

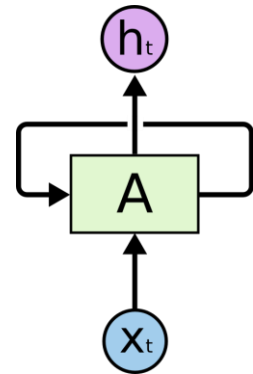
RECURRENT NEURAL NETWORKS



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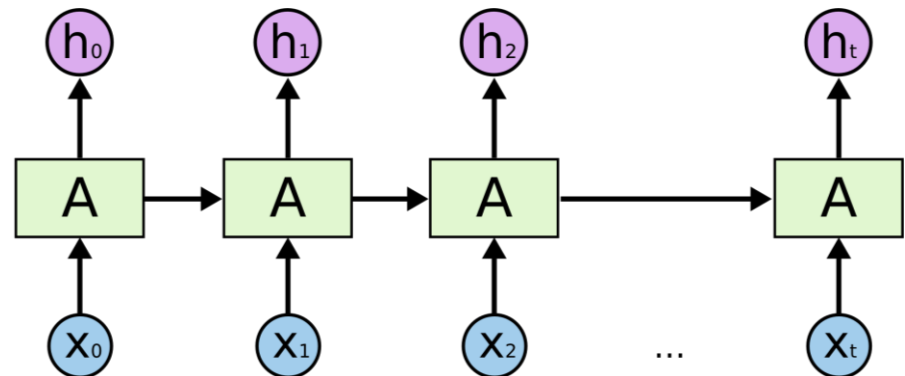
Definition.

Neural networks where the connections form a directed cycle. As opposed to feedforward neural networks.



Redefinition.

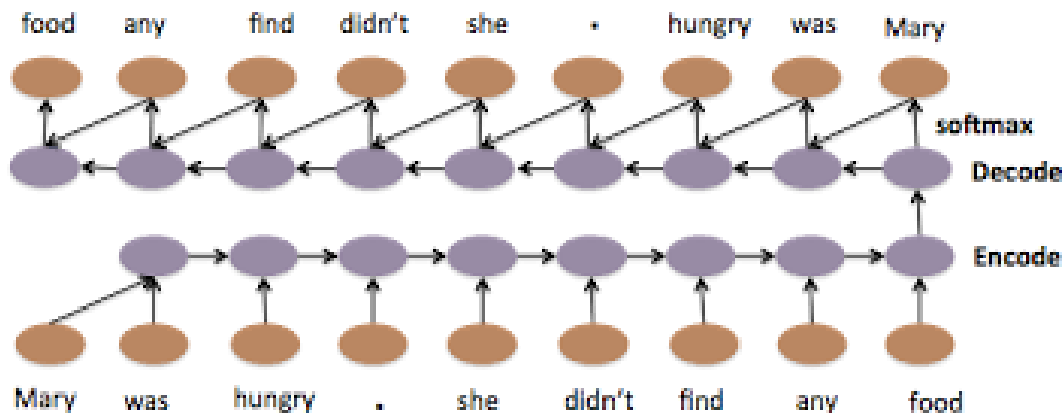
May be unrolled as a feedforward network with shared weights propagating through time.



RECURRENT NEURAL NETWORKS

Temporal State.

- RNN dynamics encode some time-varying state.
- Perfect for modeling sequences, e.g. speech recognition, natural language processing, financial modeling.



PROBLEM WITH VERY DEEP NETWORKS

Vanishing or Exploding Gradient Problem.

- Difficult to train very deep networks because of issues with backpropagation algorithm
- Error signals $\delta^{(l)}$ either shrink exponentially, or grow exponentially with the number of layers

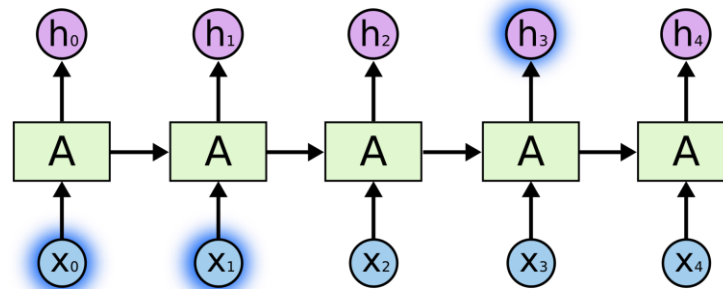
Recurrent Neural Networks are Very Deep!

