

# Agenda for today

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## Quiz

Talk more about independent set + activity

Computing optimal choices (not just value)

A non-recursive formulation for dynamic programming

YOU should use  $PC(j)$  in  
 $OPT(j)$

- (a) (4 points) Suppose that the houses are located at (in yards)  $x_1 = 100$ ,  $x_2 = 350$ ,  $x_3 = 1000$ ,  $x_4 = 1100$ ,  $x_5 = 1200$ ,  $x_6 = 1300$ , and  $c_1$  through  $c_6$  are 5, 8, 2, 12, 3, 7. What is the maximum amount of candy that you can get? Which houses should you visit?

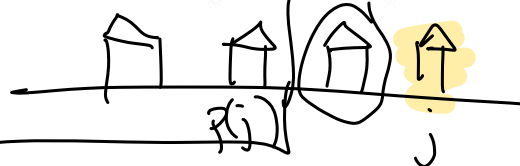
20 pieces = OPT(6)

visit 2 and 4

$$c_2 = 8 \quad c_4 = 12$$

OPT(j) = optimal # candies using a compatible subset of houses 1 to j

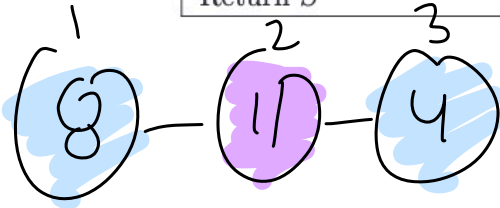
- (b) (5 points) Let  $\text{OPT}(j)$  denote the number of candy pieces that you can get using all houses up to house  $j$ . Also, for house located at  $x_j$ , let  $p(j)$  be the earlier house that is closest to house  $j$  but more than 300 yards away. Fill in part of the recursive definition of  $\text{OPT}(j)$  below.



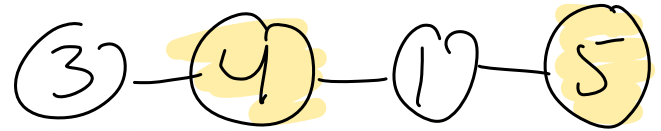
$$\text{OPT}(j) = \begin{cases} 0 & \text{if } j = 0 \\ \max(\text{OPT}(j-1), c_j + \text{OPT}(p(j))) & \text{if } j > 0. \end{cases}$$

- (a) (3 points) Give an example to show that the following algorithm *does not* always find an independent set of maximum total weight.

Start with  $S$  equal to the empty set  
 While some node remains in  $G$ :  
   Pick a node  $v_i$  of maximum weight and add it to  $S$   
   Delete  $v_i$  and its neighbors from  $G$   
 Return  $S$

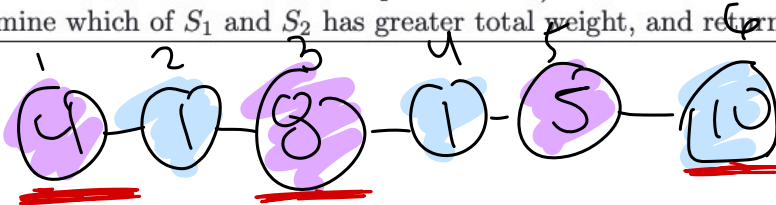


wt 11      wt 12



- (b) (3 points) Give an example to show that the following algorithm also *does not* always find an independent set of maximum total weight.

Let  $S_1$  be the set of all  $v_i$  where  $i$  is odd  
 Let  $S_2$  be the set of all  $v_i$  where  $i$  is even  
 (Note that  $S_1$  and  $S_2$  are both independent sets)  
 Determine which of  $S_1$  and  $S_2$  has greater total weight, and return this one

