

# Gated RNNs

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# How hard is it to train an RNN?

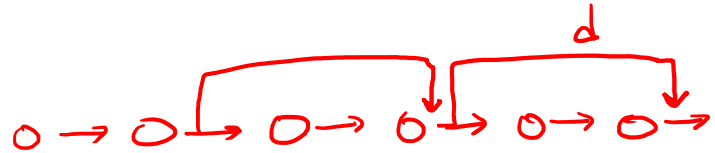
- Slow to train (TBPTT helps)
- RNN can suffer exploding/vanishing gradient
- First or early memory or info get lost through the time step

# Remedies

- ReLU activation function
- Truncated BPTT
- Clip gradient
- Use learning rate scheduling
- Add residual connection
- Change architectures- LSTM, GRU

# Long-term dependencies

- Skip connections



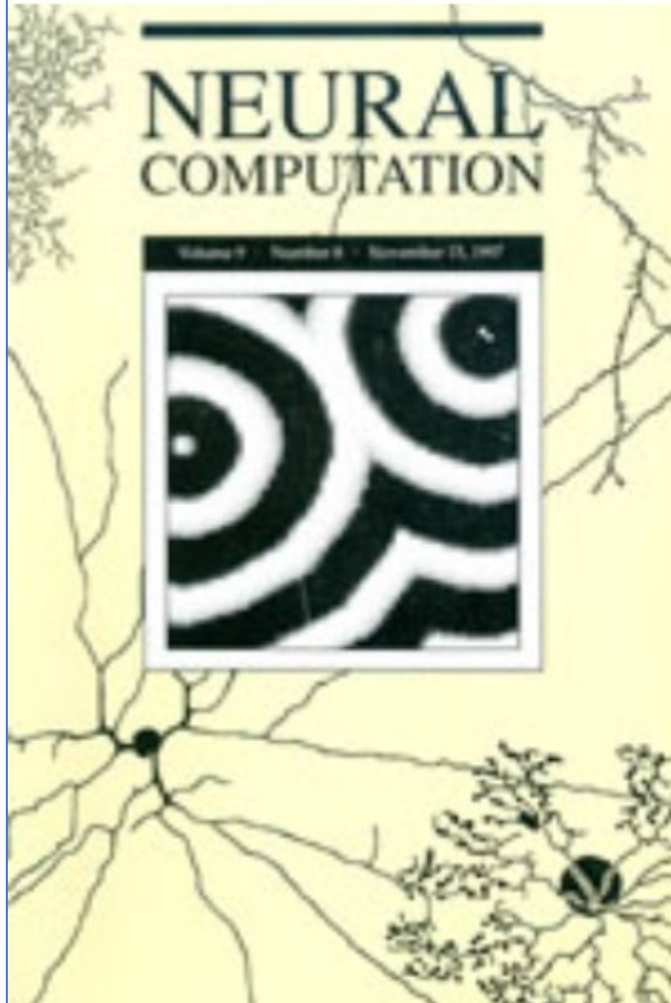
$$|W|^N \rightarrow |W|^{N/d}$$

- Leaky units

$$\underline{h_t = f(WX_t + Uh_{t-1})}$$

$$\underline{h_t = \alpha h_{t-1} + (1-\alpha) \cdot f(WX_t + Uh_{t-1})}$$

# Long Short-Term Memory cell



## Long Short-Term Memory

Sepp Hochreiter and Jürgen Schmidhuber

Posted Online March 13, 2006

<https://doi.org/10.1162/neco.1997.9.8.1735>

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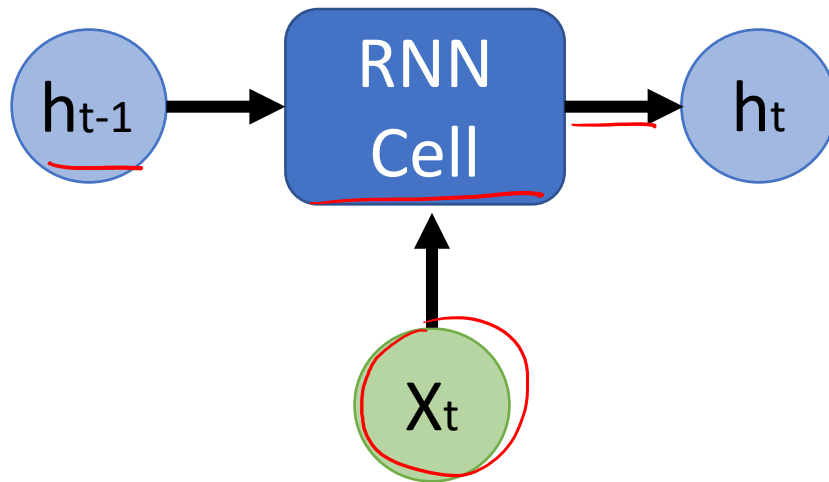
### Neural Computation

