# **Midterm Examination**

CS 540: Introduction to Artificial Intelligence

July 15, 2004

LAST (FAMILY) NAME:	SOLUTION	
FIRST NAME:		

Problem	Score	Max Score
1		15
2		16
3		10
4		12
5		18
6		21
7		8
Total		100

### 1. [15] **Decision Tree Learning**

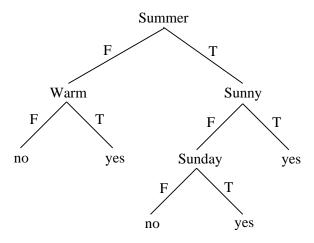
(a) [6] Describe briefly two (2) different causes of overfitting when learning a decision tree using the MaxGain criterion.

Noisy attribute values and irrelevant attributes.

(b) [3] Suppose we generate a training set from a given "correct" decision tree and then use the decision tree learning algorithm with MaxGain to that training set. Is it the case that the algorithm is guaranteed to construct the same "correct" decision tree as the training set size goes to infinity? Briefly explain why or why not.

No, because even if all possible training examples are eventually used, there is not, in general, a unique decision tree, so a different, but logically equivalent, decision tree may be created because the MaxGain criterion picks a different attribute at a given node than the "correct" tree has.

(c) [3] Given the following decision tree for making a binary decision about whether or not to go on a bike ride, write a single sentence in Propositional Logic that expresses the same information, i.e., when to go on a bike ride.



 $(\neg Summer \land Warm) \lor (Summer \land Sunny) \lor (Summer \land \neg Sunny \land Sunday)$ 

(d) [3] In a problem where each example has n binary attributes, to select the best attribute for a decision tree node at depth k, where the root is at depth 0, how many attributes must be compared?

## 2. [16] Search Methods

Consider the 3-puzzle problem, which is a simpler version of the 8-puzzle where the board is 2 x 2 and there are three tiles, numbered 1, 2, and 3. There are four operators, which move the blank **up**, **down**, **left**, and **right**, and these operators are applied *in this order* for all uninformed searches and in case of sibling ties for other searches. Break other ties by increasing time on OPEN. The cost of each operator is 1. The start and goal states are

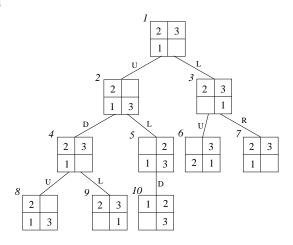
Start		Go		
2	3	1	2	
1			3	

(a) [4] Draw the entire state space for this problem, labeling nodes and arcs clearly.

Letting B123 mean the state where the blank and the tiles 1, 2, and 3 are in the upper-left, upper-right, lower-left, and lower-right corners, respectively, and U means the operation of moving the blank up, etc., the state space contains two disconnected cycles of 12 states each:

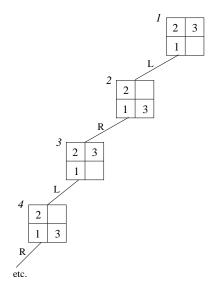
(b) [12] Assuming there is no checking for repeated states of any kind, draw the search trees produced by each of the following search methods. For each node in a tree, label it with a number indicating when it was removed from the OPEN list (and expanded or detected as a goal node). If a method does not find a solution, show the part of the search tree and then explain why no solution is found.

#### (i) Breadth-First search

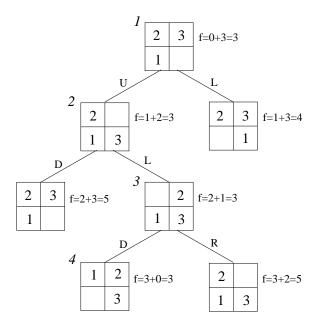


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(ii) Depth-First search



(iii) A\* search with the heuristic equal to the number of misplaced tiles



3. [10] Evaluation Functions for Heuristic Search

Say we define the evaluation function for use in a heuristic search problem as

$$f(n) = (1-w)g(n) + wh(n)$$

where g(n) is the cost of the best path found from the start state to state n, h(n) is an admissible heuristic function that estimates the cost of a path from n to a goal state, and  $0.0 \le w \le 1.0$ .

