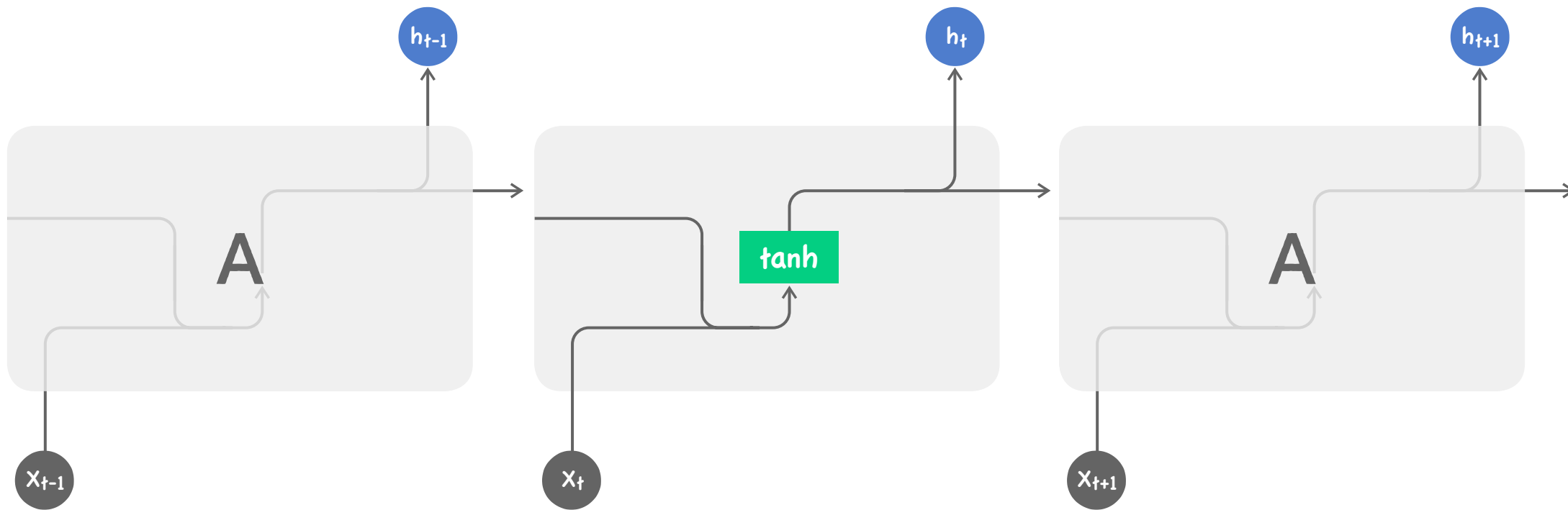


LECTURE 12-2

LSTM INTRODUCTION

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<http://hunkim.github.io/ml>

