

Approximation Algorithms

Lecture 1: Introduction and Vertex Cover

Part I: Organizational

Organizational

Lectures: Zoom (in German or English)

Synchronous (key material)

More technical lectures via inverted classroom

Tutorials: One exercise sheet per lecture

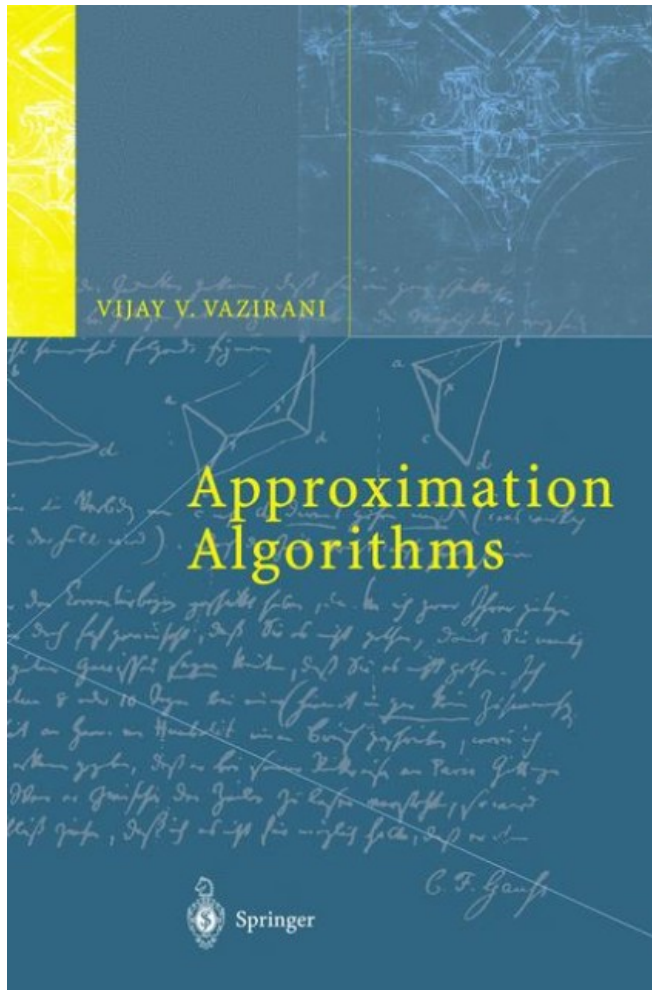
Solving assignments and presenting solutions

Tuesdays 10:15 - 11:45 (SE I)

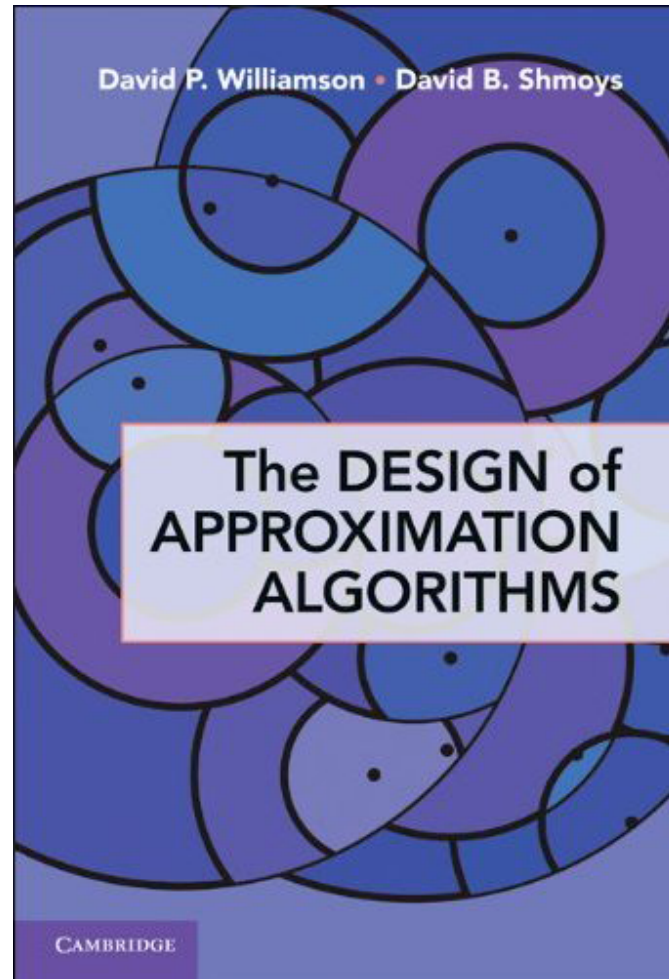
Bonus (+0.3 on final grade) for $\geq 50\%$ points

