UMBC CMSC 471 Final Exam, 22 May 2017

Please write all of your answers on this exam. The exam is closed book and has seven problems that add up to 130 points. You may use a simple calculator if needed. Use the final scratch sheet as needed. You have the two hours to work on this exam. Good luck.

1. True/False (10 points) Circle T or F for each statement

T F Sound inference algorithms never infer false statements from true ones.

T F Valid sentences are true in all models.

T F Backpropogation is a technique for adjusting weights in a simple, single-layer perceptron.

T F A drawback of a naive bayes classifier is that it requires many joint probability tables.

T F A Support Vector Machine classifier can only be used directly to do binary classification.

T F A single layer perceptron is only capable of learning linearly separable patterns.

T F The ID3 decision tree induction algorithm is guaranteed to find the optimal decision tree consistent with a given training set.

T F Random variables A and B are independent if $p(A \land B) = p(A|B) \times p(B)$

T F If the Blackbox planner finds a plan, it is guaranteed to be an optimal one, i.e., there is no other plan that has fewer steps.

T F Situation calculus is a technique to adjust the weights in a neural network during learning.
2. Short answers (10 points: 2;2;3;3)

2.1 Which of the following are classification tasks appropriate for a classification learning algorithm? Circle the letter for all that apply. [2]

(a) Predicting if a credit card transaction is fraudulent or legitimate
(b) Predicting how much it will rain tomorrow
(c) Predicting the letter of the alphabet represented by an image of a handwritten character
(d) Breaking a database of customers into clusters based on their buying patterns (where the nature of the clusters is determined automatically by the computer, not in any way provided by a human)

2.2 Given a vocabulary with three propositions, A, B and C. How many models are there for the sentence \((A \iff B) \lor B\)? [2]

   a. 6  
   b. 8  
   c. 3  
   d. 2

2.3 Given a linear Support Vector Machine (SVM) that perfectly classifies a set of training data, which training examples, if any, could be removed and still produce the exact same SVM as derived for the original training set? [3]

2.4 Under what condition, if any, would removal of exactly one training example cause the margin of a linear SVM to increase? [3]
3. Resolution Proof in Propositional Logic (30)

A classmate believes (0) UMBC students are clever; (1) if you are clever and you study, you will pass the final; (2) if you are lucky and are either clever or you study you will pass the final.

(a) Construct a KB of propositional sentences using the five propositional variables (umbc, clever, study, lucky, pass) and logical connectives (∧, ∨, ¬, →). Encode each of the three sentences into one or more sentences in conjunctive normal form (CNF). We’ve done the first one for you. (10 pts)

<table>
<thead>
<tr>
<th>#</th>
<th>English</th>
<th>CNF clauses</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>UMBC students are smart.</td>
<td>0.1 ¬UMBC ∨ smart</td>
</tr>
</tbody>
</table>
| 1 | If you are clever and you study, you will pass the final; | 1.1  
    | 1.2  
    | 1.3 |
| 2 | if you are lucky and are either clever or you study you will pass | 2.1  
    | 2.2  
    | 2.3 |

(b) How many models are there for this KB in which UMBC and pass are both True? Recall that a model is an assignment of true and false to each variable such that the KB is consistent? (5 points)
(c) Show a resolution refutation proof that you will pass the final given the facts in your KB and the additional facts that you are a UMBC student and you study. Start with the negation of what’s to be proved, add the given facts and any statements from your KB needed for the proof. Use resolution to derive an empty clause, i.e. a contradiction (⊥). The table to the right shows an example proof (15 points).

<table>
<thead>
<tr>
<th>step</th>
<th>action</th>
<th>result</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>assume</td>
<td>¬ pass</td>
</tr>
<tr>
<td>2</td>
<td>given</td>
<td>¬PvQ</td>
</tr>
<tr>
<td>3</td>
<td>given</td>
<td>P</td>
</tr>
<tr>
<td>4</td>
<td>resolve 2,3</td>
<td>Q</td>
</tr>
<tr>
<td>5</td>
<td>resolve 1,4</td>
<td>⊥</td>
</tr>
</tbody>
</table>

Sample proof of Q given P→Q and P
4. English to logic (25 points, 3 each plus one freebie)

For each English sentence below, write the letter or letters corresponding to its best or closest sentence in first order logic. If none of the FOL sentences correspond to the English sentence, write *none* in the box.