Volume 9 | Issue 8 | November 15, 1997

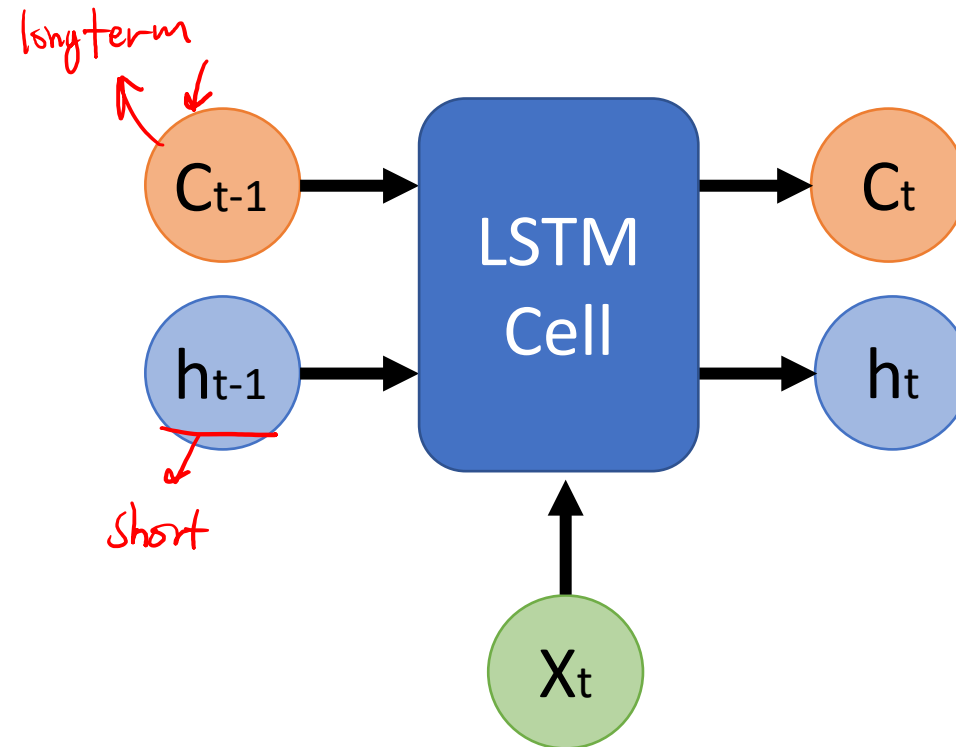
p.1735-1780

# What is LSTM cell?

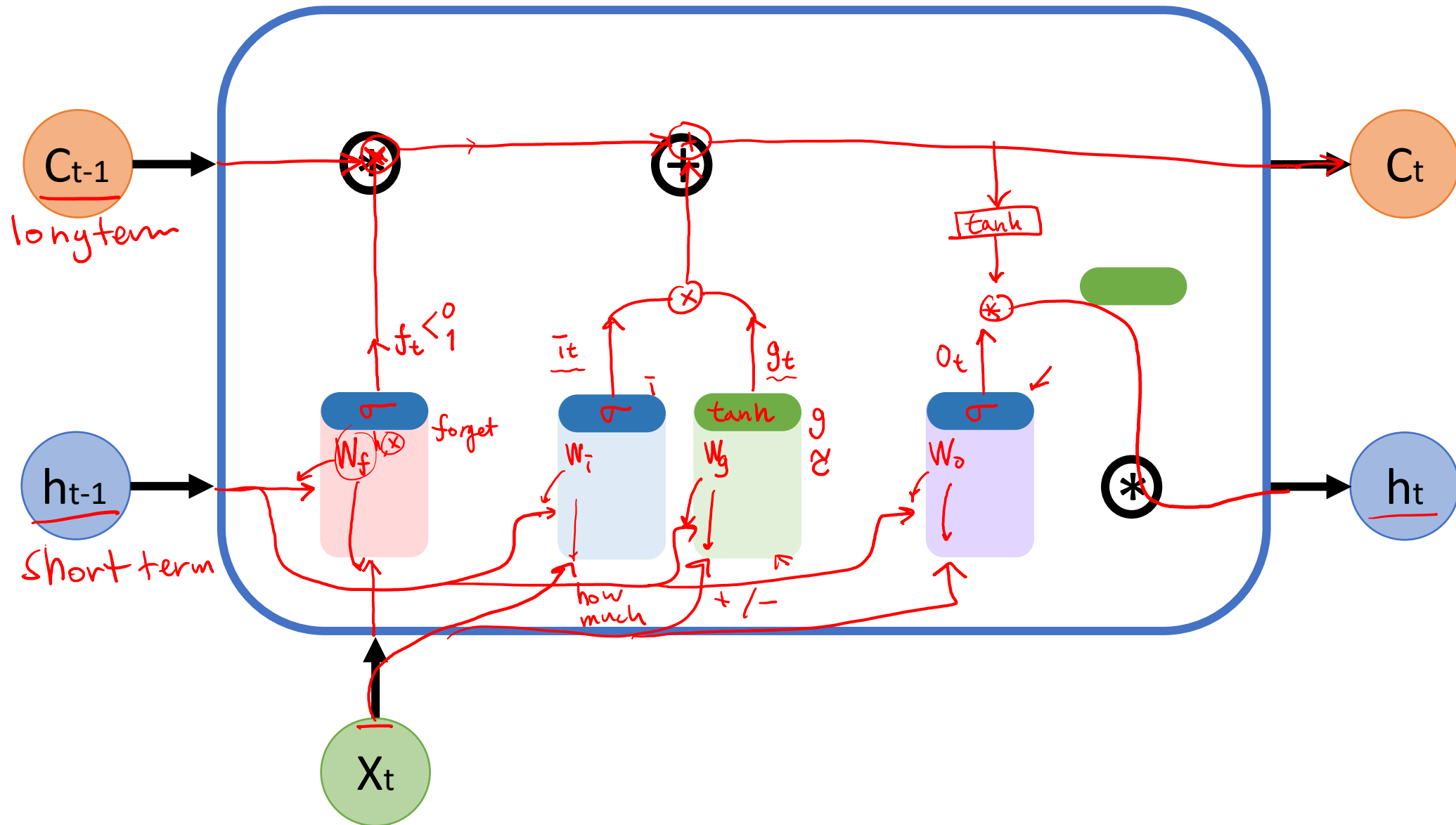
