Ada (Gradient	Boost Ou	ıiz: Attemı	pt review
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Started on	Wednesday, 29 October 2025, 12:29 PM
	Finished
	Wednesday, 29 October 2025, 12:33 PM
	4 mins 36 secs
Marks	13.00/15.00
Grade	86.67 out of 100.00
_	
Question 1	
Complete	
Mark 1.00 out of 1.00	
In AdaBoost, how is t	ne weight of a weak learner determined?
a. Using gradie	nt descent
b. Using the for	mula $\alpha = (1/2)\ln((1-\epsilon)/\epsilon)$
C. By accuracy a	alone
d. Randomly in	itialized each time
Question 2	
Complete	
Mark 1.00 out of 1.00	
The function E(v) - E	$prev(x) + \eta \cdot h(x)$ indicates:
me function r (x) = r_	prev(x) + 1/11(x) mulcates.
a. Updating we	ights in AdaBoost
b. Bagging mod	
	rrection step in Gradient Boosting
d. Cross-validat	
u. Cross vandat	ion step
Question 3	
Complete	
Mark 1.00 out of 1.00	
AM - 1 1	
What does the learning	ng rate η control in Gradient Boosting?
a. The number	of residuals generated
	tion of each weak learner
od. The number	ot teatures used
a. The numberb. Tree depth	ng rate η control in Gradient Boosting? of residuals generated tion of each weak learner of features used

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Question 4	
	4
Mark 100 a	out of 1.00
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What d	does the term "adaptive" in AdaBoost refer to?
Wildt G	aces the term dauptive invadablest reter to.
○ a.	Modifying the kernel function dynamically
○ b.	
○ c.	Adjusting model depth at each iteration
d.	Updating the weights of samples based on errors
Question	5
Complete	
Mark 1.00 d	out of 1.00
What h	nappens to the sample weights after an AdaBoost iteration?
a.	Weights of misclassified samples increase
O b.	Weights of correctly classified samples increase
○ c.	Weights remain unchanged
() d.	All weights are halved
Question (6
Complete	
Mark 0.00 d	out of 1.00
IVIAIR 0.00 C	001.00
What is	s a key reason Gradient Boosting may overfit?
○ a.	Too few trees
O b.	Using regularization
○ b.○ c.	Using regularization Too many deep trees
○ b.○ c.	Using regularization
○ b.○ c.	Using regularization Too many deep trees
b.c.d.	Using regularization Too many deep trees Too small learning rate
b. c. d.	Using regularization Too many deep trees Too small learning rate
b.c.d.	Using regularization Too many deep trees Too small learning rate
b. c. d.	Using regularization Too many deep trees Too small learning rate
b. c. d.	Using regularization Too many deep trees Too small learning rate
D. C. d. Question 7 Complete Mark 1.00 c	Using regularization Too many deep trees Too small learning rate 7 out of 1.00
D. C. d. Question 7 Complete Mark 1.00 c	Using regularization Too many deep trees Too small learning rate
Question 7 Complete Mark 1.00 o	Using regularization Too many deep trees Too small learning rate 7 out of 1.00 s a residual in Gradient Boosting?
Question a Complete Mark 1.00 o	Using regularization Too many deep trees Too small learning rate 7 out of 1.00 s a residual in Gradient Boosting? The number of iterations completed
D. C. Question 7 Complete Mark 1.00 of What is	Using regularization Too many deep trees Too small learning rate 7 out of 1.00 s a residual in Gradient Boosting? The number of iterations completed The gradient of the feature vector
Description of the complete Mark 1.00 of the	Using regularization Too many deep trees Too small learning rate 7 out of 1.00 s a residual in Gradient Boosting? The number of iterations completed The gradient of the feature vector The difference between actual and predicted values
Description of the complete Mark 1.00 of the	Using regularization Too many deep trees Too small learning rate 7 out of 1.00 s a residual in Gradient Boosting? The number of iterations completed The gradient of the feature vector
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Question 8
Complete
Mark 1.00 out of 1.00
What is the loss function minimized by AdaBoost?
a. Cross-entropy loss
b. Exponential loss
○ c. Mean squared error
○ d. Hinge loss
Question 9
Complete
Mark 1.00 out of 1.00
What is the main goal of AdaBoost?
a. To increase data dimensionality
b. To minimize the mean squared error
c. To combine multiple weak classifiers into a strong one
d. To combine multiple strong classifiers into a weak one
Question 10
Complete
Mark 1.00 out of 1.00
What type of optimization is Gradient Boosting performing?
a. Genetic optimization
○ b. Random search
c. Gradient descent in function space
○ d. Grid search
G. Gita Search
Question 11
Complete
Mark 1.00 out of 1.00
Which modern algorithms are derived from the principle of Gradient Boosting?
a Logistic Pagescripp and CVM
a. Logistic Regression and SVM
b. Random Forest and Bagging
○ c. PCA and t-SNE
d. XGBoost, LightGBM, CatBoost

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Question 12
Complete
Mark 0.00 out of 1.00
Which of the following is NOT an advantage of AdaBoost?
a. Handles noise robustly
c. Simplicity of implementation
○ d. Works well with weak learners
Question 13
Complete
Mark 1.00 out of 1.00
Which of the following is TRUE about AdaBoost vs. Gradient Boosting?
 a. AdaBoost uses weighted data; Gradient Boosting uses residuals
b. Both minimize exponential loss
C. AdaBoost can use any differentiable loss
d. Gradient Boosting works only for classification
d. Gradient boosting works only for classification
Question 14
Complete
Mark 1.00 out of 1.00
Which of the following is usually used as a weak learner in AdaBoost?
Which of the following is assumed as a weak learner in Adaboost.
a. Neural network
b. Decision stump
C. Logistic regression
○ d. K-nearest neighbors
Question 15
Complete
Mark 1.00 out of 1.00
Which statement best describes Gradient Boosting?
a. It averages independent models to reduce variance
b. It sequentially adds models that correct residual errors
c. It combines models that predict misclassified labels
O di Tre combines models dide product misclassifica lascis
d. It reduces dimensionality before modeling