

BACKPROPAGATION IN CONVOLUTIONAL LSTMS

Ankur Handa and Viorica Pătrăucean

Department of Engineering

University of Cambridge

Cambridge, CB2 1PZ

{handa.ankur, vpatrauc}@gmail.com

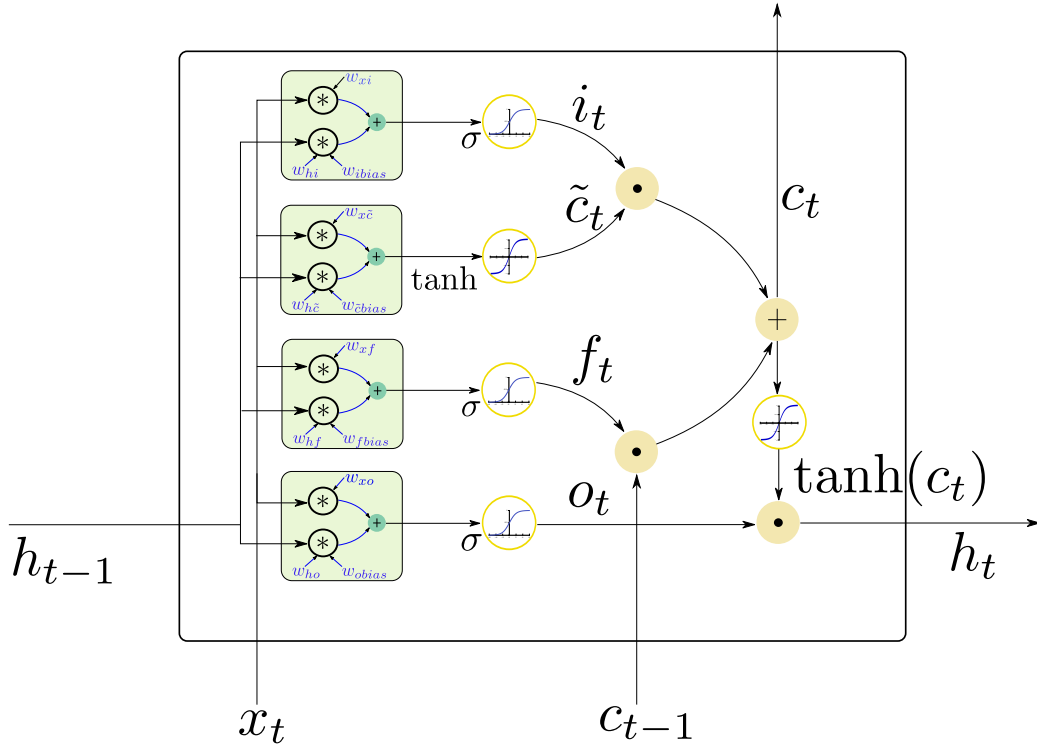


Figure 1: LSTM flow diagram. Note that h_0 and c_0 are initialised to zero. Loops and recursions are explicitly avoided for clarity.

Backpropagation derivations of convolutional LSTMs introduced in Stollenga et al. (2015); Xingjian et al. (2015); Pătrăucean et al. (2015); Romera-Paredes & Torr (2015)

1 CONVOLUTIONAL LSTM

$$\begin{aligned}
 i_t &= \sigma(x_t * w_{xi} + h_{t-1} * w_{hi} + w_{ibias}) \\
 f_t &= \sigma(x_t * w_{xf} + h_{t-1} * w_{hf} + w_{fbias}) \\
 \tilde{c}_t &= \tanh(x_t * w_{xc} + h_{t-1} * w_{hc} + w_{cbias}) \\
 o_t &= \sigma(x_t * w_{xo} + h_{t-1} * w_{ho} + w_{obias}) \\
 c_t &= \tilde{c}_t \odot i_t + c_{t-1} \odot f_t \\
 h_t &= o_t \odot \tanh(c_t)
 \end{aligned}$$

