

**10. EXTENDING TRACTABILITY** 

- finding small vertex covers
- solving NP-hard problems on trees
- circular arc coverings
- vertex cover in bipartite graphs

Lecture slides by Kevin Wayne Copyright © 2005 Pearson-Addison Wesley http://www.cs.princeton.edu/~wayne/kleinberg-tardos

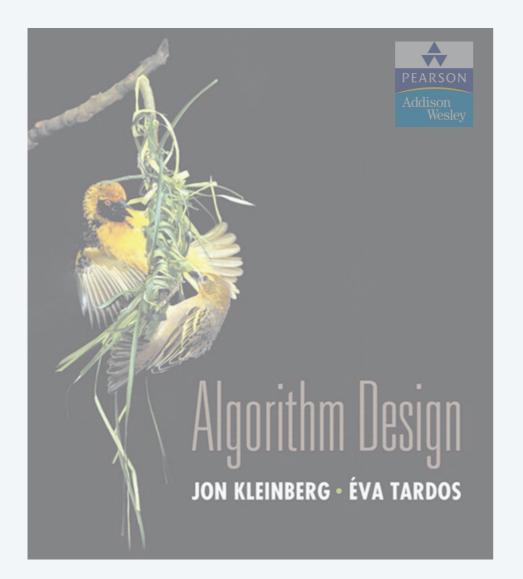
## Coping with NP-completeness

- Q. Suppose I need to solve an NP-complete problem. What should I do?
- A. Theory says you're unlikely to find poly-time algorithm.

Must sacrifice one of three desired features.

- Solve problem to optimality.
- Solve problem in polynomial time.
- Solve arbitrary instances of the problem.

This lecture. Solve some special cases of NP-complete problems.

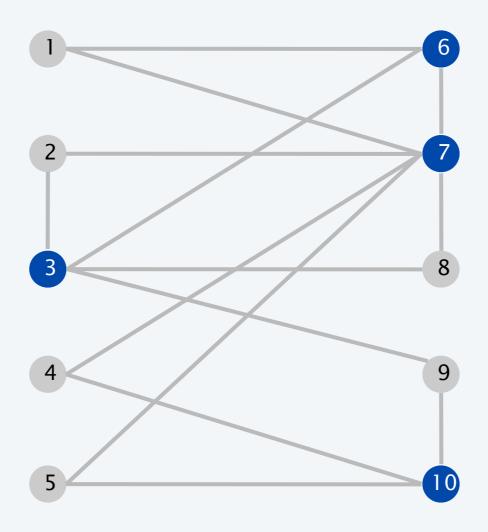


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### Vertex cover

Given a graph G = (V, E) and an integer k, is there a subset of vertices  $S \subseteq V$ such that  $|S| \le k$ , and for each edge (u, v) either  $u \in S$  or  $v \in S$  or both?



S = { 3, 6, 7, 10 } is a vertex cover of size k = 4

**Q.** VERTEX-COVER is **NP**-complete. But what if *k* is small?

#### Brute force. $O(k n^{k+1})$ .

- Try all  $C(n, k) = O(n^k)$  subsets of size k.
- Takes *O*(*kn*) time to check whether a subset is a vertex cover.

**Goal**. Limit exponential dependency on k, say to  $O(2^k k n)$ .

Ex. n = 1,000, k = 10. Brute.  $k n^{k+1} = 10^{34} \implies$  infeasible. Better.  $2^k k n = 10^7 \implies$  feasible.

**Remark.** If *k* is a constant, then the algorithm is poly-time; if *k* is a small constant, then it's also practical.

Claim. Let (u, v) be an edge of G. G has a vertex cover of size  $\leq k$  iff at least one of  $G - \{u\}$  and  $G - \{v\}$  has a vertex cover of size  $\leq k - 1$ .

delete v and all incident edges

Pf.  $\Rightarrow$ 

- Suppose G has a vertex cover S of size  $\leq k$ .
- *S* contains either *u* or *v* (or both). Assume it contains *u*.
- $S \{u\}$  is a vertex cover of  $G \{u\}$ .

### **Pf.** ⇐

- Suppose *S* is a vertex cover of  $G \{u\}$  of size  $\leq k 1$ .
- Then  $S \cup \{u\}$  is a vertex cover of G.

