

Algorithms

Dynamic
Programming
(part 4)
Trees 

Recap

- HW 3 posted
 - ↳ due Monday, Sept 28,
- don't forget groups!
- Reading as usual
- Next week: all about greed
- Afterwards, we switch to the graph chapters.

Ch. 5+6 are mainly recap:

module {
- data structures
- DFS/BFS

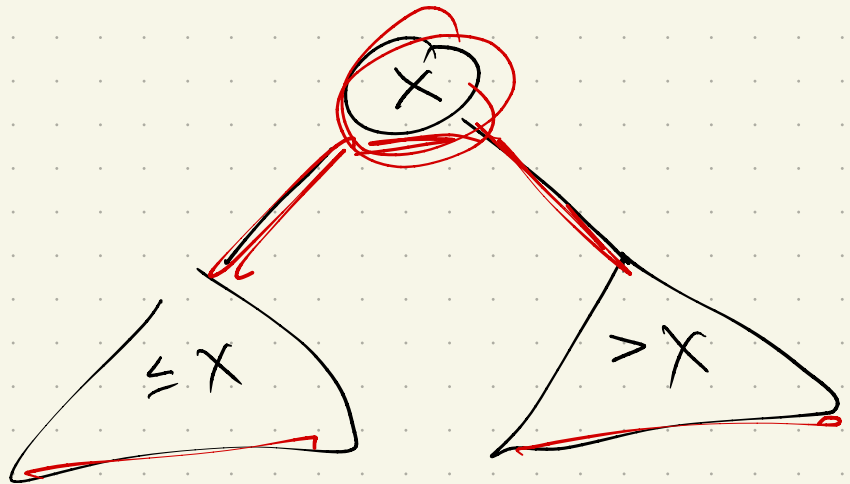
- New zoom link coming soon

Balanced search trees (again)

Recall: input: list of #s

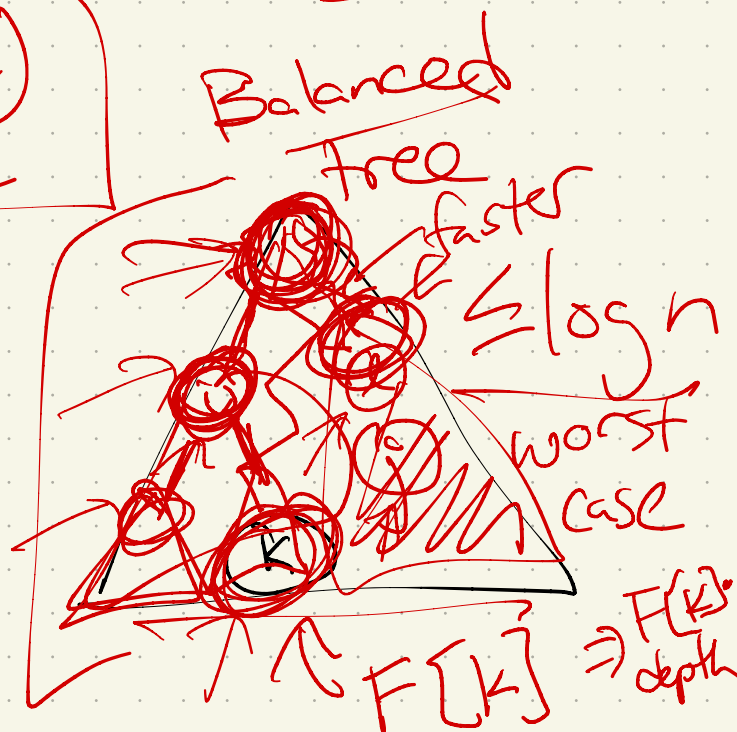
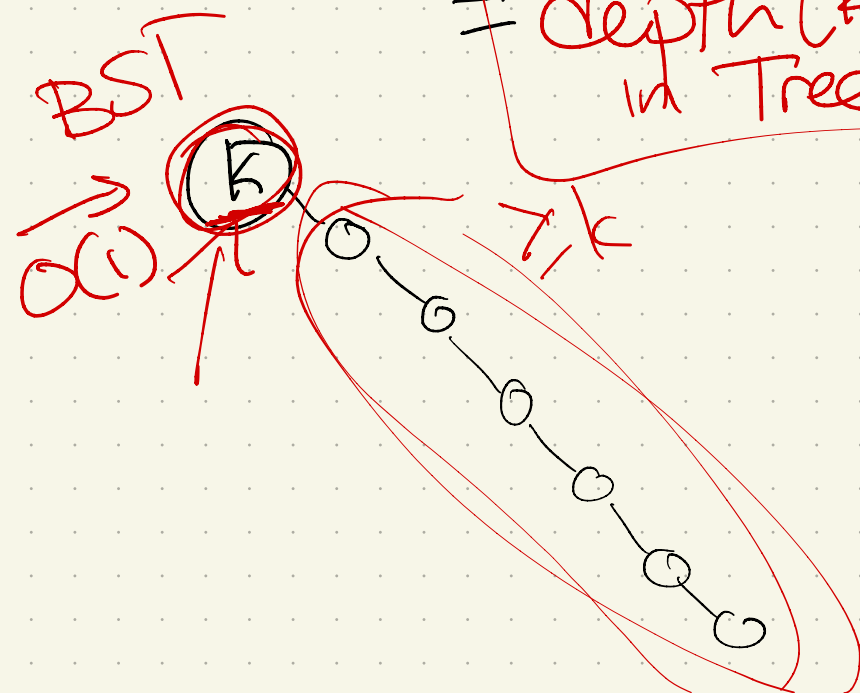
What is the "best" one?

