

# Approximation Algorithms

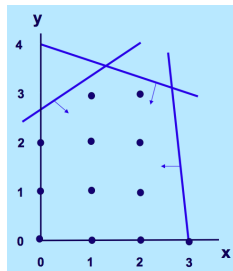
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# Motivation

- Goal:
  - Find “good” solutions for difficult problems (NP-hard).
  - Be able to quantify the “goodness” of the given solution.
- Presentation of a technique to get approximation algorithms: fractional relaxation of integer linear programs.

# Fractional Relaxation

- Reminder:
  - **Integer** Linear Programs often **hard to solve** (NP-hard).
  - Linear Programs (with **real numbers**) easier to solve (**polynomial-time** algorithms).
- Idea:
  - 1- **Relax** the integrality constraints;
  - 2- Solve the (fractional) linear program and then;
  - 3- **Round the solution** to obtain an integral solution.



# Set Cover

**Definition:** An approximation algorithm produces

- in polynomial time
- a feasible solution
- whose objective function value is close to the optimal  $OPT$ , by close we mean within a guaranteed factor of the optimal.

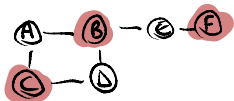
**Example:** a factor 2 approximation algorithm for the cardinality vertex cover problem, i.e. an algorithm that finds a cover of cost  $\leq 2 \cdot OPT$  in time polynomial in  $|V|$ .

# Set Cover

- **Problem:** Given a universe  $U$  of  $n$  elements, a collection of subsets of  $U$ ,  $\mathcal{S} = S_1, \dots, S_k$ , and a cost function  $c : S \rightarrow Q^+$ , find a minimum cost subcollection of  $S$  that covers all elements of  $U$ .
- **Model numerous classical problems** as special cases of set cover: vertex cover, minimum cost shortest path...
- **Definition:** The **frequency** of an element is the number of sets it is in. The **frequency of the most frequent element** is denoted by  $f$ .
- Various **approximation algorithms** for set cover achieve one of the two factors  $O(\log n)$  or  $f$ .

$$\begin{aligned} \min \quad & \sum_{i=1}^k x_i \cdot c_i \\ \text{s.t.} \quad & \sum_{i \in \theta} x_i \leq 1, \forall \theta \in U \\ & x_i \geq 0, \forall i=1, \dots, k \end{aligned}$$

## Fractional relaxation



$$U = \{ \overline{AB}, \overline{AC}, \dots, \overline{EF} \}$$

$$S_A = \{ \overline{AB}, \overline{AC} \}$$

$$S_B = \{ \overline{AB}, \overline{BD}, \overline{BE} \}$$

$$\vdots$$

$$S_F = \{ \overline{EF} \}$$

Write a linear program to solve vertex cover.

Var.:  $x_S = 1$  if  $S$  picked in  $\mathcal{C}$ ,  
 $x_S = 0$  otherwise

$$\min \quad \sum_{S \in \mathcal{S}} c(S) x_S$$

s. t.

$$\begin{aligned} \sum_{S: \theta \in S} x_S &\leq 1 & (\forall \theta \in U) \\ x_S &\in \{0, 1\} & (\forall S \in \mathcal{S}) \end{aligned}$$

Var.:  $1 \geq x_S \geq 0$

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to fix

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Integer  
BP

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$$x_S \geq 0 \quad (\forall S \in \mathcal{S})$$



# Fractional relaxation

- The (fractional) optimal solution of the relaxation is a **lower bound** of the optimal solution of the original integer linear program.
- **Example** in which a fractional set cover may be cheaper than the optimal integral set cover:  
Input:  $U = \{e, f, g\}$  and the specified sets  $S_1 = \{e, f\}$ ,  $S_2 = \{f, g\}$ ,  $S_3 = \{e, g\}$ , each of unit cost.
  - An **integral cover of cost 2** (must pick two of the sets).
  - A **fractional cover of cost 3/2** (each set picked to the extent of 1/2).

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$$\frac{1}{2} + \frac{1}{2} + \frac{1}{2} = \frac{3}{2}$$

# A simple rounding algorithm

Algorithm:

- 1- Find an optimal solution to the LP-relaxation.
- 2- (Rounding) Pick all sets  $S$  for which  $x_S \geq 1/f$  in this solution.

$f=2$  IM are case

- **Theorem:** The algorithm achieves an **approximation factor of  $f$**  for the set cover problem.
- **Proof:**
  - 1) **All elements are covered.**  $e$  is in at most  $f$  sets, thus one of this set must be picked to the extent of at least  $1/f$  in the fractional cover.
  - 2) **The rounding process increases  $x_S$  by a factor of at most  $f$ .** Therefore, the cost of  $\mathcal{C}$  is at most  $f$  times the cost of the fractional cover.

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$f=3$

$$x_1^{app} = 0.1 \rightarrow x_1 = 0$$

$$x_2^{app} = 0.9 \rightarrow x_2 = 1$$

$$x_3^{app} = \frac{1}{3} \rightarrow x_3 = 1$$

// AT MOST YOU INCREASE BY A FACTOR  $f$

$$OPT_f \leq OPT \leq f \cdot OPT_f$$

# Randomized rounding

- **Idea:** View the optimal fractional solutions as **probabilities**.
- **Algorithm:**
  - Flip coins with biases and round accordingly ( $S$  is in the cover with probability  $x_S$ ).
  - Repeat the rounding  $O(\log n)$  times.
- This leads to an  **$O(\log n)$  factor randomized approximation algorithm**. That is
  - The set is covered with high probability.
  - The cover has expected cost:  $O(\log n)OPT$ .

# Take Aways

- Fractional relaxation is a method to obtain for some problems:
  - **Lower bounds** on the optimal solution of an integer linear program (minimization).  
**Remark:** Used in Branch & Bound algorithms to cut branches.
  - **Polynomial approximation algorithms** (with rounding).
- Complexity:
  - **Integer linear programs** are often **hard**.
  - (Fractional) **linear programs** are quicker to solve (**polynomial time**).