10. EXTENDING TRACTABILITY

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

Algorithm Design

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Coping with NP-completeness

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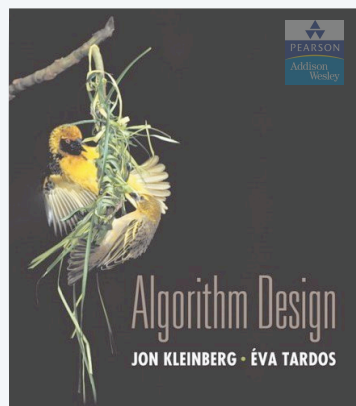A. Theory says you're unlikely to find poly-time algorithm.

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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SECTION 10.1

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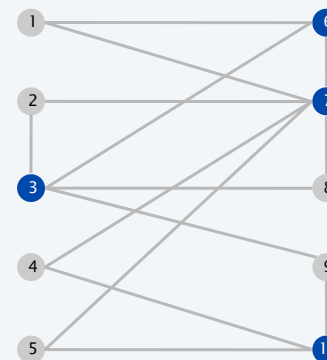
S = { 3, 6, 7, 10 } is a vertex cover of size k = 4

Finding small vertex covers

Q. VertexCover 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 $O(k\,n)$ 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. $k\,n^{k+1} = 10^{34} \Rightarrow$ infeasible.
Better. $2^k\,k\,n = 10^7 \Rightarrow$ feasible.

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

Finding small vertex covers

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

delete v and all incident edges

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

Pf. $\Leftarrow$
- Suppose S is a vertex cover of $G - \{u\}$ of size $\le k - 1$.
- Then $S \cup \{u\}$ is a vertex cover of G. ∎

Claim. If G has a vertex cover of size k, it has $\le k\,(n - 1)$ 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 $\le k$ in $O(2^k\,kn)$ 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
}
```

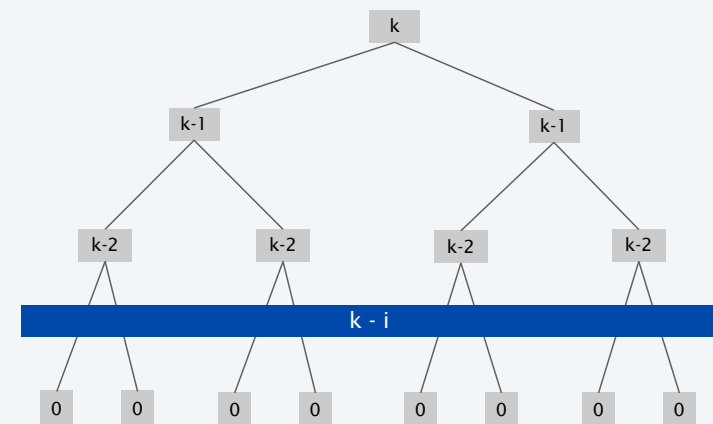Pf.
- Correctness follows from previous two claims.
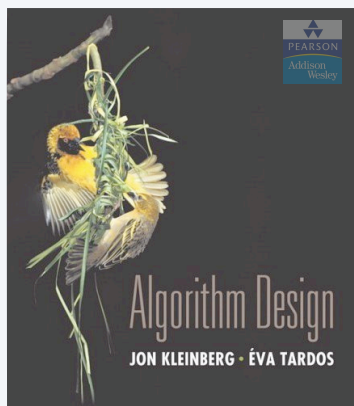- There are $\le 2^{k+1}$ nodes in the recursion tree; each invocation takes $O(kn)$ time. ∎

Finding small vertex covers: recursion tree

$$T(n, k) \le \begin{cases} c & \text{if } k = 0 \\ cn & \text{if } k = 1 \\ 2T(n, k-1) + ckn & \text{if } k > 1 \end{cases} \Rightarrow T(n, k) \le 2^k\,ckn$$
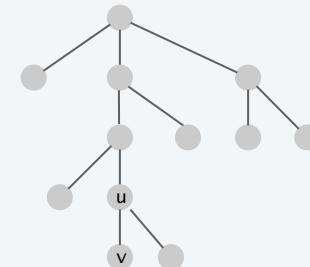
10. EXTENDING TRACTABILITY

PEARSON
Addison
Wesley

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SECTION 10.2

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. ∎

Independent set on trees: greedy algorithm

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
}
```

