

CMPSCI 611: Advanced Algorithms

Lecture 5: Greedy Algorithms and Matroids

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Subset Systems

Definition

A *subset system* $S = (E, \mathcal{I})$ is a finite set E with a collection \mathcal{I} of subsets E such that:

if $A \in \mathcal{I}$ and $B \subset A$ then $B \in \mathcal{I}$

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2. E is the edges of a graph and \mathcal{I} is the acyclic subsets of edges
3. E is the edges of a graph and \mathcal{I} are the **matchings**, i.e., subsets of edges such that no two edges share a vertex

Generic Problem and Greedy Algorithms

Problem Given a subset system $S = (E, \mathcal{I})$ and weight function $w : E \rightarrow \mathbb{R}^+$, find $A \in \mathcal{I}$ such that $w(A) = \sum_{e \in A} w(e)$ is maximized.

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Terminology: Solution $A \in \mathcal{I}$ is a **maximum** if $w(A) \geq w(A')$ for all other $A' \in \mathcal{I}$. Solution $A \in \mathcal{I}$ is **maximal** if there doesn't exist $e \in E - A$ such that $A + e \in \mathcal{I}$.

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Let $E = \{e_1, e_2, e_3\}$, $\mathcal{I} = \{\{e_1, e_2\}, \{e_2, e_3\}, \{e_1\}, \{e_2\}, \{e_3\}, \{\}\}$, and $w(e_1) = 3$, $w(e_2) = 1$, and $w(e_3) = 4$. The greedy algorithm returns

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Example (Maximum Weight Matching)

E is the edges of a graph and \mathcal{I} are the **matchings**. Greedy does not work.

Matroid Definition and Theorem

Definition

Subset system (E, \mathcal{I}) has the **exchange property** if

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Theorem

Given a subset system (E, \mathcal{I}) , the following statements are equivalent:

- 1. Greedy algorithm returns optimal solution for any weight function.*
- 2. The subset system obeys the exchange property, i.e., it's a matroid.*

Matroid implies Greedy Algorithm is Optimal

- ▶ Proof by contradiction: Assume (E, \mathcal{I}) is a matroid and let

greedy solution: $A = \{e_1, e_2, \dots, e_k\}$

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- ▶ But then $w(f_t) \geq w(f_s)$ and hence $w(f_t) > w(e_s)$. This is a contradiction since greedy algorithm picked e_s rather than f_t

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$$w(e) = \begin{cases} m + 2 & \text{if } e \in A \\ m + 1 & \text{if } e \in B - A \\ 1/(2n) & \text{otherwise} \end{cases}$$

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- ▶ Greedy algorithm returns A with weight at most $(m + 2)m + 1/2$ but a better solution is B with weight at least $(m + 1)^2$

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