

## 2006 Comprehensive Examination Solutions

### Artificial Intelligence

**1. Propositional Constraint Satisfaction.** (20 points) Taken from a Final exam in CS188 at UCB.

(a)  $n + 1$  solutions. Once any  $X_i$  is true, all subsequent  $X_j$ s must be true. Hence, each solution consists of  $i$  falses followed by  $n - i$  trues for  $i = 0, \dots, n$ .

(b) Quadratic in  $n$ . Consider what part of the complete binary tree is explored during the search. The algorithm must follow all solutions sequences, which themselves cover a quadratic-sized portion of the tree. Failing branches are those trying a false after a preceding variable is assigned true. Such conflicts are detected immediately.

(c) True. Use the Forward Chaining Algorithm in Russell and Norvig.

(d) True. Directed arc consistency in Russell and Norvig.

**2. Logic.** (20 points) Adapted from a problem in CS157 at Stanford.

- (a) True
- (b) False
- (c) True
- (d) False
- (e) False
- (f) True
- (g) False
- (h) False
- (i) False
- (j) True

**3. Resolution.** (20 points) Adapted from a problem in CS157 at Stanford.

- |  |                    |
|--|--------------------|
| 1. $\{\neg p(x, y), q(x, y, f(x, y))\}$        | Premise a          |
| 2. $\{\neg r(y, z), q(a, y, z)\}$              | Premise b          |
| 3. $\{r(y, z), \neg q(a, y, z)\}$              | Premise b          |
| 4. $\{p(x, g(x)), q(x, g(x), z)\}$             | Premise c          |
| 5. $\{\neg r(x, y), \neg q(x, w, z)\}$         | Negated Goal       |
| 6. $\{\neg q(a, x, y), \neg q(x, w, z)\}$      | 3, 5               |
| 7. $\{q(x, g(x), f(x, g(x))), q(x, g(x), z)\}$ | 1, 4               |
| 8. $\{\neg q(g(a), w, z)\}$                    | 6, 7 (factoring 7) |
| 9. $\{\}$                                      | 7, 8 (factoring 7) |

**4. Bayes Nets.** Taken from a Final exam in CS188 at UCB.

(a) Assertions (2) and (3) are implied by the structure of the net; assertion (1) is not.

$$(b) p(b, i, \neg m, g, j) = p(b) * p(\neg m) * p(i | b, \neg m) * p(g | b, i, m) * p(j | g) \\ = 0.9 * 0.9 * 0.5 * 0.8 * 0.9 = 0.2916$$

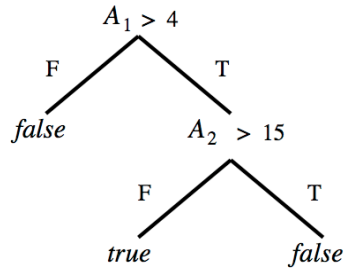
(c) Since B, I, M are true in the evidence, we can treat G as having a prior of 0.9 and look at the submodel with just G and J.

$$p(j | b, g, m) = p(j | g) * p(g) = 0.9 * 0.9 = 0.81$$

That is, the probability of going to jail is 0.81.

**5. Learning.** Taken from a Final exam in CS188 at UCB.

(a) One possibility follows.



(b) With 2 examples of each kind, the initial entropy is 1 bit. After the test, we have one subset with counts 0, 1 and one subset with counts 2, 1. Hence the information gain is as follows.

$$\begin{aligned} & 1 - \left( \frac{1}{4} * 0 + \frac{3}{4} * \left( -\frac{1}{3} \log\left(\frac{1}{3}\right) - \frac{2}{3} \log\left(\frac{2}{3}\right) \right) \right) \\ & = 1 + \frac{1}{4} * \log\left(\frac{1}{3}\right) + \frac{1}{2} * \log\left(\frac{2}{3}\right) \\ & \sim 0.3113 \text{ bits} \end{aligned}$$

(c) False. A test-once tree can with one attribute creates exactly two regions on the real line, whereas the data may alternate along the line.

(d) True. A test-many tree can define arbitrarily small hyper-rectangles, each containing exactly one example.