

Through the Lens of Sequence Submodularity

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Classical Submodularity

Ω finite set, $F : 2^\Omega \rightarrow \mathbb{R}$ submodular

$$F(A \cup \{\sigma\}) - F(A) \geq F(B \cup \{\sigma\}) - F(B), \quad A \subseteq B \subseteq \Omega, \sigma \in \Omega$$

Common property that expresses a diminishing return condition (e.g. coverage, entropy, mutual information, detection probability)

Maximization via **greedy algorithm** has provable good performance:
 $A^n = \{\sigma_1, \dots, \sigma_n\}, \sigma_n = \operatorname{argmax}_{\sigma \in \Omega} F(A^{n-1} \cup \{\sigma\})$

Theorem

$F : 2^\Omega \rightarrow \mathbb{R}$ is monotonic and submodular. Let O^T be a maximizing set for F among the sets of cardinality T . Then,

$$F(A^T) \geq \left(1 - \frac{1}{e}\right) F(O^T)$$

Extension to Sequences

Ω finite set, $\mathbb{H}(\Omega)$ set of sequences over Ω , $F : \mathbb{H}(\Omega) \rightarrow \mathbb{R}$

- ▶ *backward monotonic* if $F(\sigma \perp S) \geq F(S)$, $S \in \mathbb{H}(\Omega)$, $\sigma \in \Omega$
- ▶ *forward submodular* if

$$F(S \perp \sigma) - F(S) \geq F(S \perp R \perp \sigma) - F(S \perp R), \quad R, S \in \mathbb{H}(\Omega), \sigma \in \Omega$$

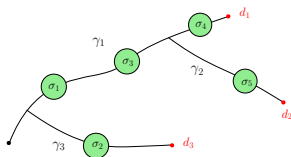
Greedy algorithm: $S^n = (\sigma_1, \dots, \sigma_n)$, $\sigma_n = \operatorname{argmax}_{\sigma \in \Omega} F(S^{n-1} \perp \sigma)$

Theorem

$F : \mathbb{H}(\Omega) \rightarrow \mathbb{R}$ is backward monotonic and forward submodular. Let O^T be a maximizing sequence for F among the sequences of length T . Then,

$$F(S^T) \geq \left(1 - \frac{1}{e}\right) F(O^T)$$

Search & Tracking



UAV is tasked with searching for a moving target and tracking it to destination. To find the target, the UAV selects a set of flight search patterns Ω based on the possible paths Γ that the target might follow.

$$\sigma \in \Omega \rightarrow \begin{cases} t(\sigma) & \text{reference time} \\ \phi_\sigma & \text{detection probability} \end{cases}$$

$S = (S_1, \dots, S_n)$, $S_i \in \Omega$, $F(S)$ detection probability

Problem: Minimize the average detection time

$$\min_{|S| \leq T} \sum_{k=1}^T t(S_k) [F(S|_1^k) - F(S|_1^{k-1})] + K(1 - F(S))$$
$$\max_{|S| \leq T} \sum_{k=1}^T (K - t(S_k)) [F(S|_1^k) - F(S|_1^{k-1})]$$

Recommender Systems

Ω set of movies, $g : \Omega \rightarrow [0, 1]$ user's satisfaction probability

\mathcal{T} set of topics, $t(\sigma) \subseteq \mathcal{T}$ subset of topics covered by $\sigma \in \Omega$

S sequence of movies generated by the recommender system

$F(S) = |\cup_k t(S_k)|$ number of topics covered by S

User chooses a topic t in \mathcal{T} uniformly at random, picks the first item S_k in the sequence for which $t \in t(S_k)$ and is satisfied with probability $g(S_k)$

Problem: Maximization of the user's satisfaction probability

$$\max_{|S| \leq T} \frac{1}{|\mathcal{T}|} \sum_{k=1}^T g(S_k) [F(S|_1^k) - F(S|_1^{k-1})]$$