Claim. If G has a vertex cover of size k, it has  $\leq k (n - 1)$  edges. Pf. Each vertex covers at most n - 1 edges.

## Finding small vertex covers: algorithm

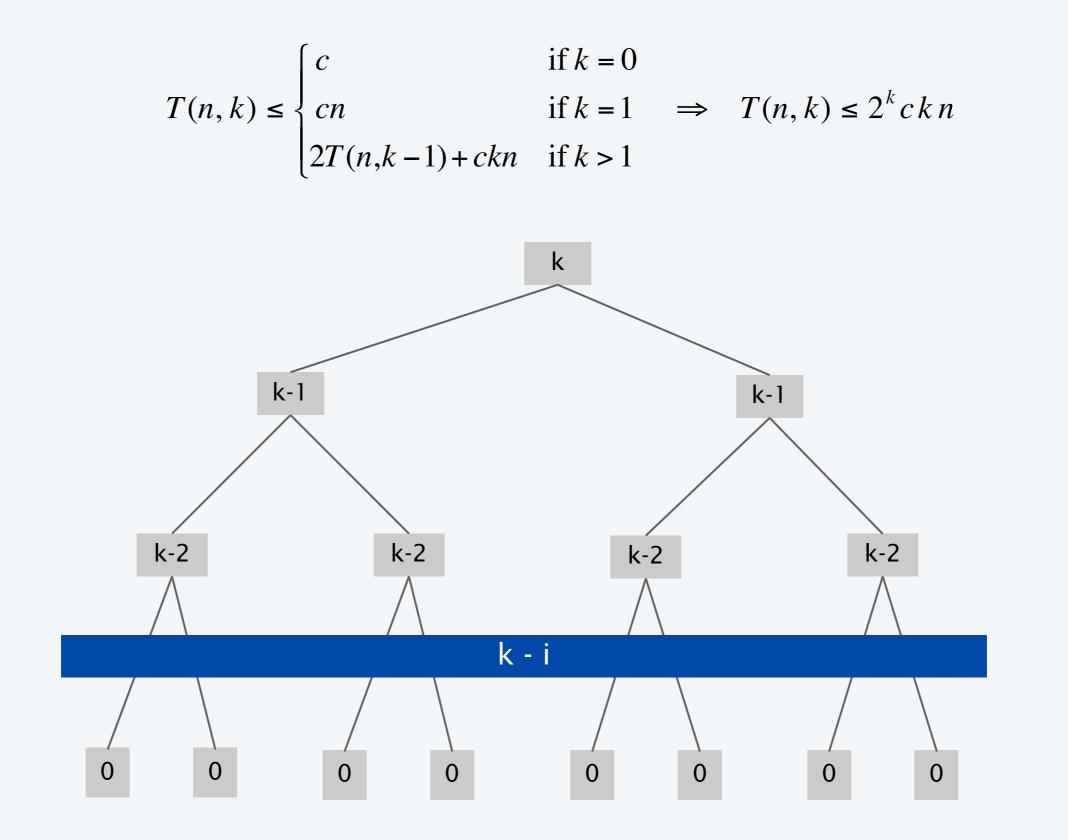
Claim. The following algorithm determines if G has a vertex cover of size  $\leq k$  in O(2<sup>k</sup> kn) time.

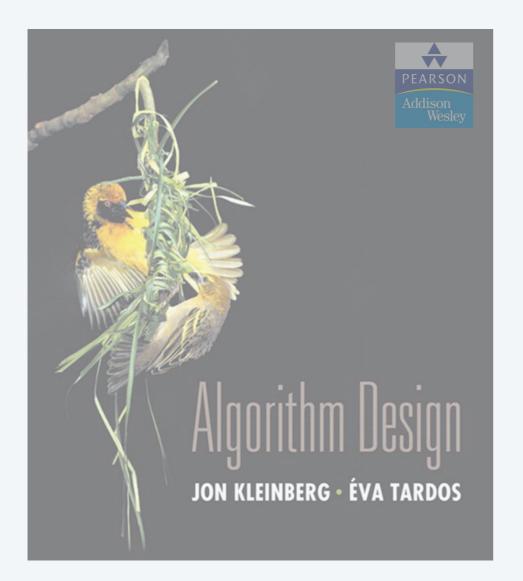
```
Vertex-Cover(G, k) {
    if (G contains no edges) return true
    if (G contains ≥ kn edges) return false
    let (u, v) be any edge of G
    a = Vertex-Cover(G - {u}, k-1)
    b = Vertex-Cover(G - {v}, k-1)
    return a or b
}
```

