

## Practice Questions: Submodular functions

CS768 Learning With Graphs      Prof. Abir De

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### 1) Submodularity, marginal gains, and diminishing returns

Let  $V$  be a finite ground set and  $f : 2^V \rightarrow \mathbb{R}$ . Recall the following definitions:

- $f$  is **submodular** if  $f(A) + f(B) \geq f(A \cup B) + f(A \cap B)$  for all  $A, B \subseteq V$ .
- The **marginal gain** is  $\Delta_f(e | S) := f(S \cup \{e\}) - f(S)$ .
- $f$  is **monotone** if  $A \subseteq B \Rightarrow f(A) \leq f(B)$ , and **normalized** if  $f(\emptyset) = 0$ .

(a) Prove that  $f$  is submodular if and only if it satisfies the **diminishing returns** property: for all  $A \subseteq B \subseteq V$  and  $e \in V \setminus B$ ,

$$\Delta_f(e | A) \geq \Delta_f(e | B).$$

(b) Using the diminishing returns property, determine whether the following functions  $f : 2^V \rightarrow \mathbb{R}$  are submodular:

- $f(S) = |S|$
- $f(S) = \min\{|S|, k\}$  for some constant  $k$ .
- $f(S) = |S|^2$

### 2) Modular and Supermodular functions, and properties of submodular functions

(a) Prove the following **closure properties**: if  $f$  and  $g$  are submodular functions, then  $f + g$ ,  $\alpha f$  (for  $\alpha \geq 0$ ), and  $f + c$  (for constant  $c$ ) are also submodular. Additionally, show that the restriction of a submodular function to a subset of the ground set preserves submodularity.

(b) A function  $m$  is **modular** if  $m(S) = \sum_{i \in S} w_i$  for some weights  $w_i$ . Prove that a function is modular if and only if it is both submodular and supermodular.

(c) Let  $g : \mathbb{Z}_{\geq 0} \rightarrow \mathbb{R}$  be a nondecreasing concave function. Prove that the set function defined by  $f(S) = g(|S|)$  is submodular.

(d) Define the **complement-transform** of  $f$  by  $f^c(S) = f(V \setminus S)$ . Prove that  $f$  is submodular if and only if  $f^c$  is submodular.

(e) Prove that if  $f$  is a normalized, monotone, and submodular function, then it is **subadditive**:  $f(A \cup B) \leq f(A) + f(B)$  for all  $A, B \subseteq V$ .

### 3) Lovász extension

Let  $V = \{1, 2, \dots, n\}$  and  $f : 2^V \rightarrow \mathbb{R}$  with  $f(\emptyset) = 0$ . For  $x \in [0, 1]^n$ , the **Lovász extension** is defined as:

$$\hat{f}(x) := \sum_{k=1}^n (x_{(k)} - x_{(k+1)}) f(S_k),$$

where  $x_{(1)} \geq x_{(2)} \geq \dots \geq x_{(n)}$  are the sorted coordinates,  $x_{(n+1)} = 0$ , and  $S_k = \{(1), \dots, (k)\}$ .

(a) Show that  $\hat{f}(\mathbf{1}_S) = f(S)$  for all  $S \subseteq V$ , where  $\mathbf{1}_S$  is the indicator vector of  $S$ .

(b) Prove that  $f$  is submodular if and only if its Lovász extension  $\hat{f}$  is convex on  $[0, 1]^n$ .

### 4) Greedy maximization under a cardinality constraint

Let  $f : 2^V \rightarrow \mathbb{R}_{\geq 0}$  be a normalized, monotone, and submodular function. Consider the problem  $\max\{f(S) : |S| \leq k\}$ . Let  $S^*$  be an optimal solution and  $S_t$  be the set produced by the greedy algorithm after  $t$  iterations.

(a) Prove the key greedy inequality:

$$\max_{e \in V \setminus S_t} \Delta_f(e \mid S_t) \geq \frac{f(S^*) - f(S_t)}{k}.$$

(b) Let  $G_t = f(S_t)$ . Using the inequality from part (a), derive a recurrence for  $G_t$  and solve it to show that:

$$f(S_k) \geq \left(1 - \left(1 - \frac{1}{k}\right)^k\right) f(S^*).$$

Conclude that the greedy algorithm provides a  $1 - 1/e$  approximation guarantee.

## 5) The Multilinear Extension

Let  $f : 2^V \rightarrow \mathbb{R}$  be a submodular function. The **multilinear extension**  $F : [0, 1]^V \rightarrow \mathbb{R}$  is defined as  $F(x) = \mathbb{E}[f(R_x)]$ , where  $R_x$  is a random set including each element  $i$  independently with probability  $x_i$ .

(a) Prove the **Partial Derivative Identity**:

$$\frac{\partial F}{\partial x_i}(x) = \mathbb{E}[\Delta_f(i \mid R_x \setminus \{i\})].$$

(b) Prove that if  $f$  is monotone, then  $F$  is monotone in each coordinate.

(c) Prove that if  $f$  is submodular,  $F$  satisfies **DR-submodularity** (Continuous Diminishing Returns): for any  $x \leq y$  coordinate-wise,  $\nabla F(x) \geq \nabla F(y)$ .