## Gates and Losses

t=0,1,...,T temporal order

d=0,1,...,D spatial order of ConvLSTM module

**X** = input to ConvLSTM module

C = cell gate

 $\mathbf{H} = \text{hidden state}$ 

W = feedforward convolutional neural net

 $X_t^d$ , i.e. the output of ConvLSTM module at depth d and time t, depends on:

**X**<sub>t</sub><sup>d-1</sup> input to ConvLSTM module **H**<sup>d</sup><sub>t-1</sub> temporal hidden state

Cd<sub>t-1</sub> temporal cell gate

$$\begin{split} i_t^d &= \sigma(W_{d,xi}(\mathbf{X}_t^{d-1}) + W_{d,hi}(\mathbf{H}_{t-1}^d)), \\ f_t^d &= \sigma(W_{d,xf}(\mathbf{X}_t^{d-1}) + W_{d,hf}(\mathbf{H}_{t-1}^d)), \\ \tilde{C}_t^d &= tanh(W_{d,xc}(\mathbf{X}_t^{d-1}) + W_{d,hc}(\mathbf{H}_{t-1}^d)), \\ \mathbf{C}_t^d &= f_t^d \circ \mathbf{C}_{t-1}^d + i_t^d \circ \tilde{C}_t^d. \\ o_t^d &= \sigma(W_{d,xo}(\mathbf{X}_t^{d-1}) + W_{d,ho}(\mathbf{H}_{t-1}^d)), \\ \mathbf{H}_t^d &= o_t^d \circ tanh(\mathbf{C}_t^d), \\ \mathbf{X}_t^d &= \mathbf{H}_t^d, \end{split}$$

$$\mathbf{L} = \sum_{t=1}^{T} \gamma^t \mathbf{L_t}, \text{ where } \mathbf{L_t} = -log \frac{e^{\mathbf{H_t^D}[C]}}{\sum_j e^{\mathbf{H_t^D}[j]}}.$$