- We want RNNs to learn long-term dependencies
- But for a long time, nobody knew how to train them

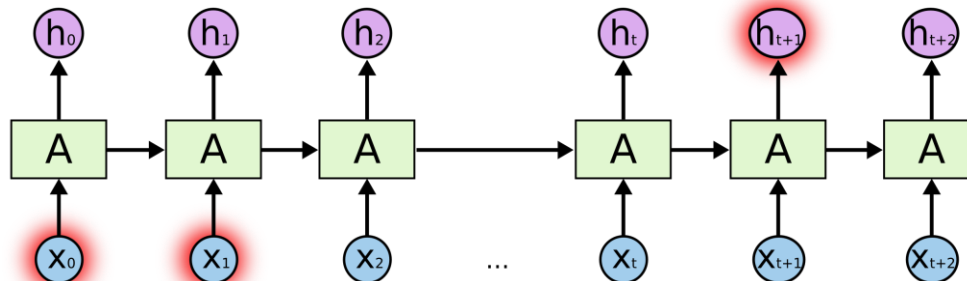


LONG-TERM DEPENDENCIES

“the clouds are in the *sky*”

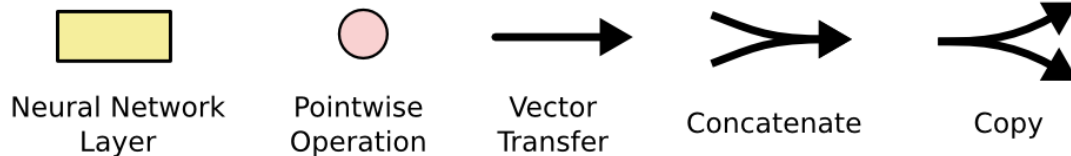
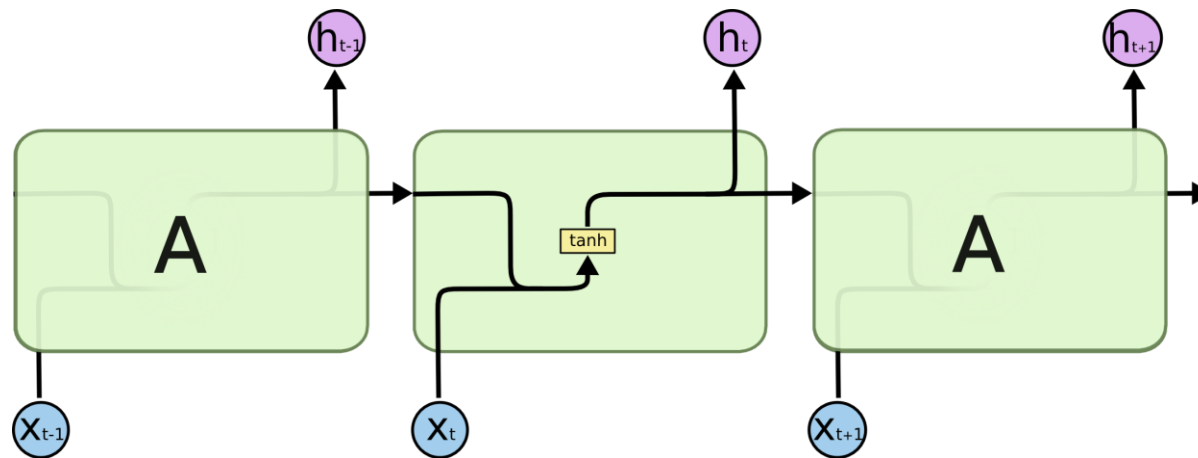


I grew up in France... I speak fluent *French*.



STANDARD RNN

Single Layer RNN.



