

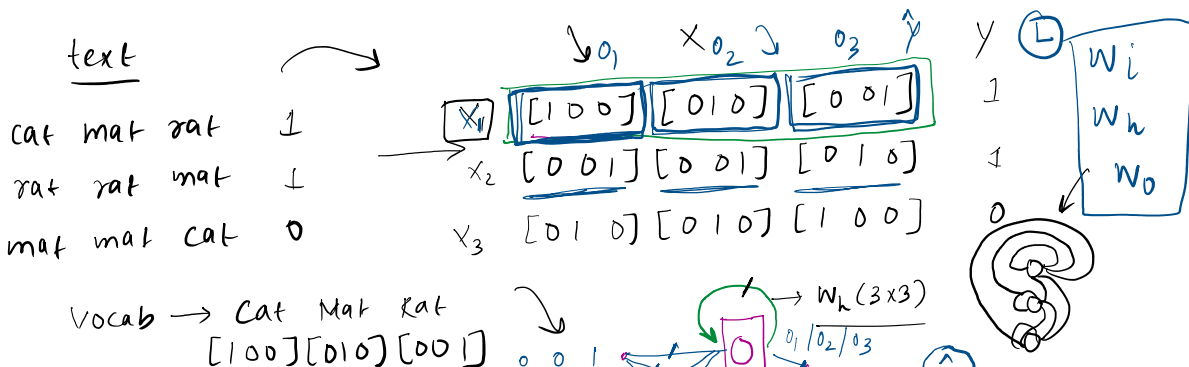
RNN Intro \rightarrow RNN \rightarrow practical \downarrow
types of RNN