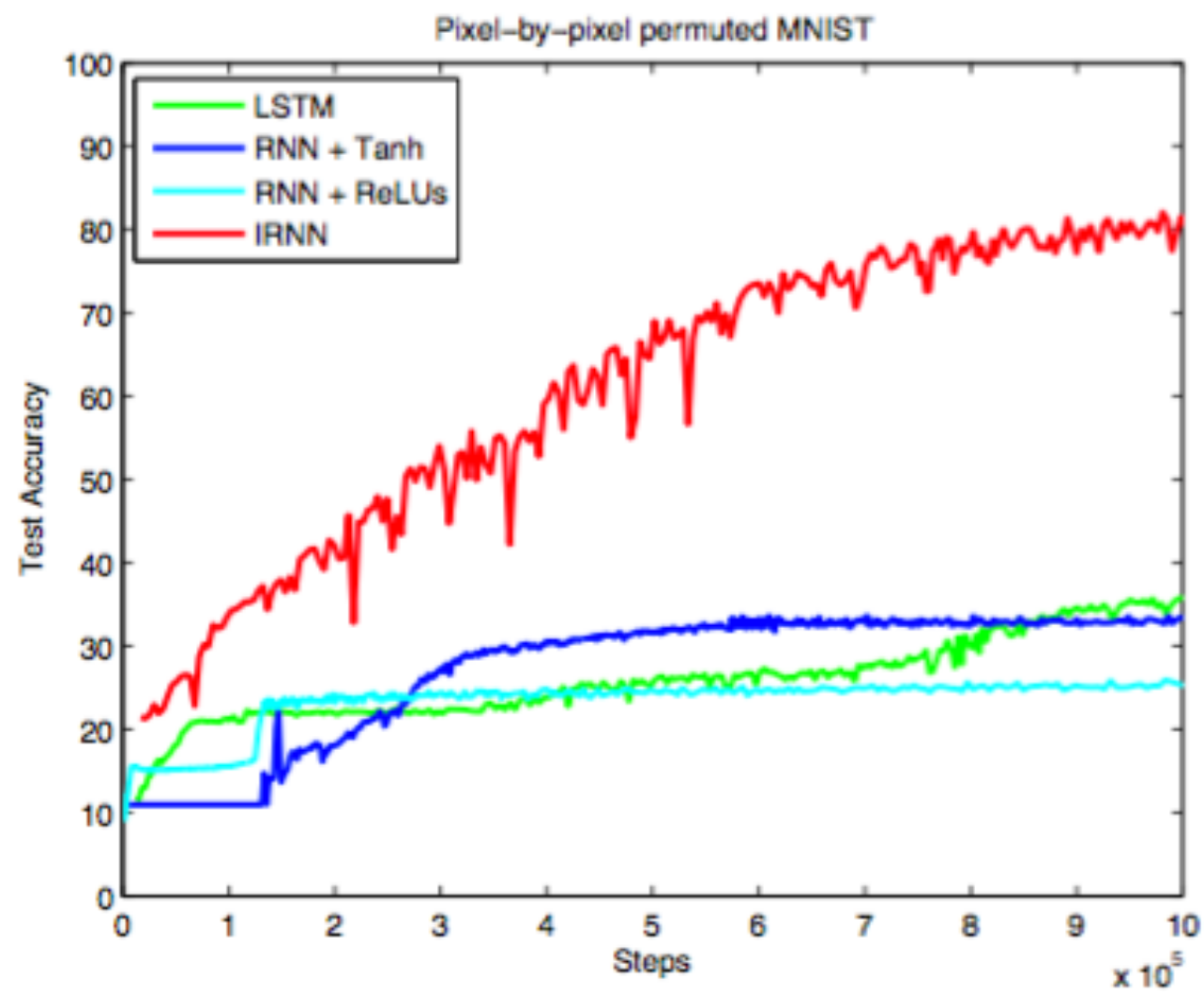


Given list of word **vectors** : $x_1, \dots, x_{t-1}, x_t, x_{t+1}, \dots, x_T$

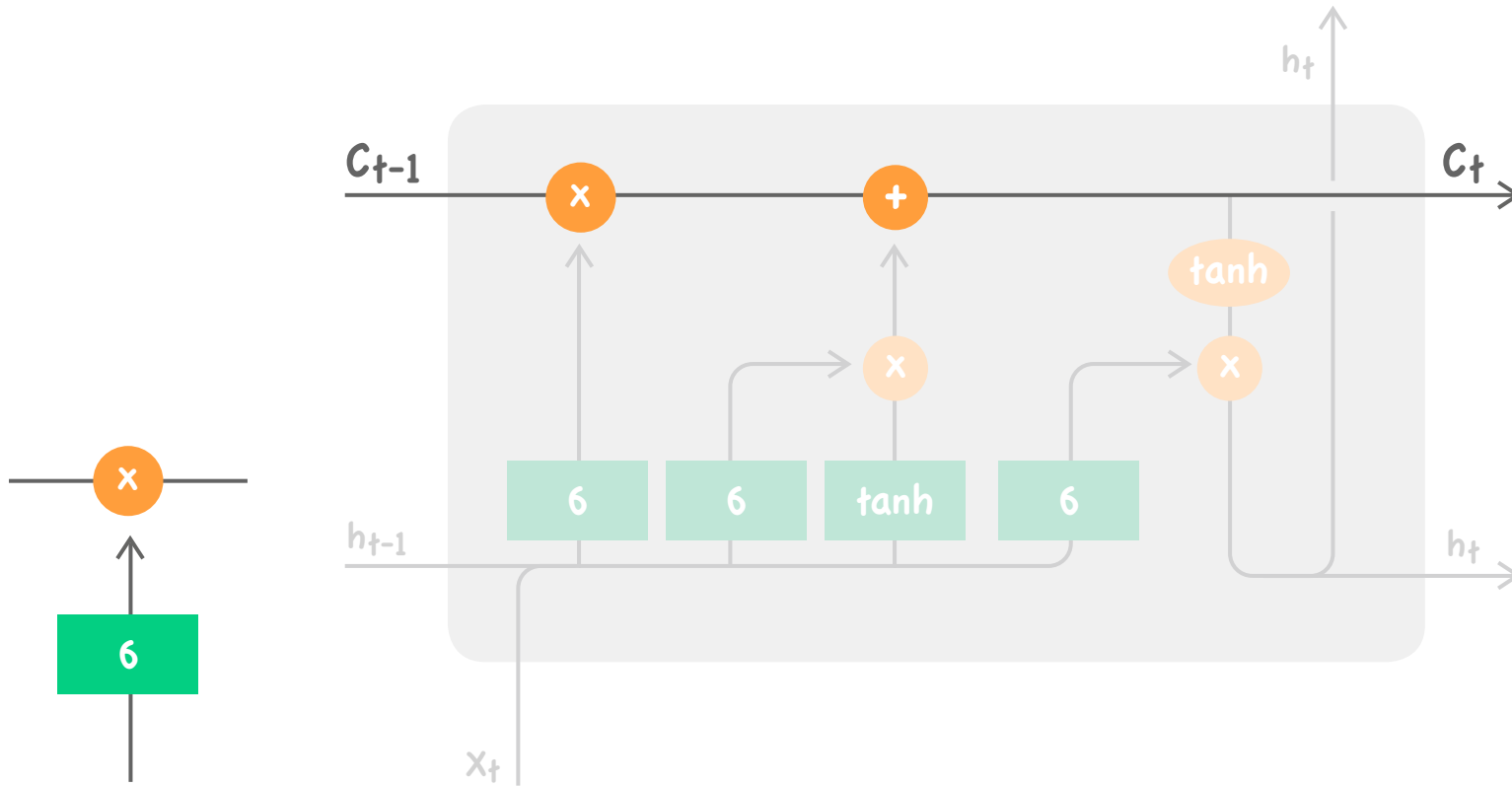
At a single time step : $h_t = \phi(W^{(hh)}h_{t-1} + W^{(hx)}x_{[t]})$

