



Approximation Algorithms

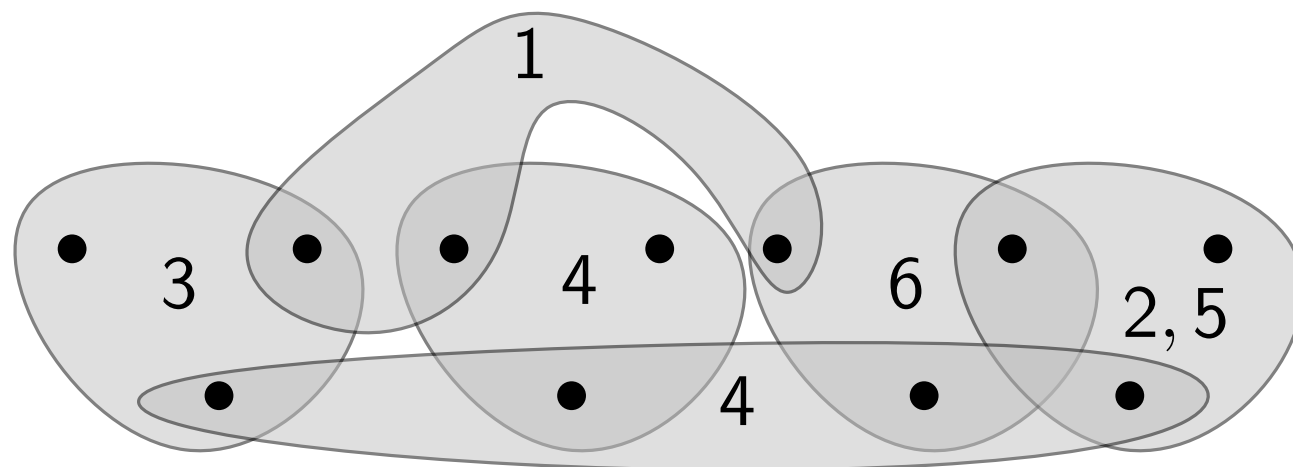
Lecture 5: LP-Based Approaches for
Set Cover

Joachim Spoerhase

SETCOVER – as an ILP

$$\begin{aligned} \min \quad & \sum_{S \in \mathcal{S}} c_S x_S \\ \text{s.t.} \quad & \sum_{S \ni e} x_S \geq 1 \quad e \in U \\ & x_S \in \{0, 1\} \quad S \in \mathcal{S} \end{aligned}$$

- Ground set U
- Set Families $\mathcal{S} \subseteq 2^U$ with $\bigcup \mathcal{S} = U$
- Costs $c: \mathcal{S} \rightarrow \mathbb{Q}^+$



Find a cover $\mathcal{S}' \subseteq \mathcal{S}$ of U with minimum cost.

SETCOVER – LP-Relaxation

$$\begin{array}{ll} \min & \sum_{S \in \mathcal{S}} c_S x_S \\ \text{s.t.} & \sum_{S \ni e} x_S \geq 1 \quad e \in U \\ & \boxed{x_S \geq 0} \quad S \in \mathcal{S} \end{array}$$

LP-Rounding

LP-Rounding: Approach I

$$\begin{array}{ll} \min & \sum_{S \in \mathcal{S}} c_S x_S \\ \text{s.t.} & \sum_{S \ni e} x_S \geq 1 \quad e \in U \\ & x_S \geq 0 \quad S \in \mathcal{S} \end{array}$$

LP-Rounding-I(U, \mathcal{S}, c)

Find an optimal solution \mathbf{x} for the LP-Relaxation

Round each x_S with $x_S > 0$ to 1

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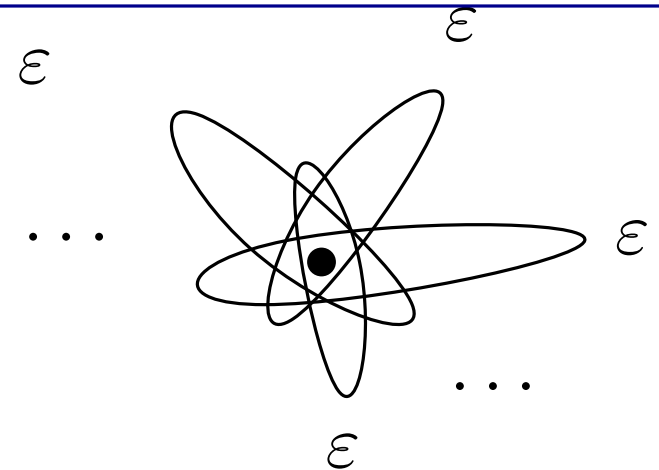
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- scaling factor is arbitrarily large :-)