#### Pf.

- Correctness follows from previous two claims.
- There are ≤ 2<sup>k+1</sup> nodes in the recursion tree; each invocation takes O(kn) time.

### Finding small vertex covers: recursion tree





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## Independent set on trees

Independent set on trees. Given a tree, find a maximum cardinality subset of nodes such that no two share an edge.

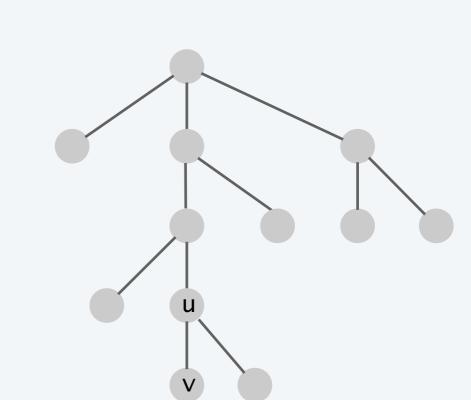
Fact. A tree on at least two nodes has at least two leaf nodes.

degree = 1

Key observation. If *v* is a leaf, there exists a maximum size independent set containing *v*.

Pf. (exchange argument)

- Consider a max cardinality independent set S.
- If  $v \in S$ , we're done.
- If  $u \notin S$  and  $v \notin S$ , then  $S \cup \{v\}$  is independent  $\Rightarrow S$  not maximum.
- If  $u \in S$  and  $v \notin S$ , then  $S \cup \{v\} \{u\}$  is independent.



Theorem. The following greedy algorithm finds a maximum cardinality independent set in forests (and hence trees).

```
Independent-Set-In-A-Forest(F) {
   S ← φ
   while (F has at least one edge) {
      Let e = (u, v) be an edge such that v is a leaf
      Add v to S
      Delete from F nodes u and v, and all edges
      incident to them.
   }
  return S ∪ { isolated vertices in F }
}
```

Pf. Correctness follows from the previous key observation. •

Remark. Can implement in O(n) time by considering nodes in postorder.

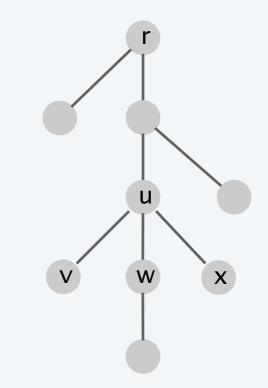
Weighted independent set on trees. Given a tree and node weights  $w_v > 0$ , find an independent set *S* that maximizes  $\sum_{v \in S} w_v$ .

Observation. If (u, v) is an edge such that v is a leaf node, then either *OPT* includes all leaf nodes incident to u.

Dynamic programming solution. Root tree at some node, say r.

- OPT<sub>in</sub>(u) = max weight independent set of subtree rooted at u, containing u.
- *OPT<sub>out</sub>*(*u*) = max weight independent set of subtree rooted at *u*, not containing *u*.

$$OPT_{in}(u) = w_u + \sum_{v \in children(u)} OPT_{out}(v)$$
$$OPT_{out}(u) = \sum_{v \in children(u)} \max \left\{ OPT_{in}(v), OPT_{out}(v) \right\}$$



children(u) = { v, w, x }

## Weighted independent set on trees: dynamic programming algorithm

Theorem. The dynamic programming algorithm finds a maximum weighted independent set in a tree in O(n) time.

