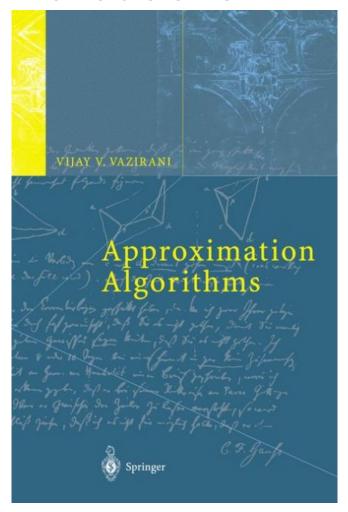




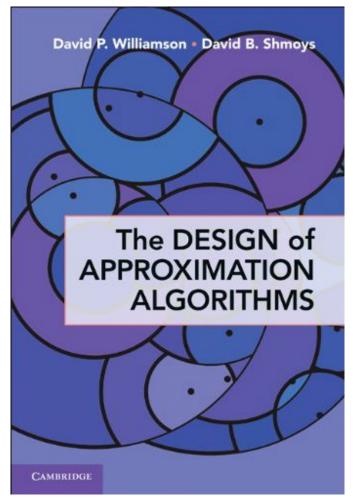
Lecture 1: Introduction & Vertex Cover

Joachim Spoerhase

#### **Textbooks**



Vijay V. Vazirani Approximation Algorithms Springer-Verlag 2003



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

"All exact science is dominated by the idea of approximation."

Bertrand Russell

#### Overview of Possible Topics

#### Combinatorial Algorithms

- Introduction
- Set Cover
- Steiner Tree and TSP
- Multiway Cut
- *k*-Center
- Shortest Superstring
- Knapsack
- Bin Packing
- Minimum Makespan Scheduling
- Euclidean TSP

#### LP-Based Algorithms

- Introduction to LP-Duality
- Set Cover via Dual Fitting
- Rounding Applied to Set Cover
- Set Cover via the Primal–Dual Schema
- Maximum Satisfiability
- Facility Location

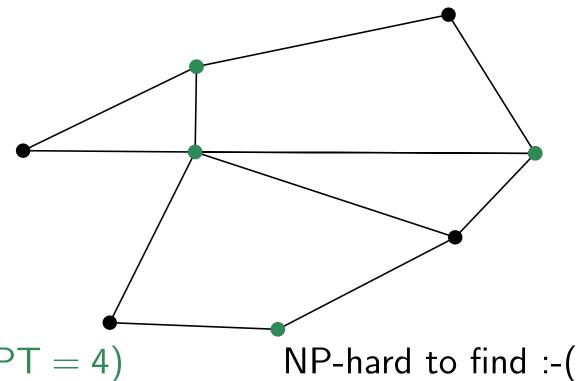
• . . .

- Many optimization problems are NP-hard (e.g. the travelling salesman problem)
- → 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 (currently) mostly from studying specific optimization problems.

# VERTEX COVER (cardinality)

**Input** Graph G = (V, E)

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

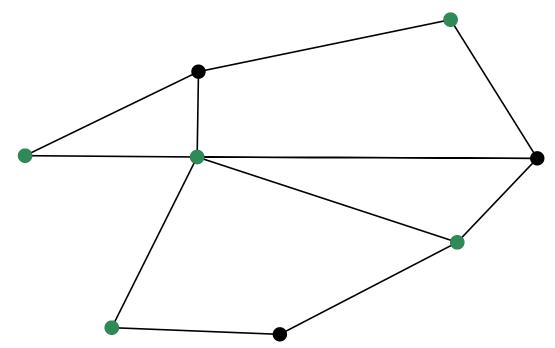


**Optimum** (OPT = 4)

# VERTEX COVER (cardinality)

Input Graph G = (V, E)

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"good" approximate solution (5/4-Approximation)

#### NP-Optimization Problem

An NP-optimization problem  $\Pi$  is given by:

- A set  $D_{\Pi}$  of **instances**. We use |I| to denote the size of an instance  $I \in D_{\Pi}$ .
- For each instance  $I \in D_{\Pi}$  there is a set  $S_{\Pi}(I) \neq \emptyset$  of **feasible solutions** for I where:
  - For each solution  $s \in S_{\Pi}(I)$ , its size |s| is polynomially bounded in |I|.
  - There is a polynomial time algorithm to decide for each pair (s,I), whether  $s \in S_{\Pi}(I)$
- A polynomial time computable objective function  $\operatorname{obj}_{\Pi}$ , which assigns a positive objective value  $\operatorname{obj}_{\Pi}(I,s)$  to a given pair (I,s).
- $\bullet$   $\Pi$  is either a minimization or maximization problem.

#### Vertex Cover NP-Optimization Problem

#### Exercise

What are the *instances*?

What are the *feasible solutions*?

What is the *objective function*?

