

# 15780: GRADUATE AI (SPRING 2017)

## Practice Midterm Exam (Solutions)

February 23, 2017

Topic	Total Score	Score
Heuristic Search	25	
VC Dimension	25	
Integer Programming	25	
Convex Optimization	25	
Total	100	

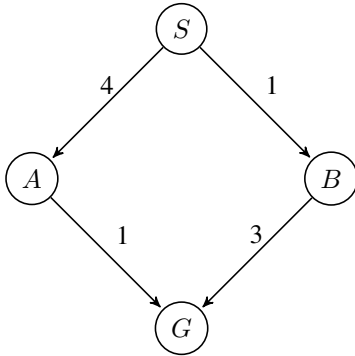
# 1 Heuristic Search [25 points]

Consider the problem of informed search with a heuristic. For each state  $x$ , let  $h^*(x)$  be the length of the cheapest path from  $x$  to a goal.

Prove or disprove the following statements:

- 1.1 [15 points] If  $h(x) = 2h^*(x)$  for all states  $x$ , then  $A^*$  tree search with the heuristic  $h$  is optimal.

**Solution:** This is false. We give a counterexample.



Note that  $f(A) = 4 + 2(1) = 6$ ,  $f(B) = 1 + 2(3) = 7$  so  $A^*$  will expand node A first. Then from node A, we have  $f(G) = 5$ , so we will expand node G and return the path S-A-G as the optimal path. However this is not the real optimal path. The real optimal path is S-B-G with a cost of 4.

- 1.2 [10 points] If  $h$  is a consistent heuristic,  $A^*$  graph search with the heuristic  $h'(x) = h(x)/2$  is optimal.

**Solution:** This is true. Since  $h$  is consistent, this means for any node  $x$  and its successor  $x'$  we know that  $h(x) \leq c(x, x') + h(x')$ . This implies  $h(x)/2 \leq c(x, x')/2 + h(x')/2$ . Since costs are nonnegative, this also implies that  $h(x)/2 \leq c(x, x') + h(x')/2$ . Thus  $h'$  is also consistent and we know that  $A^*$  graph search with a consistent heuristic is optimal.

## 2 Learning Theory [25 points]

Determine the VC dimension of the following function classes.

- 2.1 [15 points] Define  $F$  to be the set of strings of length 3 composed of the symbols 0, 1, and \*. Each  $f \in F$  acts as a pattern matcher; i.e., when applied to a binary string  $s$ , it either accepts or rejects  $s$ . For example, when we apply the schema  $f = 1**$  to the string  $s = 101$ , it accepts, and when we apply  $f$  to  $s' = 010$ , it rejects. What is the VC dimension of  $F$ ?

**Solution:** The VC dimension is 3. The set  $\{001, 010, 100\}$  can be shattered. For any set of size 4, note that if there are any two strings that differ at all three positions (call them  $s$  and  $s'$ ), then the set  $+\{s, s'\}$  can only be labeled with three wildcard characters, which also matches the rest of the strings not labeled  $+$ . Further, this means that there must be at least two pairs of strings at distance two from each other. In order to see this, put the strings on the vertices of a cube connected by edges between strings that differ from one another at exactly one position. Now, note that we can't realize this split. A pattern that matches one pair of strings must necessarily also match one string in the other pair. Concretely, this is because a pattern that matches strings  $s_1$  and  $s_2$  that differ in two positions must have two wildcards, and the third position, which is shared by  $s_1$  and  $s_2$ , must differ between  $s_3$  and  $s_4$ , meaning that one of  $s_3$  and  $s_4$  must match the pattern as well.

- 2.2 [10 points] The union of  $n$  intervals on the real line.

**Solution:** The VC dimension is  $2n$ . It's pretty clear that we can shatter  $2n$  points, as this is equivalent to essentially using one interval for every consecutive pair of adjacent points. It's also not possible to shatter  $2n + 1$  points because the assignment that alternates between  $+1$  and  $-1$  needs  $n + 1$  intervals.

### 3 Integer Programming [25 points]

Consider an undirected graph  $G = (V, E)$ . A *minimum dominating set* is a smallest subset  $S$  of  $V$  such that every node not in  $S$  is adjacent to at least one node in  $S$ . A *minimum independent dominating set* is a smallest subset  $S$  of  $V$  such that (1) every node not in  $S$  is adjacent to at least one node in  $S$  and (2) no pair of nodes in  $S$  are adjacent. In your answer, you can use  $N(i)$  to denote the set of neighbors of node  $i$  (i.e.,  $N(i)$  is a set of nodes adjacent to  $i$ ) for each node  $i \in V$ . Note that  $i \notin N(i)$ . You also can use  $(i, j) \in E$  to denote the edge between node  $i \in V$  and node  $j \in V$ .

