

Practice Questions #3

Question 1 – Why are pooling layers used in the Convolutional Neural Networks (CNNs) context?

- A. To increase the spatial dimensions of the feature maps.
- B. To apply non-linear transformations to the feature maps.
- C. To reduce the computational complexity by down-sampling the feature maps.
- D. To directly classify the input images into different categories.

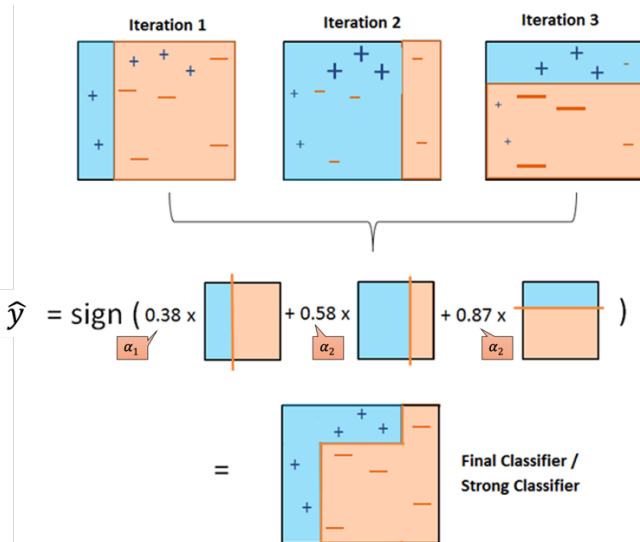
Answer Key: C

Question 2 – What is the primary difference between AdaBoost and Gradient Boosting in ensemble learning?

- A. AdaBoost changes the dataset distribution, Gradient Boosting changes the loss function.
- B. AdaBoost focuses on high bias, Gradient Boosting on high variance.
- C. AdaBoost uses fixed learning rates, Gradient Boosting uses variable learning rates.
- D. AdaBoost is for classification only, Gradient Boosting is for both classification and regression.

Answer Key: A

Question 3 – The following figure illustrates the AdaBoost algorithm's process for a binary classification problem.



Which of the following best describes the role of α (alpha) in the AdaBoost algorithm based on the weighted contribution of each weak classifier (shown by α_1 , α_2 , and α_3) in the final decision rule?

- A. The values of α represent the accuracy of each weak learner, with higher values indicating more accurate predictions.

- B. The values of α signify the learning rate, determining the speed at which the algorithm learns from the training data.
- C. The α values are coefficients that reflect the importance of each weak learner's vote in the final classifier.
- D. The α values correspond to the error rate of each weak learner, with higher values indicating a higher misclassification rate.

Answer Key: C

Question 4 – In the context of Boosting, which statement correctly describes the role and impact of the error rate (ϵ) in the AdaBoost algorithm?

- A. A higher error rate (ϵ) will result in a larger voting factor (α) value, leading to a greater influence of the corresponding weak classifier in the final model.
- B. The error rate (ϵ) has no impact on the update of weights for training points, as the update relies solely on the classifier's predictions.
- C. If the error rate (ϵ) is 0.5, the weak classifier's voting factor (α) will be zero, meaning its influence on the final model is neutral.
- D. A smaller error rate (ϵ) will cause a more significant decrease in the weights of correctly classified points, potentially leading to overfitting.

Answer Key: C

Question 5 – What is the role of the learning rate in the gradient-boosting algorithm?

- A. It controls the contribution of each new tree to the final model, allowing for more robust predictions by reducing the risk of overfitting.
- B. It determines the number of trees to be included in the model, directly affecting the algorithm's complexity.
- C. It adjusts the weight given to misclassified points in each iteration, similar to the weight adjustment in AdaBoost.
- D. It is used to increase the step size of the gradient descent optimization, speeding up the model's convergence.

Answer Key: A

Question 6 – What is the primary reason for using the SoftMax activation function in the output layer of a neural network for a multi-class classification problem?

- A. Because it normalizes the output of each neuron to a range of $[0, 1]$, which is useful for binary classification problems.
- B. Because it ensures the output of each neuron is always a positive value, simplifying the optimization problem.
- C. Because it converts the outputs into a probability distribution, where the sum of the probabilities of all classes equals 1.

D. It amplifies the differences between the neuron's outputs, making the network's decisions more distinct and easier to interpret.

Answer Key: C

Question 7 – In a feed-forward neural network, which of the following statements accurately describes the role of the learning rate (η) and its impact on the weight update process during training?

- A. The learning rate (η) determines the size of the weight adjustments; a very high learning rate can cause the network to converge rapidly to suboptimal weights due to overshooting the minimum of the loss function.
- B. The learning rate (η) is inversely proportional to the weight updates, meaning that as the learning rate increases, the magnitude of the weight updates decreases.
- C. The learning rate (η) has no effect on the training process and is kept constant to ensure that the weights do not change, which might lead to instability in the network.
- D. The learning rate (η) is adjusted after each iteration of weight updates to ensure that the network weights diverge, facilitating the exploration of the loss function landscape.

Answer Key: A

Question 8 – What is the primary purpose of stacking multiple convolutional layers sequentially in a Convolutional Neural Network (CNN)?

- A. To progressively reduce the size of the input image to a form suitable for classification.
- B. Allow the network to detect a wide range of features, from simple to complex, across the input image.
- C. To enable the network to construct a complex hierarchy of features, starting from basic edges in the early layers to intricate patterns in the deeper layers.
- D. To enhance the network's capability to identify the location and scale of features within the image, contributing to more accurate classifications.

Answer Key: C

Question 9 – How do Convolutional Neural Networks (CNNs) process multi-channel (e.g., color RGB) images during convolutional operations?

- A. CNNs first convert multi-channel images into grayscale to reduce computational complexity before processing them through convolutional layers.
- B. CNNs treat each channel separately, with distinct filters for each channel. The results from each are then combined before being passed to the next layer.
- C. CNNs process multi-channel images by applying a set of filters that span all channels, allowing the network to learn features that are activated across the entire depth of the input volume.

D. To ensure the network learns the basic features from each channel, CNNs require pre-training on single-channel data before they can be applied to multi-channel images.

Answer Key: C

Question 10 – What is the significance of feature maps at various layers in the training of convolutional neural networks (CNNs), and how do they contribute to the network's ability to classify images?

- A. Feature maps are identical copies of the input image passed through each layer to ensure the network has multiple instances of the data for robust learning.
- B. Each feature map represents a specific version of the input image after being convolved with a unique filter, which allows the network to analyze the image at different scales and orientations.
- C. Feature maps in CNNs represent binary masks directly highlighting the regions of interest in an input image, simplifying the task for the fully connected layers.
- D. The feature maps at deeper layers represent increasingly abstract and composite features, synthesizing information from previous maps to aid in recognizing complex patterns within images.

Answer Key: D