

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

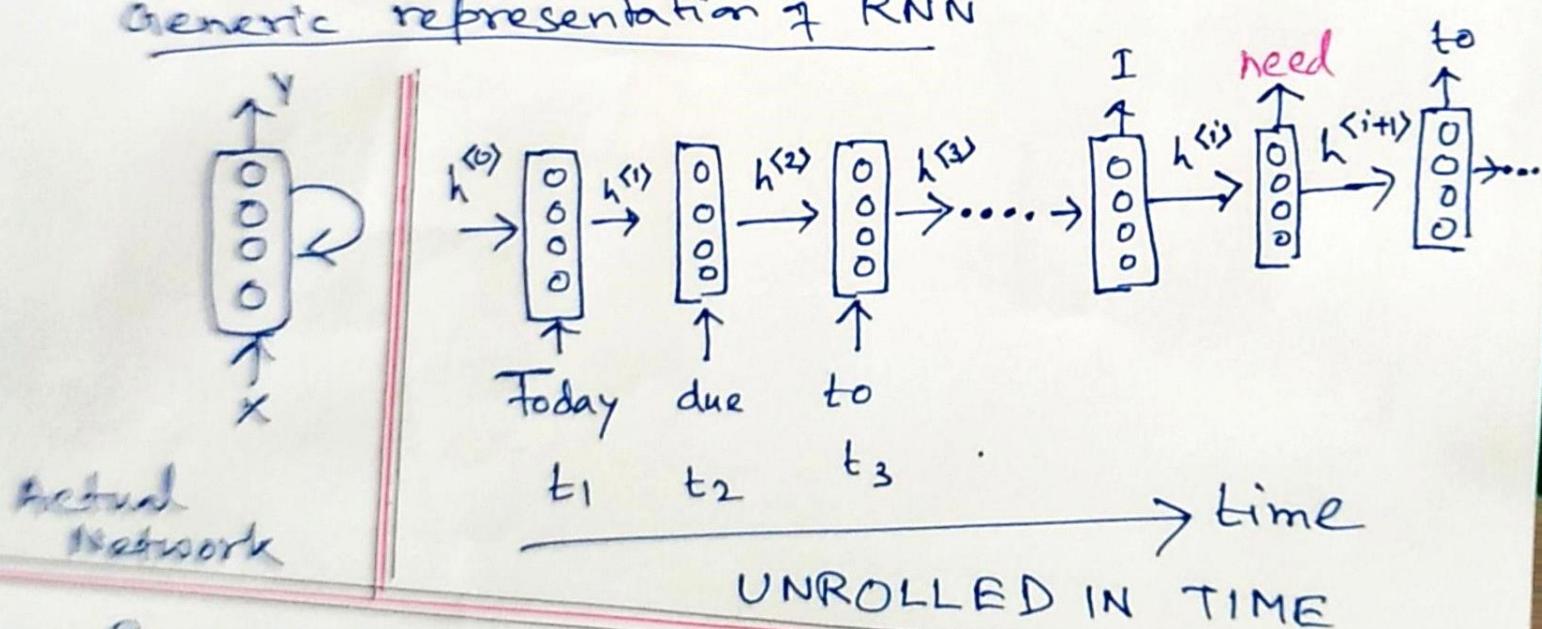
Last year & due to current COVID-19 situation, I

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

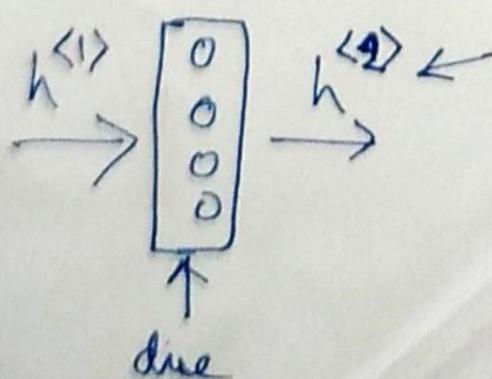
Today, due to current COVID-19 situation, I need to take a loan

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.