**4.1 Every person plays some game.**

A. $\forall x \forall y \text{Person}(x) \land \text{Game}(y) \land \text{Plays}(x, y)$
B. $\forall x \exists y \text{Person}(x) \land \text{Game}(y) \land \text{Plays}(x, y)$
C. $\forall x \forall y \text{Person}(x) \Rightarrow (\text{Game}(y) \land \text{Plays}(x, y))$
D. $\forall x \exists y \text{Person}(x) \Rightarrow (\text{Game}(y) \land \text{Plays}(x, y))$

**4.2 All games are fun.**

A. $\forall x \text{Game}(x) \land \text{Fun}(x)$
B. $\forall x \text{Game}(x) \Rightarrow \text{Fun}(x)$
C. $\exists x \text{Game}(x) \land \text{Fun}(x)$
D. $\exists x \text{Game}(x) \Rightarrow \text{Fun}(x)$

**4.3 For every game, there is a person that plays that game.**

A. $\forall x \exists y \text{Game}(x) \land \text{Person}(y) \land \text{Plays}(y, x)$
B. $\forall x \exists [\text{Game}(x) \land \text{Person}(y)] \Rightarrow \text{Plays}(y, x)$
C. $\forall x \exists y \text{Game}(x) \Rightarrow [\text{Person}(y) \land \text{Plays}(y, x)]$
D. $\forall x \forall y \text{Game}(x) \land \text{Person}(y) \land \text{Plays}(y, x)$

**4.4 Every person plays every game.**

A. $\forall x \forall y [\text{Person}(x) \land \text{Game}(y)] \Rightarrow \text{Plays}(x, y)$
B. $\forall x \forall y \text{Person}(x) \Rightarrow [\text{Game}(y) \land \text{Plays}(x, y)]$
C. $\forall x \forall y \text{Person}(x) \land \text{Game}(y) \land \text{Plays}(x, y)$
D. $\forall x \exists y [\text{Person}(x) \land \text{Game}(y)] \Rightarrow \text{Plays}(x, y)$

**4.5 There is some person in Pigtown who is smart.**

A. $\forall x \text{Person}(x) \land \text{In}(x, \text{Pigtown}) \land \text{Smart}(x)$
B. $\exists x \text{Person}(x) \land \text{In}(x, \text{Pigtown}) \land \text{Smart}(x)$
C. $\forall x [\text{Person}(x) \land \text{In}(x, \text{Pigtown})] \Rightarrow \text{Smart}(x)$
D. $\exists x \text{Person}(x) \Rightarrow [\text{In}(x, \text{Pigtown}) \land \text{Smart}(x)]$
4.6 Every person in Pigtown is smart.
A. $\forall x \text{ Person}(x) \land \text{In}(x, \text{Pigtown}) \land \text{Smart}(x)$
B. $\exists x \text{ Person}(x) \land \text{In}(x, \text{Pigtown}) \land \text{Smart}(x)$
C. $\forall x \left[ \text{Person}(x) \land \text{In}(x, \text{Pigtown}) \right] \Rightarrow \text{Smart}(x)$
D. $\exists x \text{ Person}(x) \Rightarrow \left[ \text{In}(x, \text{Pigtown}) \right] \land \text{Smart}(x)$

4.7 Some person plays every game.
A. $\exists x \forall y \left[ \text{Person}(x) \land \text{Game}(y) \right] \Rightarrow \text{Plays}(x, y)$
B. $\exists x \forall y \text{ Person}(x) \land \text{Game}(y) \land \text{Plays}(x, y)$
C. $\forall x \forall y \text{ Person}(x) \land \text{Game}(y) \land \text{Plays}(x, y)$
D. $\exists x \forall y \text{ Person}(x) \land \left[ \text{Game}(y) \Rightarrow \text{Plays}(x, y) \right]$
5. Planning in the blocks world domain (10 points)

A friend says that the standard blocks world model could be simplified by modifying the 
(on ?X ?Y) predicate to mean that block ?X is directly on top of ?Y where ?Y can either 
be another block or the table. She claims that we would no longer need the special (on-
table ?X) predicate and could replace the stack, unstack, pickup and putdown actions 
with a single move action, which would be described in PDDL as follows.