A Vania RNN cell



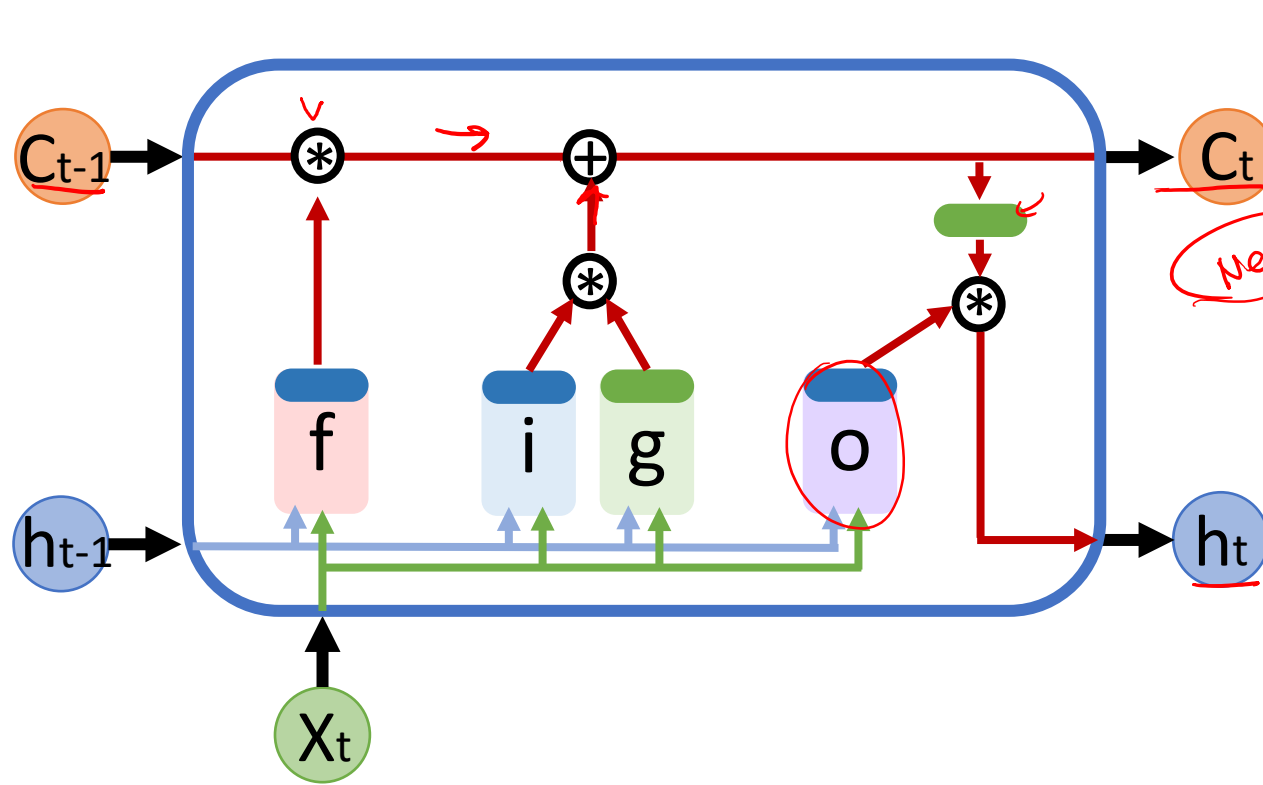
An LSTM cell



# Inside the LSTM cell



# Inside the LSTM