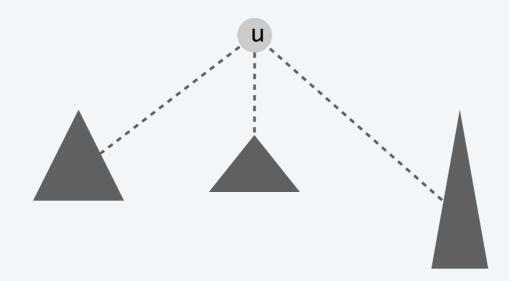
independent set itself

(not just value)

```
Weighted-Independent-Set-In-A-Tree(T) {
    Root the tree at a node r
    foreach (node u of T in postorder) {
        if (u is a leaf) {
            M_{in} [u] = W_{u}
                                              ensures a node is visited
                                                   after all its children
            M_{out}[u] = 0
        }
        else {
            M_{in} [u] = W_u + \Sigma_{v \in children(u)} M_{out}[v]
            M_{out}[u] = \Sigma_{v \in children(u)} max(M_{in}[v], M_{out}[v])
        }
    }
    return max(M<sub>in</sub>[r], M<sub>out</sub>[r])
}
```

### Context

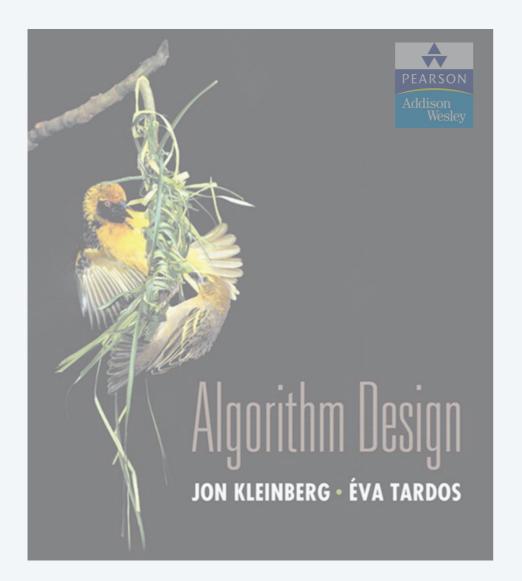
Independent set on trees. This structured special case is tractable because we can find a node that breaks the communication among the subproblems in different subtrees.



see Section 10.4 (but proceed with caution)

Graphs of bounded tree width. Elegant generalization of trees that:

- Captures a rich class of graphs that arise in practice.
- Enables decomposition into independent pieces.



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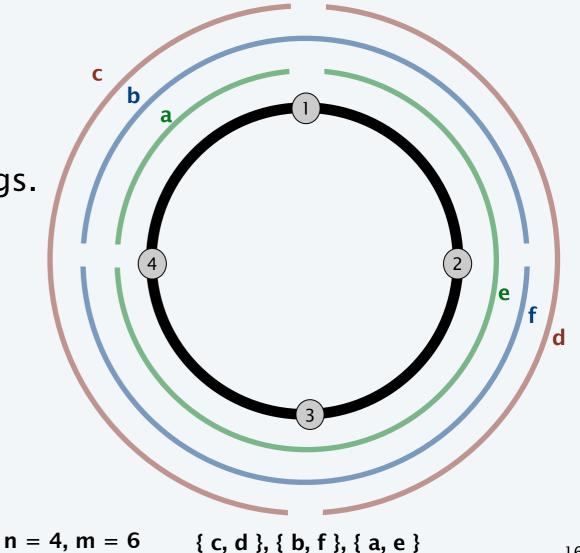
Wavelength-division multiplexing (WDM). Allows *m* communication streams (arcs) to share a portion of a fiber optic cable, provided they are transmitted using different wavelengths.

Ring topology. Special case is when network is a cycle on *n* nodes.

Bad news. NP-complete, even on rings.

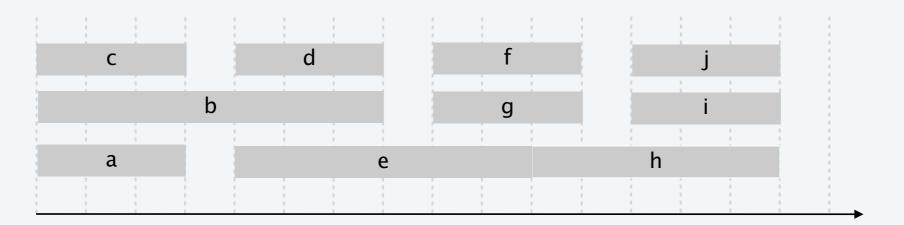
Brute force. Can determine if k colors suffice in  $O(k^m)$  time by trying all k-colorings.

