

# Submodular Maximization and Influence Maximization

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# Matroid

- A set  $(\mathcal{N}, \mathcal{I})$  is matroid if:
  1. The empty set is independent.
    - $\emptyset \in \mathcal{I}$
  2. Every subset of an independent set is independent.
    - For each  $S \subseteq T \subseteq \mathcal{N}$ , if  $T \in \mathcal{I}$ , then  $S \in \mathcal{I}$
  3. If  $A$  and  $B$  are two independent set and  $|A| > |B|$ , then there exists  $x \in A \setminus B$  such that  $B \cup \{x\}$  is an independent set.
- Ex. No cycles subgraph

# Matroid

- $w(e) \geq 0$
- The greedy algorithm can be used to find a **maximum-weight basis** of the matroid, by starting from the empty set and repeatedly adding one element at a time, at each step choosing a maximum-weight element among the elements whose addition would preserve the independence of the augmented set.
- Ex. Minimum spanning tree problem

# Submodular function

- If  $\Omega$  is a finite set, a submodular function is a set function  $f: 2^\Omega \rightarrow \mathbb{R}_+$ , where  $2^\Omega$  denotes the power set of  $\Omega$ , which satisfies one of the following equivalent definitions.
  1. For every  $X, Y \subseteq \Omega$  with  $X \subseteq Y$  and every  $x \in \Omega \setminus Y$  we have that  $f(X \cup \{x\}) - f(X) \geq f(Y \cup \{x\}) - f(Y)$ .
  2. For every,  $S, T \subseteq \Omega$  we have that  $f(S) + f(T) \geq f(S \cup T) + f(S \cap T)$ .
  3. For every  $X \subseteq \Omega$  and  $x_1, x_2 \in \Omega \setminus X$  we have that  $f(X \cup \{x_1\}) + f(X \cup \{x_2\}) \geq f(X \cup \{x_1, x_2\}) + f(X)$ .

# Submodular function

- **Monotone** submodular maximization problem
- Greedy algorithm always produces a solution whose value is at least  $1 - \left(1 - \frac{1}{k}\right)^k$  times the optimal value. This bound can be achieved for each  $k$  and has a limiting value of  $1 - \frac{1}{e}$ .
- Ex. Maximum coverage problem

$$f \text{ is monotone if}$$
$$A \subset B \Rightarrow f(A) \leq f(B)$$

# Supermodular function

- If  $\Omega$  is a finite set, a supermodular function is a set function  $f: 2^\Omega \rightarrow \mathbb{R}_+$ , where  $2^\Omega$  denotes the power set of  $\Omega$ , which satisfies one of the following equivalent definitions.
  1. For every  $X, Y \subseteq \Omega$  with  $X \subseteq Y$  and every  $x \in \Omega \setminus Y$  we have that  $f(X \cup \{x\}) - f(X) \leq f(Y \cup \{x\}) - f(Y)$ .
  2. For every,  $S, T \subseteq \Omega$  we have that  $f(S) + f(T) \leq f(S \cup T) + f(S \cap T)$ .
  3. For every  $X \subseteq \Omega$  and  $x_1, x_2 \in \Omega \setminus X$  we have that  $f(X \cup \{x_1\}) + f(X \cup \{x_2\}) \leq f(X \cup \{x_1, x_2\}) + f(X)$
- **Ex. Set cover problem** ( $(H(n))$ -approximation)

# Maximizing the spread of influence in a social network

- For an arbitrary instance of the Linear Threshold Model, the resulting influence function is **submodular**.
- For an arbitrary instance of the Independent Cascade Model, the resulting influence function is **submodular**.
- $(1 - \frac{1}{e} - \varepsilon)$ -approximation algorithm by using  $(1 + \gamma)$ -approximate values for the function to be optimized