Recap:



Time to search for k in T

$= \text{depth}(k)$
in Tree



Stepping back even more.

Suppose T holds $1 \dots n$ ~~∴ X~~
& searches are $x_1, \dots, x_m$

Some searches are "easier":

If $x_1 = x_2 = \dots = x_m = 1$,

then $T =$

IS optimal!

Why?

~~$F[n][1] \dots [n][n]$~~

So "best" can change
depending on the searches.

Balanced ~~$n/2$~~ $1 \dots n$

~~$\log_2 n$~~

~~①~~

Here: given $X[1..n] \xrightarrow{\#s}$
 (assume X is sorted) $F[1..n] \xrightarrow{\text{freq.}}$
 $n \dots 1$

element $X[i]$ will have

$F[i]$ searches.

Intuitively - want higher $F[i]$ to be closer to the root!

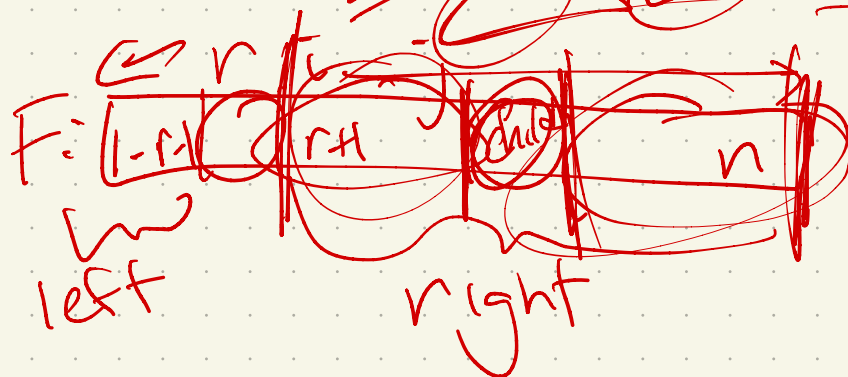
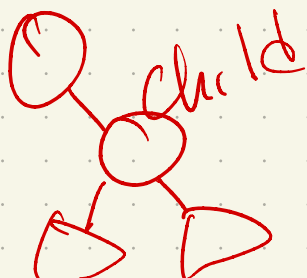
Last chapter:

$$\text{Cost}(T, f[1..n]) = \sum_{i=1}^n f[i] + \sum_{i=1}^{r-1} f[i] \cdot \# \text{ancestors of } v_i \text{ in left}(T) + \sum_{i=r+1}^n f[i] \cdot \# \text{ancestors of } v_i \text{ in right}(T)$$

pay at root

searches root $\leq r$ $\geq r$

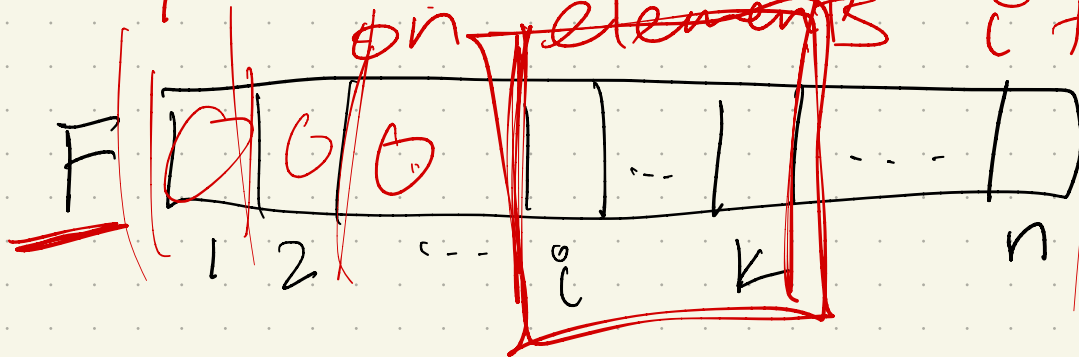
Pick r :



