

# Homework 7 for “Algorithms for Big-Data Analysis”

Note: Please write up your solutions independently. If you get significant help from others, write down the source of references. A formal mathematical proof for all your claims is required.

1. Derive the dual optimization problem for

$$\begin{aligned} \min_{w,b,\xi} \quad & \frac{1}{2} \|w\|_2^2 + C_1 \sum_{i=1}^n \xi_i + C_2 \sum_{i=1}^n \xi_i^2 \\ \text{s.t.} \quad & y_i \cdot (x_i \cdot w + b) \geq 1 - \xi_i, \forall i = 1, \dots, n \\ & \xi_i \geq 0, \forall i = 1, \dots, n \end{aligned}$$

2. Properties of Submodular Functions

- (a) Prove that any non-negative submodular function is also subadditive, i.e. if  $F : 2^X \rightarrow \mathbb{R}_+$  is submodular then  $F(S \cup T) \leq F(S) + F(T)$  for any  $S, T \subseteq X$ . Here,  $\mathbb{R}_+ = \{x \in \mathbb{R}, x \geq 0\}$ .
  - (b) Prove that a function  $F : 2^X \rightarrow \mathbb{R}_+$  is submodular if and only if for any  $S, T \subseteq X$ , the marginal contribution function  $F_S(T) = F(S \cup T) - F(S)$  is subadditive. (If the statement is not true, please either add a condition to make it correct or give a counterexample.)
3. Consider a graph  $(V, E)$ , where  $V$  is the set of nodes and  $E$  is the set of edges. Let  $S$  be a subset of  $V$  and  $V \setminus S$  be the complement of  $S$ . Define  $f(S)$  be the number of edges  $e = (u, v)$  such that  $u \in S$  and  $v \in V \setminus S$ . Prove that  $f(S)$  is submodular.
  4. Exercise 3.4 in <http://incompleteideas.net/book/RLbook2020.pdf>
  5. Exercise 3.23 in <http://incompleteideas.net/book/RLbook2020.pdf>

# Homework 7

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June 3, 2020

## 1 Problem 1

The Lagrange function can be written as

$$L(w, \xi, \lambda, \mu) = \frac{1}{2} \|w\|_2^2 + \sum_{i=1}^n C_1 \xi_i + C_2 \xi_i^2 - \lambda_i (y_i (x_i \cdot w + b) - 1 + \xi_i) - \mu_i \xi_i$$

By taking the derivative

$$\begin{aligned}\frac{\partial L}{\partial w} &= w - \sum_{i=1}^n \lambda_i y_i x_i = 0 \\ \frac{\partial L}{\partial \xi_i} &= C_1 + 2C_2 \xi_i - \lambda_i - \mu_i = 0 \\ \frac{\partial L}{\partial b} &= \sum_{i=1}^n \lambda_i y_i = 0\end{aligned}$$

we have

$$L(w^*, \xi^*, \lambda, \mu) = -\frac{1}{2} \left\| \sum_{i=1}^n \lambda_i y_i x_i \right\|_2^2 + \sum_{i=1}^n \lambda_i - \frac{(\lambda_i + \mu_i - C_1)^2}{4C_2}$$

Hence, the dual problem can be formulated as

$$\begin{aligned}\min_{\lambda, \mu} \quad & \frac{1}{2} \left\| \sum_{i=1}^n \lambda_i y_i x_i \right\|_2^2 - \sum_{i=1}^n \lambda_i + \sum_{i=1}^n \frac{(\lambda_i + \mu_i - C_1)^2}{4C_2} \\ \text{s.t.} \quad & \lambda_i, \mu_i \geq 0, \sum_{i=1}^n \lambda_i y_i = 0\end{aligned}$$

## 2 Problem 2

### 2.1 (a)

For any  $S, T \subset 2^X$ , according to the definition:

$$F(S) + F(T) \geq F(S \cup T) + F(S \cap T) \geq F(S \cup T)$$

hence  $f$  is subadditive.

### 2.2 (b)

- Sufficiency: Since  $F_S(T)$  is subadditive. We have

$$F_{A \cap B}((A - B) \cup (B - A)) \leq F_{A \cap B}(A - B) + F_{A \cap B}(B - A)$$

Hence,

$$F(A \cup B) - F(A \cap B) \leq F(A) + F(B) - 2F(A \cap B)$$

Thus,  $F$  is submodular.

