

Vertex cover

- Vertex cover
- Deterministic algorithm
- Randomized algorithm

Vertex Cover

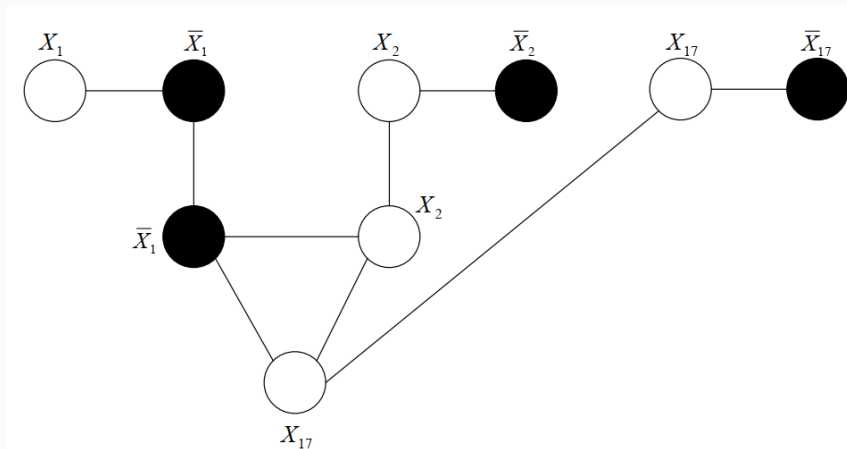
- Given: undirected graph G
- Goal: Find a minimum-cardinality subset $V_0 \subseteq V$ such that if $(u, v) \in E(G)$, then $u \in V_0$ or $v \in V_0$
- Cover edges by picking vertices
- NP-hard

Vertex Cover

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- Cover edges by picking vertices
- NP-hard
- Applications
 - Every edge forms a task, every vertex represents a person that can execute that task
 - Perform all tasks with min resources
 - Extensions: weighted vertices or hypergraphs

Vertex Cover: NP-harness proof

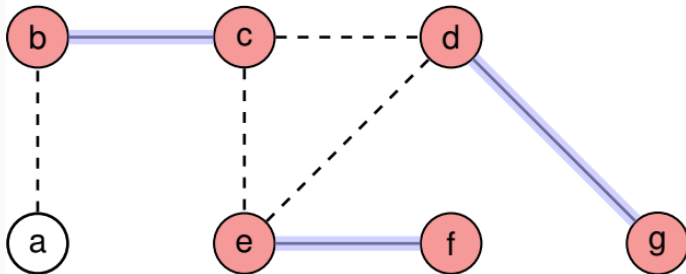
Establish a mapping between VC and 3SAT



A Greedy Approximation Algorithm

APPROX-VERTEX-COVER(G)

```
1   $C = \emptyset$   
2   $E' = G.E$   
3  while  $E' \neq \emptyset$   
4      let  $(u, v)$  be an arbitrary edge of  $E'$   
5       $C = C \cup \{u, v\}$   
6      remove from  $E'$  every edge incident on either  $u$  or  $v$   
7  return  $C$ 
```



Analysis of Greedy Approximation Algorithm

Theorem

APPROX-VERTEX-COVER is a poly-time 2-approximation algorithm

Proof

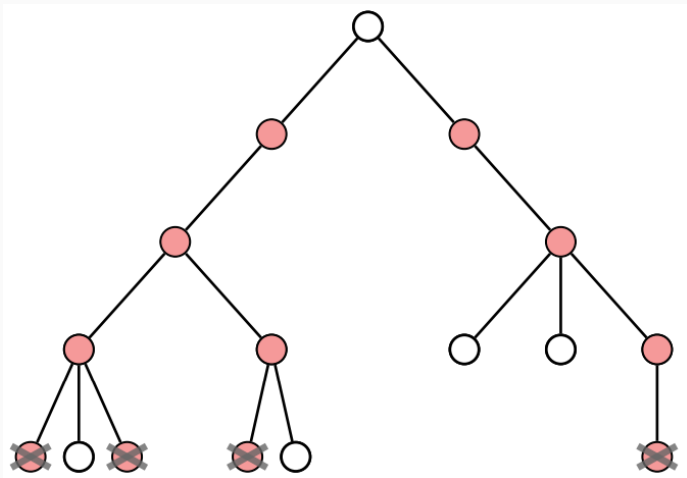
Let $A \subseteq E$ denote the set of chosen edges