3.1 [15 points] Formulate an integer linear program to find a minimum dominating set.

**Solution:** Consider the following the integer program:

$$\begin{aligned} & \text{minimize} && \sum_{i=1}^n x_i \\ & \text{subject to} && \sum_{j \in N(i) \cup \{i\}} x_j \geq 1, \forall i \in V, \\ & && \text{and } x_i = \{0, 1\}, \forall i \in V. \end{aligned}$$

If you solve this integer program,  $S = \{i \in V : x_i = 1\}$  is a minimum dominating set.

3.2 [10 points] Formulate an integer linear program to find a minimum independent dominating set.

**Solution:** Consider the following the integer program:

$$\begin{aligned} & \text{minimize} && \sum_{i=1}^n x_i \\ & \text{subject to} && \sum_{j \in N(i) \cup \{i\}} x_j \geq 1, \forall i \in V, \\ & && \text{and } x_i + x_j \leq 1, \forall (i, j) \in E, \\ & && \text{and } x_i = \{0, 1\}, \forall i \in V. \end{aligned}$$

If you solve this integer program,  $S = \{i \in V : x_i = 1\}$  is a minimum independent dominating set.

## 4 Convex Optimization [25 points]

Consider a linear program of the standard form: minimize  $\mathbf{c}^T \mathbf{x}$  such that  $\mathbf{A}\mathbf{x} \leq \mathbf{b}$ . Here  $\mathbf{x} \in \mathbb{R}^n$  is the vector of variables, and  $\mathbf{c} \in \mathbb{R}^n$ ,  $\mathbf{A} \in \mathbb{R}^{m \times n}$ , and  $\mathbf{b} \in \mathbb{R}^m$  are constants.

Prove from the definitions that this is a convex program.

**Solution:** First, we show that the objective function is linear, which we denote by  $f$ . Let  $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$ , and let  $0 \leq \theta \leq 1$ . We need to show  $f(\theta\mathbf{x} + (1 - \theta)\mathbf{y}) \leq \theta f(\mathbf{x}) + (1 - \theta)f(\mathbf{y})$ . We have

$$\begin{aligned} f(\theta\mathbf{x} + (1 - \theta)\mathbf{y}) &= \mathbf{c}^T(\theta\mathbf{x} + (1 - \theta)\mathbf{y}) = \mathbf{c}^T(\theta\mathbf{x}) + \mathbf{c}^T((1 - \theta)\mathbf{y}) \\ &= \theta\mathbf{c}^T\mathbf{x} + (1 - \theta)\mathbf{c}^T\mathbf{y} = \theta f(\mathbf{x}) + (1 - \theta)f(\mathbf{y}). \end{aligned}$$

Thus, we conclude that the desired inequality holds (in fact, it holds with equality). Next, we show that the feasible region  $\mathcal{F} = \{\mathbf{x} \in \mathbb{R}^n : \mathbf{A}\mathbf{x} \leq \mathbf{b}\}$  is convex. For this, let  $\mathbf{x}, \mathbf{y} \in \mathcal{F}$  and let  $0 \leq \theta \leq 1$ . We need to show that  $\theta\mathbf{x} + (1 - \theta)\mathbf{y} \in \mathcal{F}$  as well, which amounts to showing  $\mathbf{A}(\theta\mathbf{x} + (1 - \theta)\mathbf{y}) \leq \mathbf{b}$ . We have

$$\begin{aligned} \mathbf{A}(\theta\mathbf{x} + (1 - \theta)\mathbf{y}) &= \mathbf{A}(\theta\mathbf{x}) + \mathbf{A}((1 - \theta)\mathbf{y}) = \theta\mathbf{A}\mathbf{x} + (1 - \theta)\mathbf{A}\mathbf{y} \\ &\leq \theta\mathbf{b} + (1 - \theta)\mathbf{b} = (\theta + 1 - \theta)\mathbf{b} = \mathbf{b}. \end{aligned}$$

This completes the proof that  $\mathcal{F}$  is convex, and hence the proof that a linear program is a convex program.