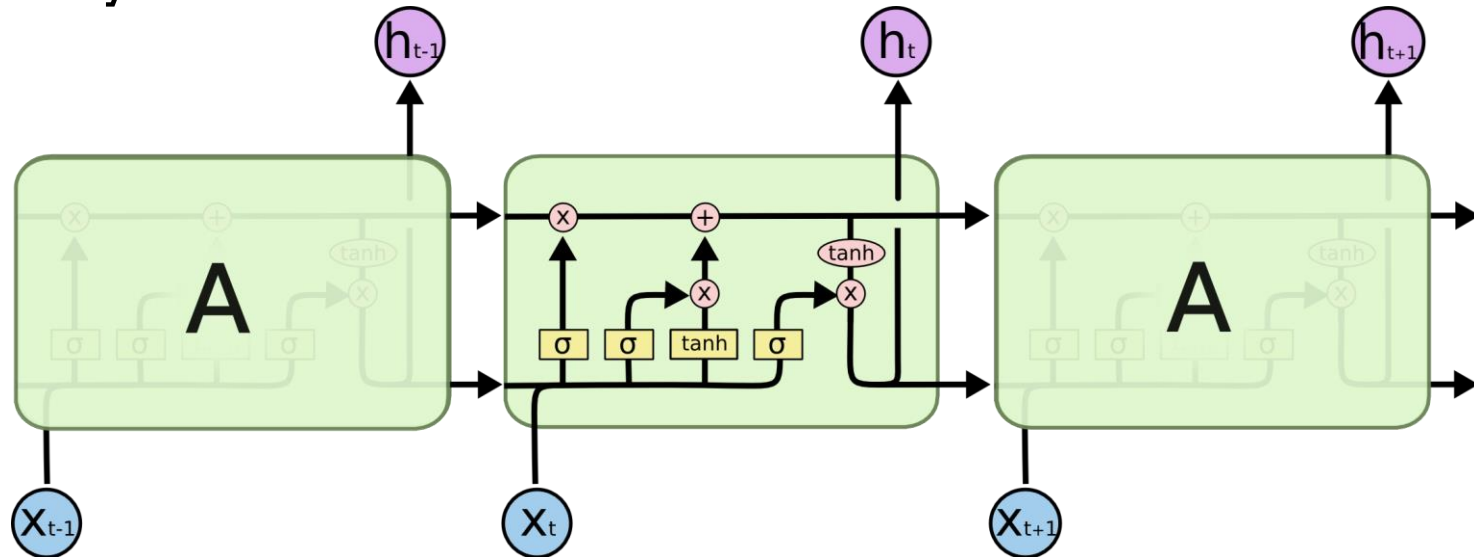


LONG SHORT TERM MEMORY (LSTM)



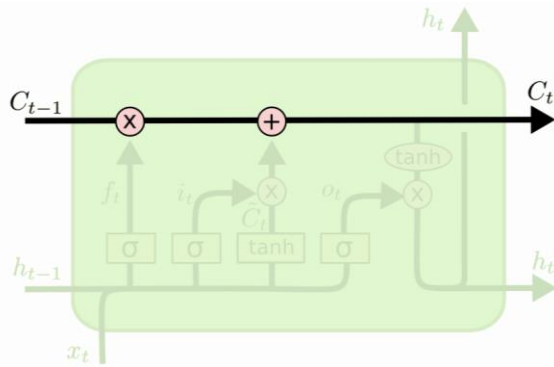
LONG SHORT TERM MEMORY

Four-Layer RNN.

