Questions Bank

Module -1

- 1. De4fine 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.

- 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

- 1. Give decision trees to represent the following boolean functions:
 - (a) A ∧~B
 - (b) A V $[B \land C]$
 - (c) A XOR B
 - (d) $[A \wedge B] \vee [C \wedge D]$
- 2. Consider the following set of training examples:

<u>Instance</u>	Classification	<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>	Green	Legs	<u>Height</u>	Smelly
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	Н	N	2	T	Y
7	Н	N	2	S	N
8	Н	N	2	T	N
9	Н	Y	2	S	N
10	Н	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 ID3Algorithm
- 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

- 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

- 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

- 1. What is Reinforcement Learning?
- 2. Explain the Q function and Q Learning Algorithm.
- 3. Describe K-nearest Neighbour learning Algorithm for continues 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 $f: Hn \rightarrow V$ 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.

15. Write the Proc	edure for estimating the o	difference in error b	etween two learning	methods.
Approximate co	onfidence intervals for th	is estimate		