- Necessity:

Since  $F_S$  is additive if and only if

$$F(A \cup C) + F(B \cup C) \geq F(C) + F(A \cup B \cup C)$$

Owing to the submodularity of  $F$ , we have

$$F(A \cup C) + F(B \cup C) \geq F(C \cup (A \cap B)) + F(A \cup B \cup C)$$

Hence, **we need an additional condition:**  $F$  is monotone. The results follows from the monotone condition

$$F(C \cup (A \cap B)) \geq F(C)$$

### 2.3 Counterexample

For a given finite, non-empty set  $X$ , we take an element  $a \in X$ , and define the function  $F(A)$ ,  $A \subset X$  as

$$F(A) = \begin{cases} |A|, & \text{if not } a \in A \\ 0 & \text{if } a \in A. \end{cases} \quad (1)$$

For any  $A \subset B \subset X$ , and  $x \in X/B$ , if  $a \in A$ , then

$$F(B \cup \{s\}) - F(B) = F(A \cup \{s\}) - F(A) = 0 \quad (2)$$

the inequality holds, and if  $a = s$

$$F(B \cup \{s\}) - F(B) = -F(B) = -|B| \leq -|A| = F(A \cup \{s\}) - F(A) \quad (3)$$

the inequality holds as well. For other cases, we have

$$F(B \cup \{s\}) - F(B) \leq 1 = F(A \cup \{s\}) - F(A) \quad (4)$$

However, if we take  $S = T = \{a\}$ ,  $R = b \neq a$ , we have

$$F_R(S) = F_R(T) = F(R \cup T) - F(R) = -1, \quad F_R(S \cup T) = F(S \cup R \cup T) - F(R) = -1 \quad (5)$$

Hence,  $F_R(S) + F_R(T) < F_R(S \cup T)$ , which means  $F_R$  is not subadditive.

### 3 Problem 3

Denote the number of edges connecting set  $A$  and  $B$  is  $L(A, B)$ . For any set  $S, T \subset V$ , denote sets  $S - T, T - S, S \cap T, V - S \cup T$  as  $A, B, C, D$  separately.

Since we have

$$\begin{aligned} f(S) &= L(A, D) + L(A, B) + L(C, D) + L(C, B) \\ f(T) &= L(A, C) + L(A, B) + L(B, D) + L(C, D) \\ f(S \cap T) &= L(A, C) + L(D, C) + L(B, C) \\ f(S \cup T) &= L(A, D) + L(B, D) + L(C, D) \end{aligned}$$

Clearly, we have

$$f(S) + f(T) - f(S \cap T) - f(S \cup T) = 2L(A, B) \geq 0$$

Hence,  $f$  is submodular.

### 4 Problem 4

According to the the transition graph of the finite MDP

$s$	$a$	$s'$	$r$	$p(s', r s, a)$
high	search	high	$r_{\text{search}}$	$\alpha$
high	search	low	$r_{\text{search}}$	$1 - \alpha$
low	search	high	$-3$	$1 - \beta$
low	search	low	$r_{\text{search}}$	$\beta$
high	wait	high	$r_{\text{wait}}$	$1$
low	wait	low	$r_{\text{wait}}$	$1$
low	recharge	high	$0$	$1$

## 5 Problem 5

According to the definition, the Bellman equation can be written as

$$q_{\star}(\text{high}, \text{search}) = \alpha(r_{\text{search}} + \gamma \max_{a'} q_{\star}(\text{high}, a')) \quad (6a)$$

$$+ (1 - \alpha)(r_{\text{search}} + \gamma \max_{a'} q_{\star}(\text{low}, a')) \quad (6b)$$

$$q_{\star}(\text{high}, \text{wait}) = r_{\text{wait}} + \gamma \max_{a'} q_{\star}(\text{high}, a') \quad (6c)$$

$$q_{\star}(\text{low}, \text{search}) = (1 - \beta)(-3 + \gamma \max_{a'} q_{\star}(\text{high}, a')) \quad (6d)$$

$$+ \beta(r_{\text{search}} + \gamma \max_{a'} q_{\star}(\text{low}, a')) \quad (6e)$$

$$q_{\star}(\text{low}, \text{wait}) = r_{\text{wait}} + \gamma \max_{a'} q_{\star}(\text{low}, a') \quad (6f)$$

$$q_{\star}(\text{low}, \text{recharge}) = \gamma \max_{a'} q_{\star}(\text{high}, a') \quad (6g)$$

where

$$\max_{a'} q_{\star}(\text{high}, a') = \max\{q_{\star}(\text{high}, \text{search}), q_{\star}(\text{high}, \text{wait})\} \quad (7a)$$

$$\max_{a'} q_{\star}(\text{low}, a') = \max\{q_{\star}(\text{low}, \text{search}), q_{\star}(\text{low}, \text{wait}), q_{\star}(\text{low}, \text{recharge})\} \quad (7b)$$