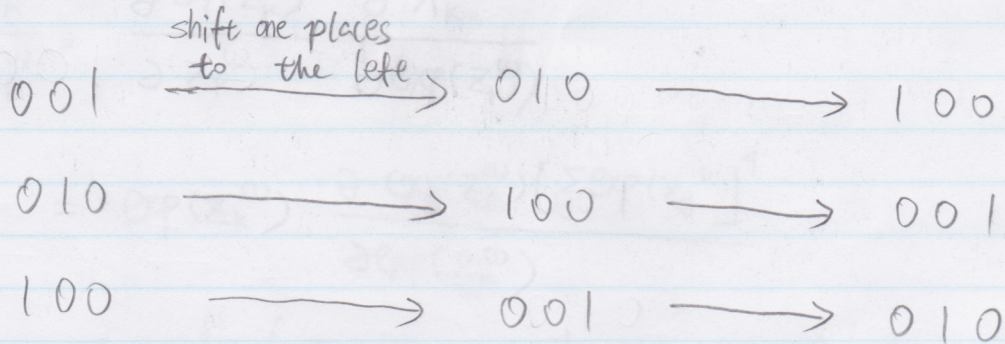


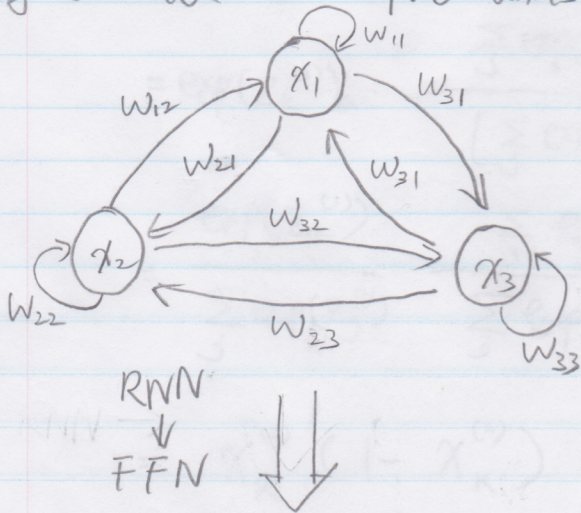
T6.

Part 1 RNN and review of FFL

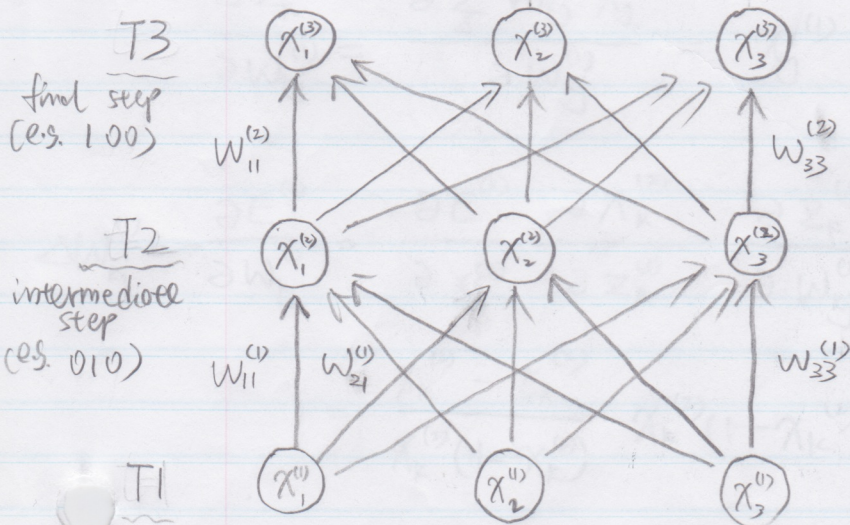
ex1 circular shift register



Fully connected 3 input units, no hidden unit



Time T:



NB: We evaluate FFN at both T2 and T3.

Forward pass in T1 → T2

$$z_k^{(1)} = \sum_j w_{kj}^{(1)} x_j^{(1)}$$

$$x_k^{(2)} = \frac{\exp(z_k^{(1)})}{\sum_j \exp(z_j^{(1)})}$$

$$E^{(2)} = -\sum_{k=1}^3 t_k^{(2)} \log x_k^{(2)}$$

Back-prop T2 → T1:

$$\frac{\partial E^{(2)}}{\partial w_{kj}^{(1)}} = \frac{\partial E^{(2)}}{\partial x_k^{(2)}} \frac{\partial x_k^{(2)}}{\partial z_k^{(1)}} \frac{\partial z_k^{(1)}}{\partial w_{kj}^{(1)}}$$

$$\frac{\partial E^{(2)}}{\partial x_k^{(2)}} = -\frac{\partial \sum_j t_j^{(2)} \log x_j^{(2)}}{\partial x_k^{(2)}}$$

$$= -\frac{\partial t_k^{(2)} \log x_k^{(2)}}{\partial x_k^{(2)}} - \frac{\partial (1-t_k^{(2)}) \log(1-x_k^{(2)})}{\partial x_k^{(2)}}$$

