Each element of \hat{y}_t is a number from 0 to N.

Un-selected is correct

Congratulations! You passed! Next Item 1/1 point Consider a task with input sequences of fixed length. Could RNN architecture still be useful for such task? Yes Correct RNN could be useful since it may need much less number of parameters. No 1/1 point Consider an RNN for a language generation task. \hat{y}_t is an output of this RNN at each time step, L is a length of the input sequence, N is a number of words in the vocabulary. Choose correct statements about \hat{y}_t : \hat{y}_t is a vector of length N. Correct The output at each time step is a distribution over a vocabulary, therefore the length of \hat{y}_t is equal to the vocabulary size. \hat{y}_t is a vector of length (L-t). **Un-selected is correct** \hat{y}_t is a vector of length L imes N. Un-selected is correct Each element of \hat{y}_t is either 0 or 1. Un-selected is correct Each element of \hat{y}_t is a number from 0 to 1. Elements of \hat{y}_t are probabilities so they are numbers from 0 to 1 and the sum of them equal to 1.

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RNN and Backpropagation

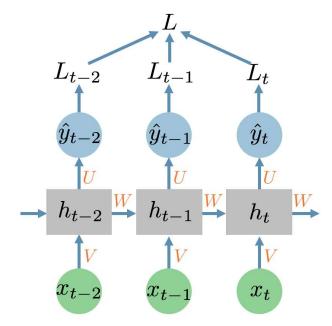
Quiz, 4 questions

point

Consider the RNN from the lecture:

$$h_t = f_t(Vx_t + Wh_{t-1} + b_h)$$

$$\hat{y}_t = f_y(Uh_t + b_y)$$



Calculate the gradient of the loss L with respect to the bias vector b_y . $\frac{\partial L}{\partial b_y} = ?$

$$\frac{\partial L}{\partial b_y} = \frac{\partial L}{\partial \hat{y}_t} \frac{\partial \hat{y}}{\partial b_t}$$

$$\frac{\partial L}{\partial b_y} = \sum_{t=0}^T \begin{bmatrix} \partial L_t \ \partial \hat{y}_t \\ \partial \hat{y}_t \ \partial b_y \end{bmatrix}$$

It is correct since b_y influence each L_t only once trough \hat{y}_t .

$$\frac{\partial L}{\partial b_y} = \sum_{t=0}^{T} \begin{bmatrix} \partial L_t \ \partial \hat{y}_t \ \partial h_t \\ \partial \hat{y}_t \ \partial h_t \ \partial b_y \end{bmatrix}$$

$$\frac{\partial L}{\partial b_y} = \sum_{t=0}^{T} \begin{bmatrix} \frac{\partial L_t}{\partial \hat{y}_t} \frac{\partial \hat{y}_t}{\partial h_t} \sum_{k=0}^{t} \frac{\partial h_k}{\partial b_y} \end{bmatrix}$$

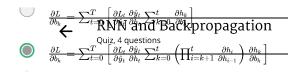
$$\frac{\partial L}{\partial b_y} = \sum_{t=0}^{T} \begin{bmatrix} \frac{\partial L_t}{\partial \hat{y}_t} \frac{\partial \hat{y}_t}{\partial h_t} \sum_{k=0}^{t} \left(\prod_{i=k+1}^{t} \frac{\partial h_i}{\partial h_{i-1}} \right) \frac{\partial h_k}{\partial b_y} \end{bmatrix}$$



Consider the RNN network from the previous question. Calculate the gradient of the loss L with respect to the bias vector b_h . $\frac{\partial L}{\partial b_h} = ?$

$$\frac{\partial L}{\partial b_h} = \frac{\partial L_t}{\partial \hat{y}_t} \frac{\partial \hat{y}_t}{\partial h_t} \frac{\partial h_t}{\partial b_h}$$

$$\frac{\partial L}{\partial b_h} = \sum_{t=0}^{T} \begin{bmatrix} \partial L_t \ \partial \hat{y}_t \ \partial h_t \\ \partial \hat{y}_t \ \partial h_t \ \partial b_h \end{bmatrix}$$



Correct

It is correct. Hidden units depend on b_h at each time step, therefore we need to backpropagate through time here.