1.1 GIVEN: δh_t , FIND δo_t AND δc_t

1.1.1 $\frac{\delta E}{\delta o_t}$

$$\begin{aligned}\frac{\delta E}{\delta o_t} &= \frac{\delta E}{\delta h_t} \cdot \frac{\delta h_t}{\delta o_t} \\ \frac{\delta E}{\delta o_t} &= \frac{\delta E}{\delta h_t} \odot \tanh(c_t)\end{aligned}$$

1.1.2 $\frac{\delta E}{\delta c_t}$

$$\begin{aligned}\frac{\delta E}{\delta c_t} &= \frac{\delta E}{\delta h_t} \cdot \frac{\delta h_t}{\delta c_t} \\ \frac{\delta E}{\delta c_t} &= \frac{\delta E}{\delta h_t} \odot o_t \odot (1 - \tanh^2(c_t))\end{aligned}$$

1.2 GIVEN: δc_t , FIND δi_t , δf_t , $\delta \tilde{c}_t$, δc_{t-1} AND δh_{t-1}

1.2.1 $\frac{\delta E}{\delta i_t}$

$$\begin{aligned}\frac{\delta E}{\delta i_t} &= \frac{\delta E}{\delta c_t} \cdot \frac{\delta c_t}{\delta i_t} \\ \frac{\delta E}{\delta i_t} &= \frac{\delta E}{\delta c_t} \odot \tilde{c}_t\end{aligned}$$

1.2.2 $\frac{\delta E}{\delta f_t}$

$$\begin{aligned}\frac{\delta E}{\delta f_t} &= \frac{\delta E}{\delta c_t} \cdot \frac{\delta c_t}{\delta f_t} \\ \frac{\delta E}{\delta f_t} &= \frac{\delta E}{\delta c_t} \odot c_{t-1}\end{aligned}$$

1.2.3 $\frac{\delta E}{\delta \tilde{c}_t}$

$$\begin{aligned}\frac{\delta E}{\delta \tilde{c}_t} &= \frac{\delta E}{\delta c_t} \cdot \frac{\delta c_t}{\delta \tilde{c}_t} \\ \frac{\delta E}{\delta \tilde{c}_t} &= \frac{\delta E}{\delta c_t} \odot i_t\end{aligned}$$

1.2.4 $\frac{\delta E}{\delta c_{t-1}}$

$$\begin{aligned}\frac{\delta E}{\delta c_{t-1}} &= \frac{\delta E}{\delta c_t} \cdot \frac{\delta c_t}{\delta c_{t-1}} \\ \frac{\delta E}{\delta c_{t-1}} &= \frac{\delta E}{\delta c_t} \odot f_t\end{aligned}$$

1.2.5 $\frac{\delta E}{\delta h_{t-1}}$

$$\frac{\delta E}{\delta h_{t-1}} = \frac{\delta E}{\delta i_t} \cdot \frac{\delta i_t}{\delta h_{t-1}} + \frac{\delta E}{\delta o_t} \cdot \frac{\delta o_t}{\delta h_{t-1}} + \frac{\delta E}{\delta f_t} \cdot \frac{\delta f_t}{\delta h_{t-1}} + \frac{\delta E}{\delta \tilde{c}_t} \cdot \frac{\delta \tilde{c}_t}{\delta h_{t-1}}$$

1.3 GIVEN: δo_t , FIND δw_{xo} , δw_{ho} AND δw_{bias}

1.3.1 $\frac{\delta E}{\delta w_{xo}}$

$$\begin{aligned}\frac{\delta E}{\delta w_{xo}} &= \sum_{t=1}^T \frac{\delta E}{\delta o_t} \cdot \frac{\delta o_t}{\delta w_{xo}} \\ \frac{\delta E}{\delta w_{xo}} &= \sum_{t=1}^T \left(\frac{\delta E}{\delta o_t} \odot \sigma'(o_t) \right) * x_t\end{aligned}$$

