Review of Last Lecture

- Massaging mixed-signed objectives
- Recognizing a min cut problem



Learning Goal

- Project selection problem and its reduction to min cut
- Turning hard constraints into punishments in the objective

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- Formally, choose a set $S \subseteq P$ that maximizes $\sum_{i \in S} p_i$, such that $R_i \subseteq S$ for every $i \in S$

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Project Selection

Adjusting the objective

- The objective is in fact mixed signed.
 - Let A be the set of projects with nonnegative profits, and B the set with negative profits, then the profit of any $S \subseteq P$ is

$$\sum_{i\in S\cap A}p_i-\sum_{i\in S\cap B}|p_i|.$$

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• To reduce to min cut, we need a minimization problem:

$$\min_{S\subseteq P}\sum_{i\in A\setminus S}p_i+\sum_{i\in S\cap B}|p_i|.$$

An illustration



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- Add source s and sink t to G; add edge from s to each $i \in A$ with capacity p_i , and edge to t from each $i \in B$ with capacity $|p_i|$.
- Now for any $S \subseteq P$, $c(S \cup \{s\}, \overline{S} \cup \{t\}) = \sum_{i \in A \setminus S} p_i + \sum_{i \in S \cap B} |p_i|$, i.e., nodes on the side of s in the cut are the selected projects.

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- If no edge in G is in the cut we output, all prerequisite requirements are satisfied!
- Let all the original edges in G carry infinite (or large enough) capacity.

Illustration of flow network



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Illustration of a cut



Last remark

• A commonly used idea when doing reductions among problems: convert "hard" constraints to "soft" ones. I.e., if in one problem, certain patterns are forbidden, then in the other problem, punish such patterns in the objective — when the punishment is high enough, such patterns are forbidden from the solution.

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- A commonly used idea when doing reductions among problems: convert "hard" constraints to "soft" ones. I.e., if in one problem, certain patterns are forbidden, then in the other problem, punish such patterns in the objective — when the punishment is high enough, such patterns are forbidden from the solution.
- (Outside this course:) Recall Lagrangian multipliers from calculus. Hard constraints are softened into punishment in the objective, and the multipliers adjust how heavy the punishment is.