## Optimum, and optimal objective value.

Let  $\Pi$  be a minimization (maximization) problem and  $I \in D_{\Pi}$  be an instance of  $\Pi$ . A feasible solution  $s^* \in S_{\Pi}(I)$  is **Optimal** when  $\operatorname{obj}_{\Pi}(I, s^*)$  is the minimum (maximum) among objective values attained by the feasible solutions of I.

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

Let  $\Pi$  be a minimization problem and  $\alpha \in \mathbb{Q}^+$ . A factor- $\alpha$ -approximation algorithm for  $\Pi$  is an efficient algorithm that provides a feasible solution  $s \in S_{\Pi}(I)$  for any instance  $I \in D_{\Pi}$  such that:

$$\frac{\mathsf{obj}_{\Pi}(I,s)}{\mathsf{OPT}_{\Pi}(I)} \leq \alpha.$$

maximization

 $\alpha:\mathbb{N}\to\mathbb{Q}$ 

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

A factor- $\alpha$ -approximation algorithm for  $\Pi$  is an efficient algorithm that provides a feasible solution  $s \in S_{\Pi}(I)$  for any instance  $I \in D_{\Pi}$  such that:

$$\frac{\operatorname{obj}_{\Pi}(I,s)}{\operatorname{OPT}_{\Pi}(I)} \stackrel{\geq}{\sim} \alpha(|I|)$$

#### Approximation Alg. for VERTEX COVER

Ideas?

- Edge-Greedy
- Vertex-Greedy (see Exercises)
- Inclusion-wise minimal vertex cover

How can we measure the quality of a feasible solution?

**Problem:** How can we estimate  $\frac{obj_{\Pi}(I,s)}{OPT}$  when it is hard to calculate OPT?

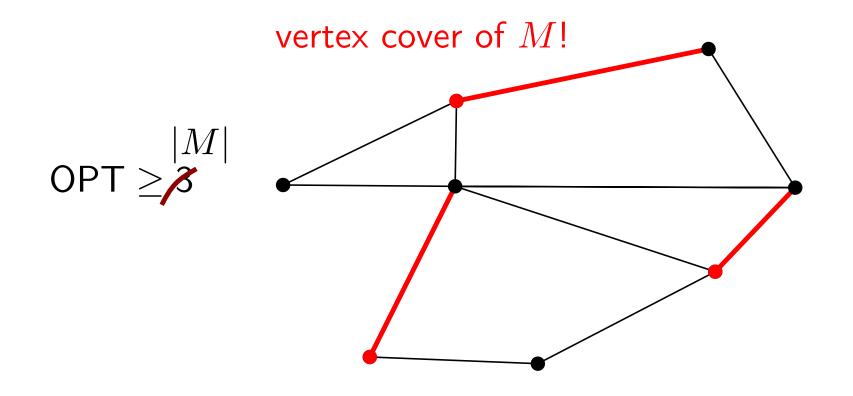
**Idea:** Find a "good" lower bound  $L \leq \mathsf{OPT}$  for  $\mathsf{OPT}$  and compare it to our approximate solution.

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

## Lower Bound by Matchings

An edge set  $M \subseteq E$  of a graph G = (V, E) is a **matching** when no vertex of G is incident to two edges in M.

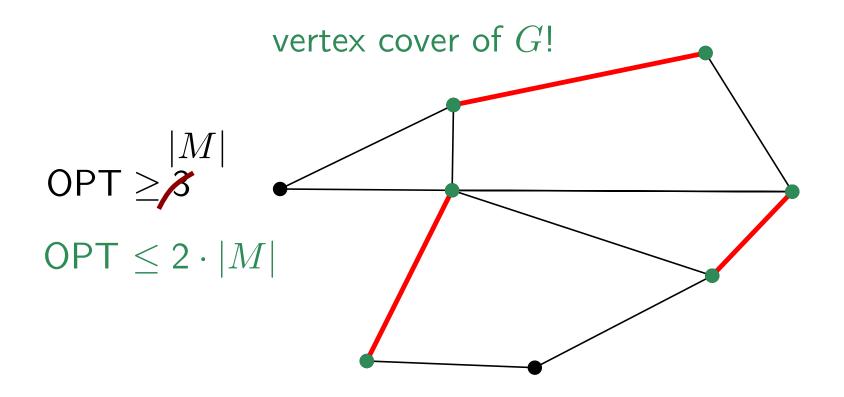
M is **maximal** when there is no matching M' with  $M' \supseteq M$ .



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#### Approximation Alg. for VERTEX COVER

```
Algorithm for Vertex Cover (G)
M \leftarrow \emptyset
foreach e \in E(G) do

if e is not adjacent to an edge in M then

M \leftarrow M \cup \{e\}
return \{u, v \mid uv \in M\}
```

**Thm 1.1** The above algorithm is a factor-2 approximation algorithm for VERTEX COVER

Next week: Set Cover