1.3.2 $\frac{\delta E}{\delta w_{ho}}$

$$\begin{aligned}\frac{\delta E}{\delta w_{ho}} &= \sum_{t=1}^T \frac{\delta E}{\delta o_t} \cdot \frac{\delta o_t}{\delta w_{ho}} \\ \frac{\delta E}{\delta w_{ho}} &= \sum_{t=1}^T \left(\frac{\delta E}{\delta o_t} \odot \sigma'(o_t) \right) * h_{t-1}\end{aligned}$$

1.3.3 $\frac{\delta E}{\delta w_{bias}}$

$$\begin{aligned}\frac{\delta E}{\delta w_{bias}} &= \sum_{t=1}^T \frac{\delta E}{\delta o_t} \cdot \frac{\delta o_t}{\delta w_{bias}} \\ \frac{\delta E}{\delta w_{bias}} &= \sum_{t=1}^T \left(\frac{\delta E}{\delta o_t} \odot \sigma'(o_t) \right)\end{aligned}$$

1.4 GIVEN: δi_t , FIND δw_{xi} , δw_{hi} AND δw_{bias}

1.4.1 $\frac{\delta E}{\delta w_{xi}}$

$$\begin{aligned}\frac{\delta E}{\delta w_{xi}} &= \sum_{t=1}^T \frac{\delta E}{\delta i_t} \cdot \frac{\delta i_t}{\delta w_{xi}} \\ \frac{\delta E}{\delta w_{xi}} &= \sum_{t=1}^T \left(\frac{\delta E}{\delta i_t} \odot \sigma'(i_t) \right) * x_t\end{aligned}$$

1.4.2 $\frac{\delta E}{\delta w_{hi}}$

$$\begin{aligned}\frac{\delta E}{\delta w_{hi}} &= \sum_{t=1}^T \frac{\delta E}{\delta i_t} \cdot \frac{\delta i_t}{\delta w_{hi}} \\ \frac{\delta E}{\delta w_{hi}} &= \sum_{t=1}^T \left(\frac{\delta E}{\delta i_t} \odot \sigma'(i_t) \right) * h_{t-1}\end{aligned}$$

1.4.3 $\frac{\delta E}{\delta w_{bias}}$

$$\begin{aligned}\frac{\delta E}{\delta w_{bias}} &= \sum_{t=1}^T \frac{\delta E}{\delta i_t} \cdot \frac{\delta i_t}{\delta w_{bias}} \\ \frac{\delta E}{\delta w_{bias}} &= \sum_{t=1}^T \left(\frac{\delta E}{\delta i_t} \odot \sigma'(i_t) \right)\end{aligned}$$

1.5 GIVEN: δf_t , FIND δw_{xf} , δw_{hf} AND δw_{fbias}

1.5.1 $\frac{\delta E}{\delta w_{xf}}$

$$\begin{aligned}\frac{\delta E}{\delta w_{xi}} &= \sum_{t=1}^T \frac{\delta E}{\delta i_t} \cdot \frac{\delta i_t}{\delta w_{xi}} \\ \frac{\delta E}{\delta w_{xi}} &= \sum_{t=1}^T \left(\frac{\delta E}{\delta i_t} \odot \sigma'(i_t) \right) * x_t\end{aligned}$$

1.5.2 $\frac{\delta E}{\delta w_{hf}}$

$$\begin{aligned}\frac{\delta E}{\delta w_{hf}} &= \sum_{t=1}^T \frac{\delta E}{\delta f_t} \cdot \frac{\delta f_t}{\delta w_{hi}} \\ \frac{\delta E}{\delta w_{hf}} &= \sum_{t=1}^T \left(\frac{\delta E}{\delta f_t} \odot \sigma'(f_t) \right) * h_{t-1}\end{aligned}$$

