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Deep learning

- Recap: GRU, RNN

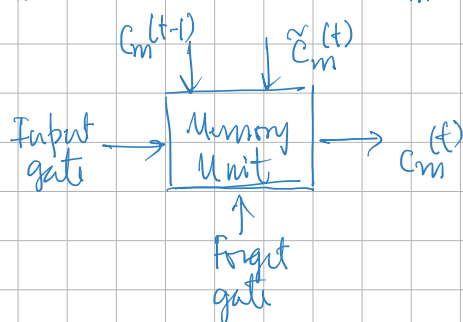
- LSTM

- RNN:  $\underline{z}_m^{(t)} = \sigma(\underline{\alpha}_m^T \underline{x}^{(t)} + \underline{\gamma}_m^T \underline{z}^{(t-1)})$

- GRU:  $\underline{z}_m^{(t)} = s_m^{(t)} \cdot \underline{z}_m^{(t-1)} + (1 - s_m^{(t)}) \cdot \tilde{\underline{z}}_m^{(t)}$

$$\tilde{\underline{z}}_m^{(t)} = \tanh(\underline{\alpha}_m^T \underline{x}^{(t)} + r_m^{(t)} \underline{\gamma}_m^T \underline{z}^{(t-1)})$$

- LSTM:



$$C_m^{(t)} = f_m^{(t)} \cdot C_m^{(t-1)} + i_m^{(t)} \tilde{C}_m^{(t)}$$

$$\tilde{C}_m^{(t)} = \tanh(\underline{\alpha}_m^T \underline{x}^{(t)} + \underline{\gamma}_m^T \underline{z}^{(t-1)})$$

$$\underline{z}_m^{(t)} = o_m^{(t)} \cdot \tanh(C_m^{(t)})$$

$s_m^{(t)}$ : update gate (long term)  
 $r_m^{(t)}$ : reset gate (short term)

$C_m^{(t)}$ :  $m^{\text{th}}$  memory unit  
 contents at time  $t$

$\tilde{C}_m^{(t)}$ :  $m^{\text{th}}$  candidate

mem. unit contents at  $t$

$f_m^{(t)}$ : forget gate unit  $m$  at time  $t$

$i_m^{(t)}$ : input gate unit  $m$  at time  $t$

$o_m^{(t)}$ : output gate unit  $m$  at time  $t$

Gates in  $m^{\text{th}}$  LSTM:

$$\begin{cases} f_m^{(t)} = \sigma(\underline{u}_m^T \underline{x}^{(t)} + \underline{v}_m^T \underline{z}^{(t-1)}) \\ i_m^{(t)} = \sigma(\underline{w}_m^T \underline{x}^{(t)} + \underline{p}_m^T \underline{z}^{(t-1)}) \\ o_m^{(t)} = \sigma(\underline{s}_m^T \underline{x}^{(t)} + \underline{d}_m^T \underline{z}^{(t-1)}) \end{cases}$$

$$\Theta = \{ \underline{\alpha}_m, \underline{\gamma}_m, \underline{\delta}_m, \underline{u}_m, \underline{v}_m, \underline{w}_m, \underline{p}_m, \underline{s}_m, \underline{d}_m \}$$

- Again, LSTM models have the ability to "remember" inputs in the long term and in the near term.

- Generative Models

Ex1: Let  $X$  be a continuous RV with CDF  $F_X(x)$ . What is the distribution of the RV  $Y$  which is defined as  $Y = F_X(X)$ ? (Theorem of a VAE)

Ex2: Let  $X \sim U[0, 1]$ . Let  $Y = g(X)$ . Find  $g(\cdot)$  such that  $Y \sim f_d(y)$  i.e.  $Y$  follows a desired distribution. (Theorem of a VAE)

$$\text{Aul: } P(Y \leq y) = P(F_X(X) \leq y)$$

$$\text{Case } 0 \leq y \leq 1 = P(X \leq F_X^{-1}(y))$$

$$= \int_{-\infty}^{F_X^{-1}(y)} f_X(x) dx$$

$$= F_X(F_X^{-1}(y)) - \underbrace{F_X(-\infty)}_{\rightarrow 0}$$

$$F_Y(y) = y$$