

## 6. 门控循环神经网络 (GRU)

$$h_t = z_t \odot h_{t+1} + (1 - z_t) \odot g(x_t, h_{t+1}; \theta)$$

$$z_t = \sigma(W_z x_t + U_z h_{t+1} + b_z)$$

$$\tilde{h}_t = g(x_t, h_{t+1}; \theta)$$

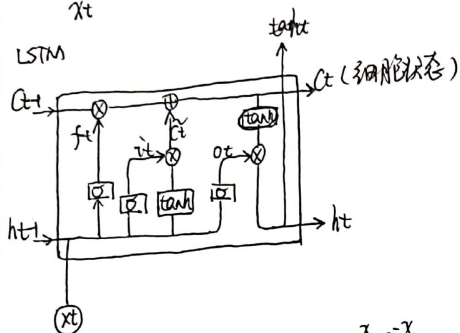
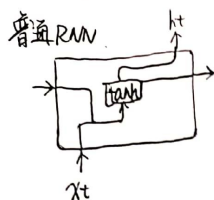
$$= \tanh(W_h x_t + U_h (r_t \odot h_{t+1}) + b_h)$$

$r_t$  为重置门: 是否忽略  $h_{t+1}$

$$r_t = \sigma(W_r x_t + U_r h_{t+1} + b_r)$$

# LSTM (长短期记忆网络)

避免长期依赖问题

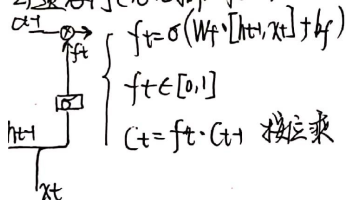


1)  $\sigma$ : sigmoid  $\sigma(x) = \frac{1}{1+e^{-x}}$ ,  $\tanh x = \frac{e^x - e^{-x}}{e^x + e^{-x}}$   
 $1 - \sigma(x) = -\tanh(\frac{x}{2})$

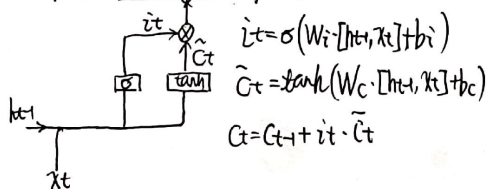
⊗ 按位乘

⊕ 按位加

2) 遗忘门 (忘记输入中  $C_t$  中不需要的信息)



3) 记忆门 (决定  $h_{t-1}$  和  $x_t$  哪些需要保留)



更新细胞状态

$$C_t = f_t \cdot C_{t-1} + i_t \cdot \tilde{C}_t$$

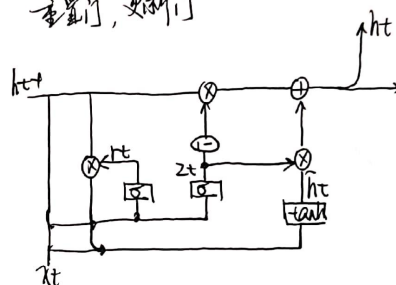
输出门

$$o_t = \sigma(W_o \cdot [h_{t-1}, x_t] + b_o)$$

$$h_t = o_t \cdot \tanh(C_t)$$

变体:

GRU: 整合 LSTM 的  $C_t$  和  $h_t$  重置门, 更新门



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

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

$$\tilde{h}_t = \tanh(W \cdot [r_t \cdot h_{t-1}, x_t])$$

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