CS7641 ML Practice Quiz

Module SL 1: Decision Trees

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Question 1

Which of the following statements correctly describe the characteristics of decision trees and neural networks?

- A. Decision trees are limited to linear functions, while neural networks can compute both linear and non-linear functions.
- B. Neural networks, unlike decision trees, cannot represent Boolean functions such as AND, OR, and XOR.
- C. The ID3 algorithm used in decision trees prefers shorter trees with good splits near the top.
- D. Decision trees can handle continuous attributes by creating branching factors equal to the number of possible values.
- E. Neural networks always require a sigmoid activation function for the neurons.

Question 2

Regarding supervised learning tasks, which of the following statements are true?

- A. Classification tasks map inputs to continuous values, while regression tasks map inputs to discrete labels.
- B. Regression to the mean is a phenomenon observed in decision trees when dealing with continuous attributes.
- C. In regression, variance can be used as a measure instead of information gain when dealing with continuous outputs.
- D. Both decision trees and neural networks can be used for classification and regression tasks.
- E. Neural networks can only handle discrete inputs and outputs.

Question 3

Which of the following best describe the expressiveness and limitations of decision trees?

- A. Decision trees can represent the XOR function using a single node.
- B. The n-version of XOR in decision trees has a size exponential in the number of attributes.
- C. Decision trees cannot handle continuous attributes without modification.
- D. A decision tree for the n-version of OR grows linearly with the number of attributes.
- E. Overfitting in decision trees is addressed by increasing the depth of the tree.

Question 4

Concerning the optimization of neural networks, which of the following statements are correct?

- A. Gradient descent cannot be used in neural networks as it is not differentiable.
- B. Neural networks cannot get stuck in local minima when using gradient descent.
- C. Momentum terms in the gradient can help avoid getting stuck in local minima.
- D. Neural networks should always be initialized with large random weights to ensure diversity of solutions.
- E. Restriction bias in neural networks is influenced by the number of hidden layers and nodes.

Question 5

In the context of the Perceptron Rule and gradient descent in machine learning, which of the following are true?

- A. The Perceptron Rule uses unthresholded values for weight updates.
- B. The Delta Rule and gradient descent are synonymous, both using threshold outputs.
- C. The Perceptron Rule can find a separating line in non-linearly separable datasets.
- D. Gradient descent converges only to a global optimum and not local optima.
- E. Gradient descent adjusts weights by minimizing the error between activation and target value.

Answer Key

- 1. C, D 2. C, D
- 3. B, C, D
- 4. C, E
- 5. E