- Every optimal cover C^* must include at least one endpoint of edges in A , and edges in A do not share a common endpoint: $|C^*| \geq |A|$
- Every edge in A contributes 2 vertices to $|C|$: $|C| = 2|A| \leq 2|C^*|$

Vertex Cover on Trees

There exists an optimal vertex cover which does not include any leaves

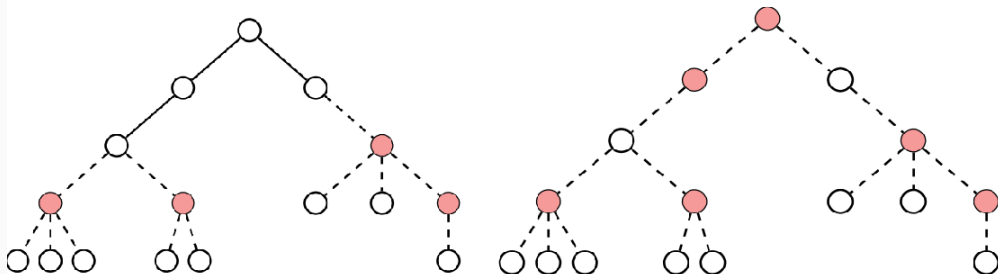
- Replace any leaf in the cover by its parent



Solving Vertex Cover on Trees

VERTEX-COVER-TREES(G)

- 1: $C = \emptyset$
- 2: **while** \exists leaves in G
- 3: Add all parents to C
- 4: Remove all leaves and their parents from G
- 5: **return** C



Can be also solved on bipartite graphs, using Max-Flows and Min-Cuts

Exact Algorithms

Find a vertex cover of size k in any simple graph

- Brute-Force Search takes $\binom{n}{k} = \Theta(n^k)$ time: How to improve?

Theorem

Consider a graph G and its edge uv . Let G_u be the graph obtained by deleting u and its incident edges. G has a vertex cover of size k iff G_u or G_v has a vertex cover of size $k - 1$

Proof

\Leftarrow : Assume G_u has a vertex cover C_u of size $k - 1$, adding u yields a vertex cover of G

\Rightarrow : Assume G has a vertex cover C of size k , which contains, say u , removing u from C yields a vertex cover of G_u

A More Efficient Search Algorithm

```
VERTEX-COVER-SEARCH( $G, k$ )  
1: If  $E = \emptyset$  return  $\emptyset$   
2: If  $k = 0$  and  $E \neq \emptyset$  return  $\perp$   
3: Pick an arbitrary edge  $(u, v) \in E$   
4:  $S_1 = \text{VERTEX-COVER-SEARCH}(G_u, k - 1)$   
5:  $S_2 = \text{VERTEX-COVER-SEARCH}(G_v, k - 1)$   
6: if  $S_1 \neq \perp$  return  $S_1 \cup \{u\}$   
7: if  $S_2 \neq \perp$  return  $S_2 \cup \{v\}$   
8: return  $\perp$ 
```

Running time: $O(2^k)$

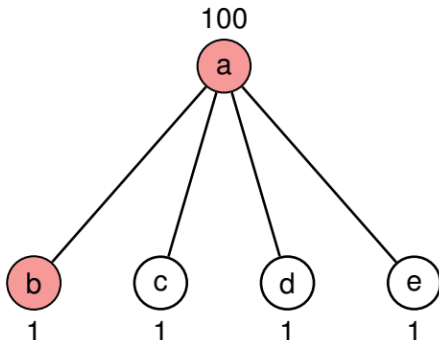
Weighted Vertex Cover

- Given: undirected, vertex-weighted graph G
- Goal: Find a minimum-weight subset $V_0 \subseteq V$ such that if $(u, v) \in E(G)$, then $u \in V_0$ or $v \in V_0$
- Cover edges by picking vertices
- NP-hard
- Applications
 - Every edge forms a task, every vertex represents a person that can execute that task
 - Weight of a vertex could be salary of a person
 - Perform all tasks with the minimal amount of resources

Greedy Algorithm from Unweighted case

APPROX-VERTEX-COVER(G)

```
1  $C = \emptyset$ 
2  $E' = G.E$ 
3 while  $E' \neq \emptyset$ 
4     let  $(u, v)$  be an arbitrary edge of  $E'$ 
5      $C = C \cup \{u, v\}$ 
6     remove from  $E'$  every edge incident on either  $u$  or  $v$ 
7 return  $C$ 
```