$$\text{OptCost}(i, k) = \begin{cases} 0 & \text{if } i > k \\ \sum_{j=i}^k f[j] + \min_{i \leq r \leq k} \left\{ \text{OptCost}(i, r-1) + \text{OptCost}(r+1, k) \right\} & \text{otherwise} \end{cases}$$

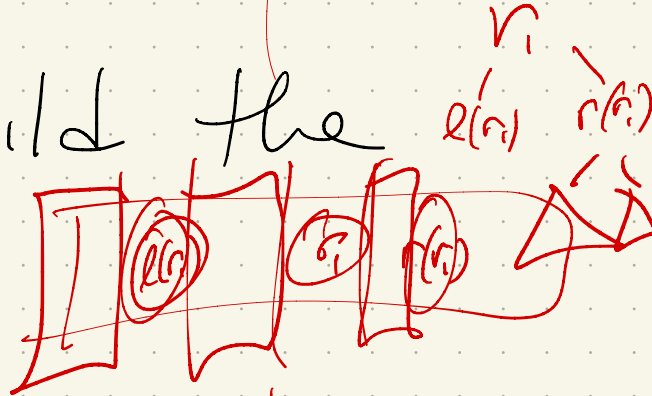
recursive

optimal cost binary tree built on elements i to k of array



(w/freq. F)

Use this to build the "best" tree:



Choose root.

input: subarray from i to j

Recursively find best left subtree, + best right subtree.

(Note: try all roots in backtracking!)

How to memoize?

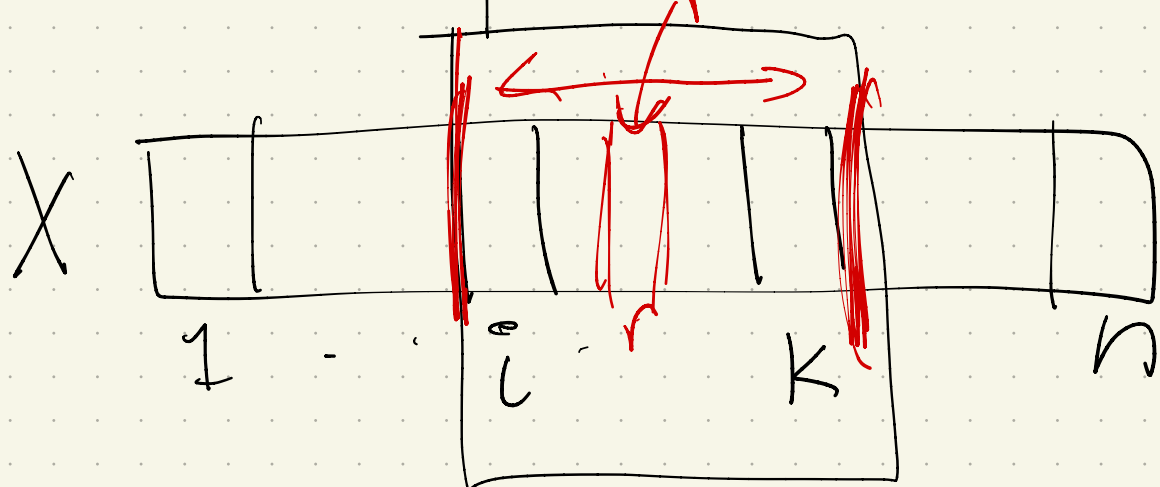
$1 \leq i \leq n$
 $1 \leq k \leq n$

not $O(1)$ of lookups

$$\text{OptCost}(i, k) = \begin{cases} 0 & \text{if } i > k \\ \sum_{j=i}^k f[j] + \min_{i \leq r \leq k} \left\{ \text{OptCost}(i, r-1) + \text{OptCost}(r+1, k) \right\} & \text{otherwise} \end{cases}$$

Store this somewhere

Remember input:



build best tree here

Everyone here pays $\sum_{j=i}^k f[j]$

so first precompute & store these sums.