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

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

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

Dynamic programming solution. Root tree at some node, say r.
- $OPT_{in}(u)$ = max weight independent set of subtree rooted at u, containing u.
- $OPT_{out}(u)$ = max weight independent set of subtree rooted at u, not containing u.

$$OPT_{in}(u) = w_u + \sum_{v \in \text{children}(u)} OPT_{out}(v)$$

$$OPT_{out}(u) = \sum_{v \in \text{children}(u)} \max\left\{OPT_{in}(v), OPT_{out}(v)\right\}$$
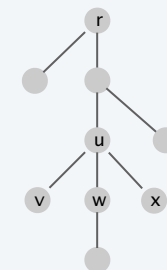
children(u) = { v, w, x }

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.

*can also find 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
            M_out[u] = 0
        }
        else {
            M_in [u] = w_u + Σ_{v∈children(u)} M_out[v]
            M_out[u] = Σ_{v∈children(u)} max(M_in[v], M_out[v])
        }
    }
    return max(M_in[r], M_out[r])
}
```

ensures a node is visited after all its children

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.
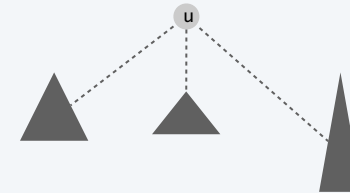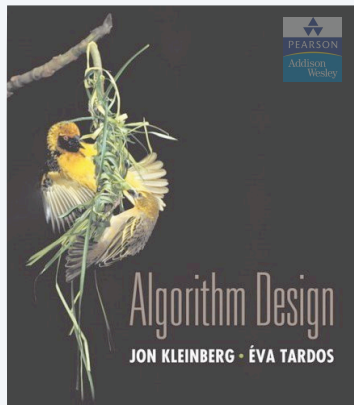

see Chapter 10.4 (but proceed with caution)

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.

SECTION 10.3

10. EXTENDING TRACTABILITY

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▸ **circular arc coverings**
▸ *vertex cover in bipartite graphs*

## Wavelength-division multiplexing

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.