Questions/Tasks during the lecture

Textbooks



Vijay V. Vazirani:
Approximation
Algorithms
Springer-Verlag, 2003.



D. P. Williamson & D. B. Shmoys:
The Design of Approximation Algorithms
Cambridge-Verlag, 2011.
<http://www.designofapproxalgs.com/>

Approximation Algorithms

„All exact science is dominated by the idea of approximation.“

– Bertrand Russell
(1872 – 1970)



Approximation Algorithms

- Many optimization problems are NP-hard (e.g. the traveling salesperson problem)
- \rightsquigarrow an optimal solution cannot be efficiently computed unless $P=NP$.
- However, good approximate solutions can often be found efficiently!
- **Techniques** for the design and analysis of approximation algorithms arise from studying specific optimization problems.

Overview

Combinatorial Algorithms

- Introduction (Vertex Cover)
- Set Cover via Greedy
- Shortest Superstring via reduction to SC
- Steiner Tree via MST
- Multiway Cut via Greedy
- k -Center via param. Pruning
- Min-Deg-Spanning-Tree & local search
- Knapsack via DP & Scaling
- Euclidean TSP via Quadtrees

LP-based Algorithms

- introduction to LP-Duality
- Set Cover via LP Rounding
- Set Cover via Primal-Dual Schema
- Maximum Satisfiability
- Scheduling und Extreme Point Solutions
- Steiner Forest via Primal-Dual

Approximation Algorithms

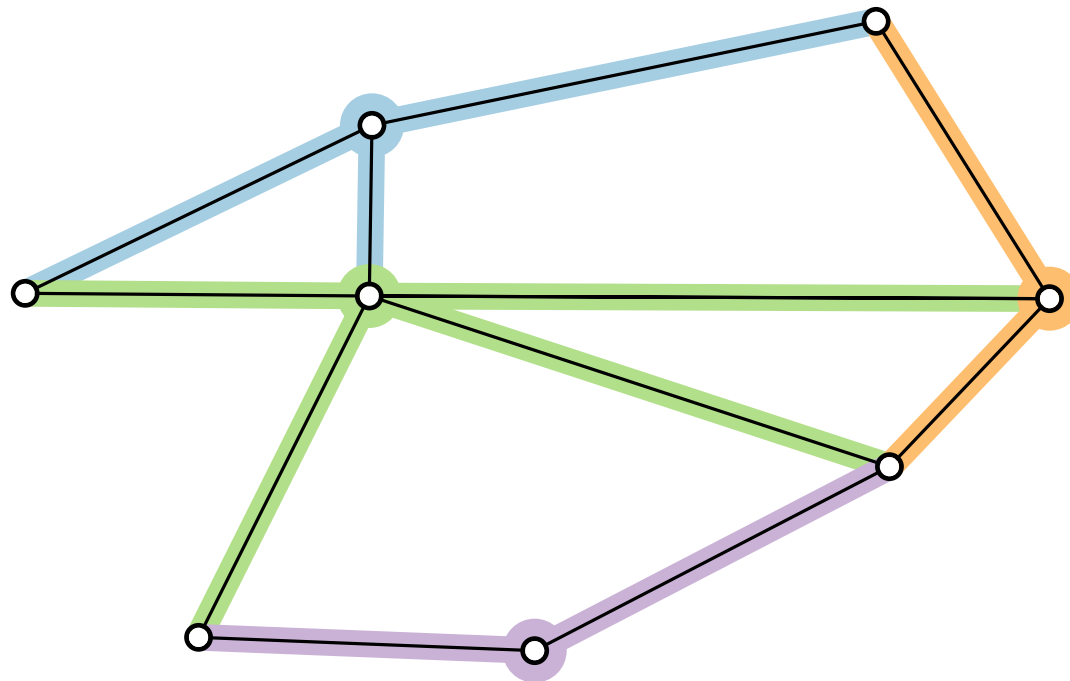
Lecture 1: Introduction and Vertex Cover

Part II: Vertex Cover (card.)

