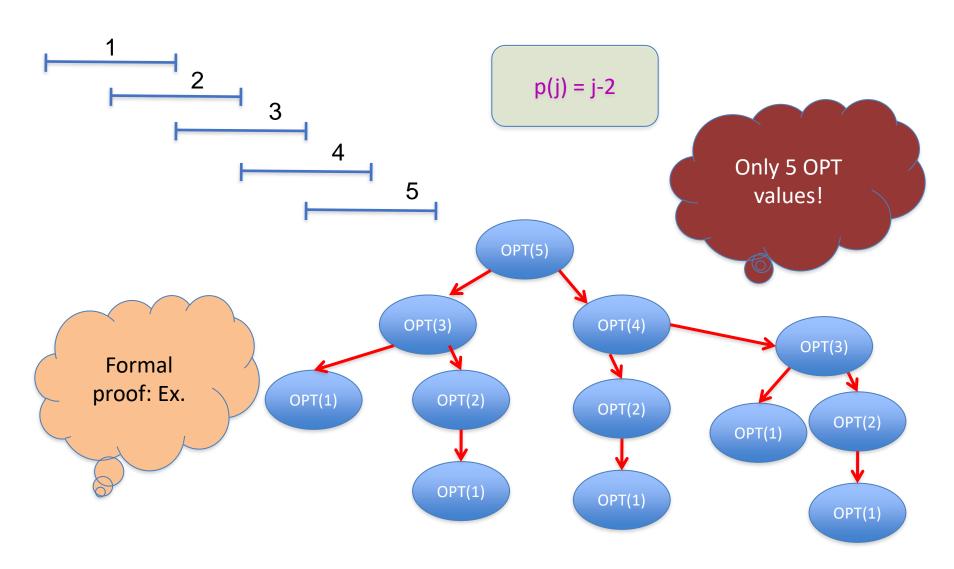
# Lecture 31

**CSE 331** 

#### A recursive algorithm

```
Proof of
                                                    correctness by
                        Correct for j=0
Compute-Opt(j)
                                                     induction on j
If j = 0 then return 0
return max { v<sub>i</sub> + Compute-Opt(p(j)), Compute-Opt(j-1) }
           = OPT(p(j))
                                      = OPT(j-1)
   OPT(j) = max \{v_i + OPT(p(j)), OPT(j-1)\}
```

### **Exponential Running Time**



## Using Memory to be smarter

Using more space can reduce runtime!

## How many distinct OPT values?

#### A recursive algorithm

Run time = O(# recursive calls)

#### Bounding # recursions

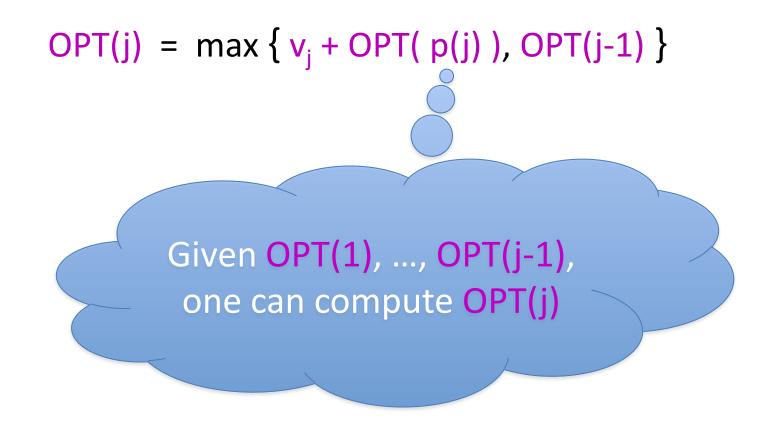
M-Compute-Opt(j)



Whenever a recursive call is made an walue is assigned

At most n values of M can be assigned

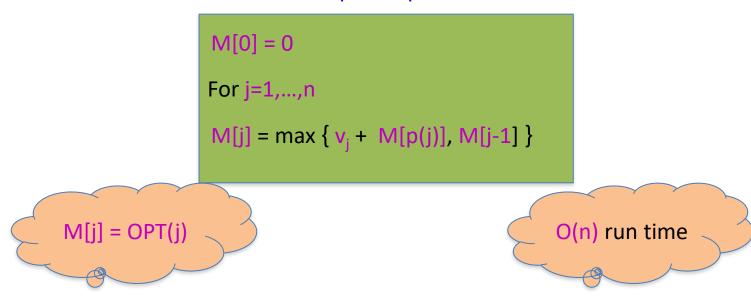
#### Property of OPT



#### Recursion+ memory = Iteration

Iteratively compute the OPT(j) values

#### Iterative-Compute-Opt



Algo run on the board...

## Reading Assignment

Sec 6.1, 6.2 of [KT]

#### When to use Dynamic Programming

There are polynomially many sub-problems

Richard Bellman

Optimal solution can be computed from solutions to sub-problems

OPT(j) = max 
$$\{v_j + OPT(p(j)), OPT(j-1)\}$$

There is an ordering among sub-problem that allows for iterative solution

#### Scheduling to min idle cycles

n jobs, ith job takes w<sub>i</sub> cycles

You have W cycles on the cloud



What is the maximum number of cycles you can schedule?

#### Subset sum problem

Input: n integers  $w_1, w_2, ..., w_n$ 

bound W

Output: subset S of [n] such that

(1) sum of w<sub>i</sub> for all i in S is at most W

(2) w(S) is maximized

## Rest of today's agenda

Dynamic Program for Subset Sum problem

Algo on the board...