Time/space: ?? space: $n \times n$ array will do

Let $F[i][k] = \sum_{j=i}^k f[j]$

Now:

$$\text{OptCost}(i, k) = \begin{cases} 0 & \text{if } i > k \\ \sum_{j=i}^k f[j] + \min_{i \leq r \leq k} \left\{ \text{OptCost}(i, r-1) + \text{OptCost}(r+1, k) \right\} & \text{otherwise} \end{cases}$$

↓

$$\text{OptCost}(i, k) = \begin{cases} 0 \\ \text{table lookup } O(1) \\ \text{precompute } O(n^2) \\ F[i][k] + \end{cases}$$

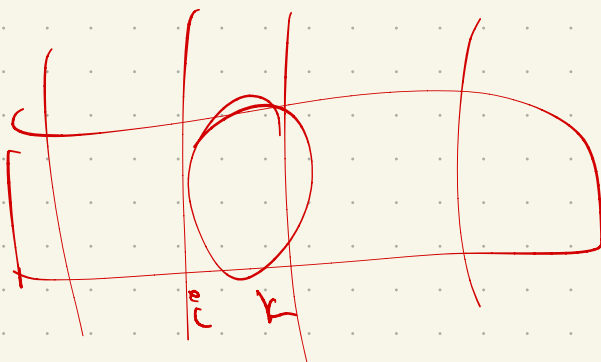
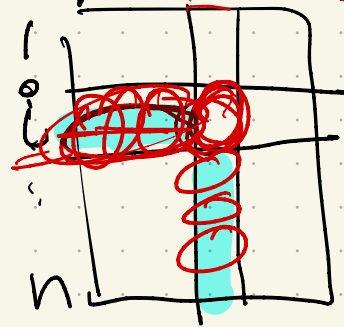
Memoize: $0 \leq i \leq k \leq n$ (letter) $O(1)$

So: 2d table!

Each $O[i][k]$ needs:

- $F[i][k]$: $O(1)$ lookup

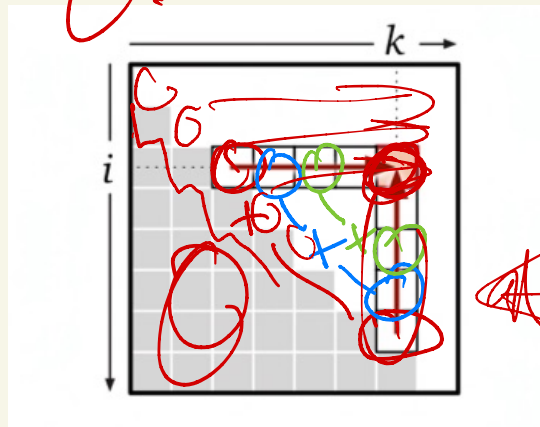
- and min of $O[i][r-1] + O[r+1, k]$ for $r \leq k$



the picture (prether):

$$\text{OptCost}(i, k) = \begin{cases} 0 & \text{if } i > k \\ F[i, k] + \min_{i \leq r \leq k} \left\{ \text{OptCost}(i, r-1) + \text{OptCost}(r+1, k) \right\} & \text{otherwise} \end{cases}$$

$F[i, k]$



I take the min

initializing F

So:

```

OPTIMALBST(f[1..n]):
  INITF(f[1..n])
  for i ← 1 to n+1
    OptCost[i, i-1] ← 0
  for d ← 0 to n-1
    for i ← 1 to n-d
      COMPUTEOPTCOST(i, i+d)
  return OptCost[1, n]
  
```

below diagonal = 0

⟨...or whatever⟩

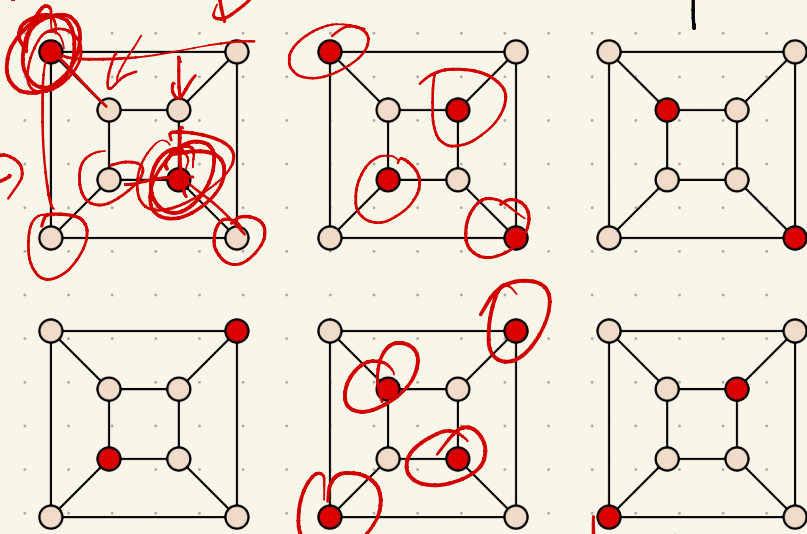
Time: $O(n)$ per table entry
 $\Rightarrow O(n^3)$