cell



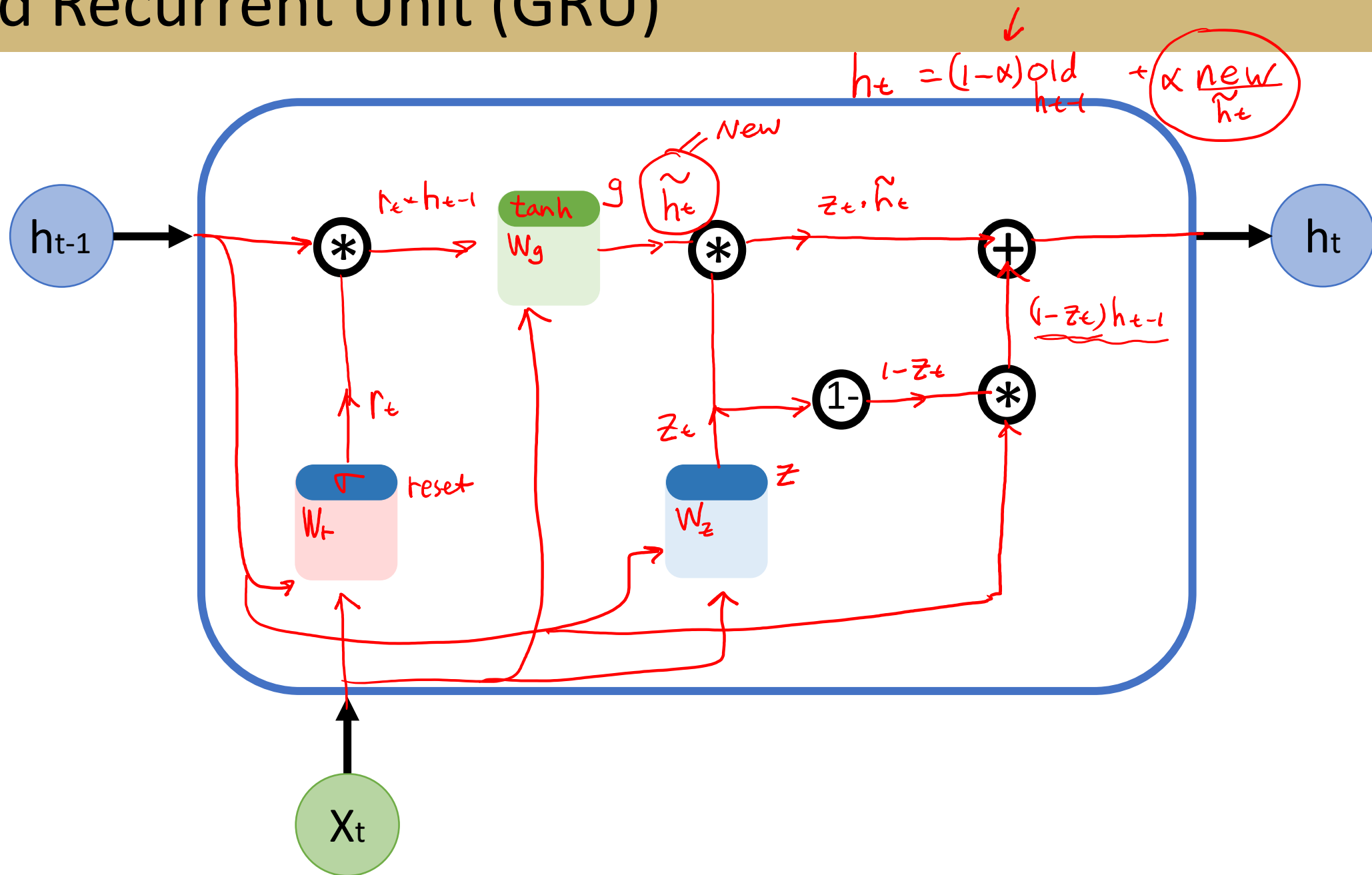
$$\begin{aligned} f_t &= \sigma(W_f \cdot [X_t, h_{t-1}] + b_f) \\ i_t &= \sigma(W_i \cdot [X_t, h_{t-1}] + b_i) \\ g_t &= \tanh(W_g \cdot [X_t, h_{t-1}] + b_g) \\ o_t &= \sigma(W_o \cdot [X_t, h_{t-1}] + b_o) \end{aligned}$$

