

Set A

CSE440: Natural Language Processing II - QUIZ 5

Date: 22 December 2025

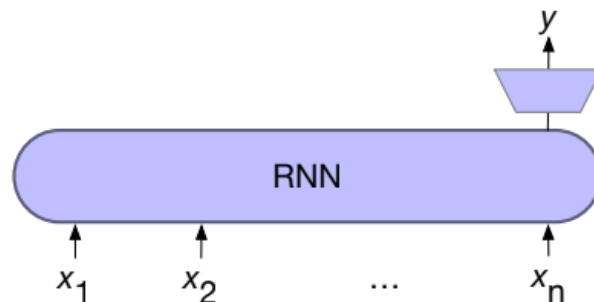
This is an MCQ quiz. Each question is worth 1. Write your answers serially on the side.

Name: _____ Student ID: _____ Section: _____

Answers:

| | |
|----|---|
| 1 | B |
| 2 | C |
| 3 | B |
| 4 | C |
| 5 | B |
| 6 | C |
| 7 | B |
| 8 | C |
| 9 | B |
| 10 | C |

- The hidden state update in a simple RNN is primarily responsible for:
 - Feature extraction only
 - Combining current input with past information
 - Encoding current input only
 - Preventing overfitting
- Which property allows an RNN to handle variable-length sequences?
 - Fixed-size hidden layers
 - One-hot encoded inputs
 - Weight sharing across time steps
 - Softmax output layer
- What happens if the forget gate outputs values close to zero in LSTM?
 - Previous memory is preserved
 - Previous cell state is mostly forgotten
 - Gradients explode
 - Output becomes zero
- In a Bidirectional RNN, how are the outputs of the forward and backward RNN combined at each time step?
 - By averaging the two outputs
 - By element-wise multiplication of the two outputs
 - By concatenating or summing the two hidden states before computing the output
 - By selecting the output with the higher activation value
- For the following image, which name is most appropriate?



- Sequence labeling
- Sequence classification
- Language modeling
- Encoder-Decoder

6. Which issue mainly prevents simple RNNs from learning long-term dependencies?

- A. Overfitting
- B. High computational cost
- C. Vanishing gradients
- D. Lack of non-linearity

7. The LSTM cell state is designed to:

- A. Store short-term memory only
- B. Preserve long-term information
- C. Replace hidden states
- D. Normalize gradients

Answer question 8 to 10 by considering the following RNN:

$$\begin{aligned}o^{(t)} &= \text{softmax}(Vh^{(t)}) \\h^{(t)} &= Uh^{(t-1)} + Wx^{(t)}\end{aligned}$$
$$U = \begin{bmatrix} 1 & 1 \\ 0 & 2 \end{bmatrix} \quad W = \begin{bmatrix} 2 & 1 & 0 \\ 1 & 0 & 1 \end{bmatrix} \quad V = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ -1 & 2 \end{bmatrix}$$
$$h^{(0)} = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \quad x^{(1)} = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$

8. What is the hidden state $h^{(1)}$?

- A. $[0 \ 1]^T$
- B. $[1 \ 2]^T$
- C. $[2 \ 0]^T$
- D. $[1 \ 0]^T$

9. Using the $h^{(1)}$ obtained in question 8, what is the pre-softmax score vector $Vh^{(1)}$?

- A. $[1 \ 2 \ 1]^T$
- B. $[2 \ 0 \ -2]^T$
- C. $[-2 \ 0 \ 2]^T$
- D. $[2 \ 2 \ -1]^T$

10. If index 0 is NN, 1 is VB, 2 is DET, what will be the prediction for $x^{(1)}$?

- A. DET
- B. VB
- C. NN
- D. Tie among NN and VB

Set B

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Name: _____ Student ID: _____ Section: _____

Answers:

| | |
|----|---|
| 1 | A |
| 2 | B |
| 3 | A |
| 4 | B |
| 5 | A |
| 6 | B |
| 7 | A |
| 8 | B |
| 9 | A |
| 10 | B |

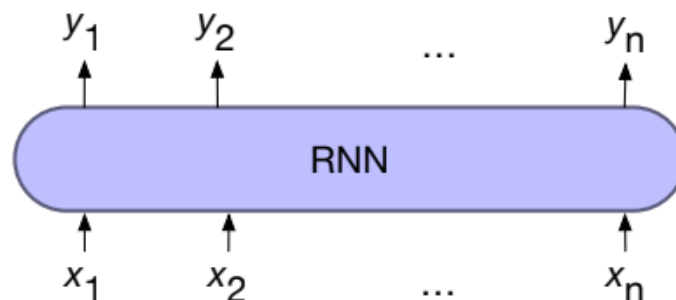
1. In a simple RNN, which parameters remain the same at every time step?
- A. Weights U , W , V
 - B. Inputs only
 - C. Hidden states only
 - D. Outputs only

2. The hidden state update in a simple RNN is primarily responsible for:
- A. Feature extraction only
 - B. Combining current input with past information
 - C. Encoding current input only
 - D. Preventing overfitting

3. Which issue mainly prevents simple RNNs from learning long-term dependencies?
- A. Vanishing gradients
 - B. Overfitting
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4. The LSTM cell state is designed to:
- A. Store short-term memory only
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5. For the following image, which name is most appropriate?



- A. Sequence labeling
 - B. Sequence classification
 - C. Language modeling
 - D. Encoder-Decoder
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- A. By concatenating or summing the two hidden states before computing the output
- B. By averaging the two outputs
- C. By element-wise multiplication of the two outputs
- D. By selecting the output with the higher activation value

Answer question 8 to 10 by considering the following RNN:

$$\begin{aligned}
 o^{(t)} &= \text{softmax}(Vh^{(t)}) \\
 h^{(t)} &= Uh^{(t-1)} + Wx^{(t)}
 \end{aligned}$$

$$U = \begin{bmatrix} 1 & 3 \\ 0 & 2 \end{bmatrix} \quad W = \begin{bmatrix} 3 & 1 & 2 \\ 1 & 0 & 1 \end{bmatrix} \quad V = \begin{bmatrix} 1 & 2 \\ 3 & 1 \\ -1 & 2 \end{bmatrix}$$

$$h^{(0)} = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \quad x^{(1)} = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$

8. What is the hidden state $h^{(1)}$?

- A. $[0 \ 1]^T$
- B. $[5 \ 2]^T$
- C. $[2 \ 5]^T$
- D. $[1 \ 0]^T$

9. Using the $h^{(1)}$ obtained in question 8, what is the pre-softmax score vector $Vh^{(1)}$?

- A. $[9 \ 17 \ -1]^T$
- B. $[1 \ 4 \ 5]^T$
- C. $[-17 \ 4 \ 12]^T$
- D. $[13 \ 2 \ -1]^T$

10. If index 0 is NN, 1 is VB, 2 is DET, what will be the prediction for $x^{(1)}$?

- A. DET
- B. VB
- C. NN
- D. Tie among NN and VB