Space: $O(n^2)$

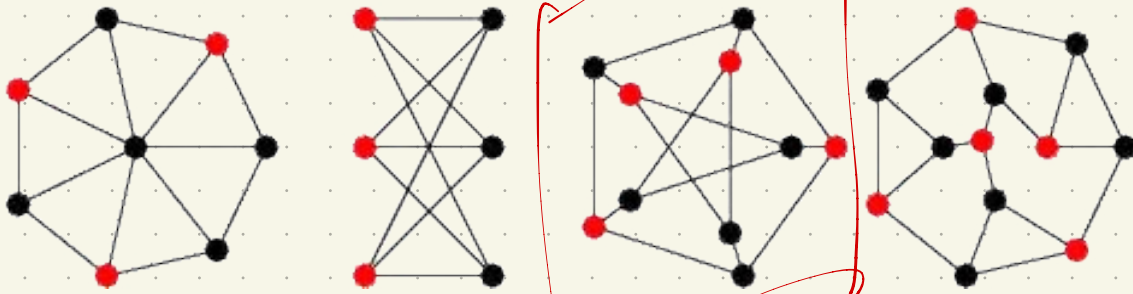
Dynamic Programming on Trees

Independent Set: ^{subset of vertices w/ 2 edges}
(nice preview of graphs)

maximal



maximal, & maximum



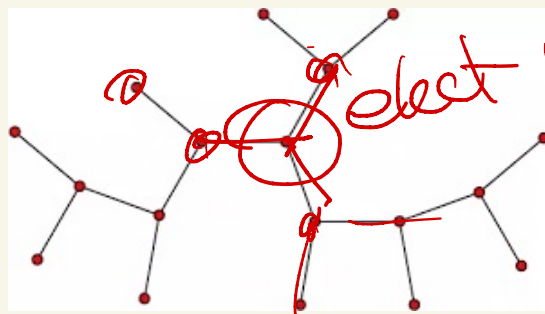
not unique

Notoriously hard!
But - can solve on simpler graphs.

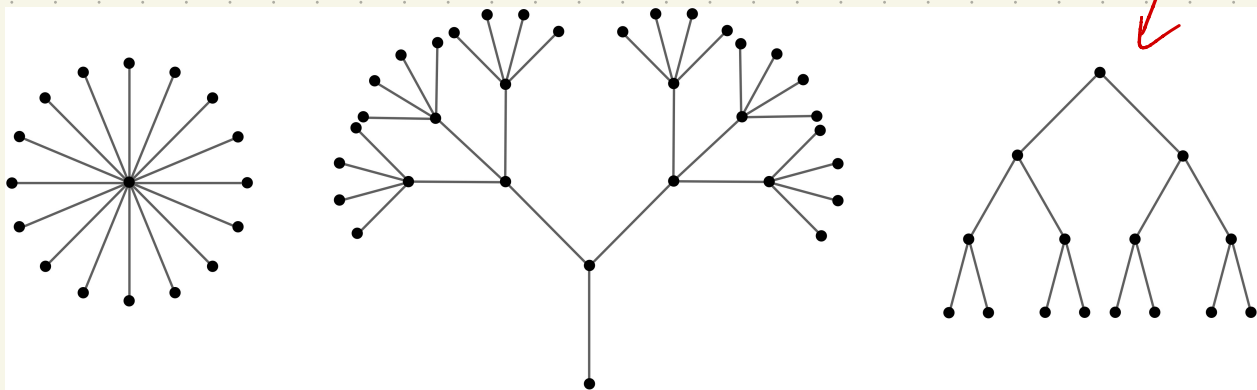
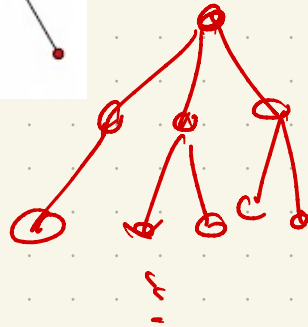
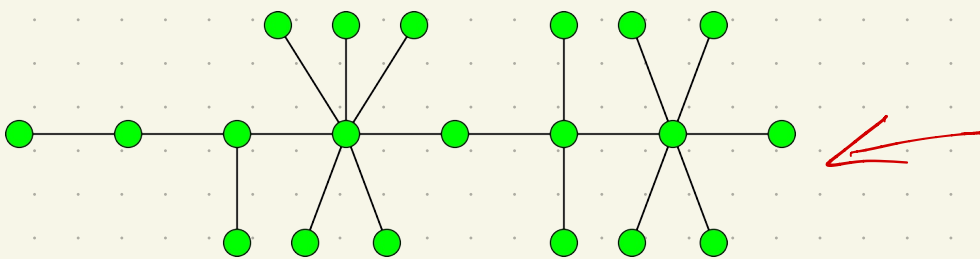
Trees: near + dear to CS majors