Optimization Problem

- ▶ Ω finite set, $\mathbb{H}(\Omega)$ set of sequences on Ω
- ▶ $g : \Omega \rightarrow \mathbb{R}^+$ any function, $F : \mathbb{H}(\Omega) \rightarrow \mathbb{R}$ permutation invariant, monotonic, and submodular
- ▶ $F_g(S) = \sum_{k=1}^n g(S_k) [F(S|_1^k) - F(S|_1^{k-1})]$ for $S = (S_1, \dots, S_n)$

$$\max_{S \in \mathbb{H}(\Omega), |S| \leq T} F_g(S)$$

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$$\max_{S \in \mathbb{H}(\Omega), |S| \leq T} F_g(S) = \max_{S \in \mathcal{I}(g), |S| \leq T} F_g(S)$$

where $\mathcal{I}(g) = \{S \in \mathbb{H}(\Omega) \mid g(S_i) \geq g(S_{i+1}) \forall i\}$

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Remark: other choices for the domain are possible (e.g. $S_i \neq S_j$ for $i \neq j$)

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F_g is forward submodular but not backward monotonic!

Generalized Greedy Algorithm

$$J : \mathbb{H}(\Omega) \rightarrow \mathbb{R}, \mathcal{I} \subseteq \mathbb{H}(\Omega), \max_{S \in \mathcal{I}, |S| \leq T} J(S)$$

The algorithm produces recursively an ordered sequence $S^T = (S_1^T, \dots, S_T^T) \in \mathcal{I}$ as follows:

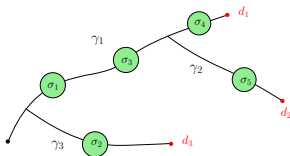
▶ $S^1 = (S_1^1)$ where $S_1^1 \in \operatorname{argmax}_{\sigma \in \Omega} J(\sigma)$;

▶ Given $S^k = (S_1^k, \dots, S_k^k) \in \mathcal{I}$, we define

$$S^{k+1} = \operatorname{argmax}_{\substack{U = (S_1^k, \dots, S_h^k, \sigma, S_{h+1}^k, \dots, S_k^k) \\ U \in \mathcal{I}}} J(U)$$

Instead of simply augmenting the sequence on the right hand side, we allow each new element to be placed in **any position** among the elements that are already present in the sequence

Example: Search & Tracking



$$\Omega = \{\sigma_1, \sigma_2, \sigma_3, \sigma_4, \sigma_5\}, t(\sigma_i) = i, K = 10$$

$$\phi_{\sigma_1} = 0.2, \phi_{\sigma_2} = \phi_{\sigma_4} = 0.5, \phi_{\sigma_3} = 0.3, \phi_{\sigma_5} = 0.95$$

$$\blacktriangleright T = 1: \operatorname{argmax}_{\sigma \in \Omega} (K - t(\sigma))F(\sigma) = \{\sigma_5\}$$

Classical greedy stops here! $\Rightarrow S = (\sigma_5)$, Detection time = 8,40s

$$\blacktriangleright T = 2: \operatorname{argmax}_{\sigma \in \Omega} [(K - t(\sigma))F(\sigma) + (K - t(\sigma_5))(F(\sigma \perp \sigma_5) - F(\sigma))] = \{\sigma_1\}$$

Generalized greedy: $\Rightarrow S = (\sigma_1, \sigma_5)$, Detection time = 7,73s

Technical Results: \mathcal{I} -Monotonicity and \mathcal{I} -Submodularity

Definition

Let $\mathcal{I} \subseteq \mathbb{H}(\Omega)$ and $J : \mathcal{I} \rightarrow \mathbb{R}$.

