

A Local Search-Based Approach for Set Covering

Jason Li (Simons Institute)

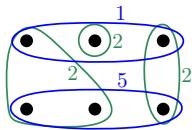
Joint with Anupam Gupta (CMU), Euiwoong Lee (UMich)

SODA 2023

January 23, 2023

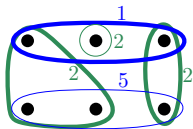
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- Set Cover: given a family of sets $S \subseteq U$ with weights $w(S)$, find a minimum weight collection of sets that cover U



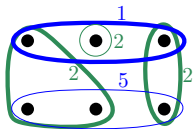
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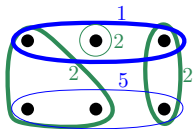
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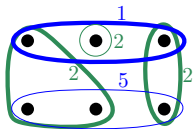
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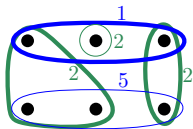
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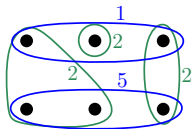
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- This talk: [local search](#) algorithm for set cover

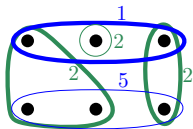
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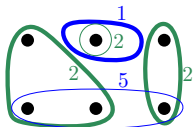


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- **Transformation:** for each original set S , add all subsets of S , each of weight $w(S)$



Local Search

- Local search for set cover:

Local Search

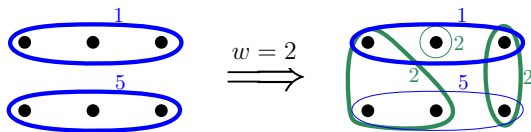
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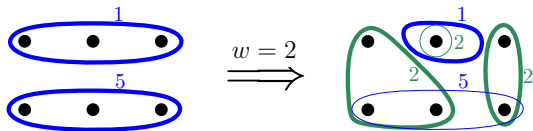
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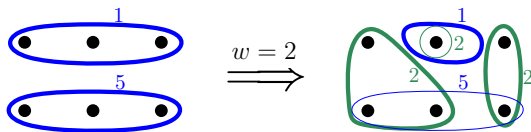
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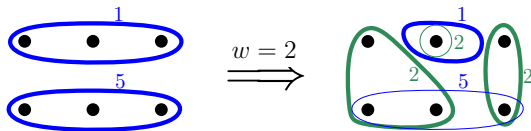
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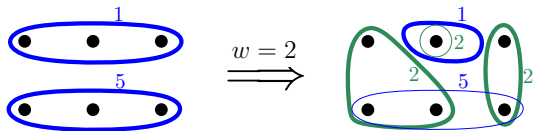
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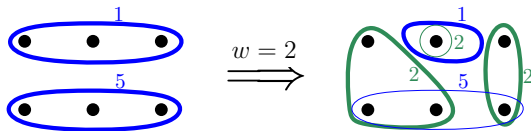


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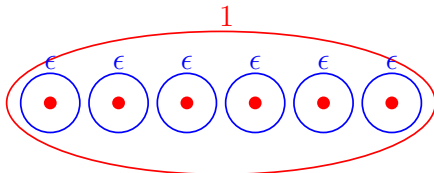
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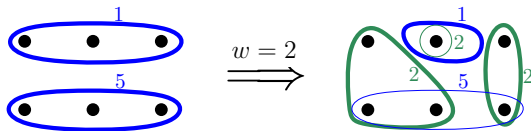
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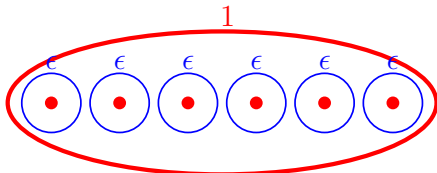
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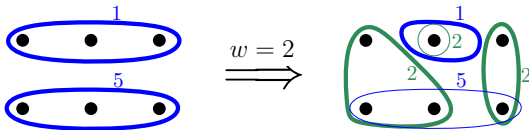
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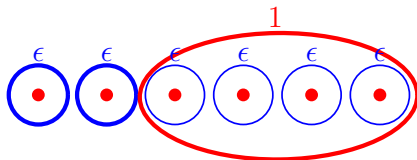
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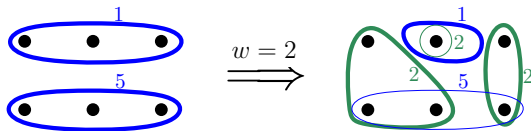
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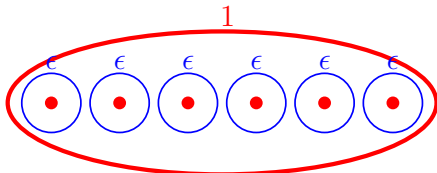
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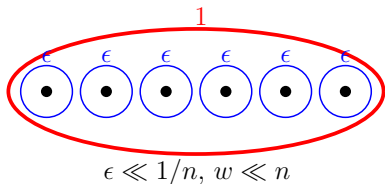
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Non-Oblivious Local Search