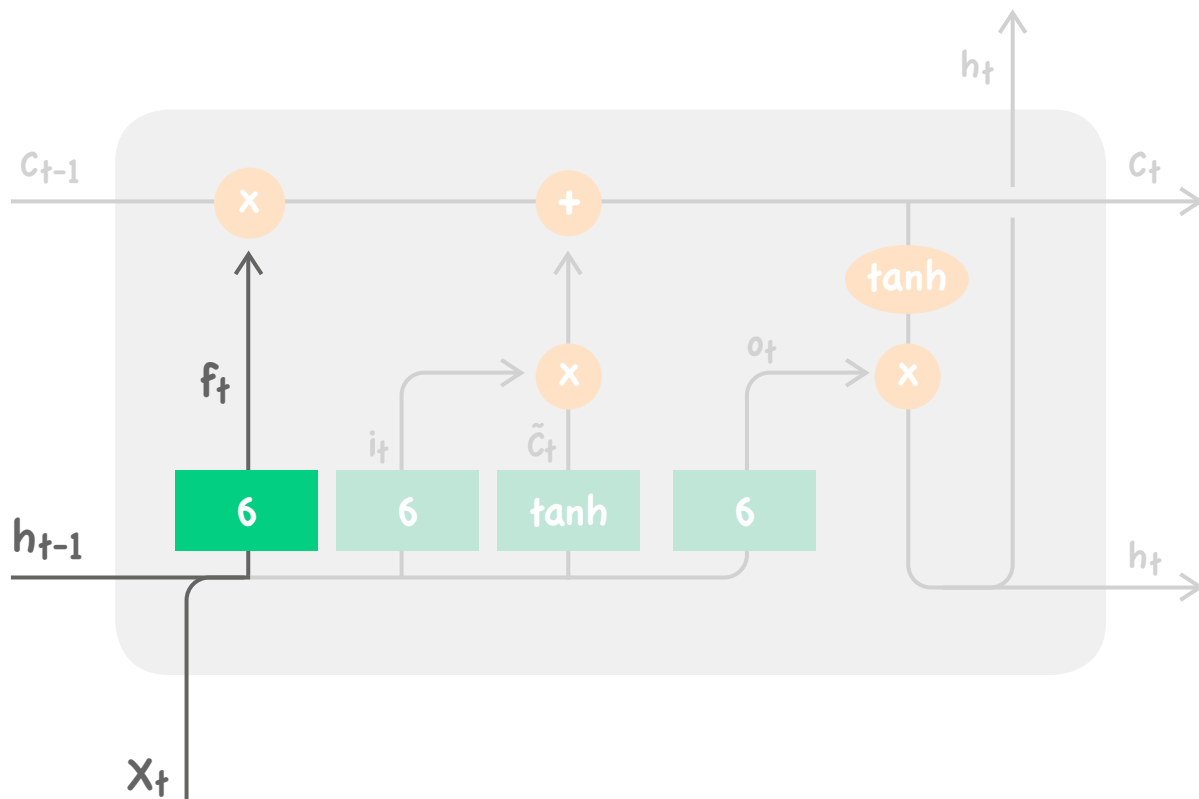
$\hat{y}_t = \text{softmax}(W^{(s)}h_t)$