- ▶ J is called *\mathcal{I} -monotonic* if for every $Q, R \in \mathcal{I}$ and $\sigma \in \Omega$ such that $Q \perp \sigma \perp R \in \mathcal{I}$, it holds:

$$J(Q \perp \sigma \perp R) \geq J(Q \perp R) \quad (1)$$

- ▶ J is called *\mathcal{I} -submodular* if for every $Q, R, S \in \mathcal{I}$ and $\sigma_1, \sigma_2 \in \Omega$ such that $Q \perp \sigma_1 R \perp \sigma_2 \perp S \in \mathcal{I}$, it holds:

$$\begin{aligned} & J(Q \perp \sigma_1 \perp R \perp \sigma_2 \perp S) - J(Q \perp \sigma_1 \perp R \perp S) \\ & \leq J(Q \perp R \perp \sigma_2 \perp S) - J(Q \perp R \perp S) \end{aligned} \quad (2)$$

Technical results: Fully Extendable Sets

$Q, R \in \mathbb{H}(\Omega)$, $Q \leq R$ means that Q can be obtained from R eliminating some elements

Definition

A subset $\mathcal{I} \subseteq \mathbb{H}(\Omega)$ is called *fully extendable* if the following conditions are satisfied:

1. For every $\sigma \in \Omega$, $(\sigma) \in \mathcal{I}$;
2. If $R \in \mathcal{I}$ and $Q \leq R$, then $Q \in \mathcal{I}$;
3. If $Q, R \in \mathcal{I}$, there exists $U \in \mathcal{I}$ such that $Q, R \leq U$ and $|U| \leq |Q| + |R|$

Examples:

$$\begin{aligned} \mathcal{I}(g) &= \{S \in \mathbb{H}(\Omega) \mid g(S_i) \geq g(S_{i+1}) \forall i\} \\ \mathcal{I}^d(g) &= \{S \in \mathbb{H}(\Omega) \mid g(S_i) > g(S_{i+1}) \forall i\} \end{aligned}$$

Technical Results: Main Theorems

Theorem

Function $J : \mathbb{H}(\Omega) \rightarrow \mathbb{R}$ and fully extendable set $\mathcal{I} \subseteq \mathbb{H}(\Omega)$. J is \mathcal{I} -monotonic and \mathcal{I} -submodular. Let O^T be a maximizing sequence for J among the sequences in \mathcal{I} of length T and let S^T be the result of the generalized greedy algorithm. Then,

$$J(S^T) \geq \left(1 - \frac{1}{e}\right) J(O^T)$$

Theorem

Functional $F : \mathbb{H}(\Omega) \rightarrow \mathbb{R}$ permutation invariant, monotonic, and submodular. Function $g : \Omega \rightarrow \mathbb{R}^+$ and fully extendable set $\mathcal{I} \subseteq \mathcal{I}(g)$. Then, F_g is \mathcal{I} -monotonic and \mathcal{I} -submodular.

Experimental Results: Search & Tracking

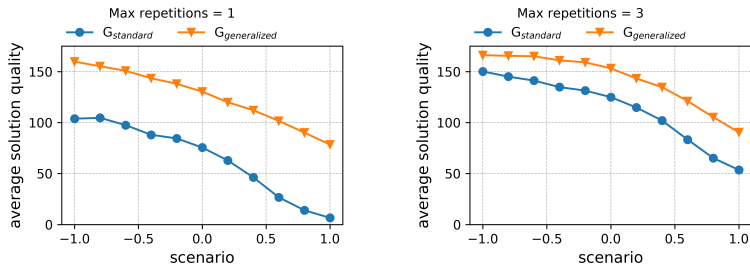


Figure: Average objective values obtained by the standard and the generalized greedy algorithms

Conclusions

- ▶ Several applicative problems involve the maximization of a **recursive functional** with the following structure:

$$F_g(S) = \sum_{k=1}^n g(S_k) \left[F(S|_1^k) - F(S|_1^{k-1}) \right] \text{ for } S = (S_1, \dots, S_n)$$

- ▶ Existing greedy algorithms do not yield strong theoretical guarantees for this functional
- ▶ We introduce an efficient **generalized greedy algorithm** that ensures finding solutions that are $(1 - \frac{1}{e})$ from the optimal
- ▶ Experiments directly show the power of the new algorithm

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