1.5.3 $\frac{\delta E}{\delta w_{fbias}}$

$$\begin{aligned}\frac{\delta E}{\delta w_{fbias}} &= \sum_{t=1}^T \frac{\delta E}{\delta f_t} \cdot \frac{\delta f_t}{\delta w_{fbias}} \\ \frac{\delta E}{\delta w_{fbias}} &= \sum_{t=1}^T \left(\frac{\delta E}{\delta f_t} \odot \sigma'(f_t) \right)\end{aligned}$$

1.6 GIVEN: $\delta \tilde{c}_t$, FIND $\delta w_{x\tilde{c}}$, $\delta w_{h\tilde{c}}$ AND $\delta w_{\tilde{c}bias}$

1.6.1 $\frac{\delta E}{\delta w_{x\tilde{c}}}$

$$\begin{aligned}\frac{\delta E}{\delta w_{x\tilde{c}}} &= \sum_{t=1}^T \frac{\delta E}{\delta \tilde{c}_t} \cdot \frac{\delta \tilde{c}_t}{\delta w_{x\tilde{c}}} \\ \frac{\delta E}{\delta w_{x\tilde{c}}} &= \sum_{t=1}^T \left(\frac{\delta E}{\delta \tilde{c}_t} \odot (1 - \tanh^2(\tilde{c}_t)) \right) * x_t\end{aligned}$$

1.6.2 $\frac{\delta E}{\delta w_{h\tilde{c}}}$

$$\begin{aligned}\frac{\delta E}{\delta w_{h\tilde{c}}} &= \sum_{t=1}^T \frac{\delta E}{\delta \tilde{c}_t} \cdot \frac{\delta \tilde{c}_t}{\delta w_{h\tilde{c}}} \\ \frac{\delta E}{\delta w_{h\tilde{c}}} &= \sum_{t=1}^T \left(\frac{\delta E}{\delta \tilde{c}_t} \odot (1 - \tanh^2(\tilde{c}_t)) \right) * h_{t-1}\end{aligned}$$

1.6.3 $\frac{\delta E}{\delta w_{\tilde{c}bias}}$

$$\begin{aligned}\frac{\delta E}{\delta w_{\tilde{c}bias}} &= \sum_{t=1}^T \frac{\delta E}{\delta f_t} \cdot \frac{\delta f_t}{\delta w_{fbias}} \\ \frac{\delta E}{\delta w_{\tilde{c}bias}} &= \sum_{t=1}^T \left(\frac{\delta E}{\delta \tilde{c}_t} \odot (1 - \tanh^2(\tilde{c}_t)) \right)\end{aligned}$$

ACKNOWLEDGEMENTS

<http://arunmallia.github.io/writeups/nn/lstm/index.html#/>
[https://wiki.inf.ed.ac.uk/twiki/pub/MLforNLP/WebHome/
lstm_intro.pdf](https://wiki.inf.ed.ac.uk/twiki/pub/MLforNLP/WebHome/lstm_intro.pdf) FeedforwardSequentialMemoryNetworks:
ANewStructuretoLearnLong-termDependency

REFERENCES

- Patraucean, Viorica, Handa, Ankur, and Cipolla, Roberto. Spatio-temporal video autoencoder with differentiable memory. *CoRR*, abs/1511.06309, 2015. URL <http://arxiv.org/abs/1511.06309>.
- Romera-Paredes, Bernardino and Torr, Philip H. S. Recurrent instance segmentation. *CoRR*, abs/1511.08250, 2015.
- Stollenga, Marijn F, Byeon, Wonmin, Liwicki, Marcus, and Schmidhuber, Juergen. Parallel multi-dimensional lstm, with application to fast biomedical volumetric image segmentation. In *Advances in Neural Information Processing Systems*, 2015.
- Xingjian, SHI, Chen, Zhouong, Wang, Hao, Yeung, Dit-Yan, Wong, Wai-kin, and Woo, Wang-chun. Convolutional lstm network: A machine learning approach for precipitation nowcasting. In *Advances in Neural Information Processing Systems*, 2015.