

PreLab:-

1.What is the primary objective of the backpropagation algorithm in the context of training neural networks?

The primary objective of backpropagation is to minimize the loss function by calculating gradients and updating the model's weights to improve prediction accuracy.

2.Briefly describe the computations involved in the forward pass of the backpropagation algorithm. What is the output of the forward pass?

The forward pass computes the activations of each layer by passing inputs through the network, ultimately producing the predicted output.

3.What is the purpose of the backward pass in the backpropagation algorithm? How does it contribute to updating the model parameters?

The backward pass calculates the gradients of the loss function with respect to each weight using the chain rule, guiding the adjustment of weights to reduce the loss.

4.Explain the role of the chain rule in backpropagation.

The chain rule is used to compute the gradient of the loss with respect to each weight by propagating errors backward through the network layers.

VIVA:-

1.How do different activation functions impact the computations in the forward pass and gradients in the backward pass of the backpropagation algorithm?

Activation functions determine the non-linearity in the forward pass and affect the gradient values in the backward pass, influencing how effectively the network learns.

2.Neural networks use non-linear activation functions. Why is non-linearity crucial for the success of backpropagation, and how does it help in capturing complex relationships in data?

Non-linearity allows neural networks to model complex, non-linear relationships in data, enabling backpropagation to adjust weights effectively across layers to capture these intricacies.

3.In the context of backpropagation, what role does the learning rate play during the parameter update step?

The learning rate controls the size of weight updates; it needs to be balanced to ensure stable and efficient convergence during training.

4.How does backpropagation contribute to the potential issue of overfitting, and what regularization techniques can be employed during training to address this concern?

Backpropagation can lead to overfitting by minimizing loss too precisely on the training data; regularization techniques like dropout, L2 regularization, and early stopping help mitigate this risk.