# **Dynamic Programming**

- (Weighted) Independent Set in Graphs
- Weighted Independent Sets in Path
- Dynamic Programming Formulation
- Implementation and Backtracking

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### Max weight independent set in path graph

$$\text{OPT-VAL}(k) = \max \begin{cases} w_1 & \text{if } k = 1 \\ \max\{w_1, w_2\} & \text{if } k = 2 \\ \text{OPT-VAL}(k-2) + w_k & \text{if } v_k \in \text{OPT-SET}(k) \\ \text{OPT-VAL}(k-1) & \text{if } v_k \notin \text{OPT-SET}(k) \end{cases}$$

#### **Algorithm** Recursive OPT-VAL(n)

ightharpoonup Only computes OPT-VAL(n), will extend it to get OPT-SET(n) Exponential runtime due to unnecessary repeated calls

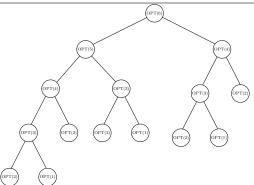
#### Recall the dynamic programming paradigm

- Express optimal solution in terms of optimal solution to smaller subproblems
- Identify repetition in the above recursion
- Use memoization or bottom-up computation

### Max weight independent set in path graph

### **Algorithm** Recursive OPT-VAL(n)

```
\label{eq:function} \begin{array}{l} \text{function } \textsc{opt-val}(k) \\ \textsc{if } k = 1 \text{ then} \\ \textsc{return } w_1 \\ \textsc{else } \textsc{if } k = 2 \text{ then} \\ \textsc{return } \max\{w_1, w_2\} \\ \textsc{else} \\ \textsc{return } \max\{\textsc{opt-val}(k-1), \textsc{opt-val}(k-2) + w_k\} \end{array}
```



Store OPT-VAL(k) for small k in memo M and use without recomputing

### **Algorithm** Recursive OPT-VAL(n) with memoization

```
function OPT-VAL(k)

if M[k] is empty then

if k=1 then

M[k] \leftarrow w_k

else if k=2 then

M[k] \leftarrow \max\{w_1, w_2\}

else

M[k] \leftarrow \max\{w_k + \text{OPT-VAL}(k-2), \text{OPT-VAL}(k-1)\}

return M[k]
```

- In one call to OPT-VAL(·) one memo entry  $M[\cdot]$  gets filled
- A memo entry is filled only once; total calls to OPT-VAL(·) is n
- Number of ops in a call to OPT-VAL(·) is O(1) + some recursive calls

#### Runtime of OPT-VAL(n) is O(n)

- Avoid overhead of recursive calls
- Write the code bottom up

#### ▶ Unwind the recursion

- Solve smaller problems first and then bigger
- Until we solve the original problem

### **Algorithm** Bottom-Up Computation of OPT-VAL(n)

```
M[1] \leftarrow w_1
M[2] \leftarrow \max\{w_1, w_2\}
for i = 3 to n do
M[i] \leftarrow \max\{M[i-2] + w[i], M[i-1]\}
return M[n]
```

- Both above algorithms give only value of the solution, OPT-VAL(n)
- How to find the solution itself, WIS? OPT-SET(n)
- Like M[k] = OPT-VAL(k), also maintain OPT-SET(k)
- Just add  $v_k$  or not depending on which branch yields larger value
- Wastes a lot of space

$$OPT-VAL(k) = \max \begin{cases} OPT-VAL(k-2) + w_k & \text{if } v_k \in OPT-SET(k) \\ OPT-VAL(k-1) & \text{if } v_k \notin OPT-SET(k) \end{cases}$$

- We can only remember the branch which gives higher value
- Backtrack when OPT-VAL(n) is computed
- Or Scan  $M[\cdot]$  again and find the branch

■ OPT-SET(k) either contains  $v_k$  or not ■ If OPT-VAL(k-2) +  $w_k \ge$  OPT-VAL(k-1), then  $v_k \in$  OPT-SET(k) ■ else  $v_k \notin$  OPT-SET(k)

#### **Algorithm** Max WIS in $P_n$

```
OPT-VAL(n)
                                                                       \triangleright M[i] = \text{OPT-VAL}(i)
function OPT-SET(n)
   S \leftarrow \emptyset
   i \leftarrow n
   while i > 1 do
      if M[i-2] + w_i \ge M[i-1] then
         S \leftarrow S \cup \{v_i\}
         i \leftarrow i - 1
      else
         i \leftarrow i - 1
   return S
```