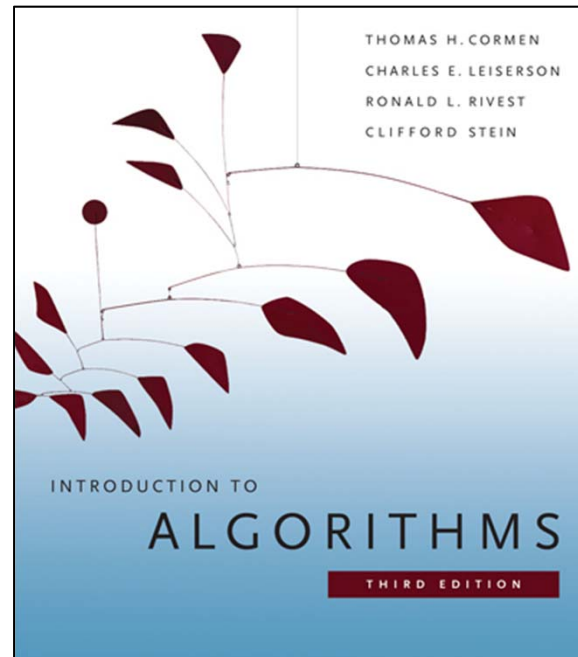


6.006

Introduction to Algorithms



Lecture 21: Dynamic Programming IV

Prof. Erik Demaine

Today

- Piano fingering
- Platform video games
- Structural dynamic programming
- Vertex cover
- Widget layout

Recall: What is **Dynamic Programming?**

- “Controlled” brute force / exhaustive search
- Key ideas:
 - **Subproblems:** like original problem, but smaller
 - Write solution to one subproblem in terms of solutions to smaller subproblems – *acyclic*
 - **Memoization:** remember the solution to subproblems we’ve already solved, and re-use
 - Avoid exponentials
 - **Guessing:** if you don’t know something, guess it!
(try all possibilities)

Recall:

How to Dynamic Program

Five easy steps!

1. Define **subproblems**
2. **Guess** something (part of solution)
3. Relate subproblem solutions (**recurrence**)
4. Recurse and **memoize** (top down)
or Build DP table bottom up
5. **Solve** original problem via subproblems
(usually easy)

Recall: How to Analyze Dynamic Programs

Five easy steps!

1. Define subproblems *count # subproblems*
2. Guess something *count # choices*
3. Relate subproblem solutions
analyze time per subproblem
4. ***DP running time*** = # subproblems
 · time per subproblem
5. Sometimes *additional running time*
to solve original problem

Two Kinds of Guessing

1. Within a subproblem

- Crazy Eights: previous card in trick
- Sequence alignment: align/drop one character
- Bellman-Ford: previous edge in path
- Floyd-Warshall: use vertex k ?
- Parenthesization: last multiplication
- Knapsack: include item i ?
- Tetris training: how to place piece i

2. Using additional subproblems

- Knapsack: how much space left in knapsack
- Tetris training: current board configuration

Piano Fingering



photo by Brian Richardson (seriousbri), 2009

<http://www.flickr.com/photos/seriousbri/4148739768/>

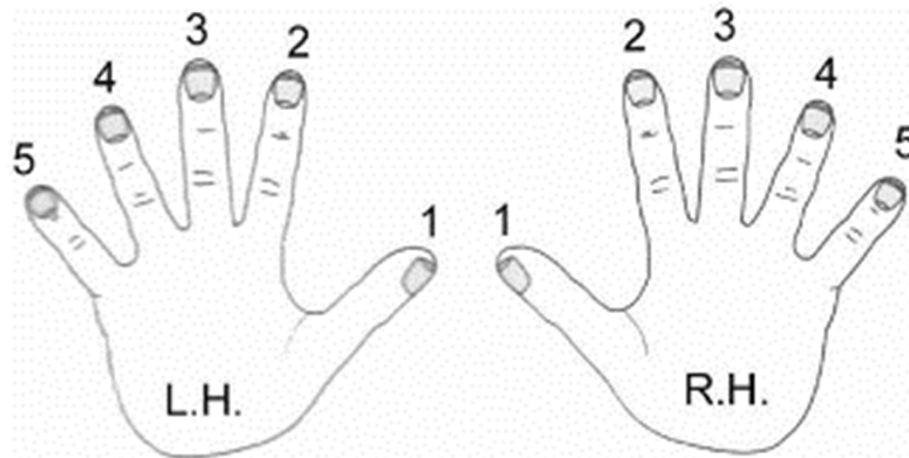
Piano Fingering

Poco Moto Ludwig van Beethoven

pp

Numbering for Right-Hand

Numbering for Left-Hand



Piano Fingering

[Parncutt, Sloboda, Clarke, Raekallio, Desain 1997;
Hart, Bosch, Tsai 2000; Al Kasimi, Nichols, Raphael 2007]

