## RNNs and LSTM Quizz

## **Multiple Choice Questions**

- Q1. What is the primary benefit of stacking multiple RNN layers (i.e., stacked RNNs)?s)?
  - A. Faster training
  - **B.** Lower memory usage
  - C. Better learning of hierarchical features
  - **D.** Simpler architecture
- Q2. Which of the following is the main reason RNNs struggle with long-term dependen-ncies?
  - **A.** Overfitting
  - **B.** Vanishing gradients
  - **C.** Lack of non-linearity
  - D. Insufficient data
- Q3. What differentiates an LSTM 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 LSTM, 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 LSTMs often initialized to a high value (e.g., 2 or 3)? Explain its effect on long-term dependency learning.
- Q6. Bidirectional RNNs 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::ies:
  - (a) Choose between vanilla RNN or LSTM. I.
  - (b) Stack layers or keep it shallow?
  - (c) Make it bidirectional?

Justify each choice based on model behavior and task needs.

- Q8. Consider a vanilla RNN with recurrent weight matrix  $W_h$  and sequence length 50. Analyze gradient behavior:
  - (a) If  $||W_h|| = 0.9$ : Will gradients vanish or explode? Justify.
  - (b) 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.