$\hat{P}(x_{t+1} = v_j | x_t, \dots, x_1) = \hat{y}_{t,j}$

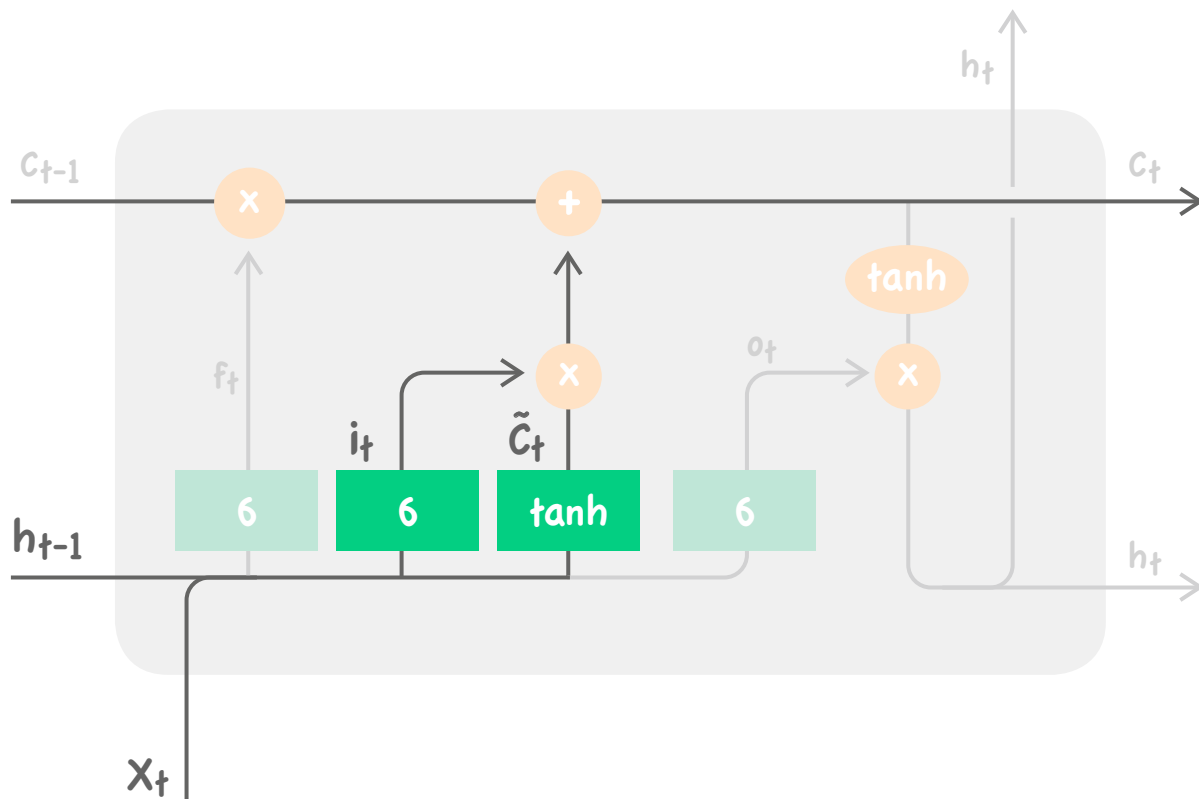


Gate



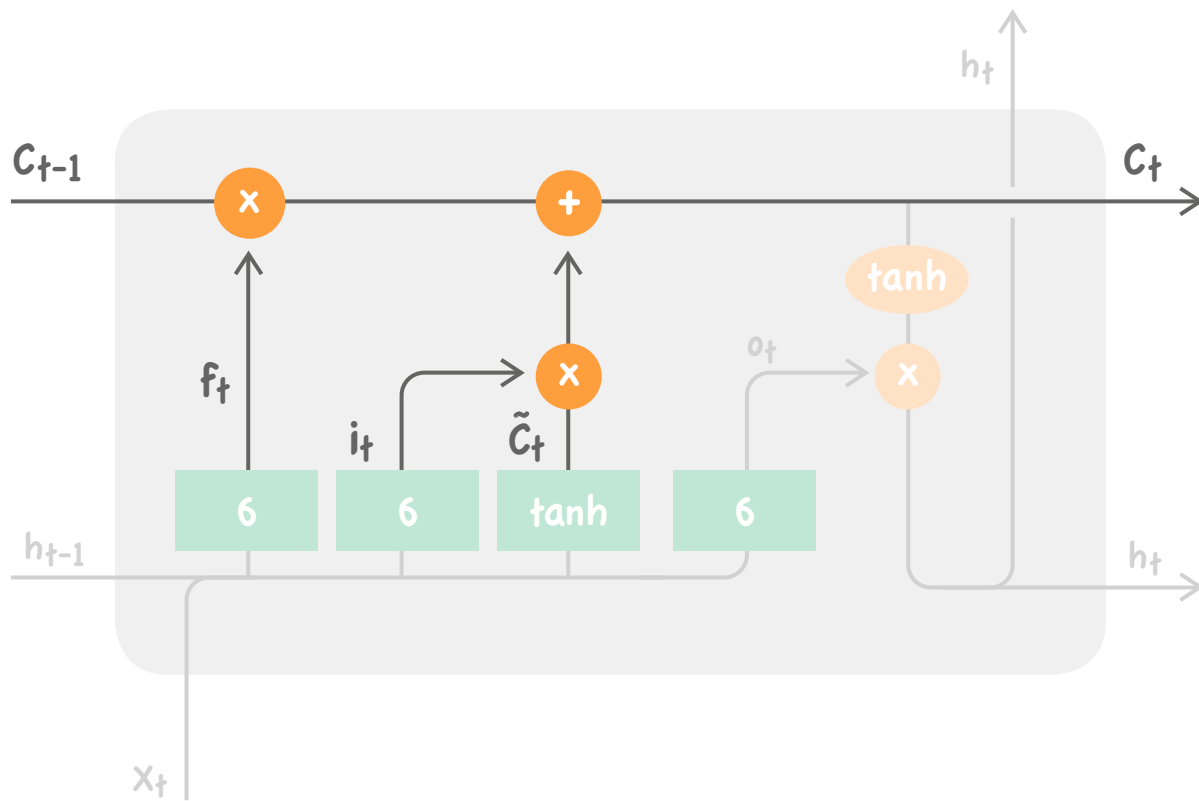


