

Questions Bank

Subject Name: Machine Learning

Subject Code: 15CS73

Sem: VII

Module -1 Questions.

1. Define the following terms:
 - a. Learning
 - b. LMS weight update rule
 - c. Version Space
 - d. Consistent Hypothesis
 - e. General Boundary
 - f. Specific Boundary
 - g. Concept
2. What are the important objectives of machine learning?
3. Explain find –S algorithm with given example. Give its application.

Table 1

Example	Sky	AirTemp	Humidity	Wind	Water	Forecast	EnjoySport
1	Sunny	Warm	Normal	Strong	Warm	Same	Yes
2	Sunny	Warm	High	Strong	Warm	Same	Yes
3	Rainy	Cold	High	Strong	Warm	Change	No
4	Sunny	Warm	High	Strong	Cool	Change	Yes

4. What do you mean by a well –posed learning problem? Explain the important features that are required to well –define a learning problem.
5. Explain the inductive biased hypothesis space and unbiased learner
6. What are the basic design issues and approaches to machine learning?
7. How is Candidate Elimination algorithm different from Find-S Algorithm
8. How do you design a checkers learning problem
9. Explain the various stages involved in designing a learning system

10. Trace the Candidate Elimination Algorithm for the hypothesis space H' given the sequence of training examples from Table 1.

$H' = \langle ?, \text{Cold}, \text{High}, ?, ?, ? \rangle \vee \langle \text{Sunny}, ?, \text{High}, ?, ?, \text{Same} \rangle$

11. Differentiate between Training data and Testing Data

12. Differentiate between Supervised, Unsupervised and Reinforcement Learning

13. What are the issues in Machine Learning

14. Explain the List Then Eliminate Algorithm with an example

15. What is the difference between Find-S and Candidate Elimination Algorithm

16. Explain the concept of Inductive Bias

17. With a neat diagram, explain how you can model inductive systems by equivalent deductive systems

18. What do you mean by Concept Learning?

Module -2 Questions.

1. Give decision trees to represent the following boolean functions:

(a) $A \wedge B$

(b) $A \vee [B \wedge C]$

(c) $A \text{ XOR } B$

(d) $[A \wedge B] \vee [C \wedge D]$

2. Consider the following set of training examples:

<u>Instance</u>	<u>Classification</u>	<u>a1</u>	<u>a2</u>
1	+	T	T
2	+	T	T
3	-	T	F
4	+	F	F
5	-	F	T
6	-	F	T

(a) What is the entropy of this collection of training examples with respect to the target function classification?

(b) What is the information gain of a2 relative to these training examples?

3. NASA wants to be able to discriminate between Martians (M) and Humans (H) based on the following characteristics: Green $\in \{N, Y\}$, Legs $\in \{2, 3\}$, Height $\in \{S, T\}$, Smelly $\in \{N, Y\}$

Our available training data is as follows:

	<u>Species</u>	<u>Green</u>	<u>Legs</u>	<u>Height</u>	<u>Smelly</u>
1	M	N	3	S	Y
2	M	Y	2	T	N
3	M	Y	3	T	N
4	M	N	2	S	Y
5	M	Y	3	T	N
6	H	N	2	T	Y
7	H	N	2	S	N
8	H	N	2	T	N
9	H	Y	2	S	N
10	H	N	2	T	Y

a) Greedily learn a decision tree using the ID3 algorithm and draw the tree.

b) (i) Write the learned concept for Martian as a set of conjunctive rules (e.g., if

(green=Y and legs=2 and height=T and smelly=N), then Martian; else if ... then Martian;...; else Human).

(ii) The solution of part b)i) above uses up to 4 attributes in each conjunction. Find a set of conjunctive rules using only 2 attributes per conjunction that still results in zero error in the training set. Can this simpler hypothesis be represented by a decision tree of depth 2? Justify.

