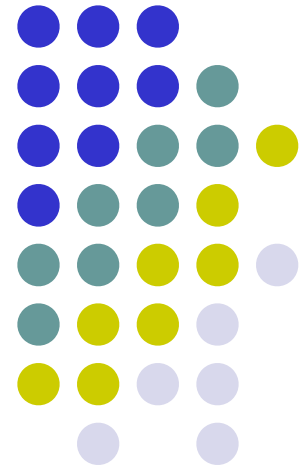
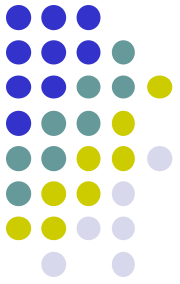


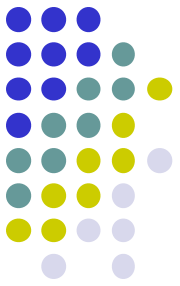
Web Algorithms

Eng. Fabio Persia, PhD





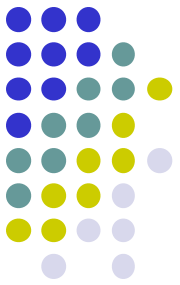
Approximation algorithms



- **QUESTION:** Suppose I need to solve an **NP**-hard problem. What should I do?
- **ANSWER.** Sacrifice one of three desired features:
 1. Solve arbitrary instances of the problem.
 2. Solve problem to optimality.
 3. Solve problem in polynomial time.

Coping strategies.

1. Design algorithms for special cases of the problem.
2. **Design approximation algorithms** or heuristics.
3. Design algorithms that may take exponential time.

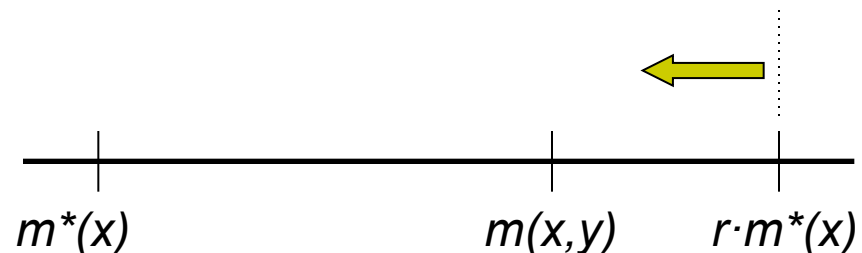


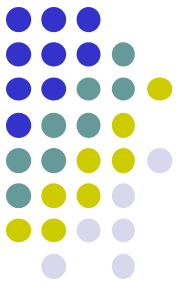
From now on we focus on NP-hard optimization problems, that is problems that cannot be solved efficiently (unless $P=NP$)

For such problems we will design algorithms able to determine solutions that are close to optimal ones, that is good "approximations"

Def: Given a minimization problem π and a number $r \geq 1$, an algorithm A is an r -approximation algorithm for π if for every input $x \in I$ it always returns an r -approximate solution, that is a feasible solution $y \in S(x)$ such that:

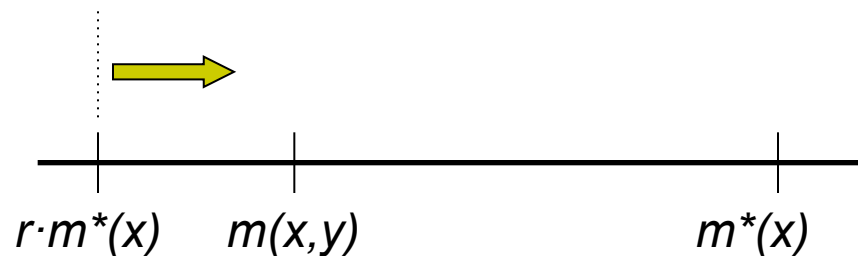
$$\frac{m(x, y)}{m^*(x)} \leq r$$



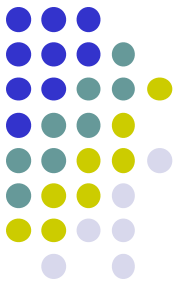


Def: Given a **maximization problem** π and a number $r \leq 1$, an algorithm A is an **r -approximation** algorithm for π if for every input $x \in I$ it always return an *r -approximate* solution, that is a feasible solution $y \in S(x)$ such that:

$$\frac{m(x, y)}{m^*(x)} \geq r$$



Determination of the approximation factor

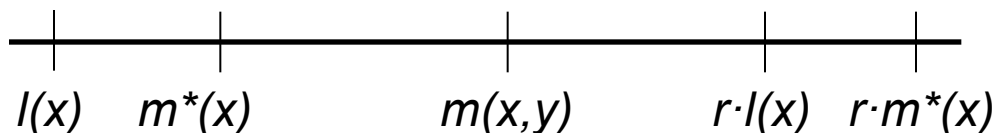


How can we determine the approximation factor r if we don't know the value m^* of an optimal solution?

For **minimization** (resp. **maximization**) problems, we compare the value of the returned solution $m(x, y)$ with a proper **lower bound** (resp. **upper bound**) $l(x)$ of $m^*(x)$.

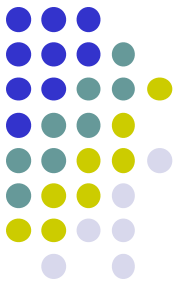
If their ratio is at most r for min or at least r for max then the algorithm is r -approximating.

$$\text{Min: if } \frac{m(x, y)}{l(x)} \leq r \quad \text{then} \quad \frac{m(x, y)}{m^*(x)} \leq \frac{m(x, y)}{l(x)} \leq r$$



Max: analogous

Approximation algorithm for Min Vertex Cover



[Definition of Min Vertex Cover](#)

Algorithm Approx Cover

Begin

$M = \emptyset$. //edges chosen by the algorithm

$U = \emptyset$. //nodes chosen in the cover

Repeat

 Select an edge $\{u, v\} \in E$.

$U = U \cup \{u, v\}$.

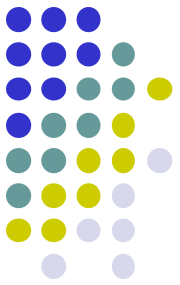
$E = E \setminus \{e \in E \mid e \text{ is incident to } u \text{ or } v\}$.

$M = M \cup \{\{u, v\}\}$.

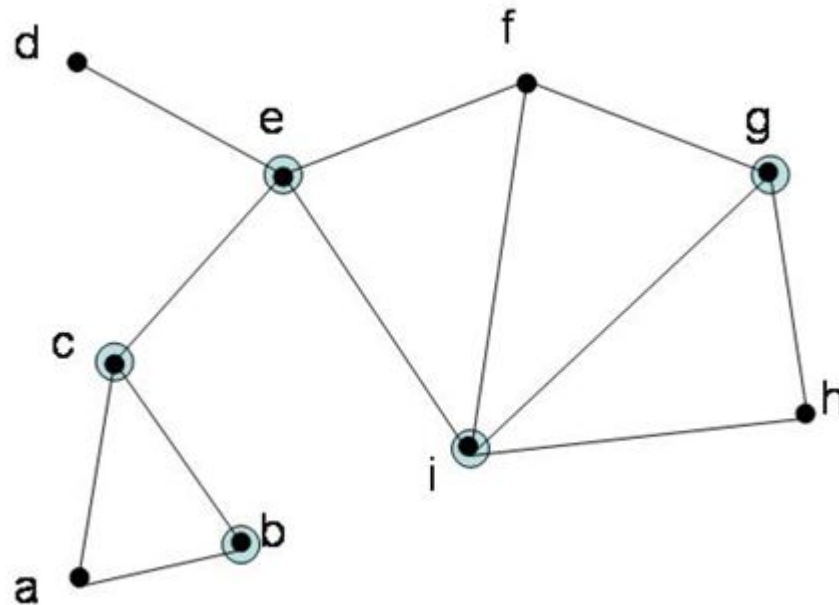
Until ($E = \emptyset$).

