

Course Examination 1st Term, 2010 - 2011

Course Code & Title : CSCI3230 Fundamentals of Artificial Intelligence

Time allowed : 2 hours 0 minutes

Student I.D. No. : Seat No. :

Answer Four out of Five questions  
Total: 100 marks

1.

- a) What is the crucial difference between the following two first-order sentences where  $R$  is a binary predicate? [3 marks]
- i.  $\forall x \exists y R(x, y).$
  - ii.  $\exists x \forall y R(x, y).$

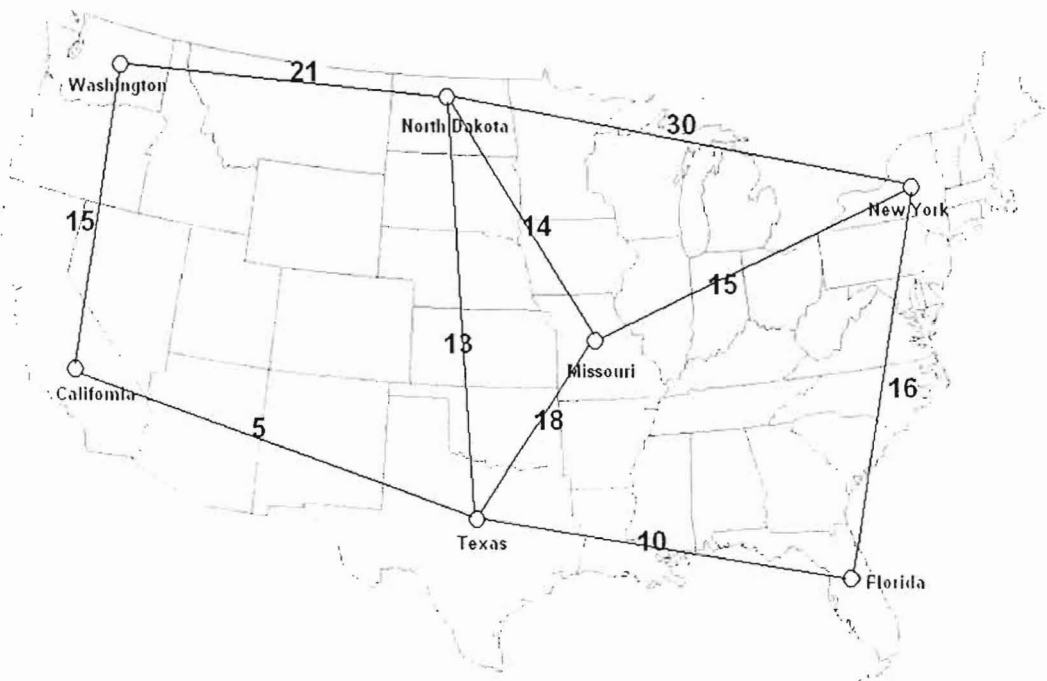
- b) Express each of the following statements (i to v) as First Order Logic sentences using the symbols below (together with  $\forall, \exists, \neg, \wedge, \vee, \rightarrow$ ): [5 marks]

Symbol	Meaning
Student(X)	X is a student
Diligent(X)	X is diligent
Lazy(X)	X is lazy
DoRevision(X,Y)	X will do revision for course Y
Tough(Y)	Y is a tough course
Pass(X,Y)	X passes the course Y

- i. Every student is either diligent or lazy.
  - ii. Diligent students will do revision for all courses.
  - iii. Lazy students cannot pass any tough course.
  - iv. *CSCI3230* is a tough course, and *Peter* is a student.
  - v. If *Peter* does not do revision for *CSCI3230*, then he cannot pass it.
- c) Convert your answers for b) into **Conjunctive Normal Form (CNF)**. [6 marks]
- d) Prove statement v from statements i to iv using *any one* of the **Forward Chaining, Backward Chaining** and **Resolution** approaches. [6 marks]
- e) What are **soundness** and **completeness** for a set of inference rules? What are their importances? [5 marks]

2.

