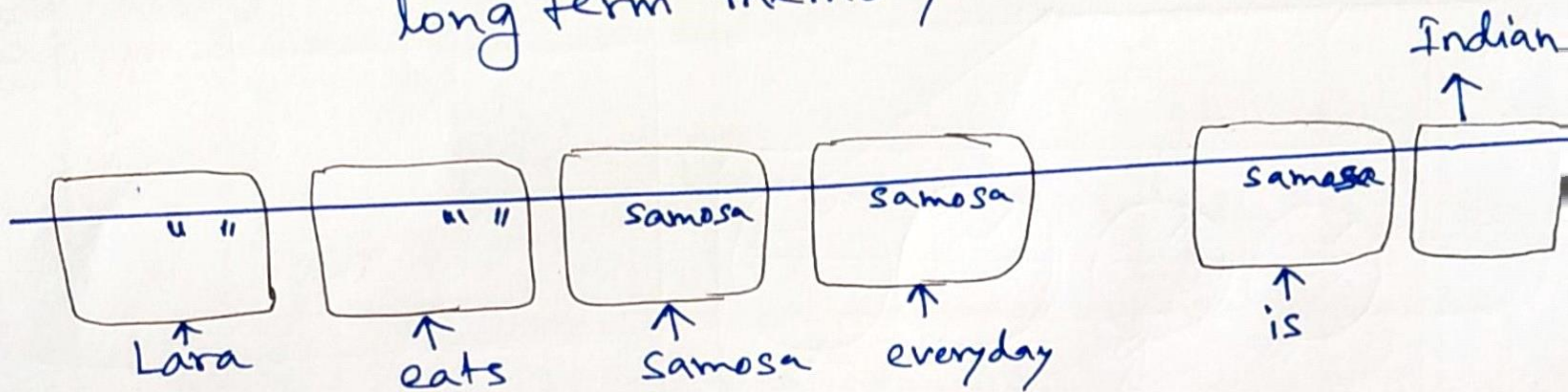
Indian →



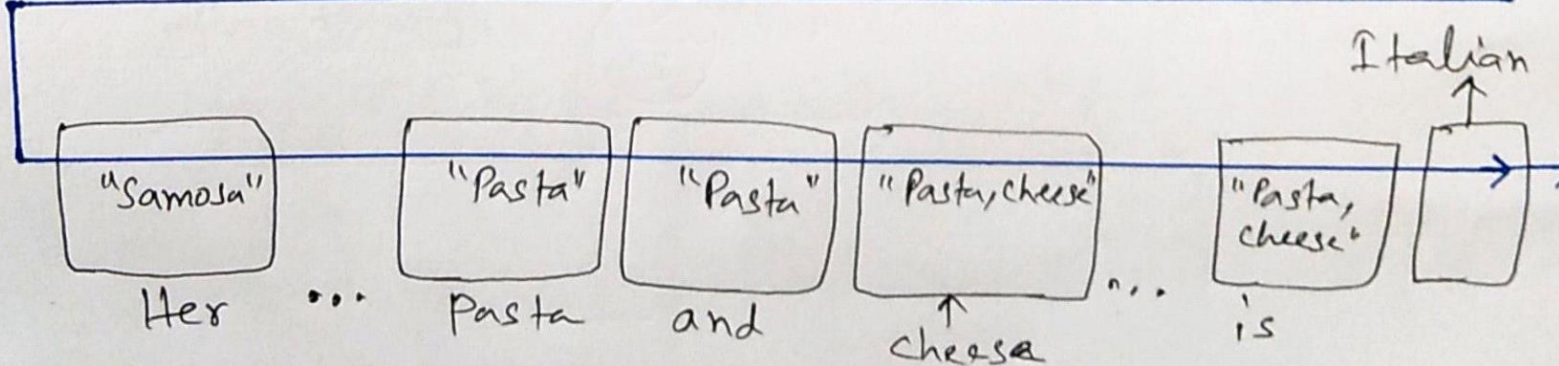
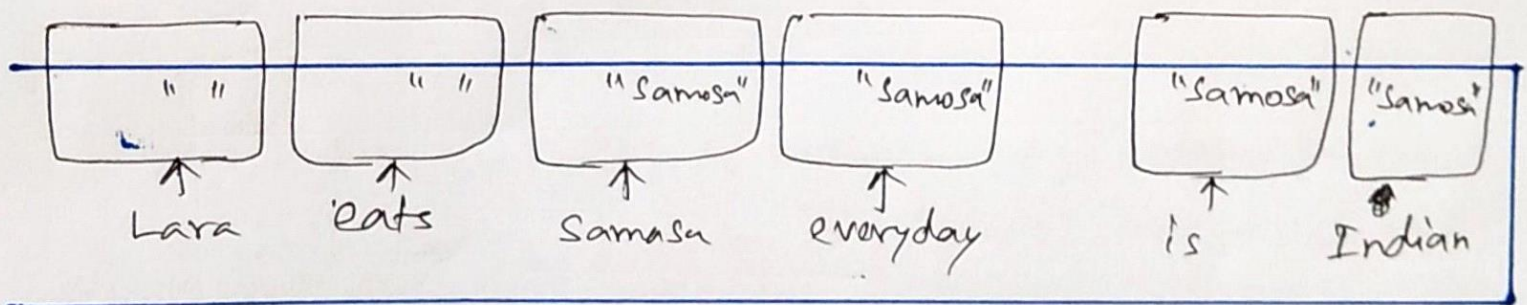
Traditional RNN will have problem here because of short term memory: by the time it comes to autocomplete, the effect of "samosa" is very fade.