**Goal.**  $O(f(k)) \cdot poly(m, n)$  on rings.



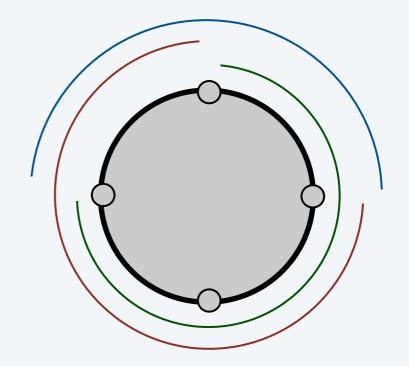
Interval coloring. Greedy algorithm finds coloring such that number of colors equals depth of schedule.

maximum number of streams at one location



#### Circular arc coloring.

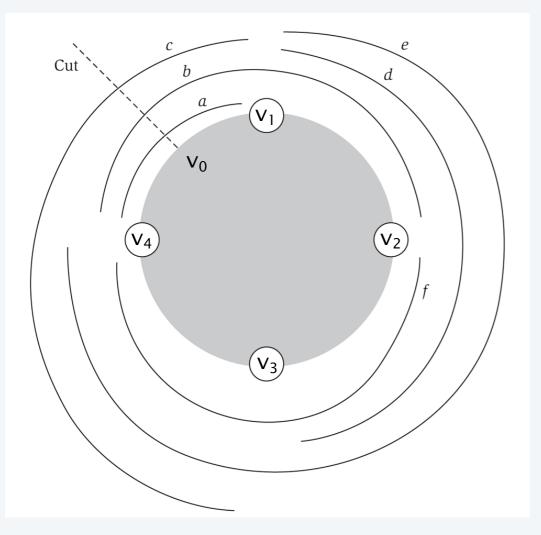
- Weak duality: number of colors  $\geq$  depth.
- Strong duality does not hold.

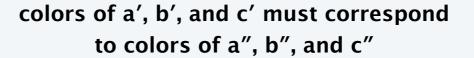


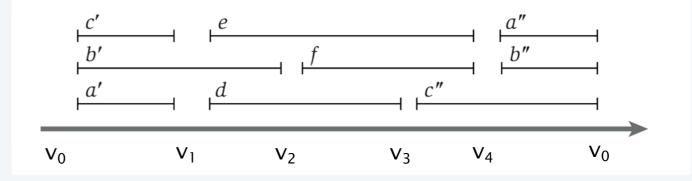
## (Almost) transforming circular arc coloring to interval coloring

Circular arc coloring. Given a set of *n* arcs with depth  $d \le k$ , can the arcs be colored with *k* colors?

Equivalent problem. Cut the network between nodes  $v_1$  and  $v_n$ . The arcs can be colored with k colors iff the intervals can be colored with k colors in such a way that "sliced" arcs have the same color.



