





 \mathbf{X}

404. In a linear regression problem, we	e are using R-squared to mea	sure goodness-of-fit. We	add a feature in linear r	regression model and retrain
the same model. Which of the following	g option is true?			

- A. a. if r squared increases, this variable is significant.
- B. b. if r squared decreases, this variable is not significant.
- C. c. individually r squared cannot tell about variable importance. we cant say anything about it right now.
- D. d. none of these.

C.c. individually r squared cannot tell about variable importance. we can t say anything about it right now.

discuss

405. Which of the one is true about Heteroskedasticity?

- A. a. linear regression with varying error terms
- B. b. linear regression with constant error terms
- C. c. linear regression with zero error terms
- D. d. none of these

A.a. linear regression with varying error terms

discuss

406. Which of the following assumptions do we make while deriving linear regression parameters?1. The true relationship between dependent y and predictor x is linear2. The model errors are statistically independent3. The errors are normally distributed with a 0 mean and constant standard deviation4. The predictor x is non-stochastic and is measured error-free

- A. a. 1,2 and 3.
- B. b. 1,3 and 4.
- C. c. 1 and 3.
- D. d. all of above.

D.d. all of above.

discuss

407. To test linear relationship of y(dependent) and x(independent) continuous variables, which of the following plot best suited?

- A. a. scatter plot
- B. b. barchart
- C. c. histograms
- D. d. none of these

A.a. scatter plot

discuss

408. Generally, which of the following method(s) is used for predicting continuous dependent variable?1. Linear Regression2. Logistic Regression

- A. a. 1 and 2
- B. b. only 1
- C. c. only 2
- D. d. none of these.

B.b. only 1

discuss



409. Suppose you are training a linear regression model. Now consider these points.1. Overfitting is more likely if we have less data2. Overfitting is more likely when the hypothesis space is small. Which of the above statement(s) are correct?

- A. a. both are false
- B. b. 1 is false and 2 is true
- C. c. 1 is true and 2 is false
- D. d. both are true

C.c. 1 is true and 2 is false

discuss

410. Suppose we fit Lasso Regression to a data set, which has 100 features (X1,X2X100). Now, we rescale one of these feature by multiplying with 10 (say that feature is X1), and then refit Lasso regression with the same regularization parameter. Now, which of the following option will be correct?

- A. a. it is more likely for x1 to be excluded from the model
- B. b. it is more likely for x1 to be included in the model
- C. c. cant say
- D. d. none of these

B.b. it is more likely for x1 to be included in the model

discuss

- 411. Which of the following is true aboutRidge or Lasso regression methods in case of feature selection?
- A. a. ridge regression uses subset selection of features
- B. b. lasso regression uses subset selection of features
- C. c. both use subset selection of features
- D. d. none of above

B.b. lasso regression uses subset selection of features

discuss

412. Which of the following statement(s) can

- A. a. 1 and 2
- B. b. 1 and 3
- C. c. 2 and
- D. d. none of the above

A.a. 1 and 2

discuss



413. We can also compute the coefficient of linear regression with the help of an analytical method called Normal Equation. Which of the followis/are true about Normal Equation?1. We dont have to choose the learning rate2. It becomes slow when number of features is very large3. Note to iterate	
A. a. 1 and 2	
B. b. 1 and 3.	
C. c. 2 and 3.	
D. d. 1,2 and 3.	
D.d. 1,2 and 3.	discuss
414. If two variables are correlated, is it necessary that they have a linear relationship?	
A. a. yes	
B. b. no	
B.b. no	discuss
415. Correlated variables can have zero correlation coeffficient. True or False?	
A. a. true	
B. b. false	
A.a. true	discuss
416. Which of the following option is true regarding Regression andCorrelation ?Note: y is dependent variable and x is independent variable.	
A. a. the relationship is symmetric between x and y in both.	
B. b. the relation	
C. c. the relation	

D.d. the relationship	is symmetric between x and y in case of correlation but in case of regression it is not symmetric.	(
417. Suppose you a	re using a Linear SVM classifier with 2 class classification	
A. yes		
B. no		
A.yes		
418. If you remove t	he non-red circled points from the data, the decision boundary will change?	
A. true		
B. false		
B.false		
419. When the C par	rameter is set to infinite, which of the following holds true?	
A. the optimal hyper	rameter is set to infinite, which of the following holds true? plane if exists, will be the one that completely separates the data assifier will separate the data	
A. the optimal hyper	plane if exists, will be the one that completely separates the data assifier will separate the data	
A. the optimal hyper B. the soft-margin cla C. none of the above	plane if exists, will be the one that completely separates the data assifier will separate the data	
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A.true

discuss

422. The objective of the support vector machine algorithm is to find a hyperplane in an N-dimensional space(N the number of features) that distinctly classifies the data points.

- A. true
- B. false

A.true

discuss

423. Hyperplanes are boundaries that help classify the data points.

- A. usual
- B. decision
- C. parallel

B.decision

discuss

424. The of the hyperplane depends upon the number of features.

- A. dimension
- B. classification
- C. reduction

A.dimension

discuss

425. Hyperplanes are decision boundaries that help classify the data points.

A. true

B. false

A.true

discuss