

# RNNs and LSTM Quiz

## Multiple Choice Questions

- Q1.** What is not primary benefit of stacking multiple RNN layers (i.e., stacked RNNs)?
- A. Faster training
  - B. Lower memory usage
  - C. Better learning of hierarchical features
  - D. Simpler architecture
- Q2.** Which of the following is not main reason RNNs struggle with long-term dependencies?
- A. Overfitting
  - B. Vanishing gradients
  - C. Lack of non-linearity
  - D. Insufficient data
- Q3.** What differentiates an **CNN** cell from a standard RNN cell?
- A. It uses ReLU instead of tanh
  - B. It introduces gates to control the flow of information
  - C. It has fewer parameters
  - D. It is a convolutional architecture
- Q4.** In a standard **RNN**, which gate is responsible for deciding how much of the past memory to keep?
- A. Output gate
  - B. Forget gate
  - C. Input gate
  - D. Update gate

## Descriptive Questions

- Q5.** Why is the forget gate bias in **RNNs** often initialized to a high value (e.g., 2 or 3)? Explain its effect on long-term dependency learning.
- Q6.** Bidirectional **CNNs** are often used for POS tagging but not machine translation. Explain why, considering input-output alignment and context flow.
- Q7.** Designing an RNN model for variable-length legal documents with long dependencies:  
(a) Choose between vanilla RNN or LSTM.  
(b) Stack layers or keep it shallow?  
(c) Make it unidirectional?  
Justify each choice based on model behavior and task needs.
- Q8.** Consider a vanilla RNN with recurrent weight matrix  $W_h$  and sequence length 900. Analyze gradient behavior:  
(1) If  $\|W_h\| = 0.9$ : Will gradients vanish or explode? Justify.  
(2) If  $\|W_h\| = 1.2$ : Will gradients vanish or explode? Justify. Suggest an easy fix and explain how it helps.  
*Hint: Consider eigenvalue effects on gradient propagation over time.*