

NLP LSTM

1. RNN: $h_t = \sigma(W_{ih}x_t + b_{ih} + W_{hh}h_{t-1} + b_{hh})$

LSTM: $i_t = \sigma(W_{ii}x_t + b_{ii} + W_{hi}h_{t-1} + b_{hi})$

$f_t = \sigma(W_{if}x_t + b_{if} + W_{hf}h_{t-1} + b_{hf})$

$g_t = \tanh(W_{ig}x_t + b_{ig} + W_{hg}h_{t-1} + b_{hg})$

$o_t = \sigma(W_{io}x_t + b_{io} + W_{ho}h_{t-1} + b_{ho})$

$C_t = f_t \odot C_{t-1} + i_t \odot g_t$

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

Assumes weights for i_t , f_t and g_t have weights of 0 and very large biases. (i.e. $W_{ii}=0$, $W_{hi}=0$, $W_{if}=0$, $W_{hf}=0$, $W_{ig}=0$, $W_{hg}=0$)

Under this assumption, $i_t \rightarrow 1$, $f_t \rightarrow 1$, $g_t \rightarrow 1$.

$C_t = 1 \odot C_{t-1} + 1 \odot 1 = C_{t-1} + 1$

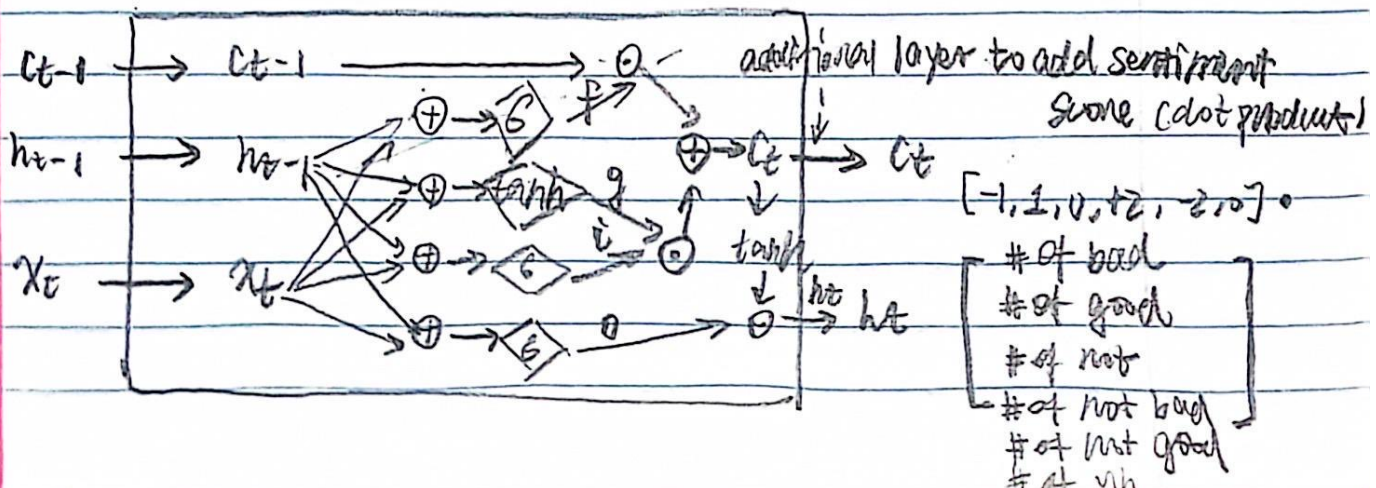
assume C_0 is close to infinity, C_t would be close to infinity as well.

in this case, $\tanh(C_t)$ would be close to 1. ($C_t \rightarrow 1$)

$h_t = o_t = \sigma(W_{io}x_t + b_{io} + W_{ho}h_{t-1} + b_{ho})$

make $W_{io}=W_{ih}$, $W_{hh}=W_{ho}$, $b_{ih}=b_{io}$, $b_{hh}=b_{ho}$, then h_t of LSTM under these assumptions would equal to RNN in a very simple form.

2. The input ^{size} of the LSTM is 6. ($M=6$), hidden state size could be varied dependent on the training model, assume size of 8 for now. ($K=8$), output size is 1 ($L=1$), which is cumulative sentiment score.



as shown in the graph. I add an additional layer on top of C_t to include the sentiment score calculation. (since h_t is defined by \tanh and has range of $(-1, 1)$, not suitable for sentiment score calculation).

similar to question #1. I assume all the weights of i_t , f_t , g_t and o_t to be 0, bias to be a large number to allow the full activation of ~~input~~ input gate, forget gate, and output gates.

under this assumption, $i_t \rightarrow 1$, $f_t \rightarrow 1$, $g_t \rightarrow 1$, $o_t \rightarrow 1$.

$$C_t = C_{t-1} + 1$$

assume $C_0 = 0$ at time of 0.

then add additional layer defined as z_t to C_t . ($z_t = W_{iz} \cdot x_t + \text{bias}$)

$$C_t = C_{t-1} + 1 + W_{iz} \cdot x_t + \text{bias}$$

Note: $W_{iz} = [-1, 1, 0, +2, -2]$

$$x_t = \begin{bmatrix} \# \text{ of bad} \\ \# \text{ of good} \\ \# \text{ of not } \text{~~good~~ bad} \\ \# \text{ of not good} \\ \# \text{ of un} \end{bmatrix}$$

(in alphabetical order).

bias set to 0