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Home » Computer Science Engineering (CSE) » Machine Learning (ML) » set 1

Machine Learning (ML) solved MCQs

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Set 2 »

- 1. Application of machine learning methods to large databases is called
- A. data mining.
- B. artificial intelligence
- C. big data computing
- D. internet of things

A.data mining.

discuss

- 2. If machine learning model output involves target variable then that model is called as
- A. descriptive model
- B. predictive model
- C. reinforcement learning
- D. all of the above

B.predictive model

discuss

- 3. In what type of learning labelled training data is used
- A. unsupervised learning
- B. supervised learning
- C. reinforcement learning
- D. active learning

B.supervised learning

discuss

- 4. In following type of feature selection method we start with empty feature set
- A. forward feature selection
- B. backword feature selection
- C. both a and b??
- D. none of the above

discuss A.forward feature selection 5. In PCA the nu discuss A.true 6. PCA can be used for projecting and visualizing data in lower dimensions. A. true B. false discuss A.true 7. Which of the following is the best machine learning method? A. scalable B. accuracy C. fast D. all of the above discuss D.all of the above 8. What characterize unlabeled examples in machine learning A. there is no prior knowledge B. there is no confusing knowledge C. there is prior knowledge D. there is plenty of confusing knowledge discuss D.there is plenty of confusing knowledge 9. What does dimensionality reduction reduce? A. stochastics B. collinerity C. performance D. entropy discuss B.collinerity

10. Data used to build a data mining model.	
A. training data	
B. validation data	
C. test data	
D. hidden data	
A.training data	
11. The problem of finding hidden structure in unlabeled data is called	
A. supervised learning	
B. unsupervised learning	
C. reinforcement learning	
D. none of the above	
	dis
B. unsupervised learning	
12. Of the Following Examples, Which would you address using an supervised learning Algorithm?	
A. given email labeled as spam or not spam, learn a spam filter	
B. given a set of news articles found on the web, group them into set of articles about the same story.	
C. given a database of customer data, automatically discover market segments and group customers into different market segments	s.
D. find the patterns in market basket analysis	
A.given email labeled as spam or not spam, learn a spam filter	dis
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B.ordinal

discuss

21. PCA is

- A. forward feature selection
- B. backword feature selection
- C. feature extraction
- D. all of the above

C.feature extraction

discuss

- 22. Dimensionality reduction algorithms are one of the possible ways to reduce the computation time required to build a model.
- A. true
- B. false

A.true

discuss

- 23. Which of the following techniques would perform better for reducing dimensions of a data set?
- A. removing columns which have too many missing values
- B. removing columns which have high variance in data
- C. removing columns with dissimilar data trends
- D. none of these

A.removing columns which have too many missing values

discuss

24. Supervised learning and unsupervised clustering both require which is correct according to the statement.

(

- A. output attribute.
- B. hidden attribute.
- C. input attribute.
- D. categorical attribute

C. input attribute.

discuss

25. What characterize is hyperplance in geometrical model of machine learning?

- A. a plane with 1 dimensional fewer than number of input attributes
- B. a plane with 2 dimensional fewer than number of input attributes
- C. a plane with 1 dimensional more than number of input attributes
- D. a plane with 2 dimensional more than number of input attributes

B.a plane with 2 dimensional fewer than number of input attributes

Set 2 »

discuss

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