\rightarrow RNN \rightarrow Backprop \rightarrow BPTT

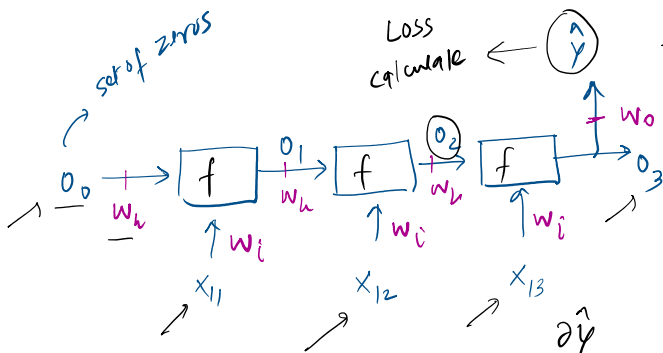
Many to One RNN
Sentiment Analysis

text \rightarrow 1/0

$\hat{y} \rightarrow \text{L}$



forward prop



$$o_1 = f(x_{i1}w_i + o_0w_h)$$

$$o_2 = f(x_{i2}w_i + o_1w_h)$$

$$o_3 = f(x_{i3}w_i + o_2w_h)$$

loss calculate \rightarrow minimize

L min gradient descent

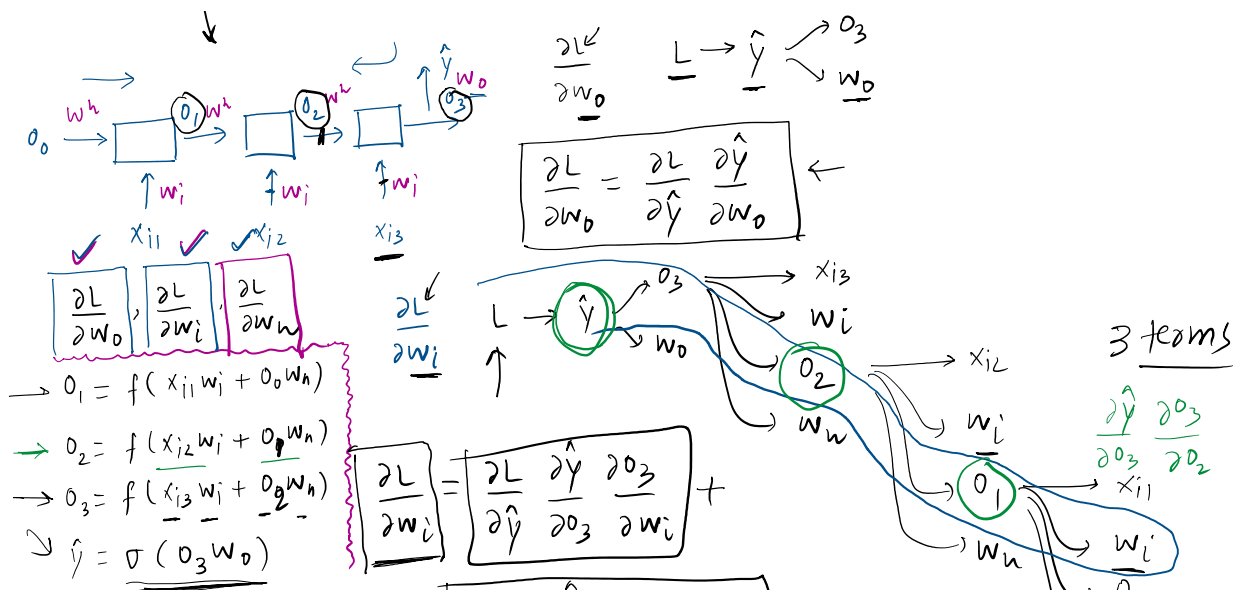
w_i, w_h, w_o

$$\begin{aligned} w_i &= w_i - \eta \frac{\partial L}{\partial w_i} \\ w_h &= w_h - \eta \frac{\partial L}{\partial w_h} \\ w_o &= w_o - \eta \frac{\partial L}{\partial w_o} \end{aligned}$$

Backprop

unfold \rightarrow 3 times





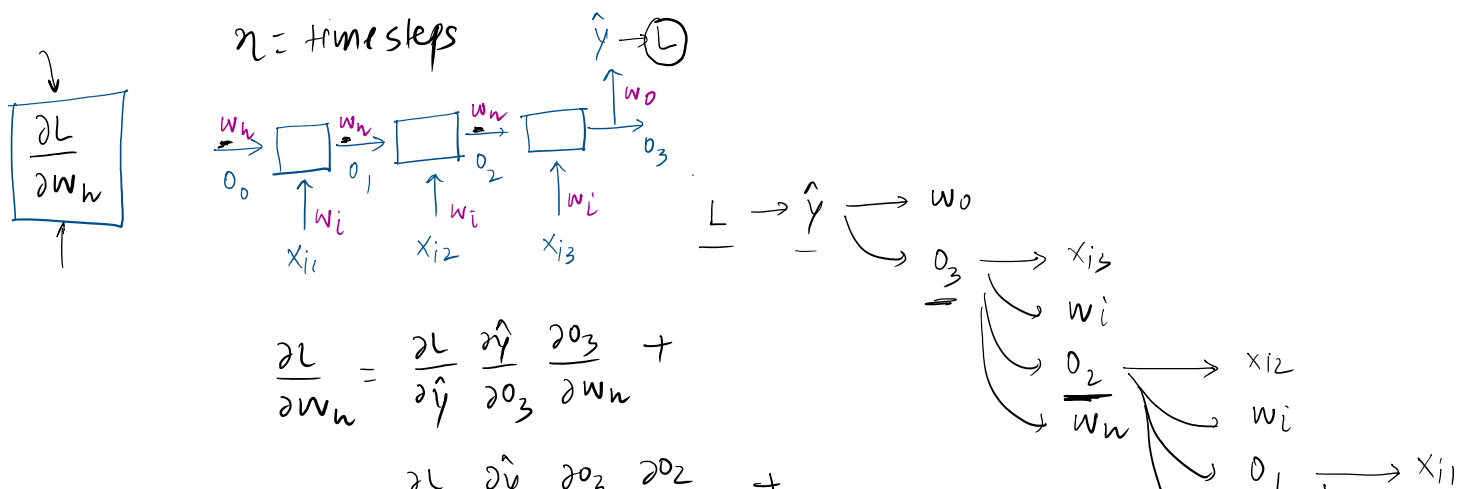
10 words
→ 10 hidden states
t=1

$$\frac{\partial L}{\partial w_i} = \sum_{j=1}^3 \frac{\partial L}{\partial \hat{y}} \frac{\partial \hat{y}}{\partial o_j} \frac{\partial o_j}{\partial w_i}$$