Not always binary!

not always
rooted



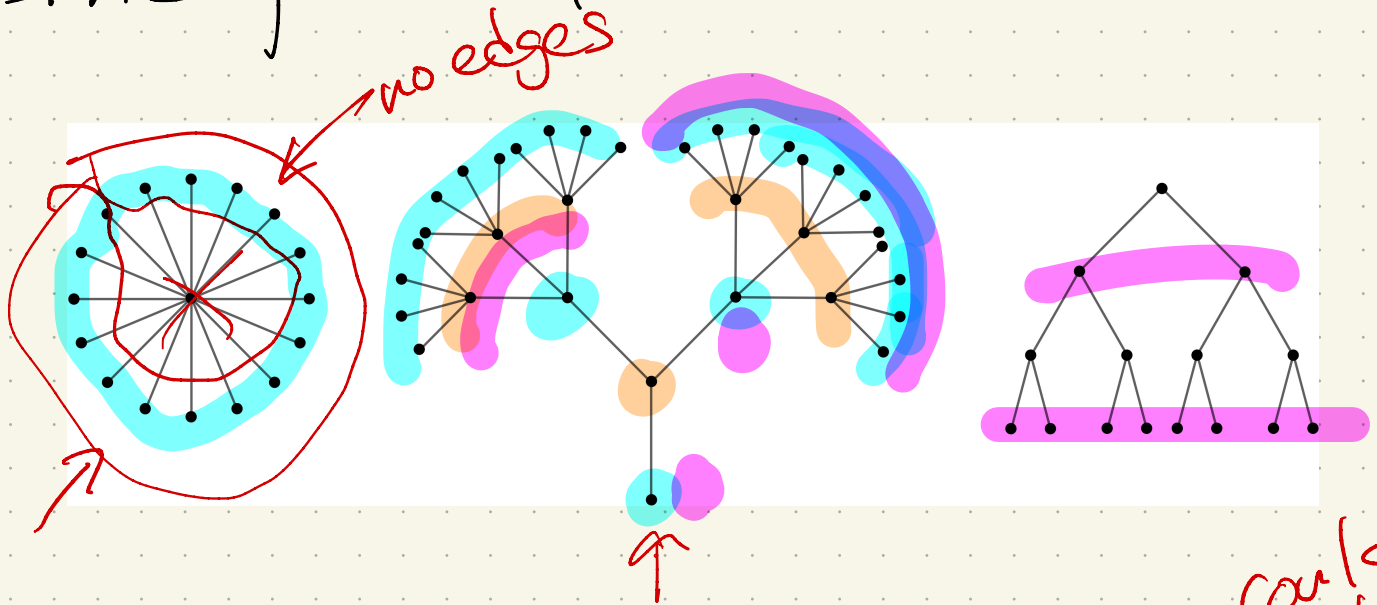
"long"
unrooted



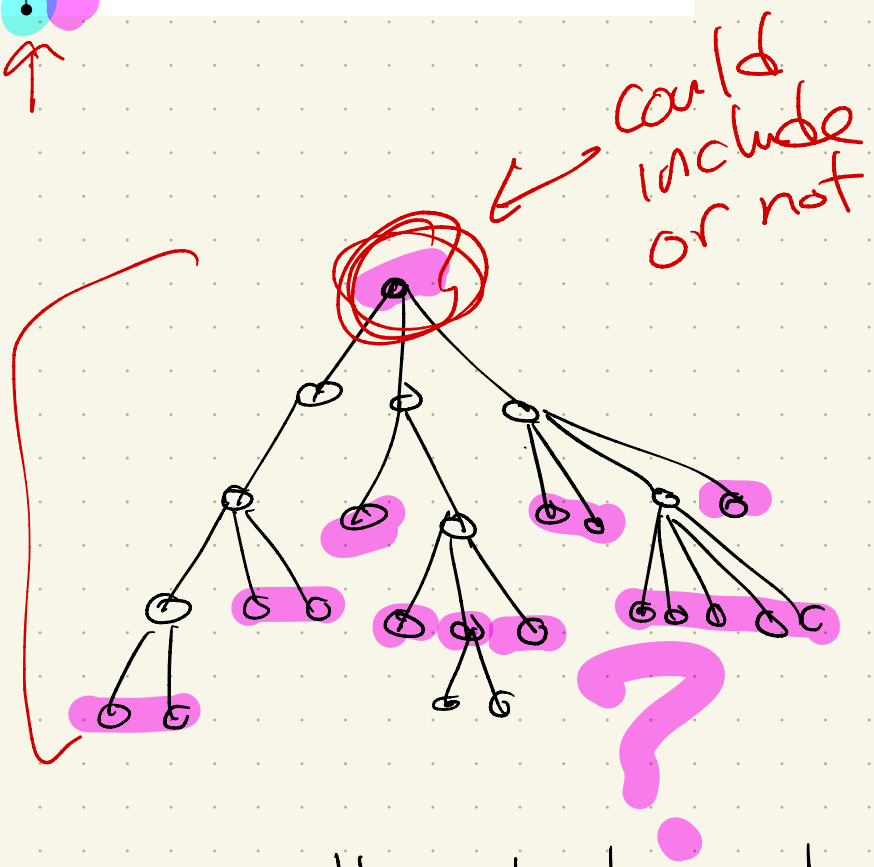
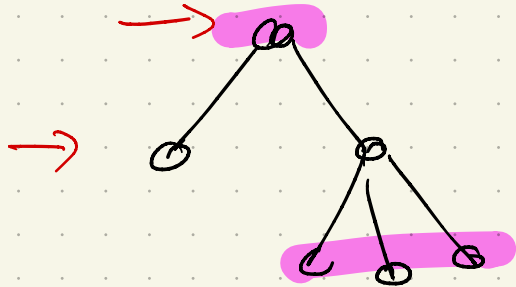
Dfn: graph w/ no cycles.