4. Discuss Entropy in ID3 algorithm with an example

5. Compare Entropy and Information Gain in ID3 with an example.

6. Describe hypothesis Space search in ID3 and contrast it with Candidate-Elimination algorithm.

7. Relate Inductive bias with respect to Decision tree learning.

8. Illustrate **Occam's razor** and relate the importance of **Occam's razor** with respect to ID3 algorithm.

9. List the issues in Decision Tree Learning. Interpret the algorithm with respect to **Overfitting the data**.

10. Discuss the effect of **reduced Error pruning** in decision tree algorithm.

11. What type of problems are best suited for decision tree learning

12. Write the steps of ID3 Algorithm
13. What are the capabilities and limitations of ID3
14. Define (a) Preference Bias (b) Restriction Bias
15. Explain the various issues in Decision tree Learning
16. Describe Reduced Error Pruning
17. What are the alternative measures for selecting attributes
18. What is Rule Post Pruning

Module -3 Questions.

- 1) What is Artificial Neural Network?
- 2) What are the type of problems in which Artificial Neural Network can be applied.
- 3) Explain the concept of a Perceptron with a neat diagram.
- 4) Discuss the Perceptron training rule.
- 5) Under what conditions the perceptron rule fails and it becomes necessary to apply the delta rule
- 6) What do you mean by Gradient Descent?
- 7) Derive the Gradient Descent Rule.
- 8) What are the conditions in which Gradient Descent is applied.
- 9) What are the difficulties in applying Gradient Descent.
- 10) Differentiate between Gradient Descent and Stochastic Gradient Descent
- 11) Define Delta Rule.
- 12) Derive the Backpropagation rule considering the training rule for Output Unit weights and Training Rule for Hidden Unit weights
- 13) Write the algorithm for Back propagation.
- 14) Explain how to learn Multilayer Networks using Gradient Descent Algorithm.
- 15) What is Squashing Function?

Module -4 Questions.

- 1) Explain the concept of Bayes theorem with an example.
- 2) Explain Bayesian belief network and conditional independence with example.
- 3) What are Bayesian Belief nets? Where are they used?
- 4) Explain Brute force MAP hypothesis learner? What is minimum description length principle

- 5) Explain the k-Means Algorithm with an example.
- 6) How do you classify text using Bayes Theorem
- 7) Define (i) Prior Probability (ii) Conditional Probability (iii) Posterior Probability
- 8) Explain Brute force Bayes Concept Learning
- 9) Explain the concept of EM Algorithm.
- 10) What is conditional Independence?
- 11) Explain Naïve Bayes Classifier with an Example.
- 12) Describe the concept of MDL.
- 13) Who are Consistent Learners.
- 14) Discuss Maximum Likelihood and Least Square Error Hypothesis.
- 15) Describe Maximum Likelihood Hypothesis for predicting probabilities.
- 16) Explain the Gradient Search to Maximize Likelihood in a neural Net.

Module -5 Questions.

1. What is Reinforcement Learning?
2. Explain the Q function and Q Learning Algorithm.
3. Describe K-nearest Neighbour learning Algorithm for continuous valued target function.
4. Discuss the major drawbacks of K-nearest Neighbour learning Algorithm and how it can be corrected
5. Define the following terms with respect to K - Nearest Neighbour Learning :
 - i) Regression ii) Residual iii) Kernel Function.
6. Explain Q learning algorithm assuming deterministic rewards and actions?
7. Explain the K – nearest neighbour algorithm for approximating a discrete – valued function $f : H_n \rightarrow V$ with pseudo code
8. Explain Locally Weighted Linear Regression.
9. Explain CADET System using Case based reasoning.
10. Explain the two key difficulties that arise while estimating the Accuracy of Hypothesis.
11. Define the following terms

a. Sample error	b. True error	c. Random Variable
d. Expected value	e. Variance	f. standard Deviation
12. Explain Binomial Distribution with an example.
13. Explain Normal or Gaussian distribution with an example.

14.Explain the Central Limit Theorem with an example.

15. Write the Procedure for estimating the difference in error between two learning methods.

Approximate confidence intervals for this estimate