$$= -\frac{t_k^{(2)}}{x_k^{(2)}} + \frac{1-t_k^{(2)}}{1-x_k^{(2)}}$$

$$= \frac{x_k^{(2)} - t_k^{(2)}}{x_k^{(2)}(1-x_k^{(2)})}$$

Back-prop cont'd

$$\frac{\partial \chi_k^{(2)}}{\partial z_k^{(1)}} = \frac{\partial \exp(z_k^{(1)})}{\partial z_k^{(1)}} \frac{\partial \chi_k^{(2)}}{\partial \exp(z_k^{(1)})}$$

$$= \exp(z_k^{(1)}) \frac{\partial \exp(z_k^{(1)}) \left[\sum_j \exp(z_j^{(1)}) \right]^{-1}}{\partial \exp(z_k^{(1)})}$$

$$= \exp(z_k^{(1)}) \left\{ \left[\sum_j \exp(z_j^{(1)}) \right]^{-1} + \exp(z_k^{(1)}) (-1) \left[\sum_j \exp(z_j^{(1)}) \right]^{-2} \right\}$$

$$= \exp(z_k^{(1)}) \cdot \frac{\sum_j \exp(z_j^{(1)}) - \exp(z_k^{(1)})}{\left[\sum_j \exp(z_j^{(1)}) \right]^2}$$

$$= \frac{\exp(z_k^{(1)})}{\sum_j \exp(z_j^{(1)})} \frac{\sum_{j \neq k} \exp(z_j^{(1)})}{\sum_j \exp(z_j^{(1)})}$$

$$= \chi_k^{(2)} (1 - \chi_k^{(2)})$$

$$\frac{\partial z_k^{(1)}}{\partial w_{kj}^{(1)}} = \frac{\partial \sum_j w_{kj}^{(1)} \chi_j^{(1)}}{\partial w_{kj}^{(1)}} = \chi_j^{(1)}$$

$$\Delta w_{kj}^{(1)} = \frac{\partial E^{(2)}}{\partial w_{kj}^{(1)}} = \frac{\partial E^{(2)}}{\partial \chi_k^{(2)}} \frac{\partial \chi_k^{(2)}}{\partial z_k^{(1)}} \frac{\partial z_k^{(1)}}{\partial w_{kj}^{(1)}}$$

$$= \frac{\chi_k^{(2)} - t_k^{(2)}}{\chi_k^{(2)} (1 - \chi_k^{(2)})} \chi_k^{(2)} (1 - \chi_k^{(2)}) \chi_j^{(1)}$$

$$= (\chi_k^{(2)} - t_k^{(2)}) \chi_j^{(1)}$$

Similarly

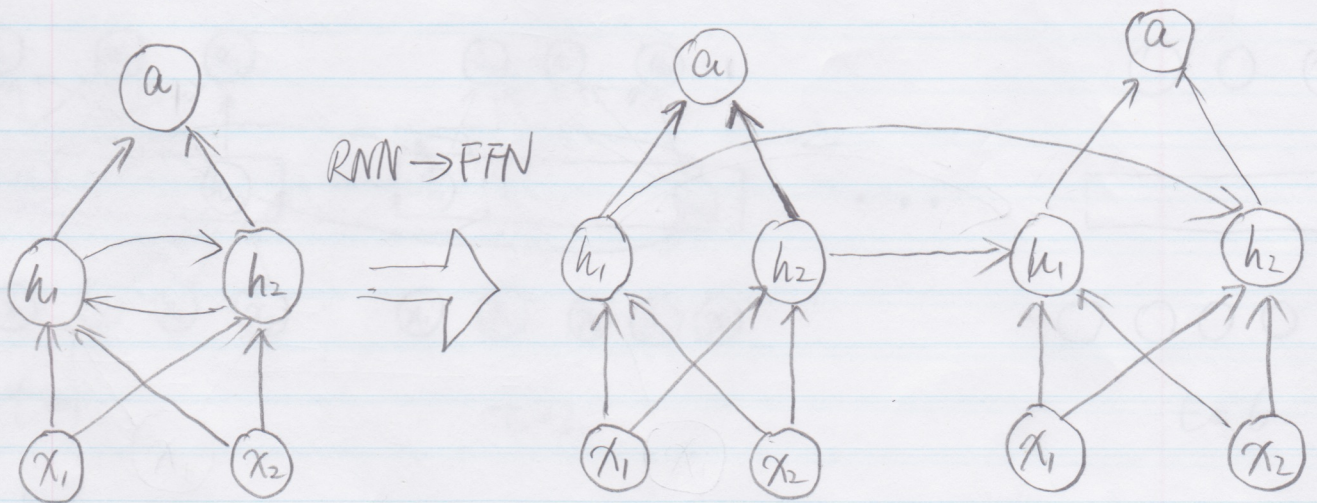
$$\Delta W_{kj}^{(2)} = (x_k^{(3)} - t_k^{(3)}) x_j^{(2)}$$