- Given musical piece to play
 - Say, sequence of single notes with right hand
 - (Can extend to both hands, multiple notes, etc.)
- Given metric $d(f, p, g, q)$ of **difficulty** going from finger f on note p to finger g on note q
 - **Crossing:** High if $1 < f < g$ and $p > q$
 - **Stretch:** High if $p \ll q$
 - **Legato:** ∞ if $f = g$
 - **Weak finger:** High if $g \in \{4, 5\}$
 - **3 \leftrightarrow 4:** High if $\{f, g\} = \{3, 4\}$
 - ...

References:

<http://www.jstor.org/pss/40285730>
<http://www.jstor.org/pss/10.1525/mp.2001.18.4.505>
<http://www.cse.unsw.edu.au/~cs9024/05s2/ass/ass01/fingering.pdf>
http://ismir2007.ismir.net/posters/ISMI_R2007_p355_kasimi_poster.pdf

Piano Fingering DP

1. **Subproblems:** for $1 \leq i \leq n$:
minimum difficulty possible for note[i :]
 2. **Guess:** finger f for note[i]
 3. **Recurrence:** $P(i) = \min(P(i + 1) + d(\text{note}[i], f, \text{note}[i + 1], \dots ??? \dots))$
for f in fingers)
- *How to know fingering for next note $i + 1$?*
 - *Guess!*

Piano Fingering DP

1. **Subproblems:** for $1 \leq i \leq n$ & finger f :
minimum difficulty possible for note[i :]
starting on finger f } $n \cdot F$
2. **Guess:** finger g for note[$i + 1$] } F choices
3. **Recurrence:** $P(i, f) = \min(P(i + 1, g) +$
 $d(\text{note}[i], f, \text{note}[i + 1], g)$
for g in fingers) } $O(F)$
4. **DP time** = # subproblems · time/subproblem
 $nF \cdot O(F) = O(nF^2)$
5. **Original problem** = $\min(P(1, f)$ for f in fingers)

MARIO
000200

1 × 01

WORLD
1-1

TIME
392

SUPER MARIO BROS.

©1985 NINTENDO



Platform Video Games

- Given entire level: objects, enemies, etc.
- Anything outside $w \times h$ screen is reset
- **Configuration** = screen state, score, velocity, ...
- Given transition function for each time step
$$\delta: (\text{config}, \text{action}) \mapsto \text{config}'$$
 - Movement, enemies, ...
- Goal: Maximize score subject to surviving and reaching goal



Platform Video Game DP

- Subproblems:** for configuration C :
best possible score starting from C $\left. \begin{array}{l} \text{\# configs.} \\ = O(1)^{w \cdot h} \\ \cdot n \cdot m \cdot S \cdot V \end{array} \right\}$
- Guess:** which action to take (if any) $\left. \right\} O(1)$
- Recurrence:**
 $P(C) = \max(P(\delta(C, A)))$ for A in actions $\left. \right\} O(1)$
 $P(\text{goal } C) = C.\text{score}; P(\text{dead } C) = -\infty$
- DP time** = $\underbrace{\text{\# subproblems}}_{O(1)^{w \cdot h} \cdot n \cdot m \cdot S \cdot V} \cdot \underbrace{\text{time/subproblem}}_{O(1)}$ — (pseudo) polynomial
- Original problem** = $P(\text{init})$ for $w \cdot h = O(\lg(nmSV))$

Cycles in Subproblems

- $C_1 \rightarrow \delta(C_1, A_1) = C_2 \rightarrow \delta(C_2, A_2) = C_3 \rightarrow \dots$
might lead to cycles
- In this problem, never helps to cycle
 - C captures entire state, including score
- So mark subproblem at start, and if cycle, ignore that subproblem
- OR: SMB timer in C , so actually no cycles

