

# 1 Parity, RNN

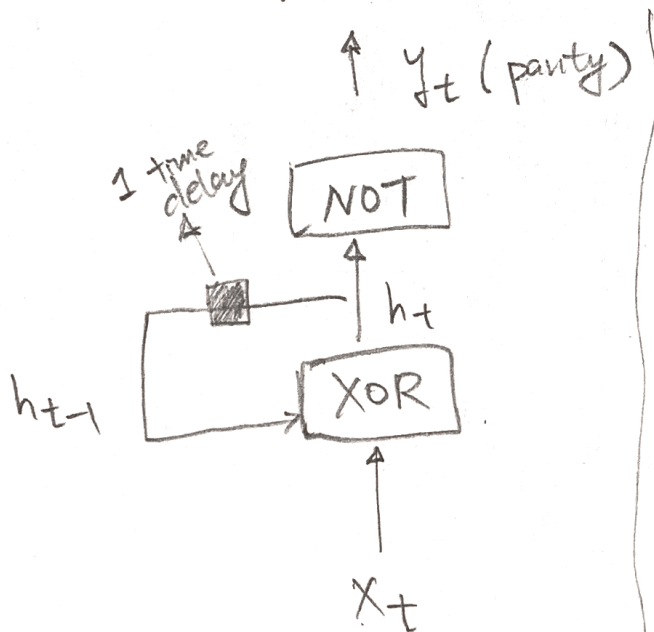
consider given example,  $X = 010110$

parity offset by 1	X	parity
$h_0$	0	1
1	1	0
0	0	0
0	1	1
1	1	0
0	0	0

this is  $XOR(c_{t-1}, X_t)$

if we set  $h_0 = 0$ , we need to use NOT before outputting.

So RNN layer has this structure



$y_t$	1	0	0	1	0	0
$X_t$	0	1	0	1	1	0
$h_{t-1}$	0	0	1	1	0	1
$h_t$	0	1	1	0	1	1

## 2 Parity, LSTM

want to find  $W_f, b_f, W_i, b_i, W_c, b_c, W_o, b_o$

s.t.  $C_t = \text{parity}$ .

if  $h_0 = 0$ , then  $C_t = \text{parity} = \text{XOR}(h_{t-1}, X_t)$

$$= (h_{t-1} \wedge \bar{X}_t) \vee (\bar{h}_{t-1} \wedge X_t) \quad (\text{from hint})$$

$$= h_{t-1} \cdot \bar{X}_t + \bar{h}_{t-1} \cdot X_t$$

$$= C_{t-1} \cdot f_t + i_t \cdot \bar{C}_t \quad (\text{eq 4})$$

Let's set

$$\begin{matrix} (a) & f_t = \bar{X}_t, & (b) & i_t = \bar{h}_{t-1}, & (c) & \bar{C}_t = X_t, & (d) & C_{t-1} = h_{t-1} \end{matrix}$$

$$(a) \quad f_t = \bar{X}_t = \sigma(W_f[h_{t-1}, X_t]) = \begin{cases} 1 & \text{if } W_f[h_{t-1}, 0] > 0 \\ 0 & \text{if } W_f[h_{t-1}, 1] \leq 0. \end{cases}$$

Letting  $W_f = \begin{bmatrix} 0 \\ -2 \end{bmatrix}$  and  $b_f = 1$

will satisfy the piecewise function

$$f_t = \bar{X}_t = 1 \quad \text{if} \quad 0(h_{t-1}) - 2(0) + 1 = (1 > 0) \equiv T$$

$$= \bar{X}_t = 0 \quad \text{if} \quad 0(h_{t-1}) - 2(1) + 1 = (-1 \leq 0) \equiv T$$

$$(b) \quad i_t = \overline{h_{t-1}} = \begin{cases} 1 & \text{if } w_i [h_{t-1}, x_t] > 0 \\ 0 & \text{if } w_i [h_{t-1}, x_t] \leq 0 \end{cases}$$

to satisfy above equation,

$$\text{set } \boxed{w_i = \begin{bmatrix} -2 \\ 0 \end{bmatrix} \quad b_i = 1}$$

$$\begin{aligned} i_t &= 1 \quad \text{if } (-2(0) + 0 + 1 > 0) \equiv (1 > 0) \equiv T \\ &= 0 \quad \text{if } (-2(1) + 0 + 1 \leq 0) \equiv (-1 \leq 0) \equiv T \end{aligned}$$

$$(c) \quad \tilde{c}_t = x_t = \begin{cases} 1 & \text{if } w_c [h_{t-1}, x_t] > 0 \\ 0 & \text{if } w_c [h_{t-1}, x_t] \leq 0 \end{cases}$$

$$\text{to satisfy thrs, set } \boxed{w_c = \begin{bmatrix} 0 \\ 1 \end{bmatrix} \quad b_c = 0}$$

$$\begin{aligned} \tilde{c}_t &= 1 \quad \text{if } (0(h_{t-1}) + (1)(1) > 0) \equiv (1 > 0) \equiv T \\ &= 0 \quad \text{if } (0(h_{t-1}) + (0)(0) \leq 0) \equiv (0 \leq 0) \equiv T \end{aligned}$$

$$(d) \quad \text{from eq 6, } h_t = o_t \times \tanh(c_t)$$

$$\text{if } o_t = 1 \quad \text{then } h_t = \tanh(c_t) = \begin{cases} 1 & \text{if } c_t = 1 \\ 0 & \text{if } c_t = 0 \end{cases} = c_t$$