$$f_t = \sigma(W_f \cdot [h_{t-1}, x_t] + b_f)$$

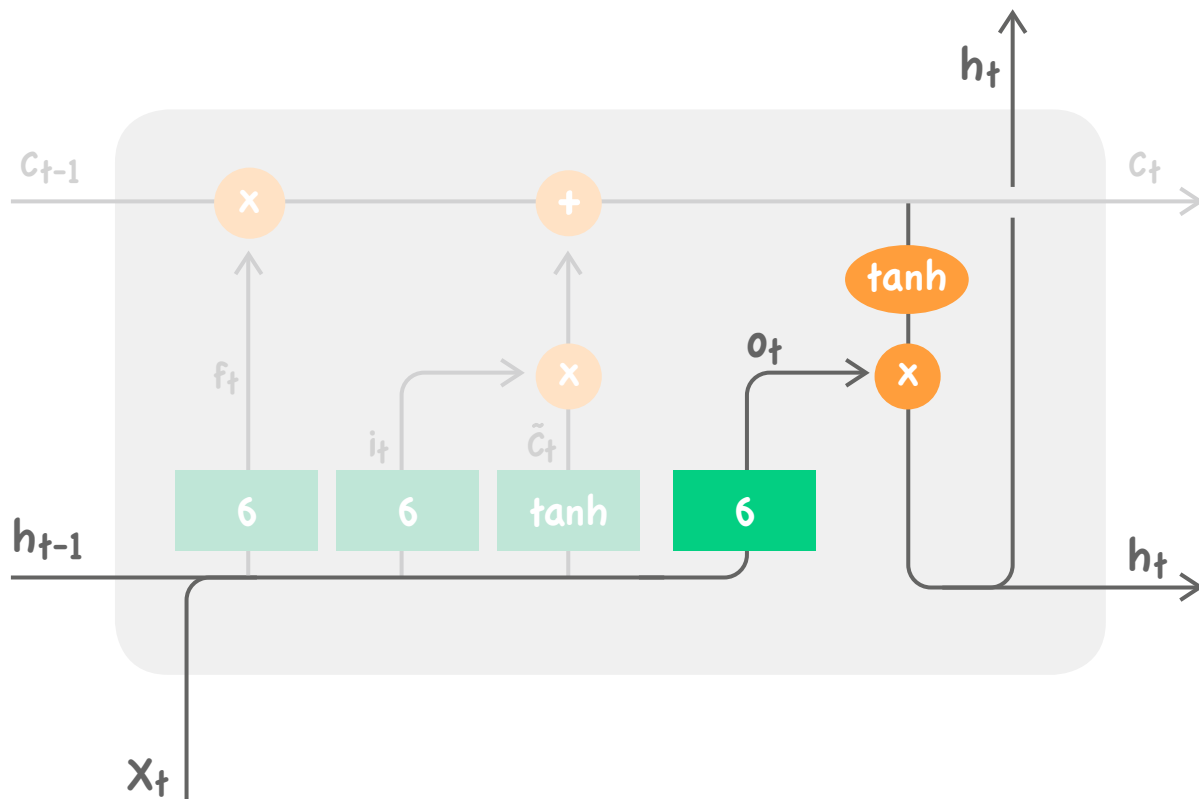


$$i_t = \sigma(W_i \cdot [h_{t-1}, x_t] + b_i)$$

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