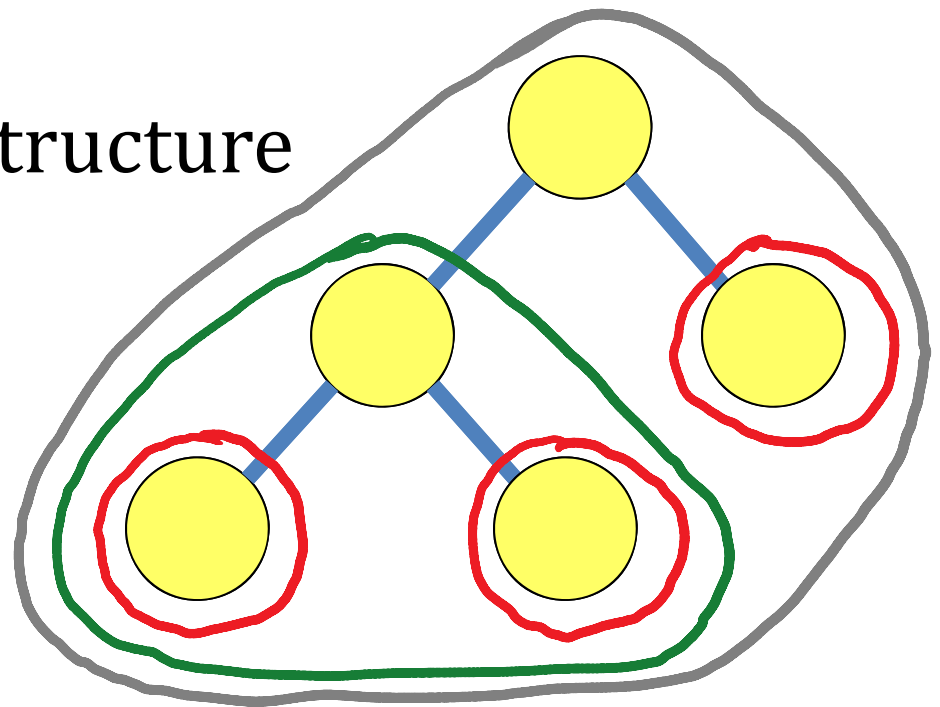
```
memo = {}  
def mario(C):  
    if C not in memo:  
        memo[C] = -∞  
        memo[C] = max(  
            mario(δ(C, A))  
            for A in actions)  
    return memo[C]
```

Structural Dynamic Programming

- Follow a combinatorial structure other than a sequence / a few sequences
 - Like structural vs. regular induction

- Main example: Tree structure

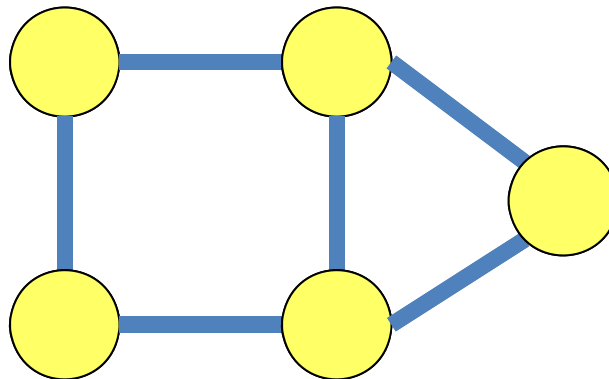
- Useful subproblems:
for every vertex v ,
subtree rooted at v