VERTEXCOVER (card.)

In: Graph $G = (V, E)$

Out: a minimum **vertex cover**: a minimum vertex set $V' \subseteq V$ such that every edge is **covered** (i.e., for every $uv \in E$, either $u \in V'$ or $v \in V'$).



Optimum ($\text{OPT} = 4$) – but in general NP-hard to find :-)

Approximation Algorithms

Lecture 1: Introduction and Vertex Cover

Part III: NP-Optimization Problem

NP-Optimization Problem

An **NP-optimization problem** Π is given by:

- A set D_Π of **instances**.

We denote the size of an instance $I \in D_\Pi$ by $|I|$.

- For each instance $I \in D_\Pi$ a set $S_\Pi(I) \neq \emptyset$ of **feasible solutions** for I such that:
 - for each solution $s \in S_\Pi(I)$, its size $|s|$ is polynomially bounded in $|I|$, and
 - for each pair (s, I) , there is a polynomial time algorithm to decide whether $s \in S_\Pi(I)$.
- A polynomial time computable **objective function** obj_Π which assigns a positive objective value $\text{obj}_\Pi(I, s) \geq 0$ to any given pair (s, I) with $s \in S_\Pi(I)$.
- Π is either a minimization or maximization problem.

VERTEXCOVER: NP-Optimization Problem

Task: Fill in the gaps for $\Pi = \text{VERTEX COVER}$.

$D_\Pi =$ Set of all graphs

For $I \in D_\Pi$: $|I| =$ Number of vertices $|V|$

$G=(V, E)$ $S_\Pi(I) =$ Set of all vertex covers of G

■ Why is $|s| \in \text{poly}(|I|)$ for every $s \in S_\Pi(I)$?

$$s \subseteq V \Rightarrow |s| \leq |V| = |I|$$

■ For a given pair (s, I) , how can we efficiently decide whether $s \in S_\Pi(I)$? Test whether all edges are covered.

$$\text{obj}_\Pi(I, s) = |s|$$

Π is Minimization problem.

Optimum and optimal objective value

maximization problem

Let Π be a minimization problem and $I \in D_\Pi$ be an instance of Π .

A feasible solution $s^* \in S_\Pi(I)$ is **optimal** if $\text{obj}_\Pi(I, s^*)$ is maximal among objective values attained by the feasible solutions of I .

The optimal value $\text{obj}_\Pi(I, s^*)$ of the objective function is also denoted by $\text{OPT}_\Pi(I)$ or simply **OPT** in context.

Approximation Algorithms

maximization problem $\alpha: \mathbb{N} \rightarrow \mathbb{Q}$

Let Π be a minimization problem and ~~$\alpha \in \mathbb{Q}^+$~~ .

A factor- α -approximation algorithm for Π is an efficient algorithm which provides for **any** instance $I \in D_\Pi$ a feasible solution $s \in S_\Pi(I)$ such that

$$\frac{\text{obj}_\Pi(I, s)}{\text{OPT}_\Pi(I)} \stackrel{\geq}{\leq} \alpha \cdot \alpha(|I|)$$

Approximation Algorithms

Lecture 1:

Introduction and Vertex Cover

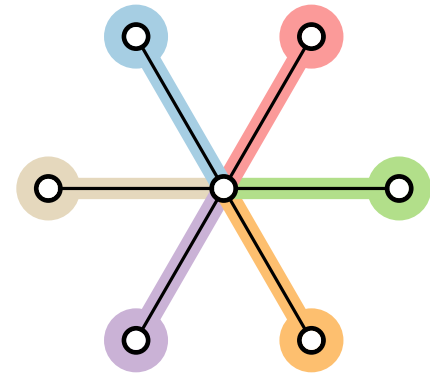
Part IV:

Approximation Algorithm for VERTEXCOVER

Approximation Alg. for VERTEXCOVER

Ideas?