(a) [9] What search algorithm do you get when

(i) 
$$w = 0.0$$

Uniform-Cost search

(ii) 
$$w = 0.5$$

A\* search

(iii) 
$$w = 1.0$$

Greedy Best-First search

(b) [1] Based on your answer to (a), for what range of values of w would you expect your algorithm is admissible?

$$0 \le w \le 0.5$$

### 4. [12] Search and Propositional Logic

The "SAT" problem is to determine if a given sentence in Conjunctive Normal Form (CNF) in Propositional Logic (PL) is true given some assignment of {True, False} to each of the symbols in the sentence. (CNF means the form of the sentence is a conjunction of disjunctions.) Say we want to solve this problem using Greedy Hill-Climbing Search. Each state corresponds to a complete assignment of True or False to each symbol. The successors of a state are all states that have exactly one symbol with a different truth value. For example, if a state were (A=True, B=False), the successor states would be (A=True, B=True) and (A=False, B=False). The evaluation function used is the number of clauses in the (CNF) sentence that are true.

(a) [3] If there are n distinct symbols in a given PL sentence (in CNF), how many neighboring states does each state have in the state space?

n

(b) [3] How many states are there in the state space?

 $2^n$  because each symbol can take one of two possible values.

- (c) [6] Given the sentence  $(\neg A \lor B \lor C) \land (A \lor \neg B \lor C) \land (A \lor B \lor \neg C) \land (A \lor B \lor C)$ , is the state (A=False, B=False, C=False)
  - (i) a goal state? Briefly explain.

No because in this case only 3 out of the 4 clauses are true.

(ii) a local, but not a global, maximum of the state space? Briefly explain.

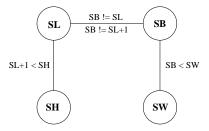
Yes because the value of the evaluation function for this state is 3 and its three neighboring states, (A=True, B=False, C=False), (A=False, B=True, C=False), and (A=False, B=False, C=True), each also have a value of 3.

### 5. [18] Constraint Satisfaction

Say you work at a factory that makes flashlights, which requires four tasks to be completed: constructing the light bulb (L), charging the battery (B), doing the wiring (W), and assembling the housing (H). Task L takes 2 hours, B takes 1 hour, W takes 2 hours, and H takes 1 hour. We'd like the total time to make each flashlight to be no more than 4 hours. Tasks can be started at the beginning of each hour, designated 1, 2, 3, and 4. Tasks can done concurrently except L and B cannot be done simultaneously, L must be done sometime before H, and B must be completed sometime before W.

This problem can be formulated as a constraint satisfaction problem (CSP) by letting the variables be the start times for each of the tasks; we'll call these times SL, SB, SW, and SH. The possible values for each variable are the possible start times, 1, 2, 3, and 4. Based on the information above, the initial possible values for each variable are  $SL=\{1,2,3\}$ ,  $SB=\{1,2,3,4\}$ ,  $SW=\{1,2,3\}$ , and  $SH=\{1,2,3,4\}$ . The constraints are SL+1 < SH, SB < SW,  $SB \ne SL$ , and  $SB \ne SL+1$ .

(a) [3] Draw the constraint graph showing all of the above information.



(b) [3] Assuming we've already chosen (as the first step in a backtracking search) SL=2, what is the domain of each variable after applying **forward checking**?

$$SB=\{1,4\}$$
,  $SW=\{1,2,3\}$ , and  $SH=\{4\}$ 

(c) [12] Now assume we have all of the original domains for each variable and we apply **arc consistency**. Show the domain of each variable after each of the following steps of removing inconsistent values by filling in the following table. "Propagate X to Y" means that values in the domain of Y should be removed if they are inconsistent with the values in the domain of X. For example, given the initial domains specified above, after "propagating SL to SH" the domain of SH will change from {1,2,3,4} to {3,4}.

Fill in the following table resulting from the given sequence of propagation steps.

Action	SL	SB	SW	SH
Initial Domain	1,2,3	1,2,3,4	1,2,3	1,2,3,4
Propagate SL to SH	1,2,3	1,2,3,4	1,2,3	3,4
Propagate SH to SL	1,2	1,2,3,4	1,2,3	3,4
Propagate SL to SB	1,2	1,3,4	1,2,3	3,4

Propagate SW to SB	1,2	1	1,2,3	3,4
Propagate SB to SW	1,2	1	2,3	3,4

### 6. [21] **Propositional Logic**

(a) [2] One way of defining that two sentences in PL,  $\alpha$  and  $\beta$ , are logically equivalent is to show that  $\alpha < -> \beta$  is a tautology. Give an alternative definition of logical equivalence in terms of entailment.

$$\alpha \mid = \beta$$
 and  $\beta \mid = \alpha$ 

(b) [8] Using the definition you gave as your answer to (a), prove that the clause  $(\neg P1 \lor \neg P2 \lor Q)$  is logically equivalent to the implication sentence  $(P1 \land P2) \to Q$  using the Resolution Refutation algorithm.