- a) Name a single point stochastic search algorithm and describe it briefly. [3 marks]
- b) The distances between cities and the Heuristic Distances (HD) to New York are shown in the U.S.A. roadmap and the table below respectively. Use **both** of the following two algorithms to search for a shortest path from **Washington** to **New York**. Show all your steps and state the paths, costs and sequence (step order numbers) clearly with a path tree.  
(In each step, you do NOT need to consider the city you have just visited)
- i. Uniform-cost search with path costs [6 marks]
- ii. A\* search (using the given Heuristic Distances table below) [8 marks]



Heuristic Estimated distances to New York			
New York	0	North Dakota	28
Washington	20	California	40
Missouri	34	Florida	45
Texas	43		

- c) Do the two algorithms used for part b) give the same shortest path, and are the given Heuristic Estimated distances admissible? Explain your answer. [4 marks]
- d) Name the heuristic searching technique used in the Multi-Layer Feed-Forward Network (MLN) back-propagation updating rule? Why is it used in MLN?[4 marks]

3.

- a) Consider the following game: there are  $N$  coins, labeled 0 to  $N-1$ , some have heads facing up, and some have tails facing up. If you choose a coin to flip, its immediate left and right neighbors (note:  $N-1$  is a neighbor of 0) will be flipped (i.e. head  $\rightarrow$  tail, tail  $\rightarrow$  head), but not the coin itself. E.g. choosing coin 2 will flip the coins 1 and 3; choosing coin 0 will flip the coins 1 and  $N-1$ . The objective of the game is to successively flip the coins so that all of the coins have heads facing up. Formulate the game as a search problem (state space, state transitions and goal state(s)).

[8 marks]



- b) Use *any heuristic* you like to solve the following instance, where  $N = 7$  and initially the coins are as follows. List which coins you successively flipped and the state transitions.

[3 marks]

0	1	2	3	4	5	6
Tail	Tail	Tail	Head	Tail	Tail	Tail

- c) In real problems, the search spaces are usually very large. Suggest **TWO** methods to handle the huge spaces.
- d) In game playing (e.g. Chess, Go, Checker), even using MiniMax with Alpha-Beta pruning, it is not guaranteed that the computer can win (at least not lose). Explain why this is so.
- e) What is **overfitting** in supervised learning? What is **Ockham's Razor**? How does it relate to overfitting? How can we detect and prevent overfitting?

[4 marks]

[4 marks]

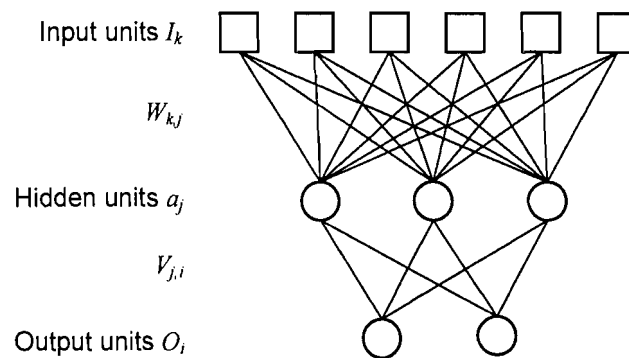
[6 marks]

4.

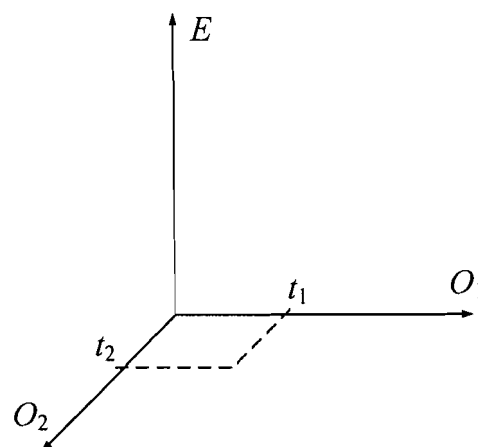
a) Why is single layer perceptron unable to solve **non-linearly separable** problems? Give an example of **non-linearly separable** problem. [3 marks]