(:action move
 :parameters (?obj ?from ?to)
 :precondition (and (on ?obj ?from)
  (clear ?obj)
  (clear ?to)
  (arm-empty))
 :effect (and (not (on ?ob1 ?from))
  (on ?obj ?to)
  (not (clear ?to)))

Assuming that this is the only change made to the blocks world model we discussed in 
class and used in the homework, will her suggestion work? Explain why it will or why it 
won't.
6. Probabilistic Reasoning (20 pts: 5;5;5;5)

Prof. Dumbledore estimates that students who finish their homework forget to submit it 1% of the time and that half the students who have not finished their homework will tell him that they forgot to submit it. He estimates that 90% of the students in his classes complete their homework.

6.1 [5] Let Forgot be a Boolean random variable indicating that the student reports that they forgot to submit their homework and Finished be a Boolean random variable indicating that the student's homework was finished. Fill in the boxes below with the probability values.

<table>
<thead>
<tr>
<th>Probability of</th>
<th>value</th>
</tr>
</thead>
<tbody>
<tr>
<td>P(Finished) =</td>
<td></td>
</tr>
<tr>
<td>P(Forgot</td>
<td>Finished) =</td>
</tr>
<tr>
<td>P(Forgot</td>
<td>¬Finished) =</td>
</tr>
</tbody>
</table>

6.2 [5] What is the probability, P(Forgot), that a student says she forgot to submit their homework? Show your work.

6.3 [5] What is the probability that a student who says she forgot to turn in their homework is telling the truth, i.e., P(Finished | Forgot)? Show your work.

6.4 [5] Fill in the missing values in the following joint probability table assuming that A and B are independent variables.

<table>
<thead>
<tr>
<th></th>
<th>A</th>
<th>¬A</th>
</tr>
</thead>
<tbody>
<tr>
<td>B</td>
<td>3/12</td>
<td>6/12</td>
</tr>
<tr>
<td>¬B</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>
7. Decision trees (25 points)

You are designing an automated system for a lumber mill to distinguish two kinds of wood, oak and pine, based on its sensed qualities. You decide to use a decision tree to do the classification using the ID3 algorithm that selects the variable at each level that maximizes the information gained. The training data is given in the table to the right.

7.1 Which attribute would information gain choose as the root of the tree? (5 pts)

<table>
<thead>
<tr>
<th>Example</th>
<th>Density</th>
<th>Grain</th>
<th>Hardness</th>
<th>Class</th>
</tr>
</thead>
<tbody>
<tr>
<td>Example#1</td>
<td>Heavy</td>
<td>Small</td>
<td>Hard</td>
<td>Oak</td>
</tr>
<tr>
<td>Example#2</td>
<td>Heavy</td>
<td>Large</td>
<td>Hard</td>
<td>Oak</td>
</tr>
<tr>
<td>Example#3</td>
<td>Heavy</td>
<td>Small</td>
<td>Hard</td>
<td>Oak</td>
</tr>
<tr>
<td>Example#4</td>
<td>Light</td>
<td>Large</td>
<td>Soft</td>
<td>Oak</td>
</tr>
<tr>
<td>Example#5</td>
<td>Light</td>
<td>Large</td>
<td>Hard</td>
<td>Pine</td>
</tr>
<tr>
<td>Example#6</td>
<td>Heavy</td>
<td>Small</td>
<td>Soft</td>
<td>Pine</td>
</tr>
<tr>
<td>Example#7</td>
<td>Heavy</td>
<td>Large</td>
<td>Soft</td>
<td>Pine</td>
</tr>
<tr>
<td>Example#8</td>
<td>Heavy</td>
<td>Small</td>
<td>Soft</td>
<td>Pine</td>
</tr>
</tbody>
</table>

7.2 Show the entire decision tree that would be constructed by ID3, i.e., by recursively applying information gain to select the roots of sub-trees after the initial decision. (10 pts)

7.3 Classify these new examples using your decision tree by finning in the last column (10)

<table>
<thead>
<tr>
<th>Example</th>
<th>Density</th>
<th>Grain</th>
<th>Hardness</th>
<th>Class</th>
</tr>
</thead>
<tbody>
<tr>
<td>Example#9</td>
<td>Light</td>
<td>Small</td>
<td>Hard</td>
<td></td>
</tr>
<tr>
<td>Example#10</td>
<td>Light</td>
<td>Small</td>
<td>Soft</td>
<td></td>
</tr>
</tbody>
</table>
Scratch sheet