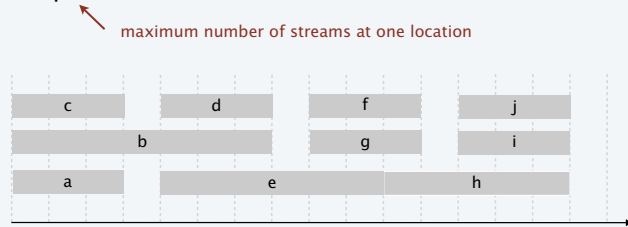**Goal.** $O(f(k)) \cdot poly(m, n)$ on rings.
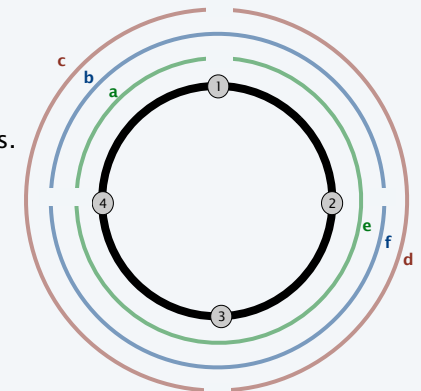
n = 4, m = 6   { c, d }, { b, f }, { a, e }

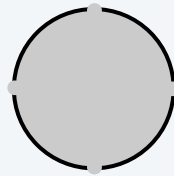## Review: interval coloring

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 ≥ depth.
- Strong duality does not hold.

max depth = 2
min colors = 3

(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.
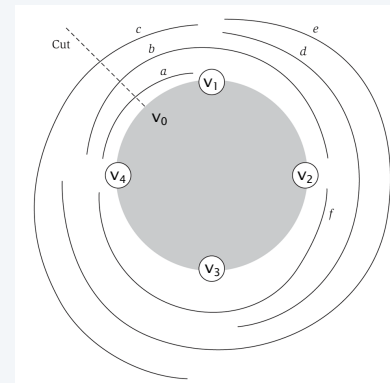
colors of a', b', and c' must correspond to colors of a", b", and c"
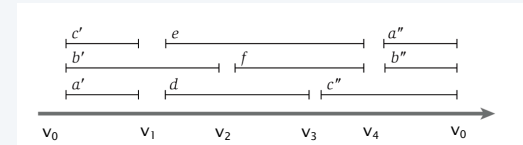

Circular arc coloring: dynamic programming algorithm

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.
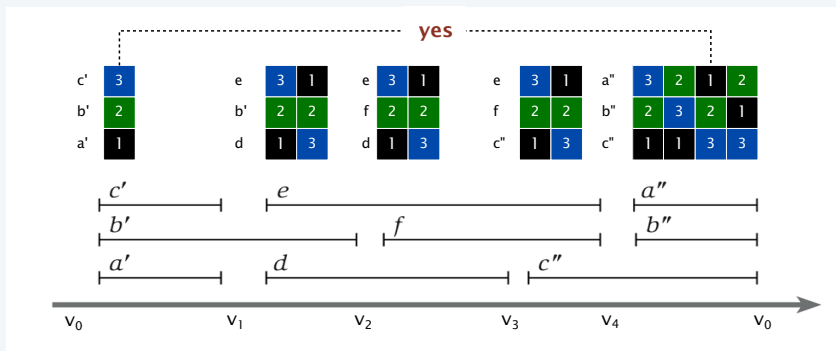
Circular arc coloring: running time

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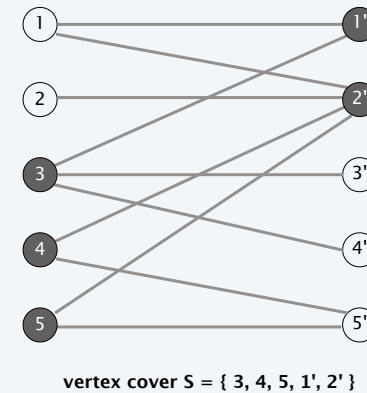
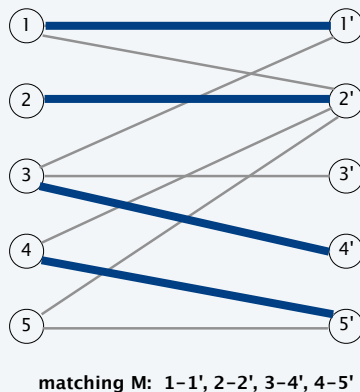vertex cover S = { 3, 4, 5, 1', 2' }

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.

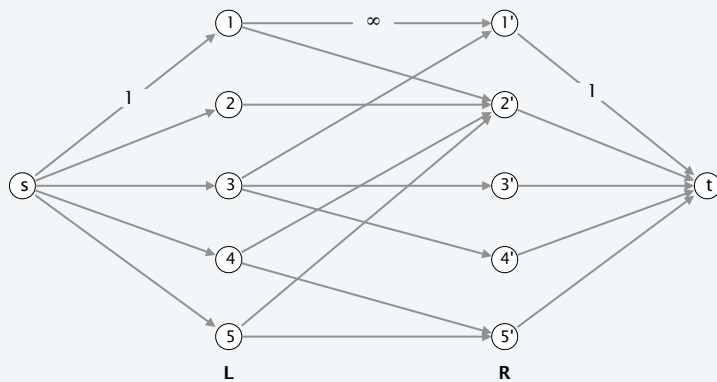matching M: 1–1', 2–2', 3–4', 4–5'
vertex cover S = { 3, 4, 5, 1', 2' }

Proof of 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.
- 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.

L   **R**

Proof of 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.
- 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.
- 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 $cap(A, B)$
  - $|M| = cap(A, B) = |L_B| + |R_A| = |S|$. ∎