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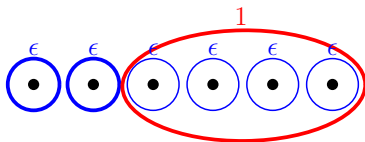
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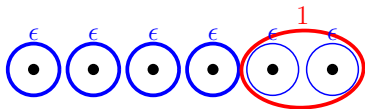
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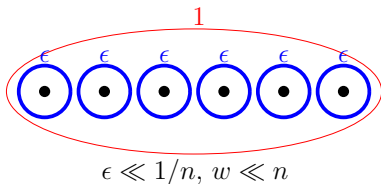
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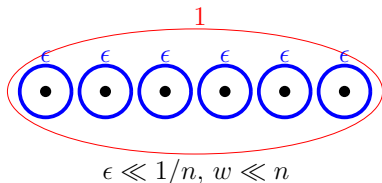
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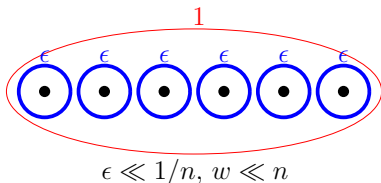
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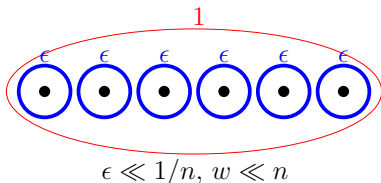
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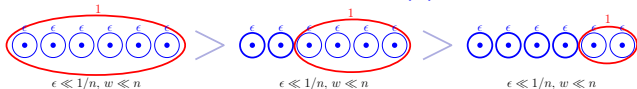
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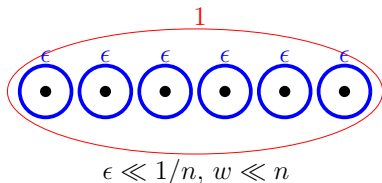


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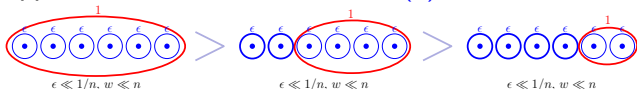


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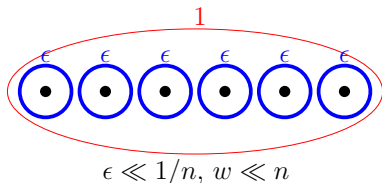
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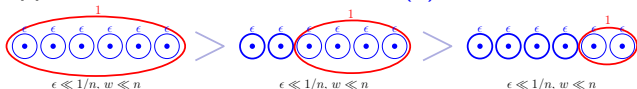
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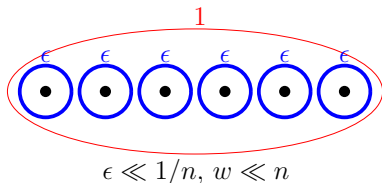
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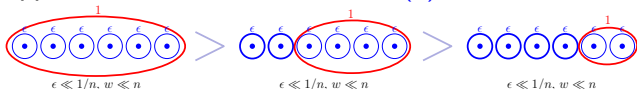
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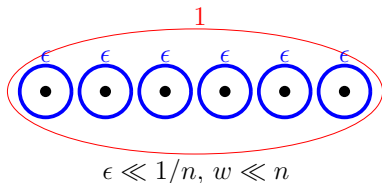
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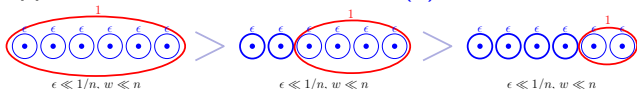
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 - Inspired by [Traub-Zenkhusen'21] on Steiner tree by non-oblivious local search
 - Φ is called the **Rosenthal** potential

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 - Adding in any $S^* \in \mathcal{C}^*$ and pruning does not improve local search:

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- $H_{|S|} - H_{|S \setminus S^*|} = \frac{1}{|S \setminus S^*| + 1} + \frac{1}{|S \setminus S^*| + 2} + \dots + \frac{1}{|S|} \geq |S \cap S^*| \cdot \frac{1}{|S|}$
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$$H_k \cdot OPT \geq \sum_{S^* \in \mathcal{C}^*} w(S^*)H_{|S^*|} \geq \sum_{S^* \in \mathcal{C}^*} \sum_{S \in \mathcal{C}} w(S) |S \cap S^*| \cdot \frac{1}{|S|}$$

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- Repeat analysis from before: $ALG \leq (1 - \Theta(1/k^2))H_k OPT$

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- Open: approximation $H_k - \Theta(1)$? (Known for unweighted set cover!)