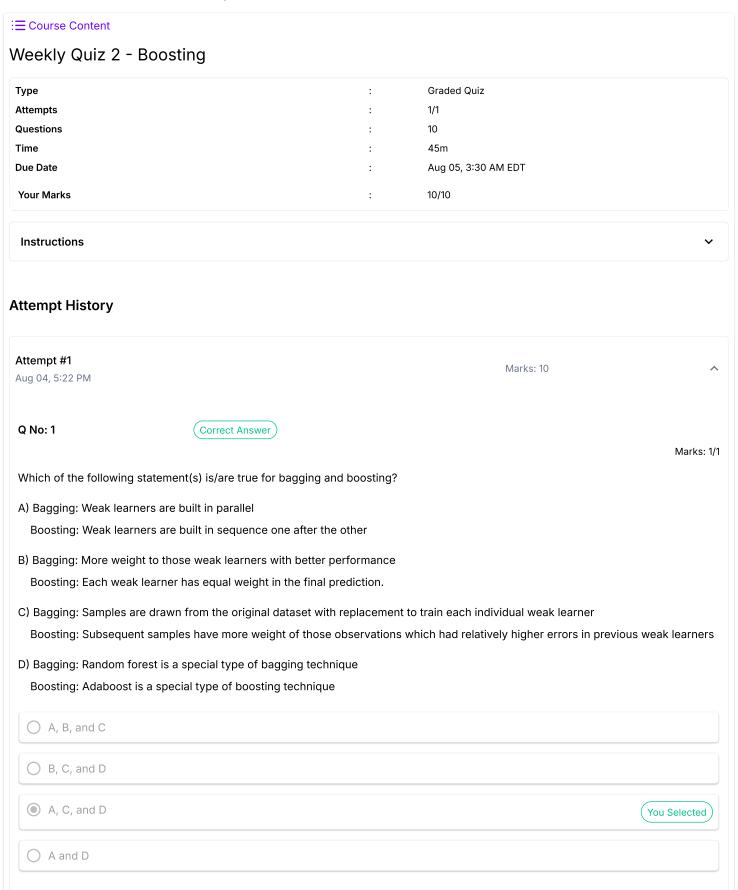








← Go Back to Advanced Machine Learning



In bagging, weak learners are built in parallel, and samples are drawn from the original dataset with replacement to train each individual weak learner. Example - Random Forest			
In boosting, weak learners are built in sequence one after the other, and subsequent samples have more weight of those observations which had relatively higher errors in previous weak learners. Example - Adaboost			
Q No: 2	Correct Answer Marks:		
Which of the following	statement(s) is/are true about the Gradient Boosting trees?		
1. Gradient Boosting tre	es work on residuals instead of changing the weights of observations.		
2. By fitting new models	s to the residuals, the overall learner gradually improves in areas where residuals are initially high.		
1 and 2	You Selected		
Only 1			
Only 2			
None			
Gradient boosting tree where residuals are in	es work by fitting new models to the residuals and thereby gradually improving the overall learner in areas uitially high.		
Q No: 3	Correct Answer Marks:		
Alpha is a hyperparamε	eter that is used while changing the weights of the samples in an AdaBoost model.		
	statement(s) is/are correct?		
	ple is decreased if it is incorrectly classified by the previous weak learner.		
	ple is increased if it is incorrectly classified by the previous weak learner.		
C) The alpha can be bo			
D) The alpha cannot be	negative		
Only A			
A and B			
B and C	You Selected		
Only D			

Q No: 4	Correct Answer	
A. Calculate the residual		Marks: 1/1
· · · · · · · · · · · · · · · · · · ·	using previous probabilities and new output values th an initial prediction for all observations	
	ing a new tree until the maximum number of estimators is reach late the output value for each leaf node	ed
$\bigcirc A \rightarrow B \rightarrow C - > D \rightarrow$	· E	
$\bigcirc B \rightarrow C \rightarrow E \rightarrow D \rightarrow A$		
$\bigcirc C \rightarrow A \rightarrow B \rightarrow D \rightarrow$	E	
	D	You Selected
The correct sequence	-	
Initialize the model with	n an initial prediction for all observations.	
Calculate the residual f	or each observation.	
Build a tree and calcula	ate the output value for each leaf node.	
Update all predictions	using previous probabilities and new output values.	
Repeat steps by creati	ng a new tree until the maximum number of estimators is reache	d.
Q No: 5	Correct Answer	
Which of the following a	lgorithms do NOT have a 'learning rate' parameter/hyperparame	Marks: 1/1 ter?
Gradient Boosting		
XGBoost		
Random Forest		You Selected
All of these		
Gradient Boosting and	XGBoost are similar algorithms and have a 'learning rate' parame	eter/hyperparameter to reduce overfitting.
0 N - 0		
Q No: 6	(Correct Answer)	Marks: 1/1
Which hyperparameter o	of XGBoost can be used to deal with the imbalance in data?	,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,

gamma
O learning_rate
scale_pos_weight You Selected
O colsample_by_node
In XGBoost, scale_pos_weight controls the balance of positive and negative weights and is used to deal with imbalanced classes.
Q No: 7 Correct Answer
Which of the following statements is false about XGBoost?
XGBoost is a boosting technique
XGBoost builds upon the idea of gradient boosting algorithm with some modifications
O It provides features that help in better computing like parallelization, cache optimization, out of core computing and distributed computing
XGBoost cannot handle missing values internally You Selected
XGBoost has the ability to handle missing values internally.
Q No: 8 Correct Answer Marks: 1/1
Which of the following statement(s) is/are correct for boosting technique?
A) Weight of observation, to be selected while building the next weak learner, increases if the observation is correctly classified B) Weight of observation, to be selected while building the next weak learner, increases, if the observation is incorrectly classified
C) Weight of observation, to be selected while building the next weak learner, decreases if the observation is correctly classified
D) Weight of observation, to be selected while building the next weak learner, decreases if the observation is incorrectly classified
O A and B
B and C You Selected
O C and D
O A and D
Weight of observation, to be selected while building the next weak learner, increases, if the observation is incorrectly classified, and the weight decreases if the observation is correctly classified.

Q No: 9	(Correct Answer)	\$ A . 1 . A //
Adaboost takes a we	eighted voting/weighted average of the weak learners for the final pre	Marks: 1/ diction.
True		You Selected
O False		
Adaboost takes a v	weighted voting/weighted average of the weak learners for the final pr	rediction.
Q No: 10	Correct Answer	
		Marks: 1/
A/l=:= = = £ 4 = = £= =		
which of the following	ng is/are true:	
	ng is/are true: ting use heterogeneous learners (different algorithms as different wea	ak learners).
I. Bagging and boos	ting use heterogeneous learners (different algorithms as different wea	ak learners).
I. Bagging and boos	ting use heterogeneous learners (different algorithms as different wea	ak learners).
Bagging and boos Stacking is a hom	ting use heterogeneous learners (different algorithms as different wea	ak learners).
1. Bagging and boos 2. Stacking is a hom Only 1	ting use heterogeneous learners (different algorithms as different wea	ak learners).
1. Bagging and boos 2. Stacking is a hom Only 1 Only 2	ting use heterogeneous learners (different algorithms as different wea	ak learners). You Selected
1. Bagging and boos 2. Stacking is a hom Only 1 Only 2 Both 1 and 2 None	ting use heterogeneous learners (different algorithms as different wea	You Selected
1. Bagging and boos 2. Stacking is a hom Only 1 Only 2 Both 1 and 2 None	ting use heterogeneous learners (different algorithms as different wea	You Selected

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