Vertex Cover

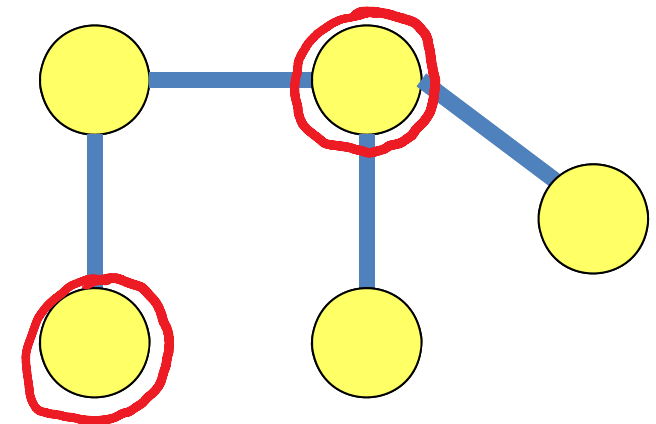
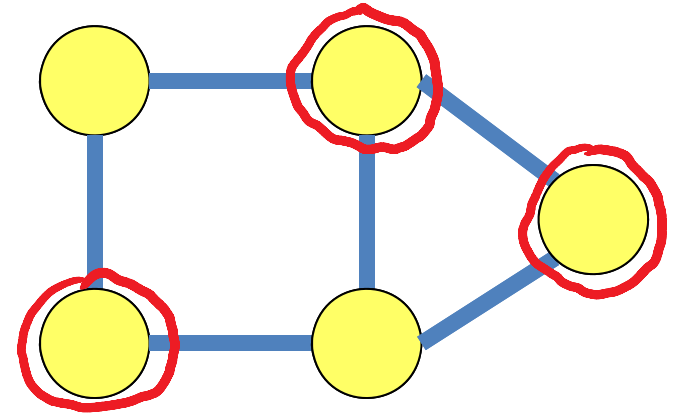
- Given an undirected graph $G = (V, E)$
- Find a minimum-cardinality set S of vertices containing at least one endpoint of every edge
 - Equivalently, find a minimum set of guards for a building of corridors, or (unaligned) streets in city

Example:



Vertex Cover Algorithms

- Extremely unlikely to have a polynomial-time algorithm, even for planar graphs (see Lecture 25)
- But polynomially solvable on trees, using dynamic programming



Vertex Cover in Tree DP

0. *Root the tree arbitrarily.*

1. **Subproblems:** for $v \in V$: size of smallest vertex cover in subtree rooted at v

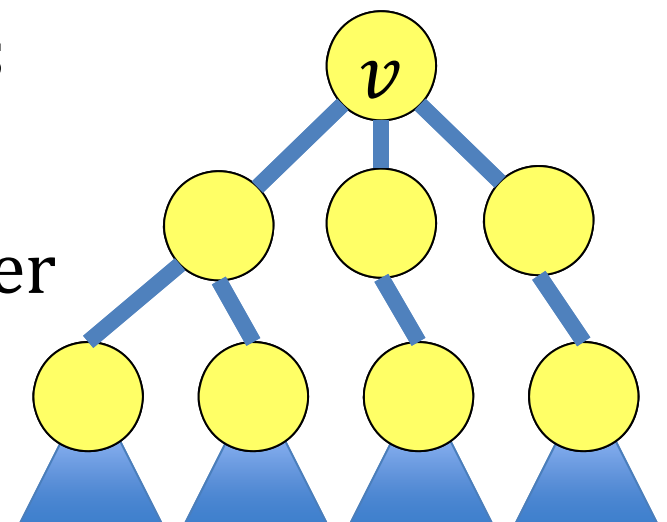
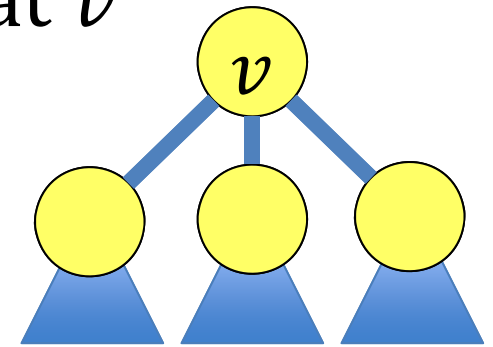
2. **Guess:** is v in the cover?

– YES:

- Cover children edges
- Left with children subtrees

– NO:

- All children must be in cover
- Left with grandchildren subtrees

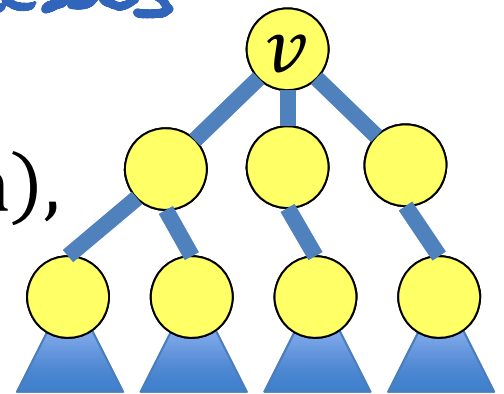


Vertex Cover in Tree DP

1. **Subproblems:** for $v \in V$: size of smallest vertex cover in subtree rooted at v $\} |V|$