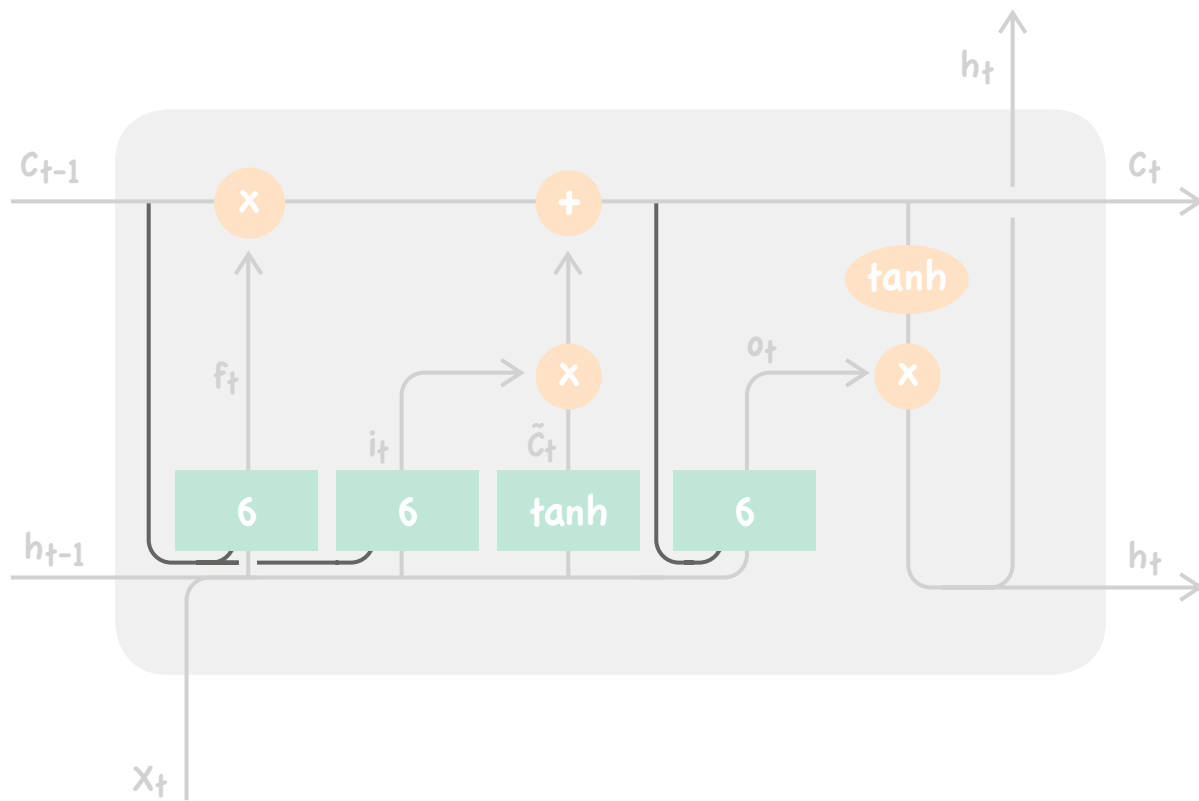


$$C_t = f_t * C_{t-1} + i_t * \tilde{C}_t$$



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

