

CS 6180 09/24

HW2 released tomorrow (will be a bit longer
so you'll have
2 weeks instead of
one)

Review from last time

- * Vanishing / Exploding gradients are a main limitation of RNNs.
- * limit the # of time steps we are considering in backpropagation (reduce the number of sigmoids we are multiplying)

losing some connections between words at the beginning of a sentence and later ones.

- * Orthogonal Init

mainly helps with exploding gradients still have the issue of the products of sigmoids.

Long Short-term memory

main idea: store some info in a cell $\vec{c}^{(t)}$
(RAM in your computer)

to keep track of relevant info.

From one time step to another

* Forget gate: some info in the cell (RAM) will be forgotten

$$\vec{f}^{(t)} = \sigma(W_f \vec{h}^{(t-1)} + U_f \vec{e}^{(t)} + \vec{b}_f)$$

* Input gate: new info that will be stored in the cell.

$$\vec{i}^{(t)} = \sigma(W_i \vec{h}^{(t-1)} + U_i \vec{e}^{(t)} + \vec{b}_i)$$

* Output gate: new info that will be stored in the hidden state

$$\vec{o}^{(t)} = \sigma(W_o \vec{h}^{(t-1)} + U_o \vec{e}^{(t)} + \vec{b}_o)$$

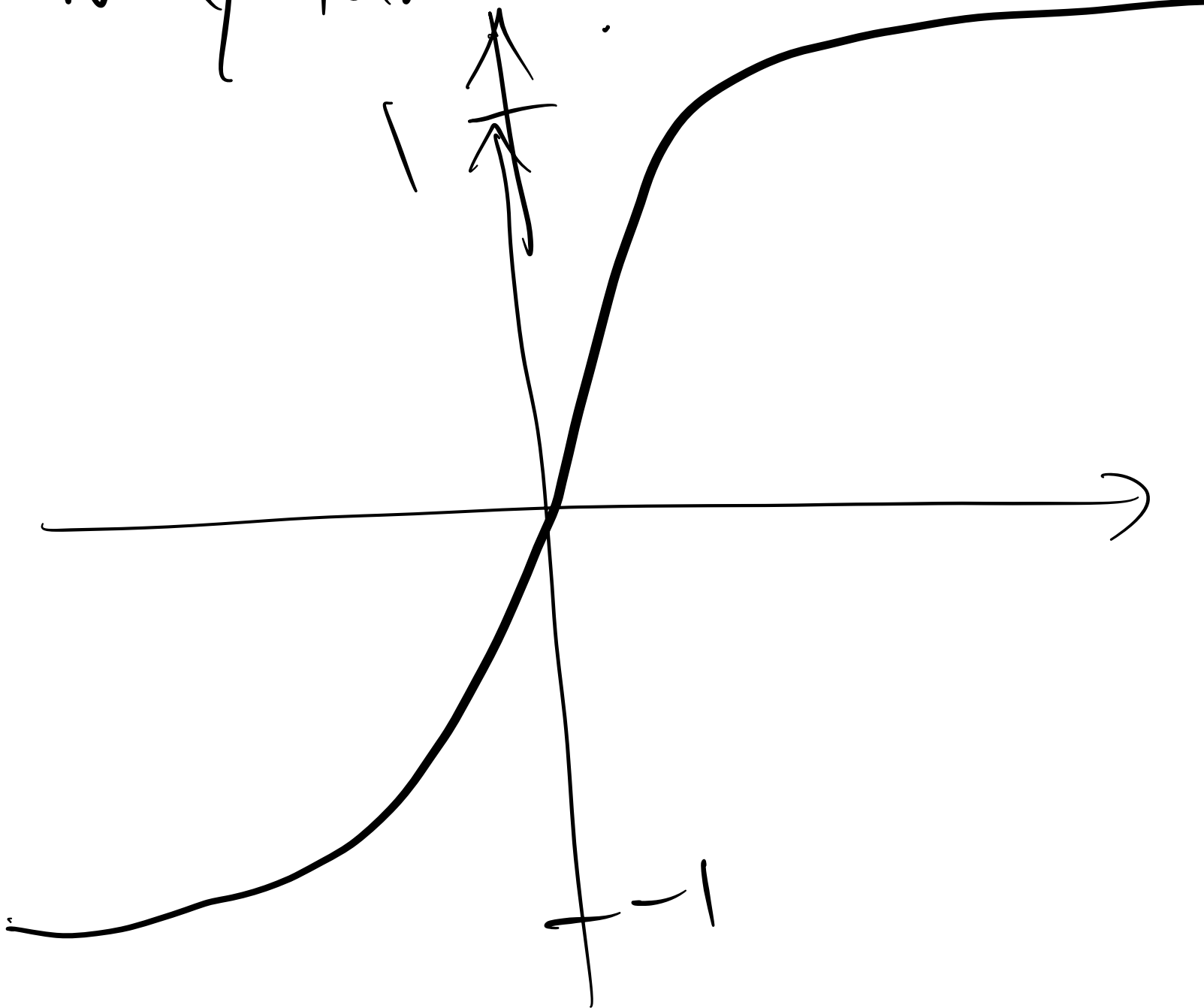
Let's first start by updating the cell

$$\vec{c}^{(t)} = \vec{f}^{(t)} \times \vec{c}^{(t-1)} + \vec{i}^{(t)} \times \vec{\tilde{c}}^{(t)}$$

↑
element wise
multiplication

$$\vec{\tilde{c}}^{(t)} = \tanh(W_c \vec{h}^{(t-1)} + U_c \vec{e}^{(t)} + \vec{b}_c)$$