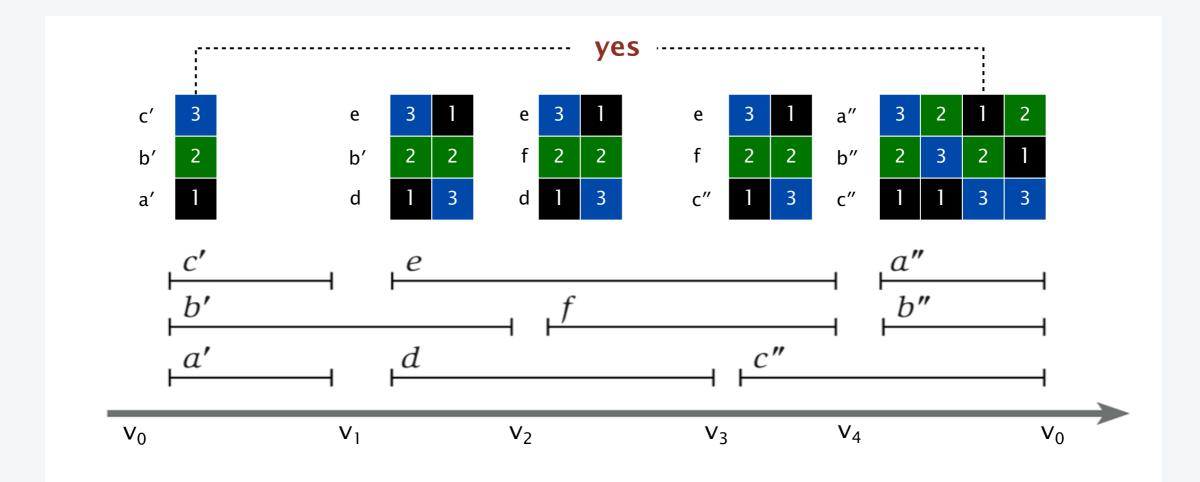




## Circular arc coloring: dynamic programming algorithm

#### Dynamic programming algorithm.

- Assign distinct color to each interval which begins at cut node  $v_0$ .
- At each node  $v_i$ , some intervals may finish, and others may begin.
- Enumerate all *k*-colorings of the intervals through  $v_i$  that are consistent with the colorings of the intervals through  $v_{i-1}$ .
- The arcs are k-colorable iff some coloring of intervals ending at cut node  $v_0$  is consistent with original coloring of the same intervals.



Running time.  $O(k! \cdot n)$ .

- The algorithm has *n* phases.
- Bottleneck in each phase is enumerating all consistent colorings.
- There are at most k intervals through v<sub>i</sub>, so there are at most k! colorings to consider.

**Remark.** This algorithm is practical for small values of k (say k = 10) even if the number of nodes n (or paths) is large.

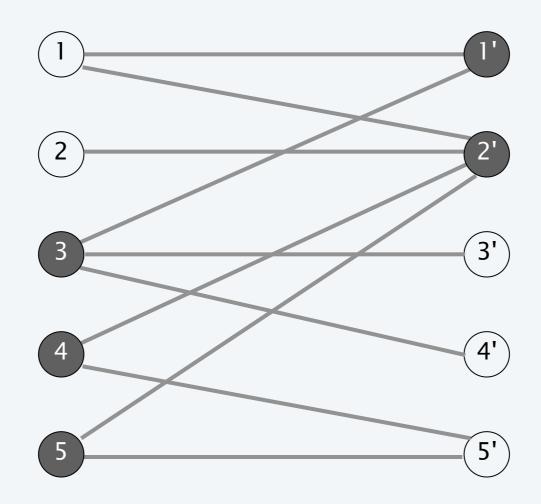
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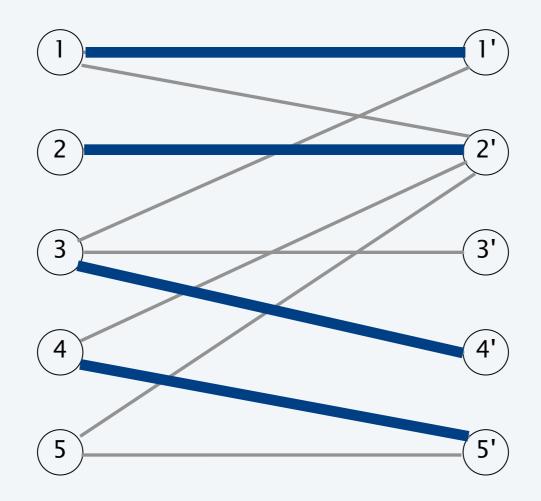


vertex cover S = { 3, 4, 5, 1', 2' }

## Vertex cover and matching

Weak duality. Let *M* be a matching, and let *S* be a vertex cover. Then,  $|M| \le |S|$ .