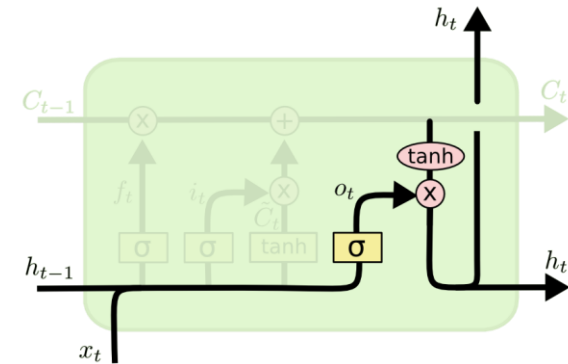


STATES AND GATES

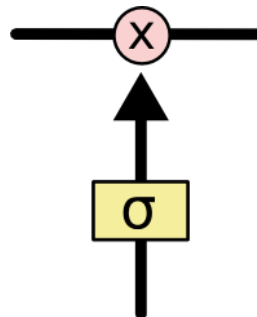
Cell State



Output State



Gates

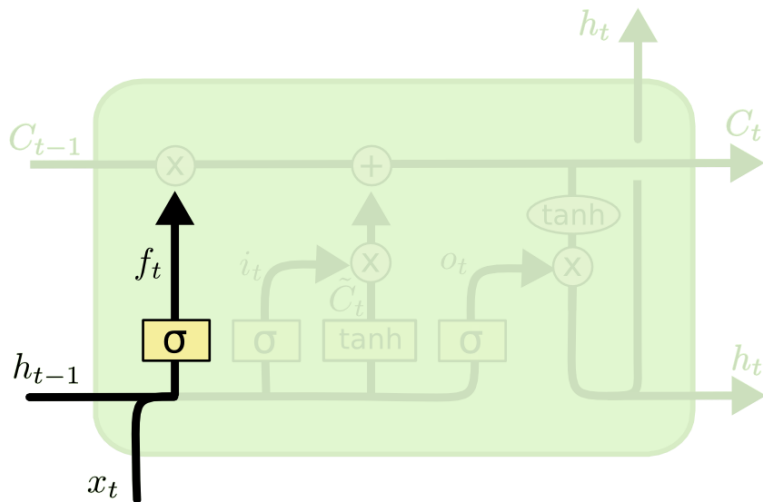


Output of sigmoid
is between 0 and 1
0 – forget past state
1 – keep past state



LAYER 1 – FORGET PAST STATE

Example. Forget gender of subject when subject changes.



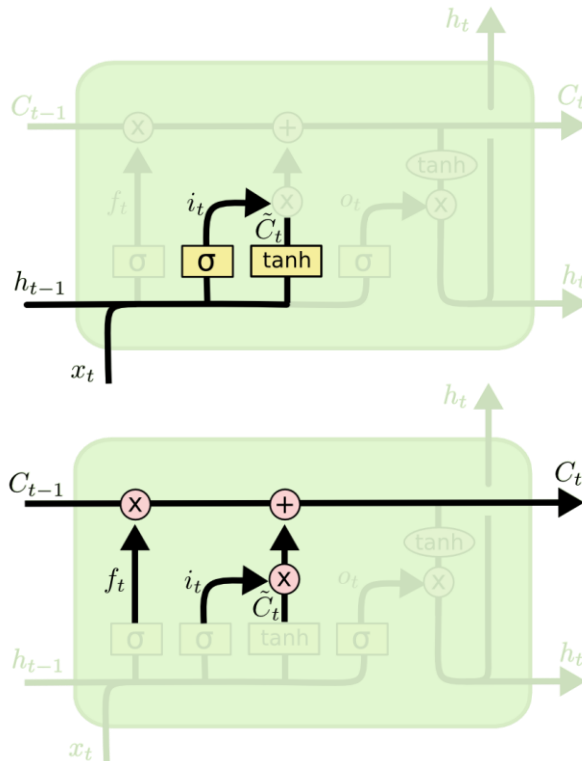
How much to forget?

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



LAYER 2,3 – ADD NEW INFORMATION

Example. Add gender of new subject.



How much to add?

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

What info to add?

$$\tilde{C}_t = \tanh(W_C \cdot [h_{t-1}, x_t] + b_C)$$

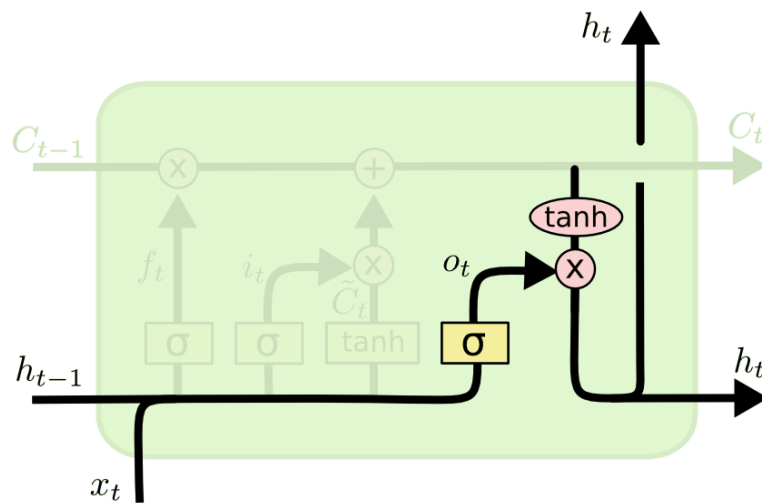
New cell state.

$$C_t = f_t * C_{t-1} + i_t * \tilde{C}_t$$



LAYER 4 – SELECTIVE OUTPUT

Example. Singular or plural form of verb? Output and forward.



How much to output?

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

What to output?

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