Update $W_{kj}^{(1)}$ and $W_{kj}^{(2)}$ with combined gradient

$$W_{kj}^{(1)*} = W_{kj}^{(1)} - \epsilon (\Delta W_{kj}^{(1)} + \Delta W_{kj}^{(2)})$$

$$W_{kj}^{(2)*} = W_{kj}^{(2)} - \epsilon (\Delta W_{kj}^{(1)} + \Delta W_{kj}^{(2)})$$

ex2. Simple conversion of a RNN with hidden units



$t=1$

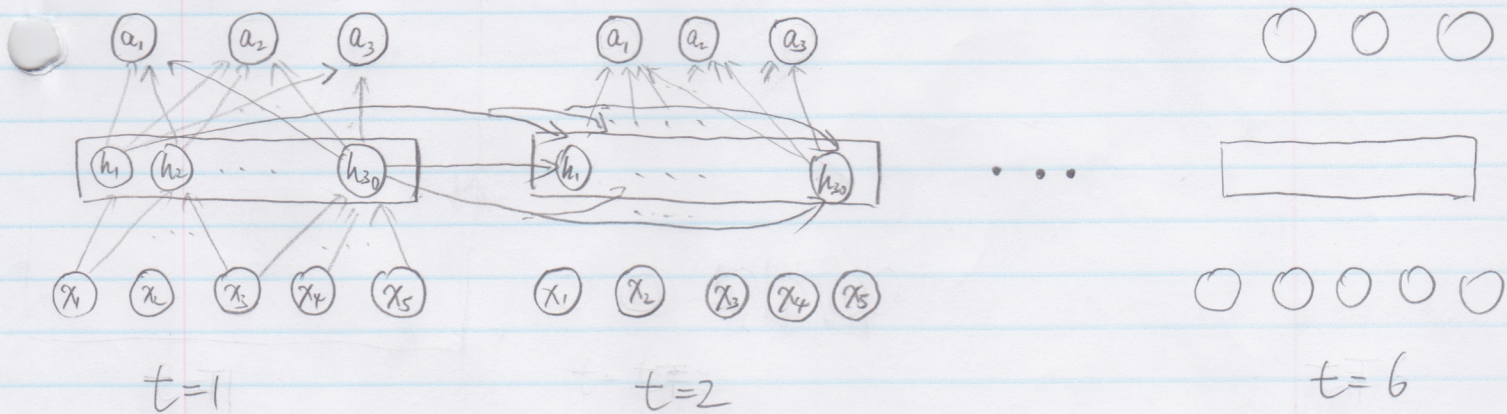
ex3. sequence completion (high level overview)

LETTER	A	B	C	D	E
Hidden Encoding	12	23	31	21	13

25 training sequences:

AA1212 . AB1223 . AC1231
 BA2312
 :
 :

Task: Given the first two letters, predict the remaining sequence



e.g. AB1223.

I/O unit:	A	B	C	D	E	Ideal output		
						a_1	a_2	a_3
time: $t=1$	✓	x	x	x	x	-	-	-
2	x	✓	x	x	x	-	-	-
3	x	x	x	x	x	✓	x	x
4	x	x	x	x	x	x	✓	x
5	x	x	x	x	x	x	✓	x
6	x	x	x	x	x	x	x	✓

Forward pass at Time t :

$$y_k^{(t)} = \sum_{kj} W_{kj}^{(t)} x_j + \sum_{kl} W_{kl}^{(t-1)} h_{kl} + b_k^{(t)}$$

$$h_k^{(t)} = \frac{\exp(y_k^{(t)})}{1 + \exp(y_k^{(t)})}$$

$$z_s^{(t)} = \sum_{sk} W_{sk} h_k^{(t)} + b_s^{(t)}$$

$$a_s^{(t)} = \frac{\exp(z_s^{(t)})}{\sum_{s'} \exp(z_s^{(t)})}$$

$$E^{(t)} = - \sum_s t_s^{(t)} \log a_s^{(t)}$$

Back-prop to get $\frac{\partial E}{\partial W_{kj}^{(t)}}$, $\frac{\partial E}{\partial W_{kl}^{(t)}}$, $\frac{\partial E}{\partial W_{sk}^{(t)}}$, $\frac{\partial E}{\partial b_{s/k}^{(t)}}$

Update:

$$\Delta W_{kj} = \sum_t \frac{\partial E^{(t)}}{\partial W_{kj}^{(t)}}$$

$$\Delta W_{kl} = \sum_t \frac{\partial E^{(t)}}{\partial W_{kl}^{(t)}}$$

$$\Delta W_{sk} = \sum_t \frac{\partial E^{(t)}}{\partial W_{sk}^{(t)}}$$

$$\Delta b_{s/k} = \sum_t \frac{\partial E^{(t)}}{\partial b_{s/k}^{(t)}}$$