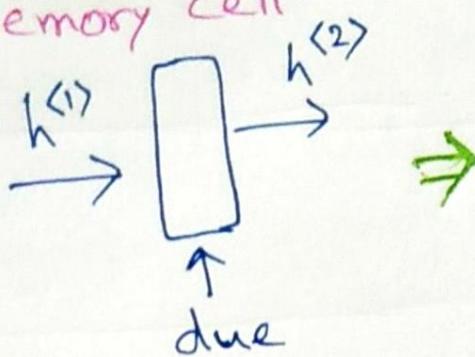


these hidden states are nothing but short term memory.

LSTM

Pg 2

Memory cell



cell state

$c^{(1)}$

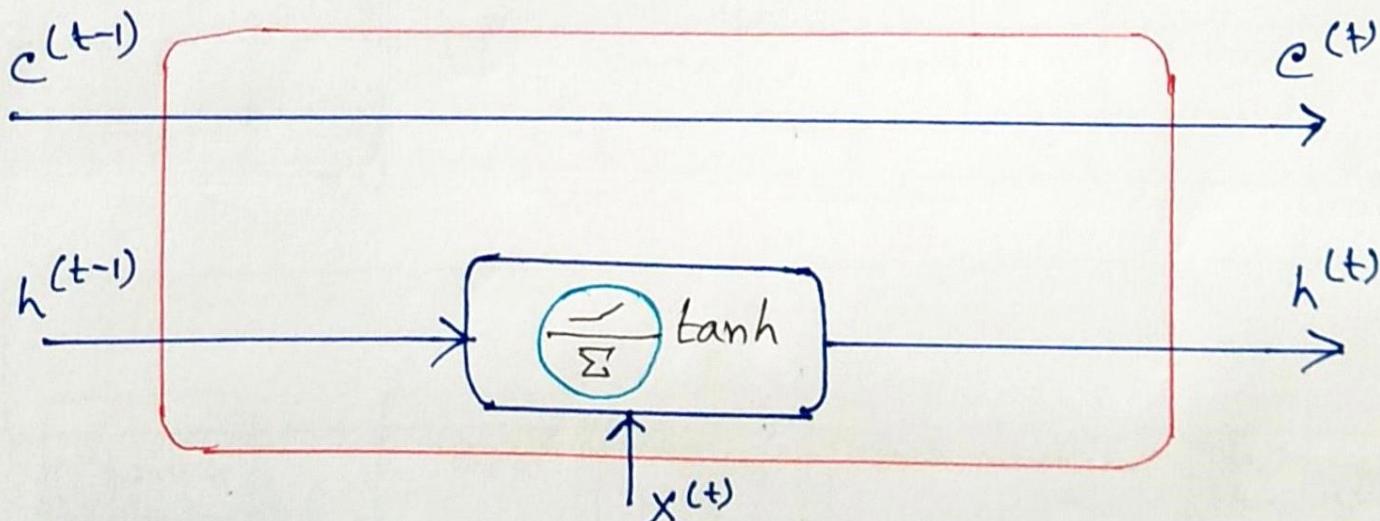
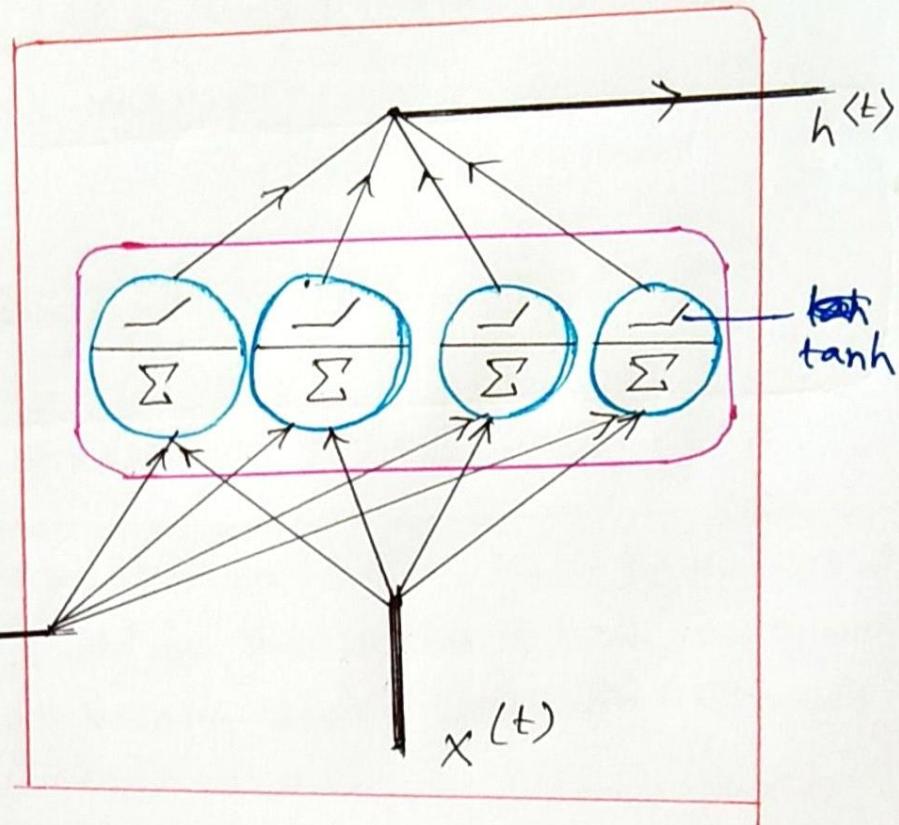
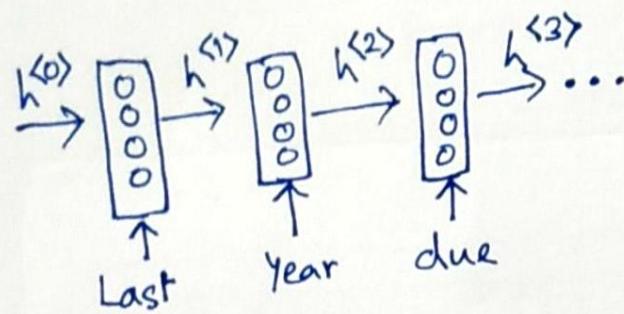
$h^{(1)}$