Here, we will "root" the tree.

Independent set in a tree:



Less clear:



So - not always "grab biggest level".

(ie - don't be greedy!!)

Recursive approach:

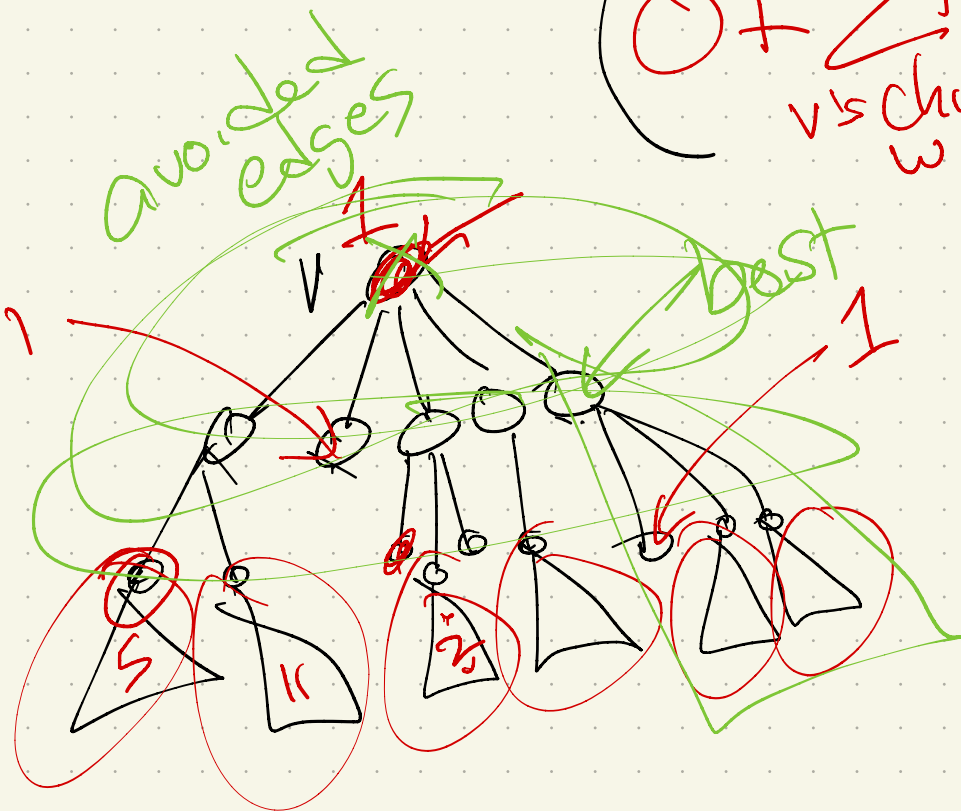
consider the root.