- Edge-Greedy
- Vertex-Greedy



Quality?

Problem: How can we estimate $\text{obj}_{\Pi}(I, s) / \text{OPT}$, when it is hard to calculate OPT ?

Idea: Find a “good” lower bound $L \leq \text{OPT}$ for OPT and compare it to our approximate solution.

$$\frac{\text{obj}_{\Pi}(I, s)}{\text{OPT}} \leq \frac{\text{obj}_{\Pi}(I, s)}{L}$$

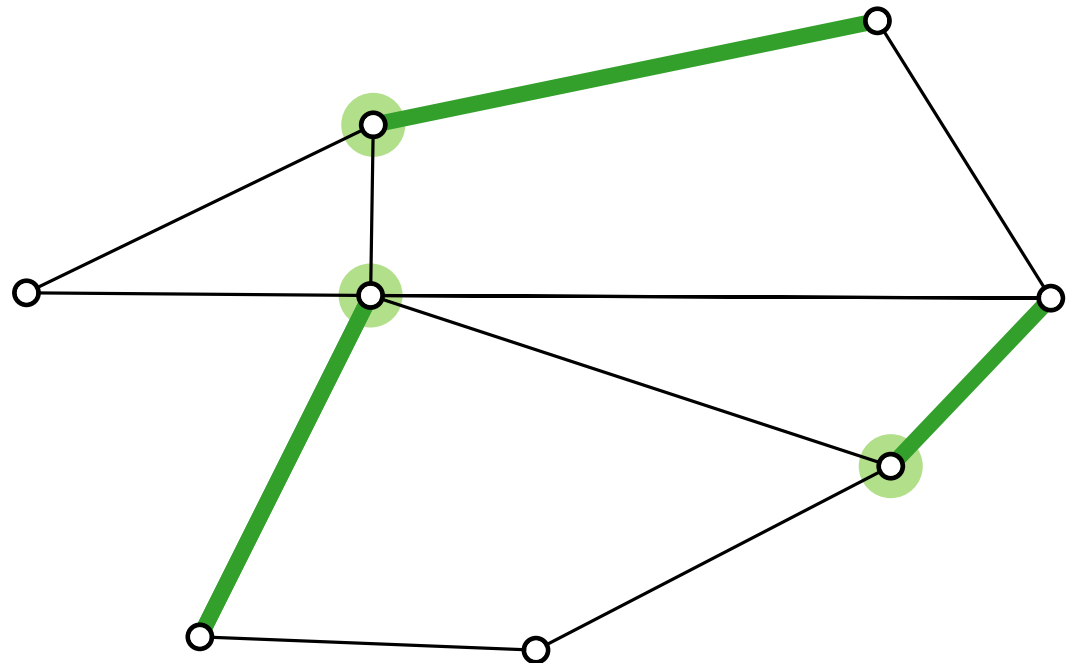
Lower Bound by Matchings

An edge set $M \subseteq E$ of a graph $G = (V, E)$ is a **matching** if no two edges of M are adjacent (i.e., share an end vertex).

M is **maximal** if there is no matching M' with $M' \supsetneq M$.