due

long term
memory
short term
memory

$c^{(2)}$

$h^{(2)}$



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

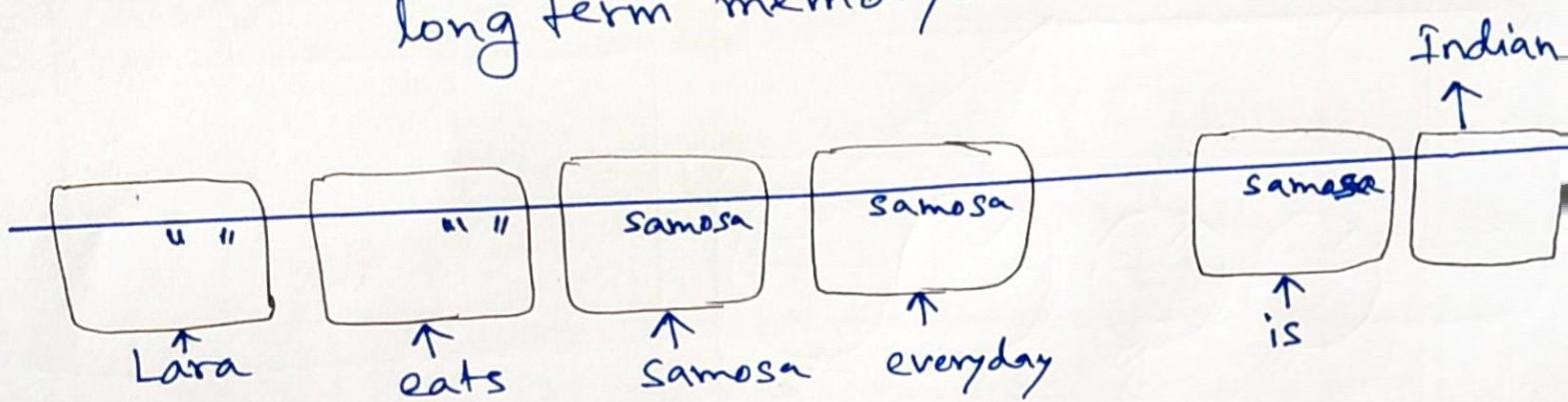
Indian →

LSTM

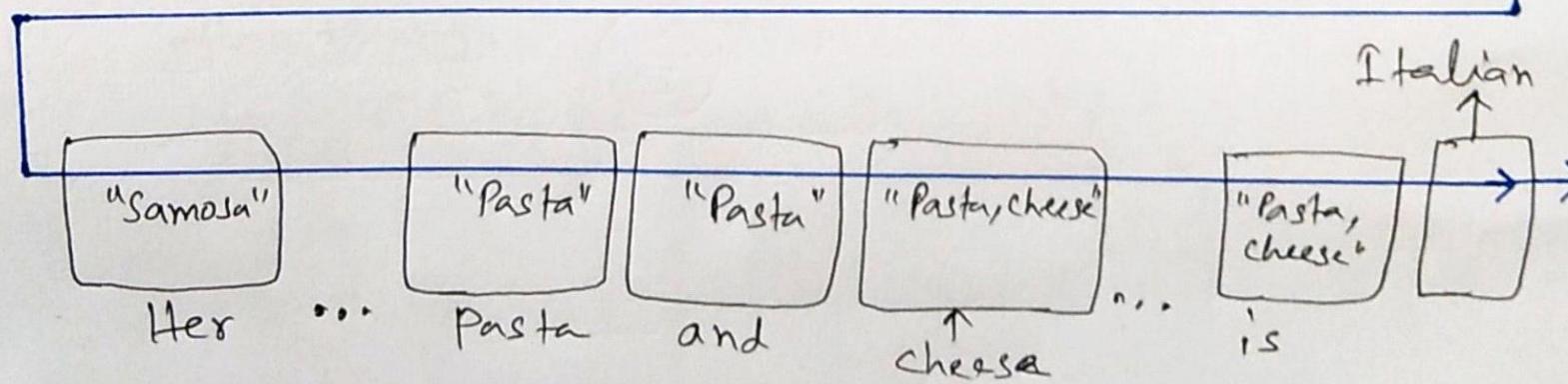
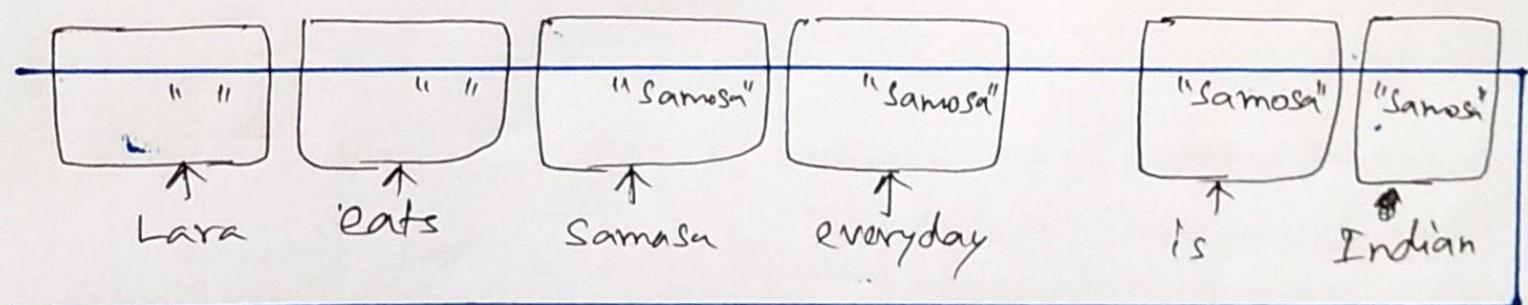
Traditional RNN will have problem here because of short term memory: by the time it comes to autocomplete, the effect of "Samosa" is very fade.

Way-out

Preserve the important keywords in long term memory.

