Algorithm Design (XV)

Extending Tractability

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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.

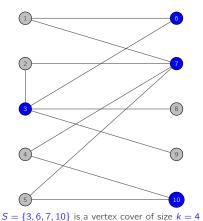
Must sacrifice one of three desired features.

- 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.

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?



Finding small vertex covers

Q. Vertex cover is **NP**-complete. But what if k is small?

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

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

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

Example. n = 1,000, k = 10. Brute. $kn^{k+1} = 10^{34} \Rightarrow \text{infeasible}$.

Better $2^k kn = 10^7 \Rightarrow$ feasible

Remark

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

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Finding small vertex covers

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 $\leq k - 1$.

Proof.



- 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 $G \{u\}$.

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- 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.

Proof. 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 $O(2^k kn)$ time.

```
VertexCover(G, k)

if G contains no edges then Return true;

if G contains \geq kn edges then Return false;

let (u, v) be any edge of G;

a = \text{VertexCover}(G - \{u\}, k - 1);

b = \text{VertexCover}(G - \{v\}, k - 1);

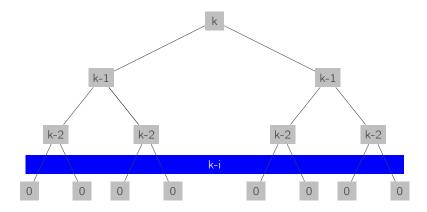
Return a or b;
```

Proof.

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

Finding small vertex covers: recursion tree

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



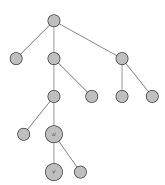
Solving NP-Hard Problems on Trees

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.

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



Proof. (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 $u \notin S$ 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 SetInForest(F)
```

 $S \leftarrow \phi$;

while F has at least one edge do

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;

end

Return *S*:

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

Weighted independent set on trees

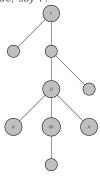
Weighted independent set on trees. Given a tree and node weights $w_v > 0$, find an independent set S that maximizes $\Sigma_{v \in S} w_v$.

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.

$$\begin{split} \textit{OPT}_{\textit{in}}(u) &= w_u + \sum_{v \in \textit{children } (u)} \textit{OPT}_{\textit{out}}(v) \\ \textit{OPT}_{\textit{out}}(u) &= \sum_{v \in \textit{children } (u)} \max \left\{ \textit{OPT}_{\textit{in}}(v), \textit{OPT}_{\textit{out}}(v) \right\} \end{split}$$



Weighted independent set on trees: dynamic programming algorithm

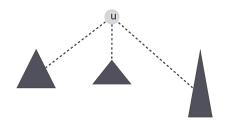
Theorem

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

```
WeightedIndependentSetInTree(T)
Root the tree at a node r:
for each node u of T in postorder do
    if u is a leaf then
        M_{in}[u] = w_u;
        M_{out}[u] = 0:
    end
    else
         M_{\text{in}}[u] = w_u + \sum_{\text{vechildren } (u)} M_{\text{out}}[v];
        M_{out}[u] = \sum_{\text{vechildren } (u)} \max(M_{in}[v], M_{\text{out }}[v]);
    end
end
Return \max(M_{in}[r], M_{out}[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.



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.

Circular Arc Coverings

Wavelength-division multiplexing

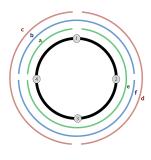
Wavelength-division multiplexing (WDM). Allows m communication streams (arcs) to share a portion of a fiber optic cable, provided they are transmitted using different wavelengths.

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.

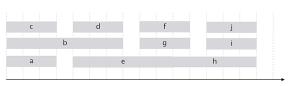
Goal. $O(f(k)) \cdot poly(m, n)$ on rings.



$$n = 4, m = 6 \quad \{c, d\}, \{b, f\}, \{a, e\}$$

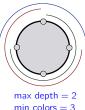
Review: interval coloring

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



Circular arc coloring.

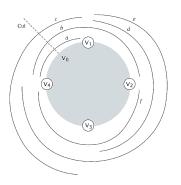
- Weak duality: number of colors > depth.
- Strong duality does not hold.



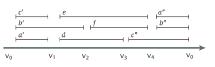
(Almost) transforming circular arc coloring to interval coloring

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.



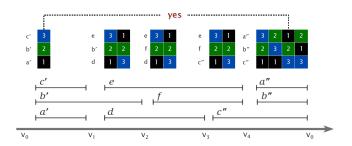
colors of a', b', and c' must correspond to colors of a", b", and c"



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

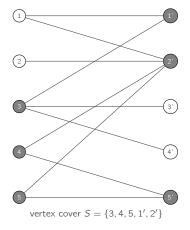
- The algorithm has *n* phases.
- Bottleneck in each phase is enumerating all consistent colorings.
- There are at most k intervals through v_i, so there are at most k!
 colorings to consider.

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

Vertex Cover in Bipartite Graphs

Vertex cover

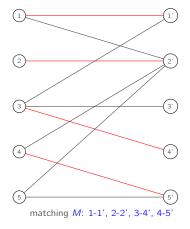
Given a graph G = (V, E) and an integer k, is there a subset of vertices $S \subseteq V$ such that $|S| \le k$, and for each edge (u, v) either $u \in S$ or $v \in S$ or both?



Vertex cover and matching

Weak duality. Let M be a matching, and let S be a vertex cover. Then, $\mid M \mid \leq \mid S \mid$.

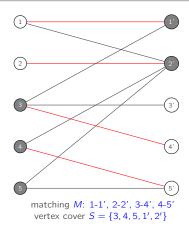
Proof. Each vertex can cover at most one edge in any matching.



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.

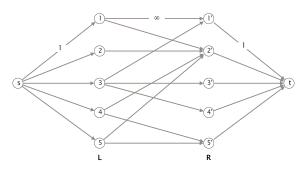


Proof of König-Egerváry Theorem

Theorem (König-Egerváry)

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- 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.



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 $\Rightarrow M = \operatorname{cap}(A, B)$
 - only edges of form (s, u) or (v, t) contribute to cap(A, B)
 - $|M| = \text{cap}(A, B) = |L_B| + |R_A| = |S|$.