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Machine Learning with Python-From Linear Models to Deep Learning

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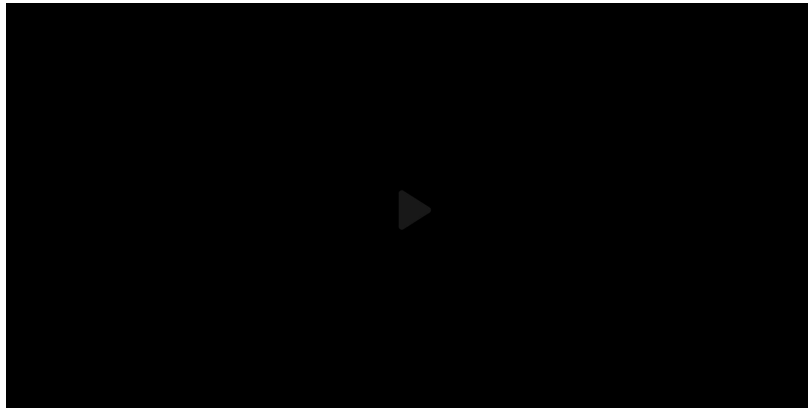
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## 5. Gating and LSTM

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## Gating and LSTM

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Learning these recurrent neural network models

is very similar to learning feed-forward neural networks.

As we turn the sentence into a vector, we make a prediction.

As a result, we get an error signal.

We back propagate that error signal



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## Gating

1/1 point (graded)

Recall that the most simple, single-layered RNN can be written in equation as:

$$s_t = \tanh(W^{s,s}s_{t-1} + W^{s,x}x_t).$$

Recognize that, in the above formulation,  $s_t$  is always overwritten with the calculated result  $\tanh(W^{s,s}s_{t-1} + W^{s,x}x_t)$ .

Now, we introduce a gate vector  $g_t$  of the same dimension as  $s_t$ , which determines "how much information to overwrite in the next state." In equation, a single-layered gated RNN can be written as:

$$\begin{aligned} g_t &= \text{sigmoid}(W^{g,s}s_{t-1} + W^{g,x}x_t) \\ s_t &= (1 - g_t) \odot s_{t-1} + g_t \odot \tanh(W^{s,s}s_{t-1} + W^{s,x}x_t). \end{aligned}$$

where the sign  $\odot$  denotes element-wise multiplication. Now, which of the following is true about the gate  $g_t$ ? (Choose all those apply.)

☐ If the  $i$ th element of  $g_t$  is 1, the  $i$ th element of  $s_t$  and that of  $s_{t-1}$  are equal

☒ If the  $i$ th element of  $g_t$  is 0, the  $i$ th element of  $s_t$  and that of  $s_{t-1}$  are equal

☐ If  $g_t$  is a vector whose elements are all 1,  $s_t$  and  $s_{t-1}$  are equal

☒ If  $g_t$  is a vector whose elements are all 0,  $s_t$  and  $s_{t-1}$  are equal



**Solution:**

Let the  $i$ th element of  $s_t, g_t, s_{t-1}$  be  $s_t^i, g_t^i, s_{t-1}^i$ .

If the  $i$ th element of  $g_t$  is 0,  $(1 - g_t^i) = 1$ , so

$$s_t^i = s_{t-1}^i.$$

Thus, if the  $i$ th element of  $g_t$  is 0, the  $i$ th element of  $s_t$  and that of  $s_{t-1}$  are equal. Also, if  $g_t$  is a vector whose elements are all 0,  $s_t$  and  $s_{t-1}$  are equal.

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You have used 1 of 2 attempts

**i** Answers are displayed within the problem

**LSTM**

1/1 point (graded)

Which of the following components of an LSTM represent the context or state? (Choose all that apply.)

☒  $c_t$

☒  $h_t$

☐  $o_t$

☐  $i_t$

**Solution:**

$c_t$  represents the memory cell, and  $h_t$  represents the visible state. Together they make up the context or state. The other two choices are the output and input gate, respectively. They simply accommodate new inputs and output predictions, and are not part of the context/state

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You have used 1 of 2 attempts

**i** Answers are displayed within the problem

**LSTM Calculations**

1/1 point (graded)

Let all the neural network's weight matrices, the hidden state, and the memory cell be a scalar 1. Let the new  $x$ -value be 5. Calculate the value of the new hidden state. Round sigmoid to 1 or 0, and round  $\tanh$  to  $-1$  or  $1$ .

$$\begin{aligned} f_t &= \text{sigmoid}(W^{f,h}h_{t-1} + W^{f,x}x_t) && \text{forget gate} \\ i_t &= \text{sigmoid}(W^{i,h}h_{t-1} + W^{i,x}x_t) && \text{input gate} \\ o_t &= \text{sigmoid}(W^{o,h}h_{t-1} + W^{o,x}x_t) && \text{output gate} \\ c_t &= f_t \odot c_{t-1} + i_t \odot \tanh(W^{c,h}h_{t-1} + W^{c,x}x_t) && \text{memory cell} \\ h_t &= o_t \odot \tanh(c_t) && \text{visible state} \end{aligned}$$

☐ -1

☒ 1

☐ 0

☐ 5



### Solution:

The forget gate is equal to  $\text{sigmoid}(6)$ , or 1. The same applies for the input and output gate. The memory cell is equal to  $1 + \tanh(1+5)$ , which is 2. The new hidden state is therefore  $\tanh(2)$ , or 1.

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You have used 1 of 2 attempts

**i** Answers are displayed within the problem

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Gating and LSTM

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- [Understanding LSTM Networks \(link\)](#)  
Resources to complement the actual lecture. Hope it helps <http://colah.github.io/posts/2015-08-Understanding-LSTMs/> <https://www.you...> 2  
 Pinned Community TA
- [Why tanh\(1+6\) in the solution?](#)  
I've got problem with understanding the solution in this part: why is there  $\tanh(1+6)$  and not again  $\tanh(1+5)$ ? 6
- [Hidden state?](#)  
The last exercise refers to "hidden state" but does not explain what quantity this term is referring to. From context, this seems to be h t... 7
- [HELP what is sigmoid\(6\)](#)  
How do you compute this sigmoid function. For example  $\text{sigmoid}(2)$ ? thank you 3

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