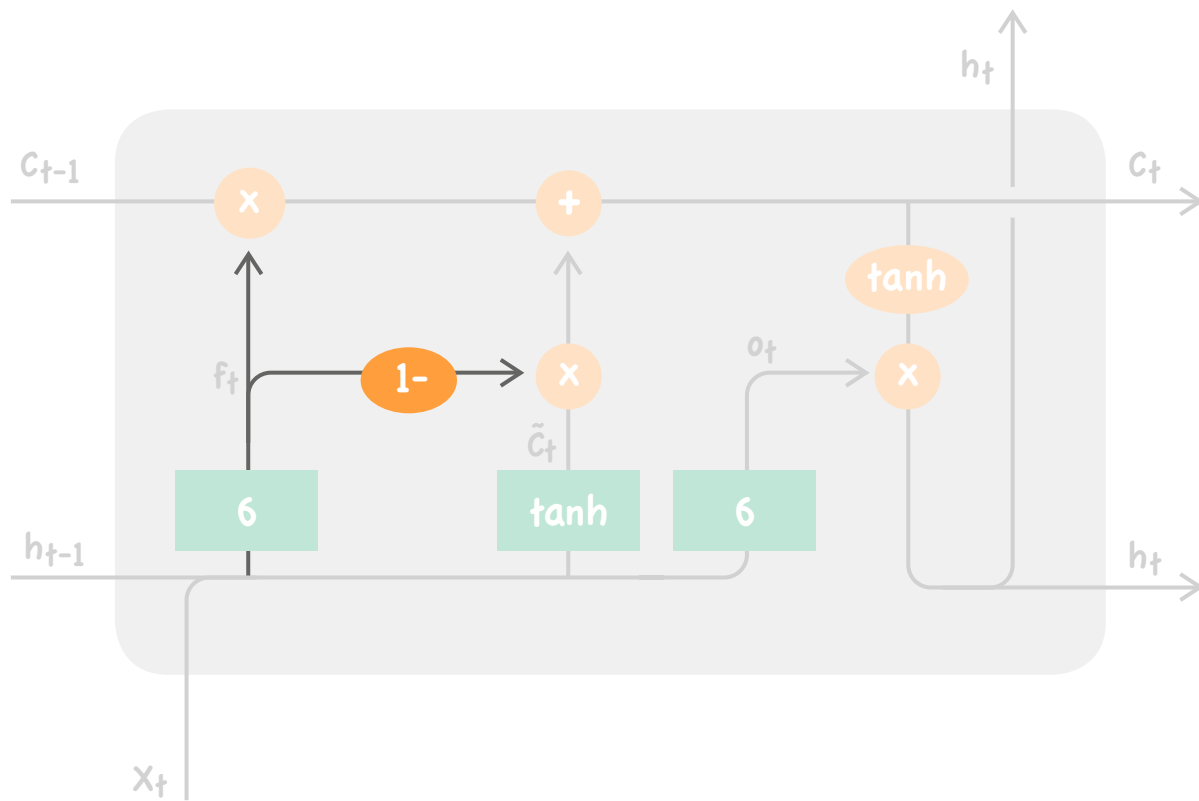
$$h_t = o_t * \tanh(C_t)$$



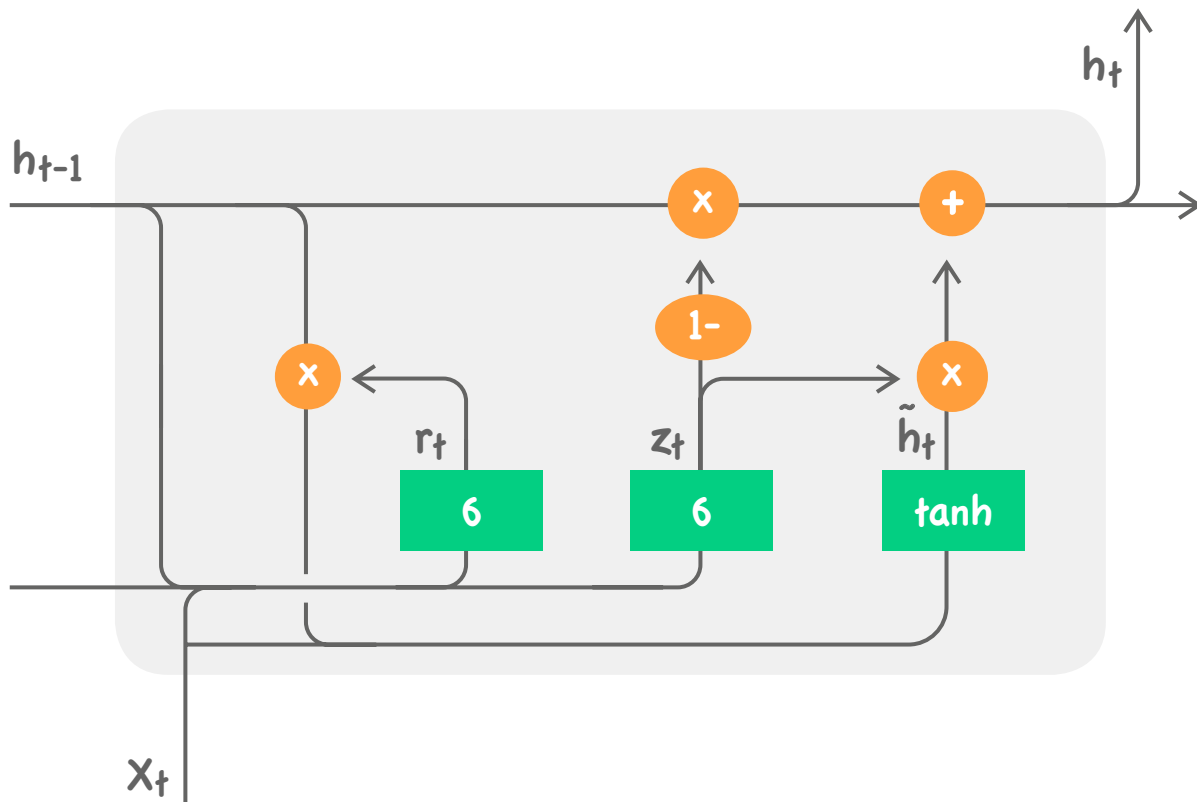
$$f_t = \sigma(W_f \cdot [C_{t-1}, h_{t-1}, X_t] + b_f)$$