$$c_t = f_t * c_{t-1} + i_t * g_t$$

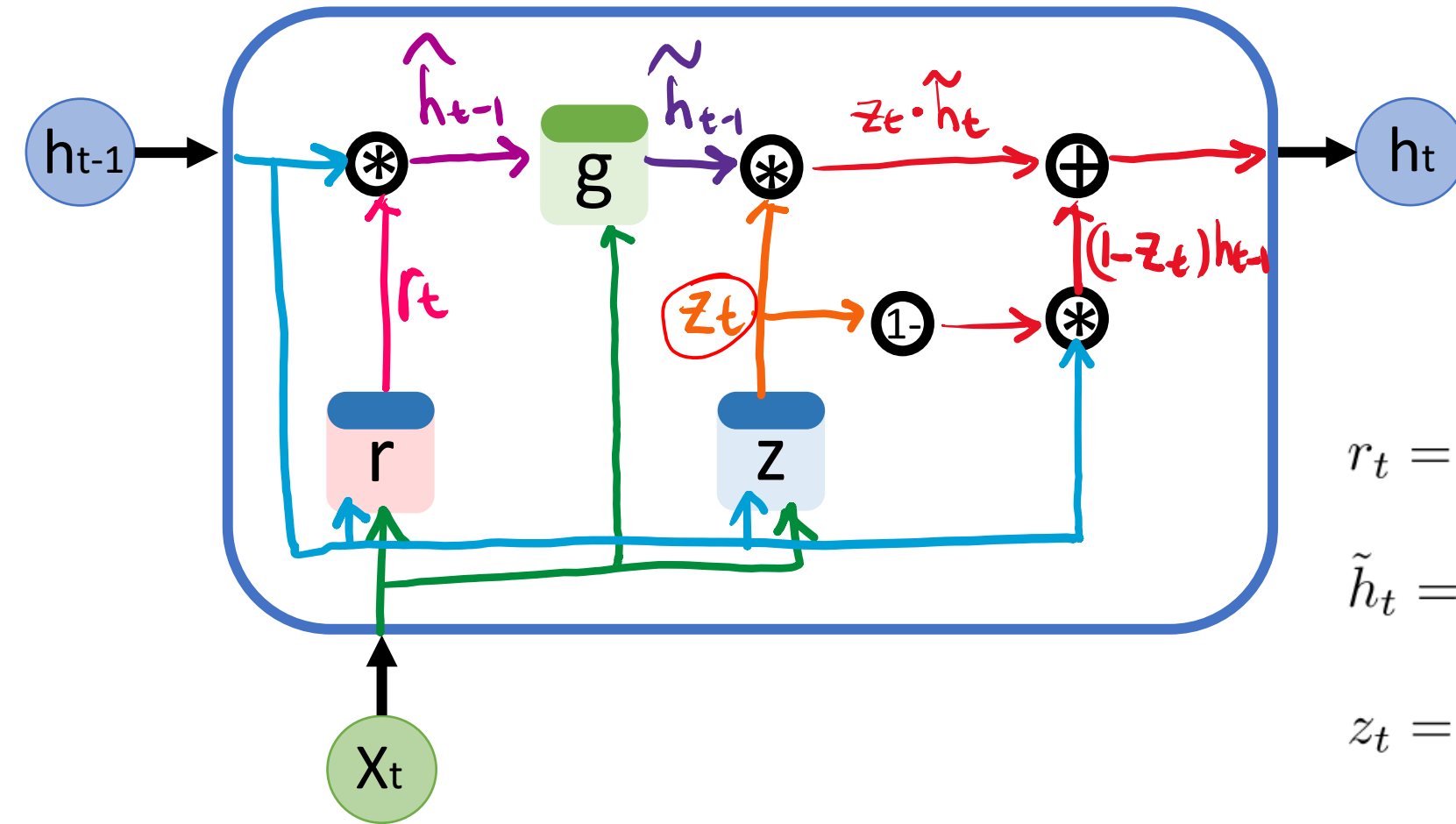
$$h_t = o_t * \tanh(c_t)$$



# Gated Recurrent Unit (GRU)



# Gated Recurrent Unit (GRU)



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

$$\tilde{h}_t = \tanh(W_g \cdot [X_t, r_t * h_{t-1}] + b_g)$$

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

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