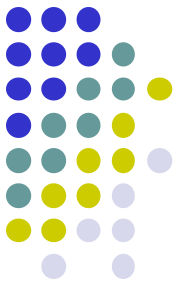
Return U .

End

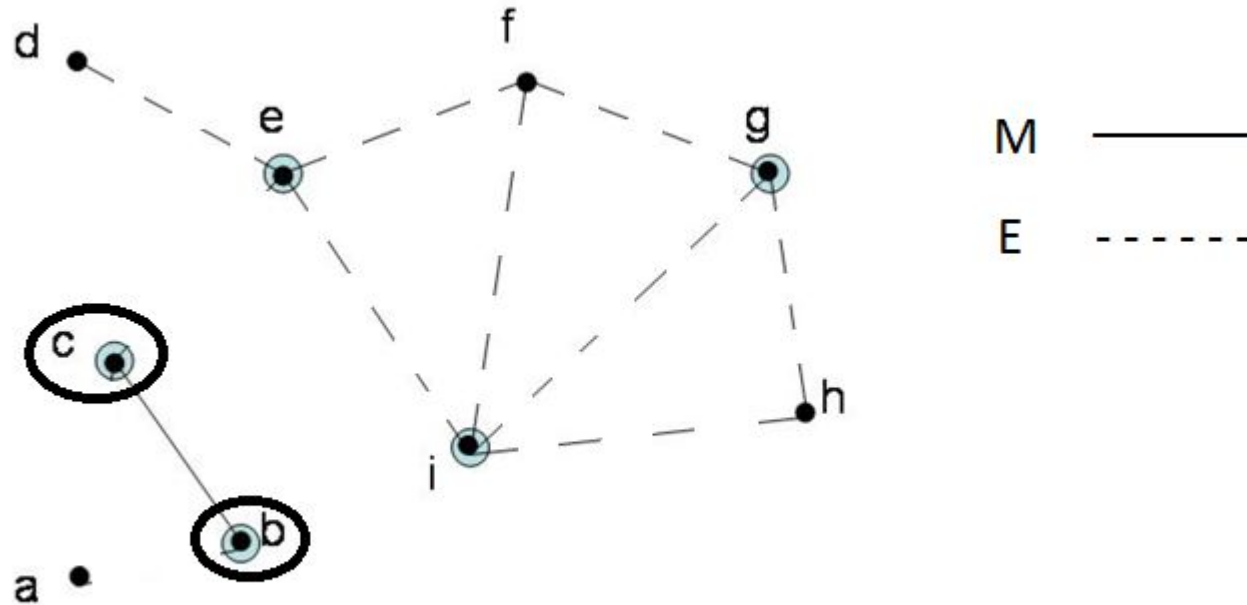


Example of execution

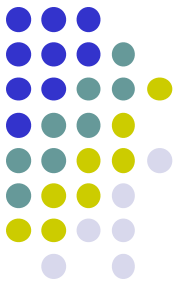




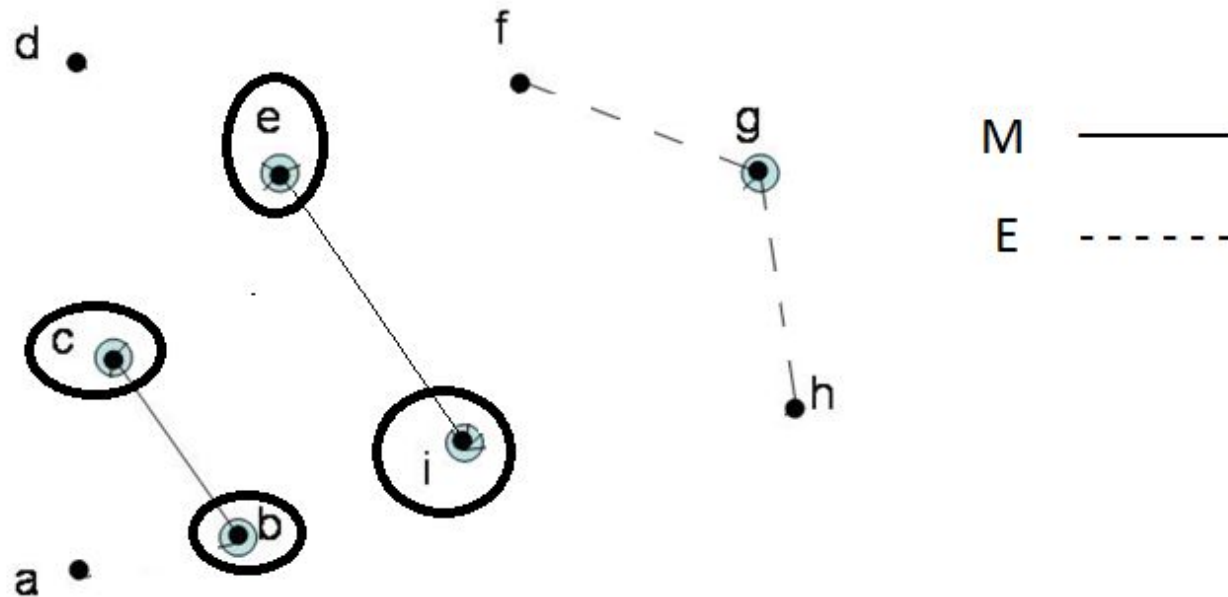
Example of execution - Step 1



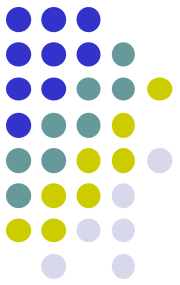
- $U = \{b, c\}$
- $M = \{\{b, c\}\}$



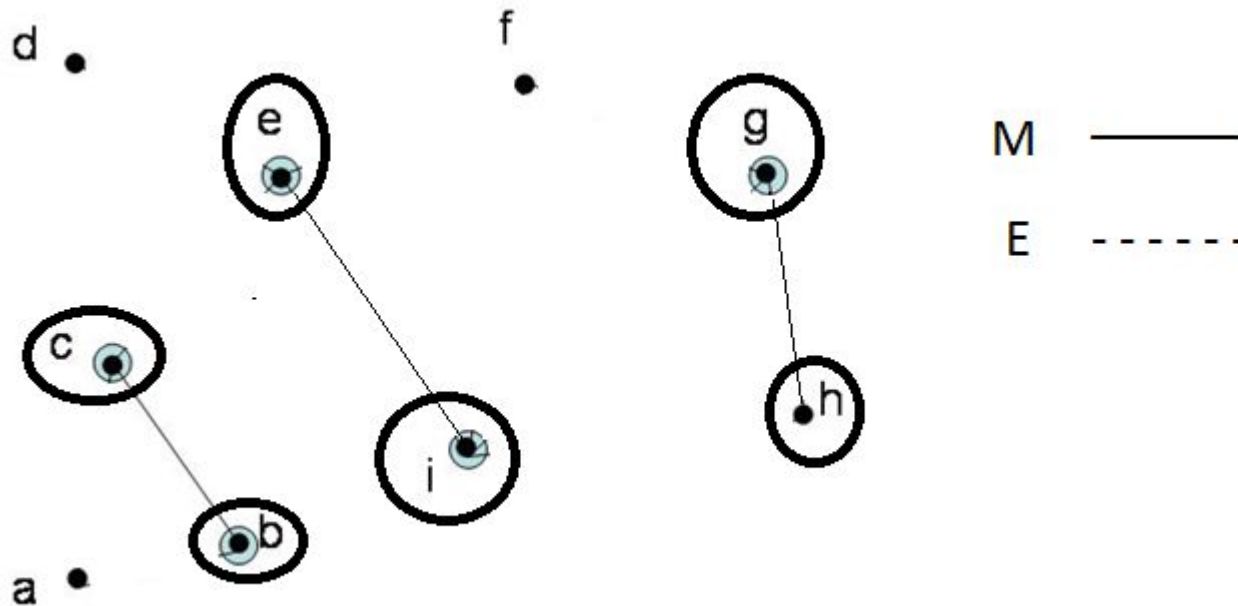
Example of execution - Step 2



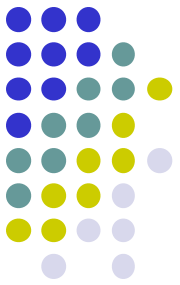
- $U = \{b, c, e, i\}$
- $M = \{\{b, c\}, \{e, i\}\}$