b) For the sigmoid function  $f(z) = \frac{1}{1 + e^{-z}}$ , find its derivative  $f'(z)$ . [3 marks]

c) Derive the **TWO** updating rules ( $\frac{\partial E}{\partial V_{j,i}}$  and  $\frac{\partial E}{\partial W_{k,j}}$ ) of weights and biases in back-propagation algorithm for Multi-Layer Feed-Forward Network (MLN). The activation function in the output units is the above sigmoid function, i.e.,  $O_i = f(u_i)$ , where  $u_i = \sum_j a_j \times V_{j,i} + B_{O_i}$ . The activation function in the hidden units is the above sigmoid function, i.e.,  $a_j = f(s_j)$ , where  $s_j = \sum_k I_k \times W_{k,j} + B_{a_j}$ . The error function is the sum of squared error, i.e.,  $E = \frac{1}{2} \sum_i (O_i - T_i)^2$  where  $T_i$  is the target value. The notations and the diagram of the MLN are shown below: [12 marks]



d) Why  $E = \frac{1}{2} \sum_i (O_i - T_i)^2$  is suitable to be used as an error function in back-propagation algorithm of MLN? Explain briefly. (Hint: Sketch the  $E$  where  $t_1$  and  $t_2$  are the target values of  $O_1$  and  $O_2$  respectively). [7 marks]

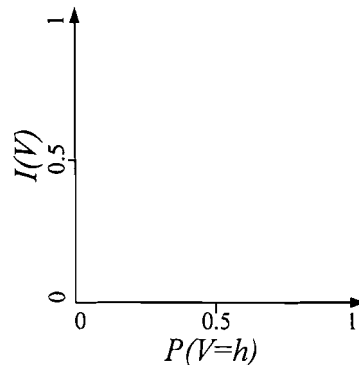


5.

- a) For a discrete random variable  $V$  with  $n$  possible values  $v_i$ 's, each with probability  $P_i$ , the Information Content ( $I$ ) is given by:

$$I(V) = \sum_{i=1}^n -P_i \log_2 P_i \text{ with the convention that } 0 \times \log_2 0 = 0$$

For a binary random variable (e.g. for flipping a coin,  $n = 2$ ),  $P(V = h)$  denotes the probability of flipping a "head". **Sketch** the graph of  $I(V)$  against  $P(V = h)$  on your answer book, and describe and explain it briefly (such as the maxima, minima and the convexity). [5 marks]



- b) For a  $K$  class classification, an attribute with  $N$  values  $A \in \{a_1, a_2, \dots, a_N\}$ , and  $v_{i,k}$  is the number of class- $k$  records having value  $a_i$  in that attribute. Write down the equation in computing the Information Gain ( $IG$ ) of the attribute. [3 marks]
- c) In constructing a Decision Tree, the attribute having the highest  $IG$  is chosen to be the splitting attribute. Explain why. [2 marks]
- d) Note that the decision tree constructed by the method in part c) may NOT be optimal, why? Suggest a method to improve the construction method. [4 marks]
- e) In binary classification, the confusion matrix consists of four elements which are True Positive (TP), False Positive (FP), True Negative (TN) and False Negative (FN). What does each mean? [4 marks]
- f) There are a number of performance measures derived from TP, FP, TN and FN:  

$$Precision = \frac{TP}{TP + FP}, \quad Recall = \frac{TP}{TP + FN}, \quad F\_measure = \frac{1}{\left(\frac{1}{Recall} + \frac{1}{Precision}\right)/2}$$
 Interpret each of them. [4 marks]
- g) In a medical classification, e.g. predicting whether a patient is suffering from cancer or not, which performance measure(s) in part f) is more important. Explain why. [3 marks]

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