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

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



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

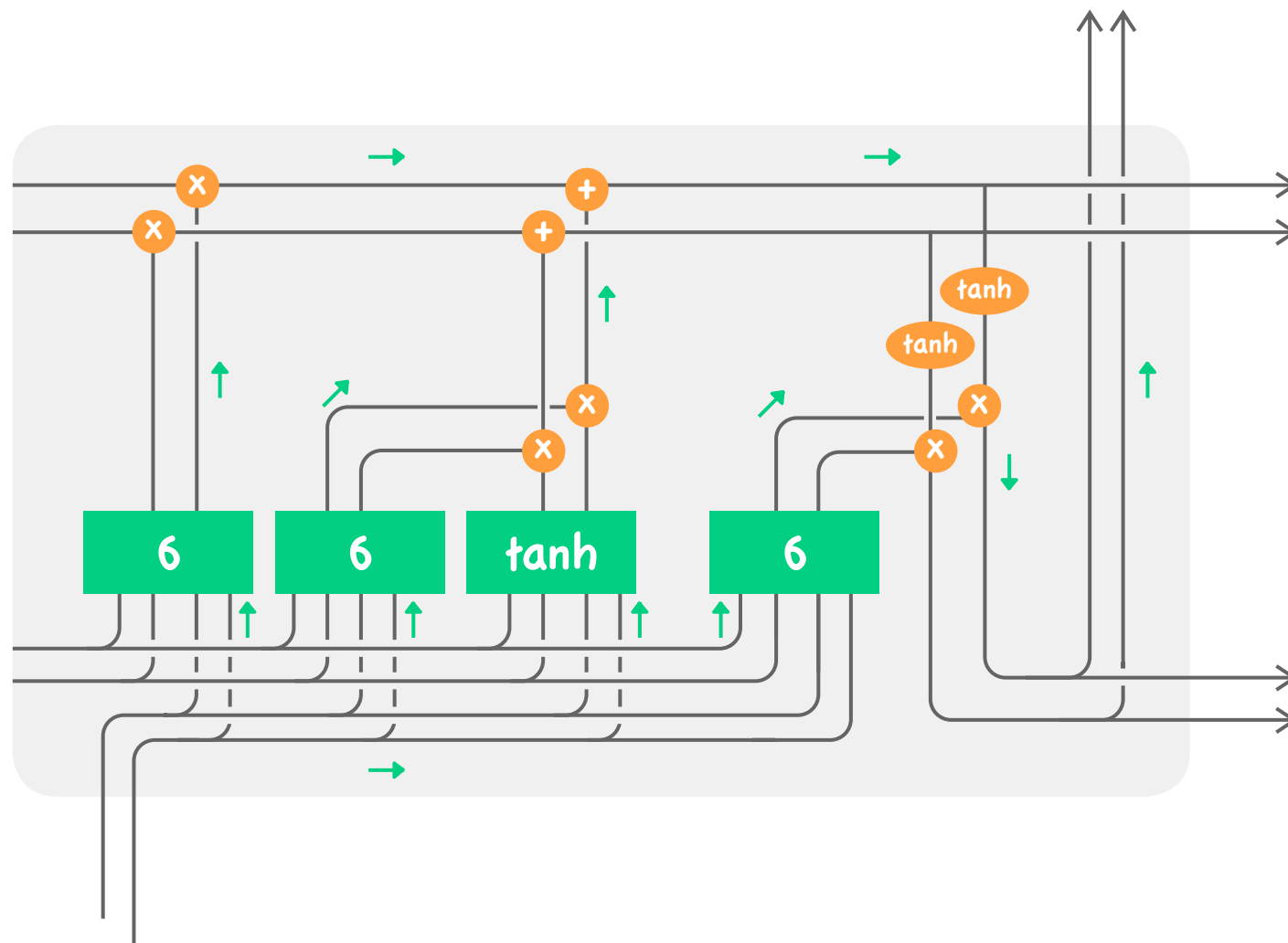


$$z_t = \sigma(W_z \cdot [h_{t-1}, x_t])$$

$$r_t = \sigma(W_r \cdot [h_{t-1}, x_t])$$

$$\tilde{h}_t = \tanh(W \cdot [r_t * h_{t-1}, x_t])$$

$$h_t = (1 - z_t) * h_{t-1} + z_t * \tilde{h}_t$$



NEXT LECTURE

RNN / LSTM CASE STUDY