2. **Guess:** is v in the cover? $\} 2 \text{ guesses}$

3. **Recurrence:** $V(v) = \min\{$
YES: $1 + \sum(V(c) \text{ for } c \text{ in } v.\text{children}),$
NO: $\text{len}(v.\text{children}) +$
 $\sum(V(g) \text{ for } c \text{ in } v.\text{children}$



for $g \text{ in } c.\text{children})\} - O(v)$

4. **DP time** = $\underbrace{\# \text{ subproblems}}_{|V|} \cdot \underbrace{\text{time/subproblem}}_{O(v) = O(V^2)}$

5. **Original problem** = $V(\text{root})$ *actually $O(v)$ because each vertex visited twice: parent & grandpar.*

Improved

Vertex Cover in Tree DP

3|V|

1. **Subproblems:** for $v \in V$ & $y \in \{\text{YES}, \text{NO}, \text{MAYBE}\}$:
size of smallest vertex cover S in subtree rooted at v
such that $[v \in S?] = y$ (unconstrained if $y = \text{MAYBE}$)

2. **Guess:** Does MAYBE = YES or NO? $\{ \leq 2 \text{ choices}$

3. **Recurrence:**

$$V(v, \text{MAYBE}) = \min\{V(v, \text{YES}), V(v, \text{NO})\} \quad - O(1)$$

$$V(v, \text{YES}) = 1 + \text{sum}(V(c, \text{MAYBE}) \text{ for } c \text{ in } v. \text{ children})$$

$$V(v, \text{NO}) = \text{sum}(V(c, \text{YES}) \text{ for } c \text{ in } v. \text{ children})$$

$\downarrow \downarrow$
 $\# \text{ children}(v)$

4. **DP time** = $\underbrace{\sum_{v \in V} 3}_{\text{# subproblems}} \cdot \underbrace{\# \text{ children}(v)}_{\text{time/subproblem}} = O(V)$

5. **Original problem** = $V(\text{root}, \text{MAYBE})$

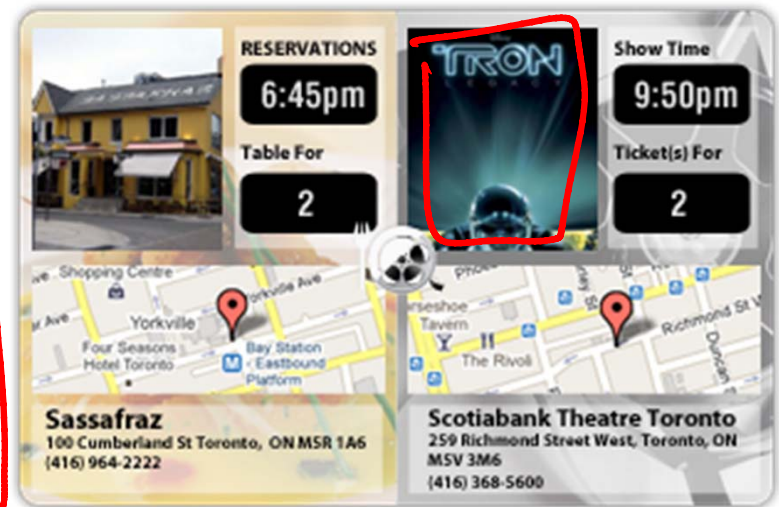
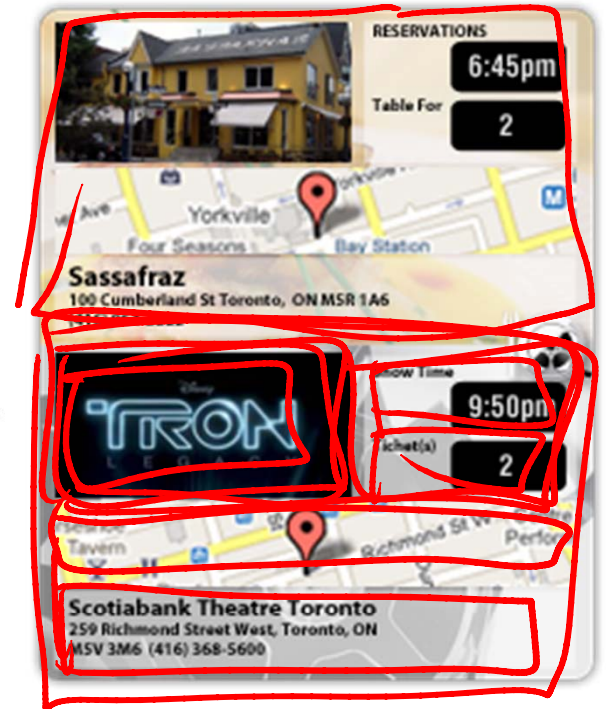
Widget Layout