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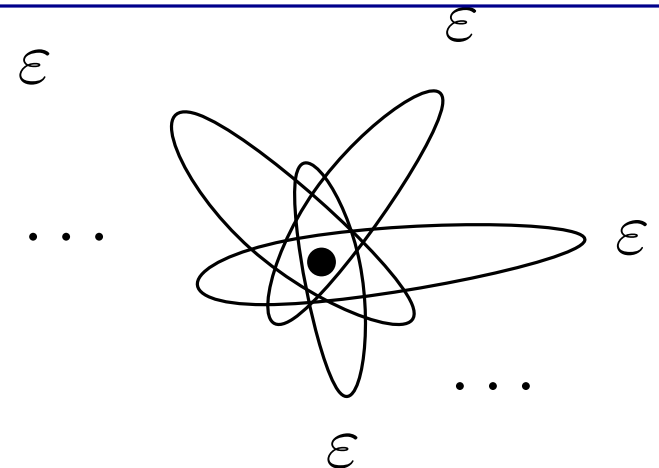
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better rounding policy?



LP-Rounding: Approach II

$$\begin{aligned} \min \quad & \sum_{S \in \mathcal{S}} c_S x_S \\ \text{s.t.} \quad & \sum_{S \ni e} x_S \geq 1 \quad e \in U \\ & x_S \geq 0 \quad S \in \mathcal{S} \end{aligned}$$

Let f be the frequency of (number of sets containing) the most frequent element.

LP-Rounding-II(U, \mathcal{S}, c)

Find an optimal solution \mathbf{x} for the LP-Relaxation

Round each x_S with $x_S \geq 1/f$ to 1; others to 0

LP-Rounding: Approach II

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Thm. LP-Rounding-II is a factor- f approximation algorithm for SETCOVER.

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Proof. (discuss).

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even better rounding?

Randomized LP-Rounding

$$\begin{array}{ll} \min & \sum_{S \in \mathcal{S}} c_S x_S \\ \text{s.t.} & \sum_{S \ni e} x_S \geq 1 \quad e \in U \\ & x_S \geq 0 \quad S \in \mathcal{S} \end{array}$$


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$$\min \sum_{S \in \mathcal{S}} c_S x_S$$

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$$x_S \geq 0 \quad S \in \mathcal{S}$$

$x_S \in [0, 1]$ in
opt. solution



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interpret x_S as a probability

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Random-LP-Rounding(U, \mathcal{S}, c)

Find an optimal solution \mathbf{x} for the LP-Relaxation

Round each x_S to 1 with probability x_S (o.w. set x_S to 0)

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Random-LP-Rounding(U, \mathcal{S}, c)

Find an optimal solution \mathbf{x} for the LP-Relaxation

Round each x_S to 1 with probability x_S (o.w. set x_S to 0)

Lemma.

Let OPT_f be the cost of an optimal solution for the above LP-Relaxation and $\mathcal{S}' \subseteq \mathcal{S}$ be a corresponding solution generated by Random-LP-Rounding. Then $E[c(\mathcal{S}')] = \text{OPT}_f$.

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Lemma.

\mathcal{S}' is not necessarily feasible!

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Random-LP-Rounding(U, \mathcal{S}, c)

Find an optimal solution \mathbf{x} for the LP-Relaxation

Round each x_S to 1 with probability x_S (o.w. set x_S to 0)

Lemma.

Let $\mathcal{S}' \subseteq \mathcal{S}$ be a result of Random-LP-Rounding. For each $u \in U$, u is covered by \mathcal{S}' with probability $1 - 1/e \approx 0.632 \dots$

Randomized LP-Rounding

Algorithm SETCOVER-Random-Rounding(U, \mathcal{S}, c)

$\mathcal{S}' \leftarrow \emptyset$

for t iterations **do**

$\mathcal{S}' \leftarrow \mathcal{S}' \cup \text{Random-LP-Rounding}(U, \mathcal{S}, c)$

return \mathcal{S}'

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Lemma.

For $t := 2 + \ln n$, the above algorithm produces an infeasible solution with probability $\leq \frac{1}{4}$.

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Lemma.

For $t := 2 + \ln n$, the above algorithm produces an infeasible solution with probability $\leq \frac{1}{4}$.

Lemma.

With probability $\leq \frac{1}{4}$, the above algorithm produces an “expensive” solution (i.e., with cost $\geq \text{OPT}_f \cdot 4t$).

Randomized LP-Rounding

Corollary.

With probability $\geq \frac{1}{2}$, our algorithm provides a feasible solution whose cost is less than $4t \cdot \text{OPT} = O(\log n) \cdot \text{OPT}$.