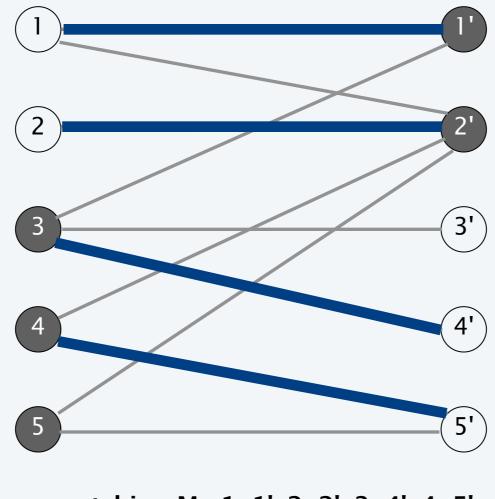
Pf. Each vertex can cover at most one edge in any matching.



matching M: 1-1', 2-2', 3-4', 4-5'

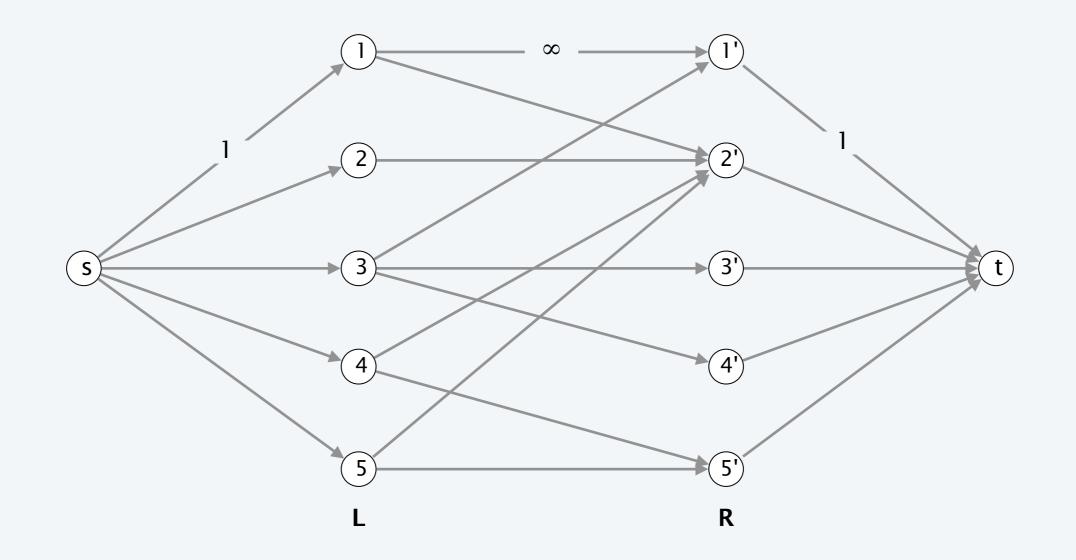
## Vertex cover in bipartite graphs: König-Egerváry Theorem

Theorem. [König-Egerváry] In a bipartite graph, the max cardinality of a matching is equal to the min cardinality of a vertex cover.



matching M: 1-1', 2-2', 3-4', 4-5' vertex cover S = { 3, 4, 5, 1', 2' } Theorem. [König-Egerváry] In a bipartite graph, the max cardinality of a matching is equal to the min cardinality of a vertex cover.

- Suffices to find matching *M* and cover *S* such that |M| = |S|.
- Formulate max flow problem as for bipartite matching.
- Let *M* be max cardinality matching and let (*A*, *B*) be min cut.



Theorem. [König-Egerváry] In a bipartite graph, the max cardinality of a matching is equal to the min cardinality of a vertex cover.

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- Formulate max flow problem as for bipartite matching.
- Let *M* be max cardinality matching and let (*A*, *B*) be min cut.
- Define  $L_A = L \cap A$ ,  $L_B = L \cap B$ ,  $R_A = R \cap A$ ,  $R_B = R \cap B$ .
- Claim 1.  $S = L_B \cup R_A$  is a vertex cover.
  - consider  $(u, v) \in E$
  - $u \in L_A$ ,  $v \in R_B$  impossible since infinite capacity
  - thus, either  $u \in L_B$  or  $v \in R_A$  or both
- Claim 2. |M| = |S|.
  - max-flow min-cut theorem  $\Rightarrow |M| = cap(A, B)$
  - only edges of form (*s*, *u*) or (*v*, *t*) contribute to *cap*(*A*, *B*)
  - $|M| = cap(A, B) = |L_B| + |R_A| = |S|.$