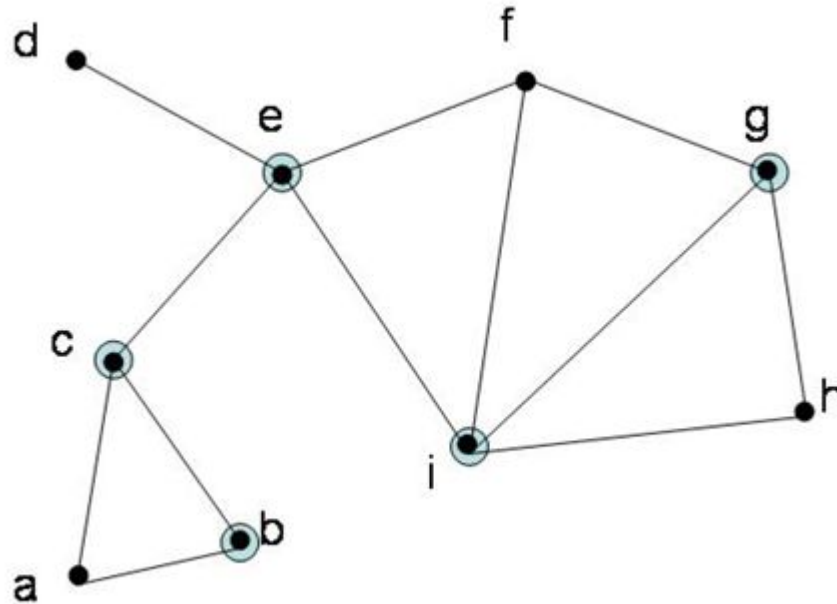
Example of execution - Step 3



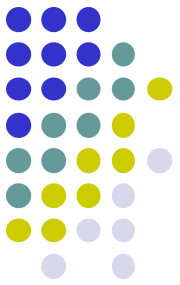
- $U = \{b, c, e, i, g, h\}$
- $M = \{\{b, c\}, \{e, i\}, \{g, h\}\}$



Optimal solution



- $U^* = \{b, c, e, i, g\}$

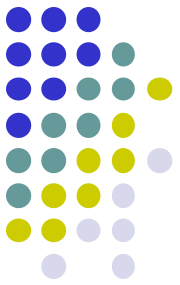


Lemma: At the end of the execution of Approx-Cover M forms a matching, that is the edges in M do not share any endpoint.

Proof: Trivially, every time an edge e is selected in M , all the edges with an endpoint in common with e are deleted from E .

Therefore, in the following steps no edge with an endpoint in common with e can be chosen by the algorithm.

□



Theorem: [Approx-Cover](#) is 2-approximating.

Proof: The value of the solution returned by the algorithm is

$$m = |U| = 2 \cdot |M|.$$

Let U^* be an optimal cover. Since the edges in M do not share any endpoint (M is a matching) and each of them must have an endpoint in U^* ,

$$m^* = |U^*| \geq |M|.$$

Therefore

$$\frac{m}{m^*} \leq \frac{2|M|}{|M|} = 2.$$

□