- Given a hierarchy of *widgets*
- **Leaf** widget = button, image, ...
 - List of possible rectangular sizes
- **Internal** widget = rectangular container

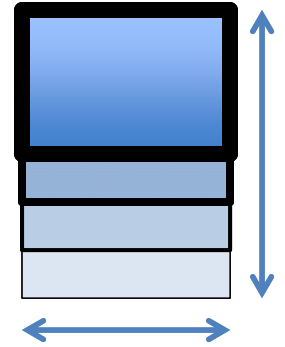


– Can join children horizontally or vertically

- Goal: Fit into a given rectangular screen

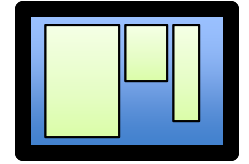


Widget Layout DP



- Subproblems:** for $v \in V$ & $0 \leq w \leq W$:
minimum h such that widget v fits into $w \times h$
- Guess:** Leaf v : Which size to use?
Internal v : Horizontal or vertical?
- Recurrence:**
 $L(\text{leaf } v, w) =$
 $\min(h' \text{ for } (w', h') \text{ in } v. \text{ sizes if } w' \leq w)$
 $L(\text{internal } v, w) =$
 $\min \{ \text{sum}(L(c, w) \text{ for } c \text{ in } v. \text{ child}),$
 $H(v, w, 1) \}$

Horizontal Layout DP



1. **Subproblems:** for $v \in V$, $0 \leq w \leq W$,
 $1 \leq i \leq \text{len}(v.\text{children})$:
 minimum h such that horizontal layout of
 $v.\text{child}[i:]$ fits into $w \times h$ rectangle

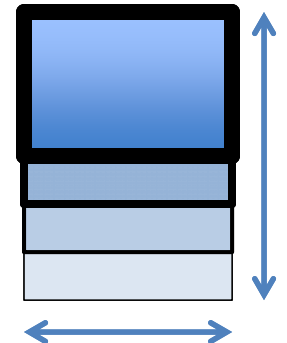
$$\left. \begin{array}{l} \sum_{v \in V} W \cdot \text{deg}(v) \\ W \cdot |E| \end{array} \right\}$$

2. **Guess:** Width $0 \leq w' \leq W$ of child i } W choices

3. **Recurrence:** $H(v, w, i) = \min(\max\{L(v.\text{child}[i], w'), H(v, w - w', i + 1)\})$
 for $1 \leq w' \leq w$ } $O(w)$

4. **DP time** = $\underbrace{\# \text{ subproblems}}_{W \cdot |E|} \cdot \underbrace{\text{time/subproblem}}_{O(w) = O(W^2 E)}$

Widget Layout DP



1. **Subproblems:** for $v \in V$ & $0 \leq w \leq W$: $\rightarrow |V| \cdot W$
 minimum h such that widget v fits into $w \times h$
2. **Guess:** Leaf v : Which size to use? $\text{deg}(v)$
 Internal v : Horizontal or vertical? α
3. **Recurrence:**
 $L(\text{leaf } v, w) = \dots \text{ for } \dots \text{ in } v. \text{ sizes } \dots$
 $L(\text{internal } v, w) = \dots \text{ for } \dots \text{ in } v. \text{ child } \dots$
} $O(\text{deg}(v))$
4. **DP time** = $\underbrace{\sum_{v \in V} W}_{\text{# subproblems}} \cdot \text{time/subproblem}$
 $\cdot O(\text{deg}(v)) = O(WE)$
5. **Original problem** = $S(\text{root}, W) \leq H$

Widget Layout Summary

- Two “levels” of dynamic programming
 1. Optimal height for given width of subtree rooted at v
 2. Optimal layout (partitioning) of children into horizontal arrangement
- Really just one bigger dynamic program
- Pseudopolynomial running time:
$$O(W^2E + WE) = O(W^2E)$$