$$\text{OPT} \geq |M|$$

Vertex cover of M



Lower Bound by Matchings

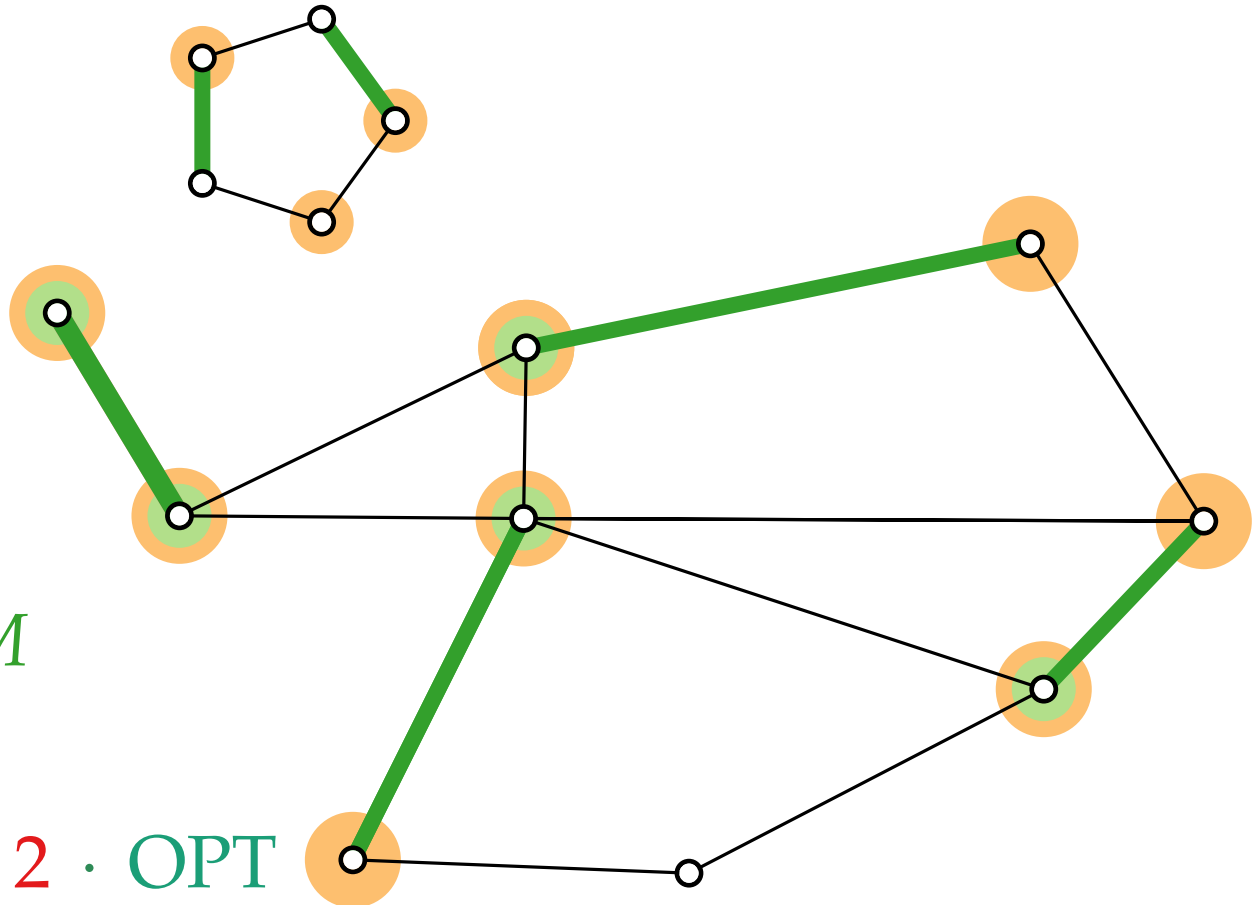
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M is **maximal** if there is no matching M' with $M' \supsetneq M$.

$$\text{OPT} \geq |M|$$
~~$$\text{OPT} = |M|?$$~~

Vertex cover of M
Vertex cover of E

$$\text{ALG} = 2 \cdot |M| \leq 2 \cdot \text{OPT}$$



Approximation Alg. for VERTEXCOVER

Algorithm VertexCover(G)

$M \leftarrow \emptyset$

foreach $e \in E(G)$ **do**

if e not adjacent to edge in M **then**
 $M \leftarrow M \cup \{e\}$

return $\{u, v \mid uv \in M\}$

Theorem. The above algorithm is a factor-2-approximation algorithm for VERTEXCOVER.

The best-known approximation factor for

VERTEXCOVER is $2 - \Theta(1/\sqrt{\log n})$

VERTEXCOVER cannot be approximated within factor 1.3606 (unless $P=NP$)

VERTEXCOVER cannot be approximated within factor $2 - \Theta(1)$, if “Unique Games Conjecture” holds.