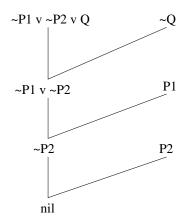
First, prove ( $\neg P1 \lor \neg P2 \lor Q$ ) |= (P1  $\land$  P2)  $\rightarrow$  Q. The premise is already in clause form. Negating the goal and converting it to clause form we get

$$\neg((P1 \land P2) \rightarrow Q) \equiv \neg(\neg(P1 \land P2) \lor Q)$$

$$\equiv \neg(\neg P1 \lor \neg P2 \lor Q)$$

$$\equiv P1 \land P2 \land \neg Q$$

So, there are 3 clauses resulting from the negation of the goal: P1, P2, and  $\neg Q$ . From these 4 clauses we now use the Resolution Refutation algorithm to prove this case:



Second, prove (P1  $\land$  P2)  $\rightarrow$  Q |= (¬P1  $\lor$  ¬P2  $\lor$  Q). Converting the premise to clause form we get

$$(P1 \land P2) \rightarrow Q \equiv \neg (P1 \land P2) \lor Q$$
  
$$\equiv \neg P1 \lor \neg P2 \lor Q$$

Converting the negation of the goal to clause form we get

$$\neg (\neg P1 \lor \neg P2 \lor Q) \equiv P1 \land P2 \land \neg Q$$

So, there are 3 clauses resulting from the negation of the goal: P1, P2, and  $\neg Q$ . These are the same 4 clauses as above, so the same proof tree results. Therefore, the 2 entailments prove the two sentences are logically equivalent.

(c) [8] (i) Using the three propositional symbols, J means "I get the job," H means "I work hard," and P means "I get promoted," convert the following English sentences into three sentences in Propositional Logic.

If I get the job and work hard, I will be promoted. I was not promoted. Thus, either I did not get the job or I did not work hard.

$$(J \land H) \to P$$
$$\neg P$$
$$\neg J \lor \neg H$$

(ii) Give an inference rule based on your sentences in (i) and then prove whether or not it is a sound rule of inference.

$$\frac{(J \land H) \rightarrow P, \neg P}{\neg J \lor \neg H}$$

To prove soundness we must show that whenever the two premises are true, the inferred sentence is also true. We do this by constructing the truth table below, which shows by the 4th, 6th, and 8th rows that yes, this inference rule is sound.

J	H	P	$J \wedge H$	$(J \wedge H) \to P$	$\neg P$	$\neg J \lor \neg H$
T	T	T	T	T	F	F
T	T	F	T	F	T	F
T	F	T	F	T	F	T
T	F	F	F	T	T	T
F	T	T	F	T	F	T
F	T	F	F	T	T	T
F	F	T	F	T	F	T
F	F	F	F	T	T	T

(d) [3] Is the following FOL sentence a tautology (aka valid), a contradiction (aka unsatisfiable), satisfiable, or none of these? Explain your answer using part or all of a truth table.

$$(A \rightarrow \neg B) \rightarrow (C \rightarrow B)$$

It is satisfiable (but not a tautology) as seen by two entries in the truth table: when A=true, B=true, and C=true, the sentence is true, but when A=true, B=false, and C=true, the sentence is false.

## 7. [8] First-Order Logic

For each of the following sentences in English, is the accompanying FOL sentence a good translation? If your answer is no, explain why not and correct it.

(a) [4] "Any course in Computer Science is harder than some courses in Psychology."

$$\forall x \ (Course(x) \land Dept(x, CS)) \rightarrow \exists y \ ((Course(y) \land Dept(y, Psychology)) \rightarrow Harder(x, y))$$

No, with 
$$\exists$$
 use  $\land$ , not  $\rightarrow$ . The correct version is  $\forall x \ (Course(x) \land Dept(x,CS)) \rightarrow \exists y \ ((Course(y) \land Dept(y,Psychology)) \land Harder(x,y))$ 

(b) [4] "If a course is harder than all courses in Math, it must be in Computer Science."

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\forall x \ Course(x) \land (\forall y \ Course(y) \land Dept(y,Math) \land Harder(x,y)) \rightarrow Dept(x,CS)
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No, with \forall use \rightarrow, not \wedge, so correct version is \forall x \ Course(x) \land (\forall y \ Course(y) \land Dept(y,Math) \rightarrow Harder(x,y)) \rightarrow Dept(x,CS)
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