Way-out

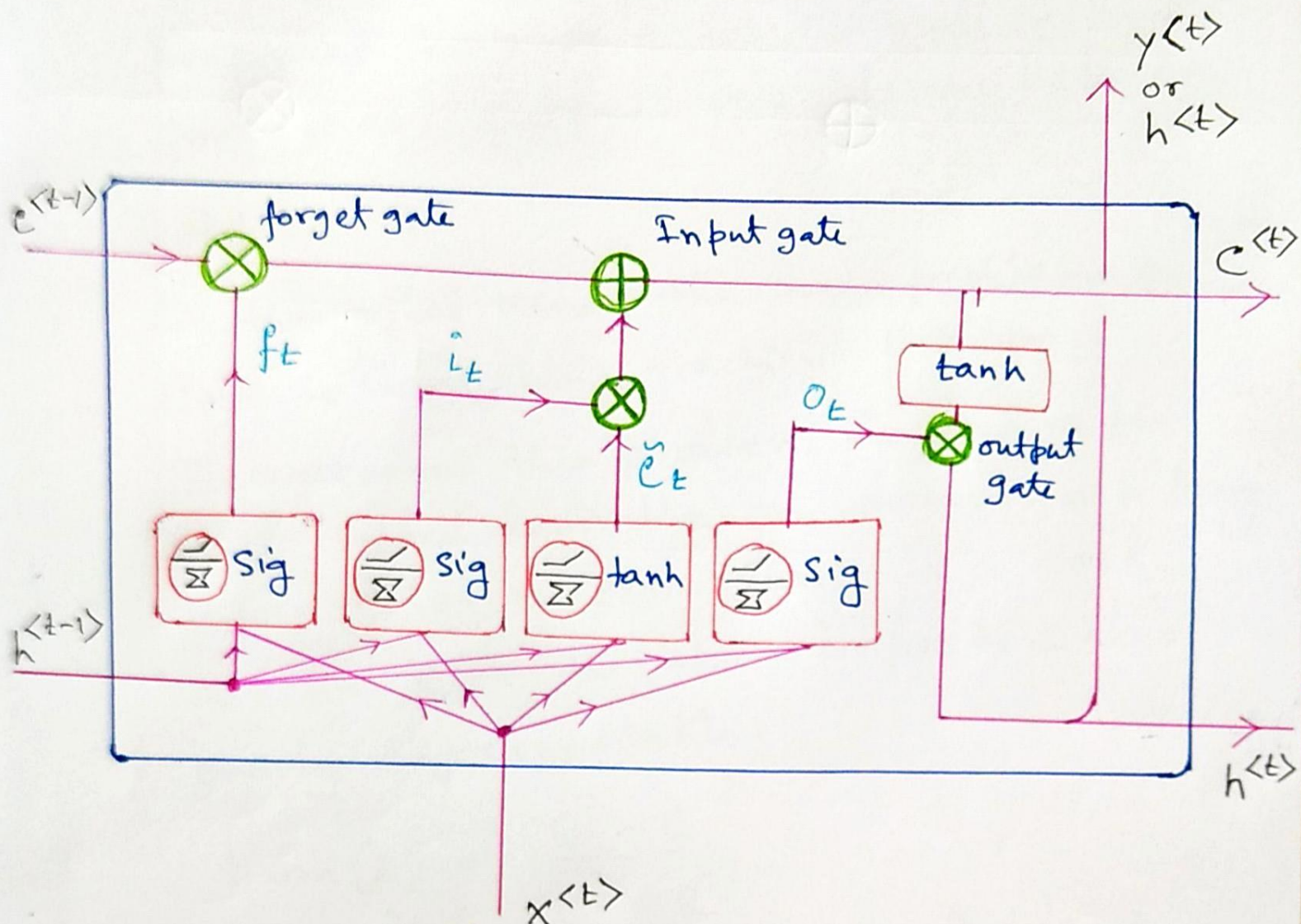
Preserve the important keywords in long term memory.



Lara eats samosa everyday, it should not be hard to guess that her favourite cuisine is Indian. Her brother Bhabin is a lover of Pasta and cheese that means Bhabin's favourite cuisine is ... (Italian).







- Cell state  $c^{(t-1)}$  or  $c^{(t)}$  allow only linear interactions.
- Sigmoid layers (outputs between 0 & 1) allows pairwise multiplication operation  $\Rightarrow$  Dictates how much to allow through.  $1 \Rightarrow$  all,  $0 \Rightarrow$  none

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

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

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

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

$$o_t = \sigma(W_o [h_{t-1}, x_t] + b_o) \parallel h_t = o_t * \tanh(c_t)$$



## Insight for LSTM sig & tanh activation fns.

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For Input & output gates sigmoid is preferred because of its normalization properties. outputs are always between 0 & 1.

On the other hand, the main problem of RNN is vanishing gradient problem.

To overcome this problem we need a fn. whose second derivative can sustain for a long before going to zero. Tanh is a good function with the above property & also easily differentiable.

$$\tanh(z) = \frac{e^z - e^{-z}}{e^z + e^{-z}}$$

$$\tanh'(z) = 1 - (\tanh(z))^2$$