Recapture Dynamic programming

Chapter 6 in Kleinberg & Tardos

Independent Set on Trees

Dynamic Programming Summary

Recipe.

- 1. Characterize structure of problem.
- 2. Recursively define *value* of optimal solution.
- 3. Compute and store *values* of optimal solution iteratively.
- 4. Construct optimal solution itself from computed information.

Dynamic programming techniques.

- Binary choice: weighted interval scheduling
- Multi-way choice: segmented least squares.
- Adding a new variable: knapsack.
- Dynamic programming over intervals (more subproblems): RNA secondary structure.

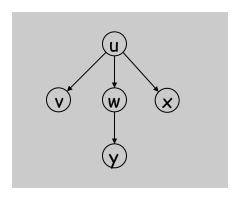


-neighbors are not allowed

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

Brute Force. O(2ⁿ)

With dynamic programming... efficiently solvable!





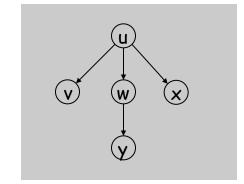
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Weighted independent set on trees. Given a tree and node weights $w_v > 0$, find an independent set S that maximizes $\Sigma_{v \in S} w_v$.

Idea. Use dynamic programming to optimize sum of weights OPT(u) for a tree with root u.

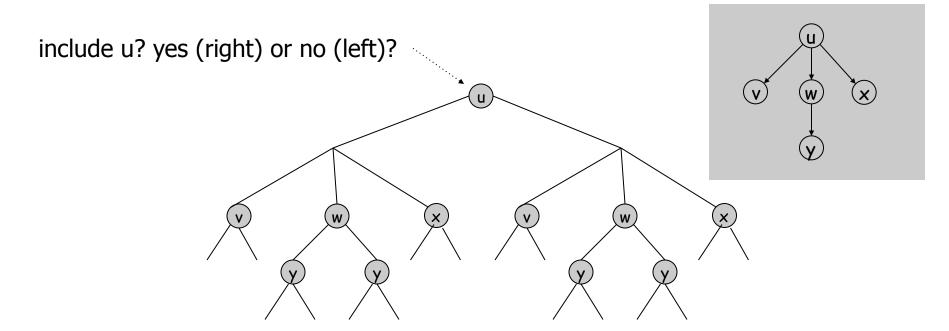
Start with defining a search tree:

- Q. Starting at root u, what are the options?
- Α.
 - 1. include u, or
 - 2. don't include u





Independent Set: Brute Force Search Tree



Observation. Number of nodes grows exponentially with problem size. Observation. Search tree may contain redundant sub-problems (e.g. y). Two types of subproblems y: where y may be chosen, or not.

Idea. Dynamic programming:

- 1. Store and reuse solutions to subproblems.
- 2. Compute these bottom-up.



neighbors are not allowed

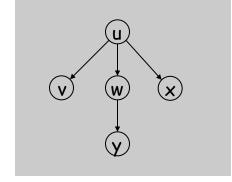
Weighted independent set on trees. Given a tree and node weights $w_v > 0$, find an independent set S that maximizes $\Sigma_{v \in S} w_v$.

Idea. Use dynamic programming to optimize sum of weights for a tree with root u.

Q. Starting at root u, what are the options?

A.

- 1. include u (and thus don't include children of u), or
- 2. don't include u (and possibly include all children of u).



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

Q. How to express the value of an optimal solution in these cases?

A.

- 1. = w_u + sum over optimal solution of children, excluding children
- 2. = sum over optimal solution of children

Idea. Use different notation for optimal solution with and without u.

Idea. Use different notation for OPT with and without u.

- OPT_{in} (u) = max weight independent set rooted at u, containing u.
- OPT_{out}(u) = max weight independent set rooted at u, not containing u.

$$OPT(u) = \max \{OPT_{in}(u), OPT_{out}(u)\}$$

The two subcases are:

- 1. include u and don't include children of u, or
- 2. don't include u and possibly include all children of u. Give recursive formulas for OPT_{in} and OPT_{out}.

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

$$\begin{aligned} OPT_{in}(u) &= w_u + \sum_{v \in \text{children}(u)} OPT_{out}(v) \\ OPT_{out}(u) &= \sum_{v \in \text{children}(u)} \left\{ OPT_{in}(v), \ OPT_{out}(v) \right\} \end{aligned}$$



Independent Set on Trees: DP Algorithm

Claim. The following dynamic programming algorithm efficiently finds a maximum weighted independent set in trees.

```
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_{in} start from bottom:
                                          ensures a node is visited after
         \mathbf{M}_{\mathrm{out}}[\mathbf{u}] = 0
                                            all its children
     else {
         M_{in}[u] = \sum_{v \in children(u)} M_{out}[v] + w_{u}
         M_{\text{out}}[u] = \sum_{v \in \text{children}(u)} \max(M_{\text{out}}[v], M_{\text{in}}[v])
return max(M<sub>in</sub>[r], M<sub>out</sub>[r])
```

- Q. What is the run time of this algorithm?
- A. Takes O(n) time since we visit nodes in postorder and examine each edge exactly once. •