Algorithms & Models of Computation CS/ECE 374, Fall 2020

14.3 Maximum Weighted Independent Set in Trees

Maximum Weight Independent Set Problem

Input Graph G = (V, E) and weights $w(v) \ge 0$ for each $v \in V$ Goal Find maximum weight independent set in G



Maximum weight independent set in above graph: {**B**, **D**}

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Maximum Weight Independent Set in a Tree

Input Tree T = (V, E) and weights $w(v) \ge 0$ for each $v \in V$ Goal Find maximum weight independent set in T



Maximum weight independent set in above tree: ??

For an arbitrary graph **G**:

- Number vertices as v_1, v_2, \ldots, v_n
- Solutions Find recursively optimum solutions without v_n (recurse on $G v_n$) and with v_n (recurse on $G v_n N(v_n)$ & include v_n).
- Saw that if graph **G** is arbitrary there was no good ordering that resulted in a small number of subproblems.

What about a tree? Natural candidate for v_n is root r of T?

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Natural candidate for v_n is root r of T? Let \mathcal{O} be an optimum solution to the whole problem.

Case $r \notin \mathcal{O}$: Then \mathcal{O} contains an optimum solution for each subtree of T hanging at a child of r.

Case $r \in \mathcal{O}$: None of the children of r can be in \mathcal{O} . $\mathcal{O} - \{r\}$ contains an optimum solution for each subtree of T hanging at a grandchild of r.

Subproblems? Subtrees of **T** rooted at nodes in **T**.

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Example



A Recursive Solution

T(u): subtree of T hanging at node uOPT(u): max weighted independent set value in T(u)

 $OPT(u) = \max \begin{cases} \sum_{v \text{ child of } u} OPT(v), \\ w(u) + \sum_{v \text{ grandchild of } u} OPT(v) \end{cases}$

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- Naive bound: $O(n^2)$ since each $M[v_i]$ evaluation may take O(n) time and there are *n* evaluations.
- Better bound: O(n). A value M[v_j] is accessed only by its parent and grand parent.

$$\begin{split} \textbf{MIS-Tree}(\textbf{\textit{T}}): \\ & \text{Let } \textbf{\textit{v}}_1, \textbf{\textit{v}}_2, \dots, \textbf{\textit{v}}_n \text{ be a post-order traversal of nodes of T} \\ & \textbf{for } i = 1 \text{ to } \textbf{\textit{n}} \text{ do} \\ & \textbf{\textit{M}}[\textbf{\textit{v}}_i] = \max \begin{pmatrix} \sum_{\textbf{\textit{v}}_i \text{ child of } \textbf{\textit{v}}_i} \textbf{\textit{M}}[\textbf{\textit{v}}_j], \\ & \textbf{\textit{w}}(\textbf{\textit{v}}_i) + \sum_{\textbf{\textit{v}}_j \text{ grandchild of } \textbf{\textit{v}}_i} \textbf{\textit{M}}[\textbf{\textit{v}}_j] \end{pmatrix} \\ & \textbf{return } \textbf{\textit{M}}[\textbf{\textit{v}}_n] \text{ (* Note: } \textbf{\textit{v}}_n \text{ is the root of } \textbf{\textit{T}} \text{ *)} \end{split}$$

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THE END

(for now)

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