Invoking an (Integer) Linear Program

Idea: Round the solution of an associated linear program

0-1 Integer Program

$$\begin{array}{ll} \text{minimize} & \sum_{v \in V} w(v)x(v) \\ \text{subject to} & x(u) + x(v) \geq 1 \quad \text{for each } (u, v) \in E \\ & x(v) \in \{0, 1\} \quad \text{for each } v \in V \end{array}$$

Linear Program

$$\begin{array}{ll} \text{minimize} & \sum_{v \in V} w(v)x(v) \\ \text{subject to} & x(u) + x(v) \geq 1 \quad \text{for each } (u, v) \in E \\ & x(v) \in [0, 1] \quad \text{for each } v \in V \end{array}$$

LP-based Algorithm

APPROX-MIN-WEIGHT-VC(G, w)

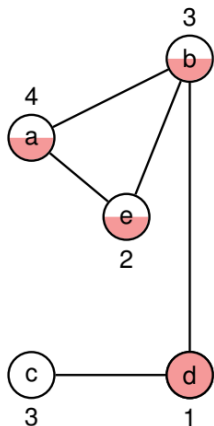
```
1   $C = \emptyset$ 
2  compute  $\bar{x}$ , an optimal solution to the linear program
3  for each  $v \in V$ 
4      if  $\bar{x}(v) \geq 1/2$ 
5           $C = C \cup \{v\}$ 
6  return  $C$ 
```

Approximation ratio

APPROX-MIN-WEIGHT-VC is a polynomial-time 2-approximation algorithm for the minimum-weight vertex-cover problem

Example

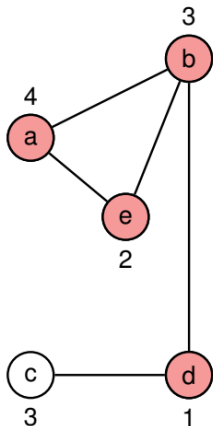
$$\bar{x}(a) = \bar{x}(b) = \bar{x}(e) = \frac{1}{2}, \bar{x}(d) = 1, \bar{x}(c) = 0$$



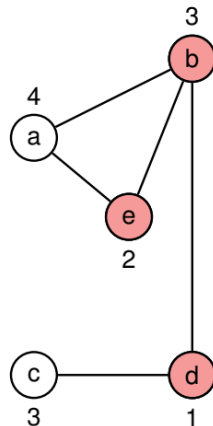
fractional solution of LP
with weight = 5.5

Rounding

$$x(a) = x(b) = x(e) = 1, x(d) = 1, x(c) = 0$$



rounded solution of LP
with weight = 10



optimal solution
with weight = 6

Approximation Ratio

Approximation ratio

- Let C^* denote an optimal solution, z^* the value of an optimal solution of the LP: $z^* \leq w(C^*)$
- We first prove that the computed set C covers all vertices
 - Consider any edge (u, v) with constraint $x(u) + x(v) \geq 1$: at least one of $x(u)$ and $x(v) \geq 1/2$, hence C covers (u, v)
- We then prove that $w(C) \leq 2z^*$

$$w(C^*) \geq z^* = \sum_{v \in V} w(v) \bar{x}(v) \geq \sum_{v \in V: \bar{x}(v) \geq 1/2} w(v) \cdot \frac{1}{2} = \frac{1}{2} W(C)$$

Randomized Algorithm

RAND-VC: For each $e = (u, v)$, if e is not covered, add u to C with probability $w_v / (w_u + w_v)$, otherwise add v

Approximation ratio

RAND-VC is a poly-time 2-approximation algorithm in expectation

Proof

- C_i : C after iteration i ; C^* : optimum
- We prove by induction $E[\sum_{v \in C_i \cap C^*} w_v] \geq E[\sum_{v \in C_i \setminus C^*} w_v]$
 - If both $u, v \in C^*$: $>$ holds; if one of $u, v \in C^*$:

$$E \left[\sum_{v \in C_i \cap C^*} w_v \right] = E \left[\sum_{v \in C_{i-1} \cap C^*} w_v \right] + \frac{w_u w_v}{w_u + w_v};$$

$$E \left[\sum_{v \in C_i \setminus C^*} w_v \right] = E \left[\sum_{v \in C_{i-1} \setminus C^*} w_v \right] + \frac{w_u w_v}{w_u + w_v}$$