Could include, or not.

Back tracking!

$$\underline{MIS}(v) = \begin{cases} 1 + \sum_{\substack{w \\ \text{v's grandchildren}}} \underline{MIS}(w) & \text{include } v \\ 0 + \sum_{\substack{w \\ \text{v's children}}} \underline{MIS}(w) & \text{don't include } v \end{cases}$$

no children



This recurrence (in code):

TREEMIS(v):

skipv $\leftarrow$ 0

for each child w of v

skipv $\leftarrow$ skipv + TREEMIS(w)

keepv $\leftarrow$ 1

for each grandchild x of v

keepv $\leftarrow$ keepv + ~~TREEMIS(x)~~

v.MIS $\leftarrow$ max(keepv, skipv)

return v.MIS

recurse on children

~~TREEMIS(x)~~

with v

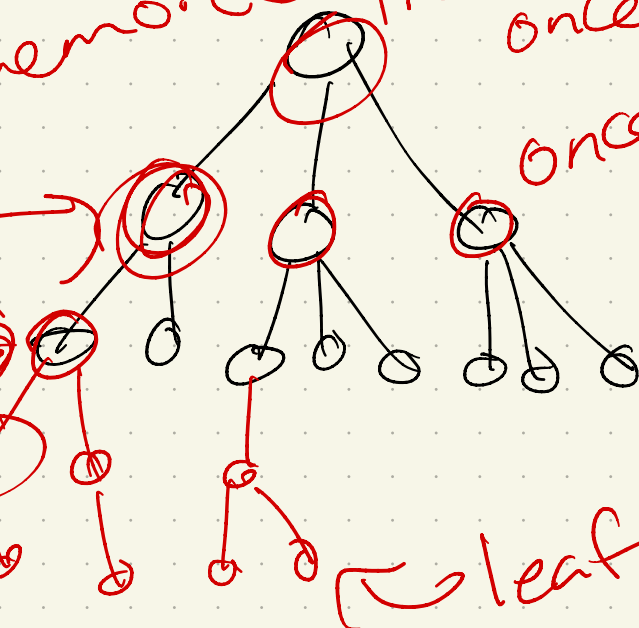
w/out v

Q: Given this recursion, are we calling any function too often?

Yes!

So memoize TREEMIS(v) once

once each
twice



once each

leaf: 1 or 0
w/out

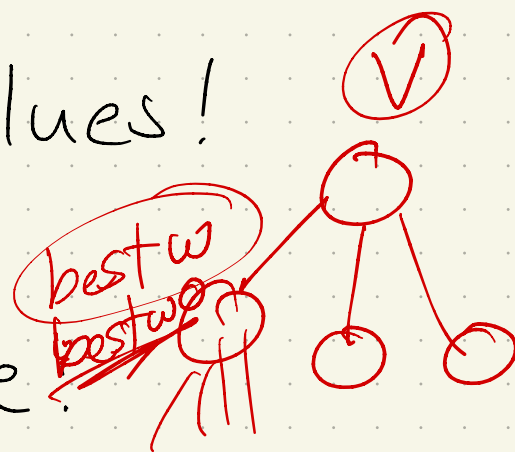
How to memoize:

Well, for each node, need the best ind set in that subtree.

Even better — 2 values!

(same big-O)

For each V , store:



— Best set with V

— Best set without V

→ post order traversal

Think data structures:

extend

$V.MIS$

Node $v = \left\{ \begin{array}{l} \text{data} \\ \text{left, right, parent} \end{array} \right.$

weight / balance factor, etc
add 2 fields with #s without

So: Use a tree for the data structure!

TREEMIS2(v):

$v.MISno \leftarrow 0$

$v.MISyes \leftarrow 1$

for each child w of v

$v.MISno \leftarrow v.MISno + \text{TREEMIS2}(w)$

$v.MISyes \leftarrow v.MISyes + w.MISno$

return $\max\{v.MISyes, v.MISno\}$

update myself

All on my answer

fields in DS node

Note: At heart, still a post-order traversal.

$\rightarrow O(n)$ on any binary tree
Space: added 2 fields per node
 $\rightarrow O(n)$