Why tanh?




~~~~~ "is fantastically bad"  
t-1 t

$$C^{(t-1)} = 0.8$$

positive  
sentiment

$$\beta^{(t)} = 1$$

$$\hat{C}^{(t)} = -0.6$$


$$\lambda^{(t)} = 1$$

$$\begin{aligned} C^{(t)} &= \beta^{(t)} * C^{(t-1)} + \lambda^{(t)} * \hat{C}^{(t)} \\ &= 1 * 0.8 + 1 * (-0.6) \\ &= 0.2 \end{aligned}$$

we need these negative values to represent an idea of negation or reversal.

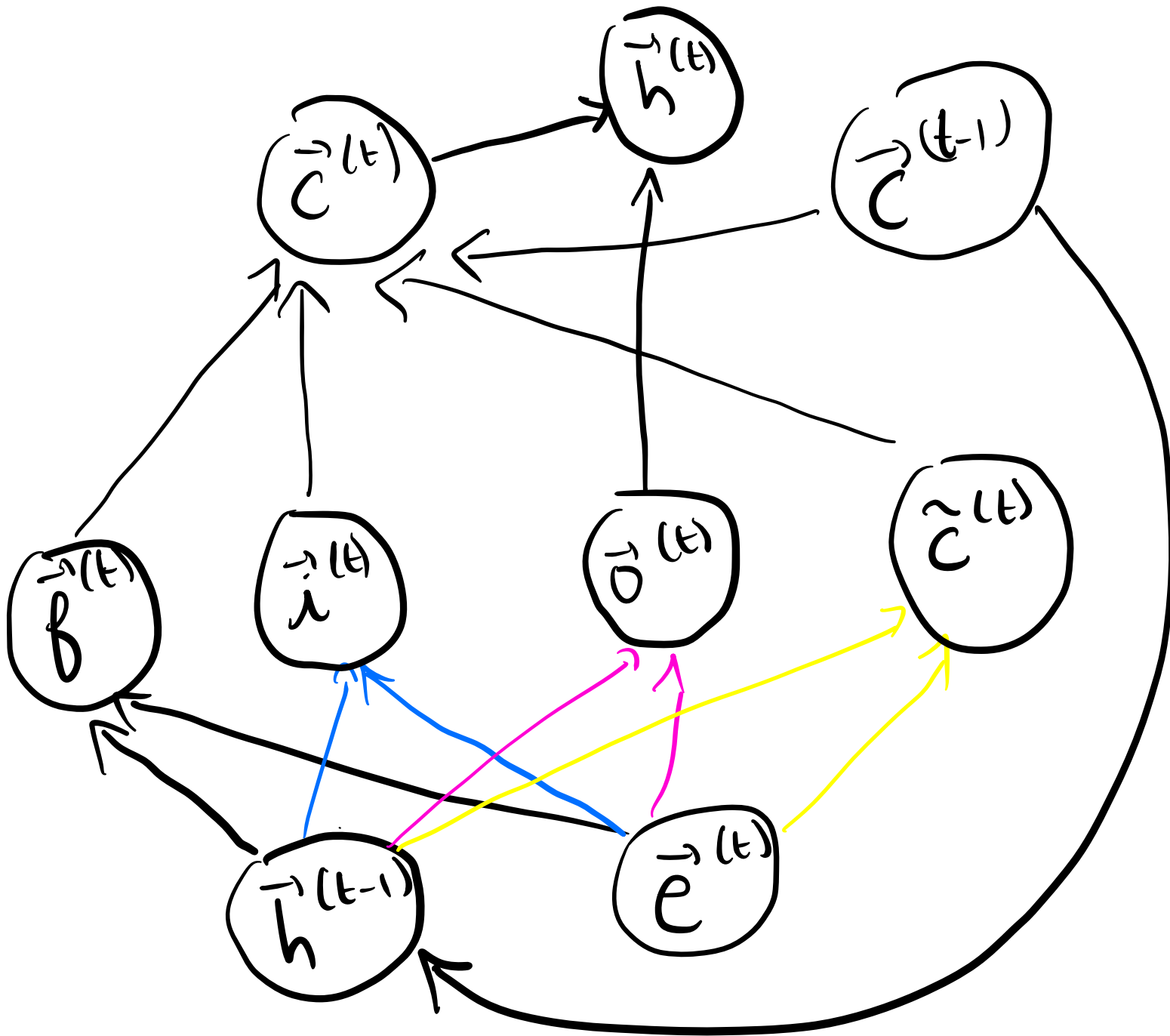
Remaining step:

Update the hidden state

$$\vec{h}^{\rightarrow}(t) = \vec{O}^{\rightarrow}(t) * \tanh(C_t)$$

{  $f(t)$  forget gate  
 $i(t)$  input gate  
 $O(t)$  output gate  
 $\tilde{C}(t)$  new info ( $\tanh$ )  
 $\vec{h}(t)$   $\vec{h}(t-1)$   $\vec{C}(t)$   $\tilde{C}(t)$

$$\begin{cases} c^{(t)} = f * c^{(t-1)} + \lambda * c^{(t-1)} \\ h^{(t)} = o^{(t)} * \tanh(c^{(t)}) \end{cases}$$



# Exercises:

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Show

$$\frac{\partial \mathcal{L}}{\partial \vec{C}^{(t-1)}} = \frac{\partial \mathcal{L}}{\partial \vec{C}^{(t)}} \cdot \vec{\beta}^{(t)}$$

then show

$$\frac{\partial \mathcal{L}}{\partial \vec{C}^{(t-K)}} = \frac{\partial \mathcal{L}}{\partial \vec{C}^{(t)}} \cdot \prod_{j=0}^{K-1} \vec{\beta}^{(t-j)}$$