Primal-Dual Schema

Recall Complementary Slackness

$$\begin{array}{ll} \min & \sum_{j=1}^n c_j x_j \\ \text{s.t.} & \sum_{j=1}^n a_{ij} x_j \geq b_i \\ & x_j \geq 0 \end{array}$$

$$\begin{array}{ll} \max. & \sum_{i=1}^m b_i y_i \\ \text{s.t.} & \sum_{i=1}^m a_{ij} y_i \leq c_j \\ & y_i \geq 0 \end{array}$$

Thm.

Let $\mathbf{x} = (x_1, \dots, x_n)$ and $\mathbf{y} = (y_1, \dots, y_m)$ be feasible solutions for the primal and dual LPs respectively. The solutions \mathbf{x} and \mathbf{y} are optimal if and only if the following conditions are met:

Primal CS:

For each $j = 1, \dots, n$: either $x_j = 0$ or $\sum_{i=1}^m a_{ij} y_i = c_j$

Dual CS:

For each $i = 1, \dots, m$: either $y_i = 0$ or $\sum_{j=1}^n a_{ij} x_j = b_i$

Relaxing Complementary Slackness

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For each $i = 1, \dots, m$: either $y_i = 0$ or $\sum_{j=1}^n a_{ij} x_j = b_i$

$$\Leftrightarrow \sum_{j=1}^n c_j x_j = \sum_{i=1}^m b_i y_i$$

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~~Primal CS:~~ relaxed primal CS

For each $j = 1, \dots, n$: either $x_j = 0$ or $\sum_{i=1}^m a_{ij} y_i = c_j$

$$c_j / \alpha \leq \sum_{i=1}^m a_{ij} y_i \leq c_j$$

Dual CS:

For each $i = 1, \dots, m$: either $y_i = 0$ or $\sum_{j=1}^n a_{ij} x_j = b_i$

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$$\Leftrightarrow \sum_{j=1}^n c_j x_j = \sum_{i=1}^m b_i y_i$$

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$$\Leftrightarrow \sum_{j=1}^n c_j x_j = \sum_{i=1}^m b_i y_i \quad \Rightarrow \quad \sum_{j=1}^n c_j x_j \leq \alpha \beta \sum_{i=1}^m b_i y_i \leq \alpha \beta \cdot \text{OPT}_{\text{LP}}$$

Primal-Dual Schema

- Start with a feasible dual and infeasible primal solution (often trivial).

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- do so until the relaxed CS conditions are met.

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- maintain that the primal solution is integer valued.

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- Start with a feasible dual and infeasible primal solution (often trivial).
- “Improve” the feasibility of the primal solution.
- simultaneously improve the obj. value of the dual solution.
- do so until the relaxed CS conditions are met.
- maintain that the primal solution is integer valued.
- The feasibility of the primal solution and relaxed CS condition provide an approximation ratio.

Relaxed CS for SETCOVER

$$\begin{array}{ll} \min & \sum_{S \in \mathcal{S}} c_S x_S \\ \text{s.t.} & \sum_{S \ni e} x_S \geq 1 \quad e \in U \\ & x_S \geq 0 \quad S \in \mathcal{S} \end{array}$$

LP

$$\begin{array}{ll} \max & \sum_{e \in U} y_e \\ \text{s.t.} & \sum_{e \in S} y_e \leq c_S \quad S \in \mathcal{S} \\ & y_e \geq 0 \quad e \in U \end{array}$$

DLP

Relaxed CS for SETCOVER

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DLP

critical set 

(non-relaxed) primal CS: $x_S \neq 0 \Rightarrow \sum_{e \in S} y_e = c_S$

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critical set

select only critical sets

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Relaxed dual CS: $y_e \neq 0 \Rightarrow \sum_{S \ni e} x_S \leq f$

Relaxed CS for SETCOVER

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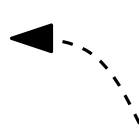
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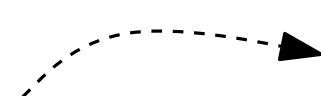
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Relaxed dual CS: $y_e \neq 0 \Rightarrow \sum_{S \ni e} x_S \leq f$

trivial for binary \mathbf{x} 

Primal-Dual-Schema for SETCOVER

PrimalDualSetCover(U, \mathcal{S}, c)

$\mathbf{x} \leftarrow 0, \mathbf{y} \leftarrow 0$

repeat

select an uncovered element e

increase y_e until a set S is critical ($\sum_{e' \in S} y_{e'} = c_S$)

select all critical sets and update \mathbf{x}

mark all elements in these sets as **covered**

until all elements are covered

return \mathbf{x}

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Thm. The above is a factor- f approximation algorithm for SETCOVER. The factor is also tight.