(d continued).

if  $O_t = 1$ , then  $1 = \sigma(W_0 \cdot [h_{t-1}, X_t] + b_0)$

$$= \begin{cases} 1 & \text{if } W_0 [h_{t-1}, X_t] + b_0 > 0 \end{cases}$$

Setting  $\boxed{W_0 = \begin{bmatrix} 0 \\ 0 \end{bmatrix}, b_0 = 1}$  ensures

$O_t = 1$ , hence ensures  $h_t = C_t$ .

So, in order for (a) (b) (c) (d) to be true,

we have:

$$W_f = \begin{bmatrix} 0 \\ -2 \end{bmatrix}, b_f = 1$$

$$W_i = \begin{bmatrix} -2 \\ 0 \end{bmatrix}, b_i = 1$$

$$W_c = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, b_c = 0$$

$$W_0 = \begin{bmatrix} 0 \\ 0 \end{bmatrix}, b_0 = 1$$

### 3) Beam Search.

Want to Show:  $\forall i$ , if  $\text{best}_{\leq i}$  exists with score  $S$

and  $S$  is better than score for highest scoring item in  $B_i$

then  $\rightarrow \exists y''$ , with  $|y''| > |\text{best}_{\leq i}|$

s.t.  $y''$  has better score than  $S$ .

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$\forall i$ , suppose  $y = \text{best}_{\leq i}$ , with score  $S$ .

and  $y_{B_i}$  be the highest scoring item in  $B_i$

with score  $S_{B_i}$  s.t.  $S_{B_i} \leq S$ .

Let's try to grow  $y_{B_i}$  into  $y''$ , with  $j$  additional Beam Search steps.

Case 1  $\text{comp}(y_{B_i}) \equiv T$ , We cannot do additional steps

so  $y''$  does not exist.

Case 2  $\text{comp}(y_{B_i}) \equiv F$ , We can grow w/  $j$  more steps.

So,  $y'' = y_{B_i} \circ y_{i+1} \circ y_{i+2} \circ \dots \circ y_{i+j}$

Consider the score for  $y''$ . Call it  $S''$

$$S'' = S_{B_i} \cdot S_{i+1} \cdot \dots \cdot S_{i+j}$$

$$= S_{B_i} \cdot P(y_{i+1} | X, y_{B_i}) \cdot P(y_{i+2} | X, y_{B_i} \circ y_{i+1}) \\ \cdot \dots \cdot P(y_{i+j} | X, y_{B_i} \circ y_{i+1} \circ \dots \circ y_{i+j-1})$$

Note: the maximum possible values for  $S_{i+1} \dots S_{i+j}$  is 1, since they are probabilities.

$$\text{thus, the max value for } S'' = S_{B_i} \cdot 1^j = S_{B_i} \\ \leq S_{B_i}$$

So,  $y''$  does not have score higher than  $S_{B_i}$ .

#### 4 | Exploding Gradients.

$$h_t = W^T h_{t-1} = W^T W^T h_{t-2} = \dots = (W^T)^t h_0$$

$W$  can be decomposed, i.e.

$$W = Q \Lambda Q^{-1}, \text{ where } \Lambda \text{ is a diagonal}$$

matrix with elements  $\lambda_1, \dots, \lambda_n$ .  $\Lambda = \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix}$ .

let  $p = t$

$$h_p = (W^T)^p h_0 \quad \text{with } p \gg 0$$

$$= \left( (Q \Lambda Q^{-1})^T \right)^p h_0 = \left( (Q^{-1})^T \Lambda^T Q^T \right)^p h_0$$

$$= \left( (Q^T)^{-1} \Lambda^T Q^T \right)^p h_0$$

$$= \left( \underbrace{(Q^T)^{-1} \Lambda^T Q^T}_{\mathbb{I}} \underbrace{(Q^T)^{-1} \Lambda^T}_{\mathbb{I}} \dots \underbrace{(Q^T)^{-1} \Lambda^T}_{\mathbb{I}} \right)$$

$$= (Q^T)^{-1} \Lambda^p Q^T = h_p$$



Consider  $\Lambda^P = \begin{bmatrix} \lambda_1^P & & 0 \\ & \ddots & \\ 0 & & \lambda_n^P \end{bmatrix}$

if  $\rho(w) > 1$ , then  $(\rho(w))^P$  explodes with  $P \gg 0$

if  $\rho(w) < 1$ , then  $(\rho(w))^P$  vanishes with  $P \gg 0$

$\Lambda^P$  contains  $(\rho(w))^P$ , so  $\Lambda^P$  explodes/vanish accordingly.

as does  $h_P = (Q^T)^T \Lambda^P Q^T$