$$= \frac{\partial L}{\partial \hat{y}} \left[\frac{\partial \hat{y}}{\partial o_1} \frac{\partial o_1}{\partial w_i} + \frac{\partial \hat{y}}{\partial o_2} \frac{\partial o_2}{\partial w_i} + \frac{\partial \hat{y}}{\partial o_3} \frac{\partial o_3}{\partial w_i} \right]$$

$$\frac{\partial L}{\partial \hat{y}} \left[\frac{\partial \hat{y}}{\partial o_2} \frac{\partial o_2}{\partial w_i} \right] = \frac{\partial L}{\partial \hat{y}} \frac{\partial \hat{y}}{\partial o_2} \frac{\partial o_2}{\partial w_i}$$

$$\frac{\partial L}{\partial w_i} = \sum_{j=1}^n \frac{\partial L}{\partial \hat{y}} \frac{\partial \hat{y}}{\partial o_j} \frac{\partial o_j}{\partial w_i}$$



$$\frac{\partial \mathcal{L}}{\partial w_h} \quad \text{of} \quad \frac{\partial \hat{y}}{\partial o_3} \quad \dots$$

$$\frac{\partial \mathcal{L}}{\partial \hat{y}} \frac{\partial \hat{y}}{\partial o_3} \frac{\partial o_3}{\partial o_2} \frac{\partial o_2}{\partial w_h} +$$

$$\frac{\partial \mathcal{L}}{\partial \hat{y}} \frac{\partial \hat{y}}{\partial o_3} \frac{\partial o_3}{\partial o_2} \frac{\partial o_2}{\partial o_1} \frac{\partial o_1}{\partial w_h}$$

$$\boxed{\frac{\partial \mathcal{L}}{\partial w_h} = \sum_{j=1}^n \frac{\partial \mathcal{L}}{\partial \hat{y}} \frac{\partial \hat{y}}{\partial o_j} \frac{\partial o_j}{\partial w_h}}$$

$\eta = \text{timesteps}$

for $j=3$

$$\frac{\partial \mathcal{L}}{\partial \hat{y}} \frac{\partial \hat{y}}{\partial o_3} \frac{\partial o_3}{\partial w_h} \rightarrow \frac{\partial \mathcal{L}}{\partial \hat{y}} \frac{\partial \hat{y}}{\partial o_3} \frac{\partial o_3}{\partial o_1} \frac{\partial o_1}{\partial w_h}$$

for $j=10$

$$\frac{\partial \mathcal{L}}{\partial \hat{y}} \frac{\partial \hat{y}}{\partial o_{10}} \frac{\partial o_{10}}{\partial o_1} \frac{\partial o_1}{\partial w_h}$$

$$\boxed{\begin{array}{c} j \\ \prod \\ t=2 \end{array} \frac{\partial o_t}{\partial o_{t-1}}}$$

$$\prod_{t=2}^{t=3} \frac{\partial o_t}{\partial o_{t-1}} = \frac{\partial o_2}{\partial o_1} \frac{\partial o_3}{\partial o_2}$$

$$o_t = f(x_{it} w_{inp} + o_{t-1} w_h)$$

$$\frac{\partial o_t}{\partial o_{t-1}} = \prod_{t=2}^j f'(x_{it} w_{inp} + o_{t-1} w_h) w_h$$

\uparrow \downarrow
 $[0-1]$