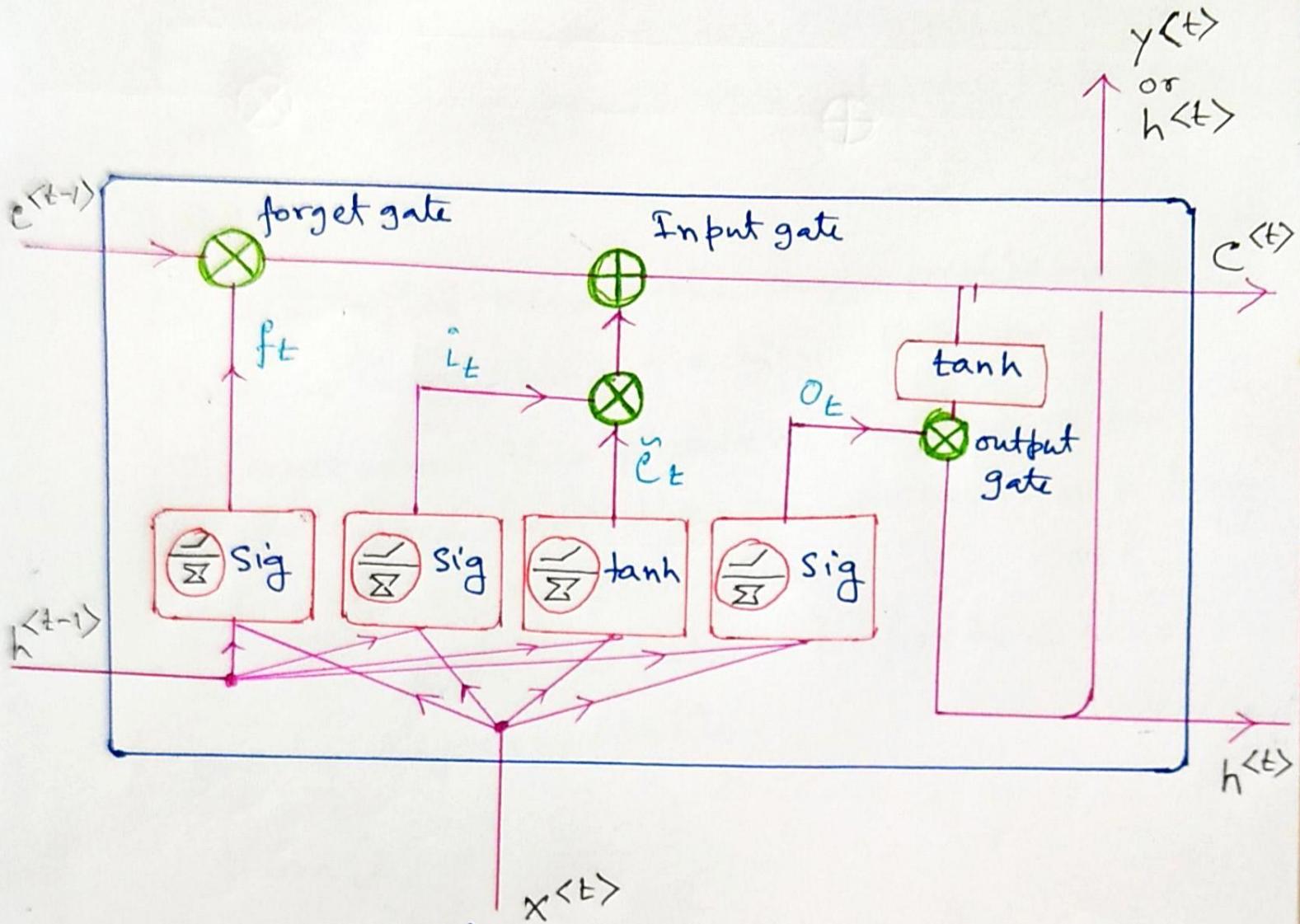


Lara eats samosa everyday, it should not be hard to guess that her favourite cuisine is Indian. Her brother Bharin is a lover of Pasta and cheese that means Bharin's favourite cuisine is (Italian).



LSTM

Pg 9



- 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

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

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

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

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

$$o_t = \sigma(w_o \cdot [h_{t-1}, x_t] + b_o) \quad || \quad h_t = o_t * \tanh(c_t)$$

LSTM

Insight for sig & tanh activation fns. Pg 5

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.

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

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