CNRNNs and LSTM Quiz

Multiple Choice Questions

- Q1. What is the 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 the main reason RNNs struggle with long-term dependen- dependencies?
 - 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. Bidirectio Q6. Bidirectional RNNs are often used for POS tagging but not machine translation. Explain why, considering input—output alignment and context flow.
- Q7. Designing Q7. Designing an RNN model for variable-length legal documents with long dependencies: 9S:
 - (a) Ch^(a) Choose between vanilla RNN or LSTM. STM.
 - (b) Stack layers or keep it shallow?
 - (c) Make it bidirectional?

Justify each choice based on model behavior and task needs.

Q8. Cons Q8. Consider a vanilla RNN with recurrent weight matrix

Analyze gradient behavior:

 W_h and sequence length 50.

- (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.