## Dynamic Programming

- · related to DE: We solve a problem by breaking down into simpler problems then combining the simpler problems.
- DP helps when we have independent overlap between the subproblems/ Simpler problems

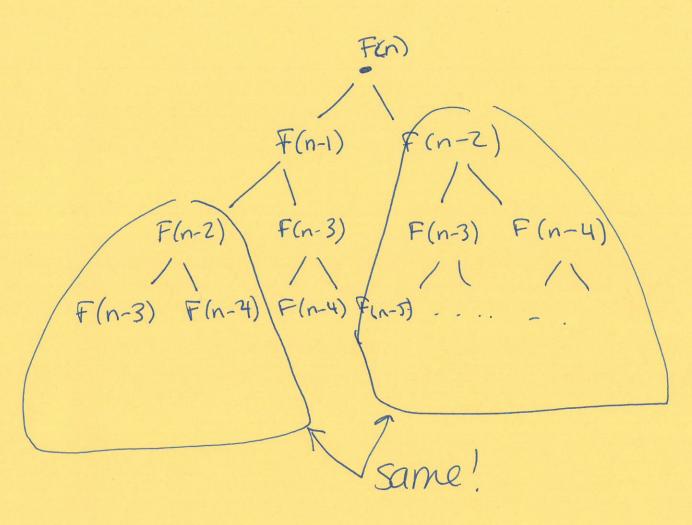
DPStock Exchange (p), n) soln = array of size n minidx  $\leftarrow 1$  } the solin for p[1...1] sol[1]  $\leftarrow 0$ for i=2 ... n if (soln [i-1] < p[i] - p[minidx]) | Sol'n [i] = p[i] - p[minidx] } else | Soln [i] \ Soln [i-1] (a) for endelse the for lap. if (p[i] <p[minidx]) minidx L endif end for return soln [n]

Runtime: O(n) linear!

lecall the Stock market problem:
p: Till price on each day
goal: find i < j to max up(j)-up(i)
Solve (L)  Solve (R)
max (solve (L), solve@(R), max(R)-min(L)
time complexity: $T(n) = 2 T(\frac{1}{2}) + \Theta(n)$ $= \Theta(n \log n)$
Another Sol'n $p:   5  2 8   4   5$ best (i) holds  best (3) = $\begin{cases} solution = 8-2=6 \\ pair (2,3) \\ minidx = 2 \end{cases}$ best [4]. = max (best [3]. sol'n, up [4] - p[minidx])  Another ex: $p = [2, 8, 1, 4, ]$ this update is now constant time.

Partial Sol'n our desired sol'n "bottom-up" approach: Start n/ the you solve the small + grow until big problem. Prob of sizen Schomatically Think of recordion top-down here! tree Soften up: Staft here! DP helps when subproblems overlap (are me same)

$$F(n) = F(n-1) + F(n-2) + \forall n \geq 2$$



$$F(5) = F(4) + F(3)$$

$$F(4) = F(3) + F(2)$$

$$F(3) = F(2) + F(1)$$

$$F(2) = F(1) + F(0)$$

Rod Cutting
· given: a rod of length n, profit array
owant: cut into pieces (ob unit multiples)
to maximize profit
• profit[i] = profit
of having a
e.q., n=3  • profit is length n
postit = [6,2,1)
© 3 (4) r
total $17$ $6+2=8$ $6+6+6$ $7=18$
The optimal sol'n
(note: sometimes opt is not unique)
opc 13
and DP solin to
in groups: Ownat in a DP solin to
$\mathcal{N}(\mathcal{M})$
(2) What in the O(2) lord
② What in the ⊙(2") brute force approach?
(Answers next time)