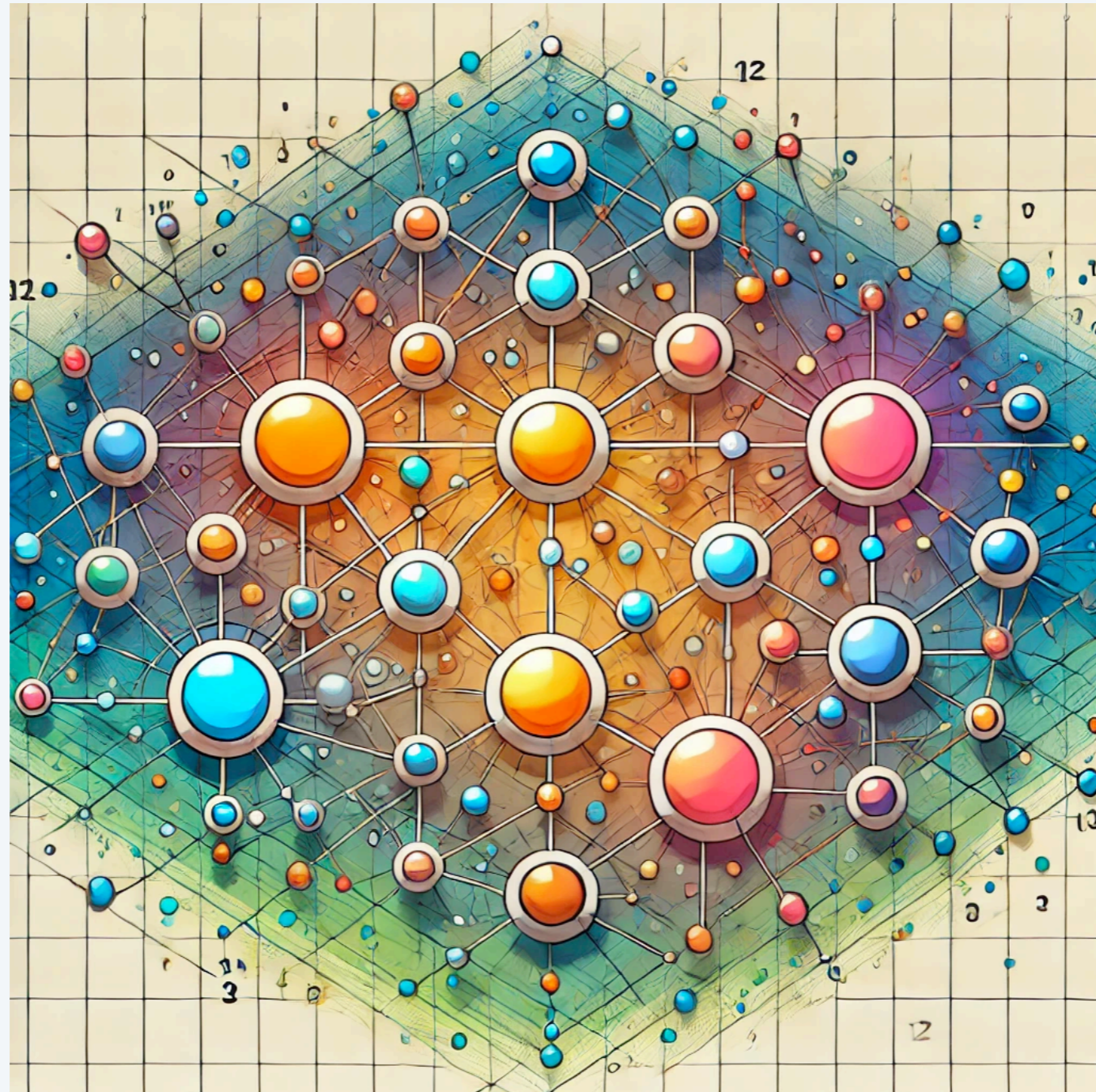


weight  $S_1$  17  
 weight  $S_2$  12

weight 22

# Activity

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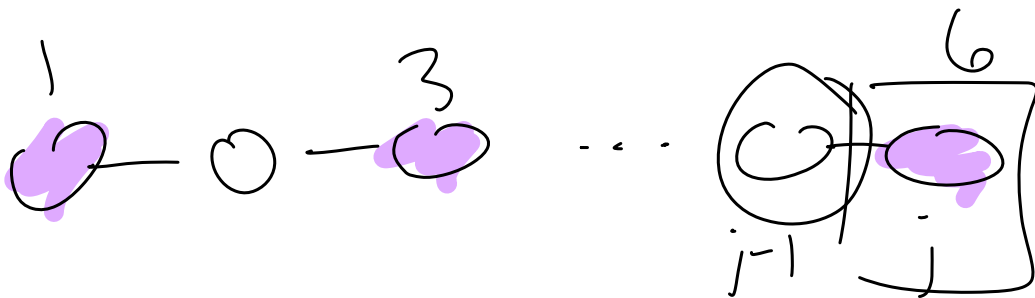


(a) Give a recurrence relation for the weight,  $\text{OPT}(j)$ , of the first  $j$  nodes in an  $n$ -node path as input to independent set. Notice that you don't need  $p(j)$  for this problem.

(b) baseline weight 1

(c) +2 for every node included

$$\text{OPT}(j) = \begin{cases} 1 & \text{if } j=0 \\ \max(\text{OPT}(j-1), w_j + 2 + \text{OPT}(j-2)) & \text{if } j > 0 \end{cases}$$



# Weighted interval scheduling: finding a solution

FIND-SOLUTION( $j$ )

IF ( $j = 0$ )

    RETURN  $\emptyset$ .

ELSE IF ( $w_j + M[p[j]] > M[j-1]$ )

    RETURN  $\{j\} \cup \text{FIND-SOLUTION}(p[j])$ .

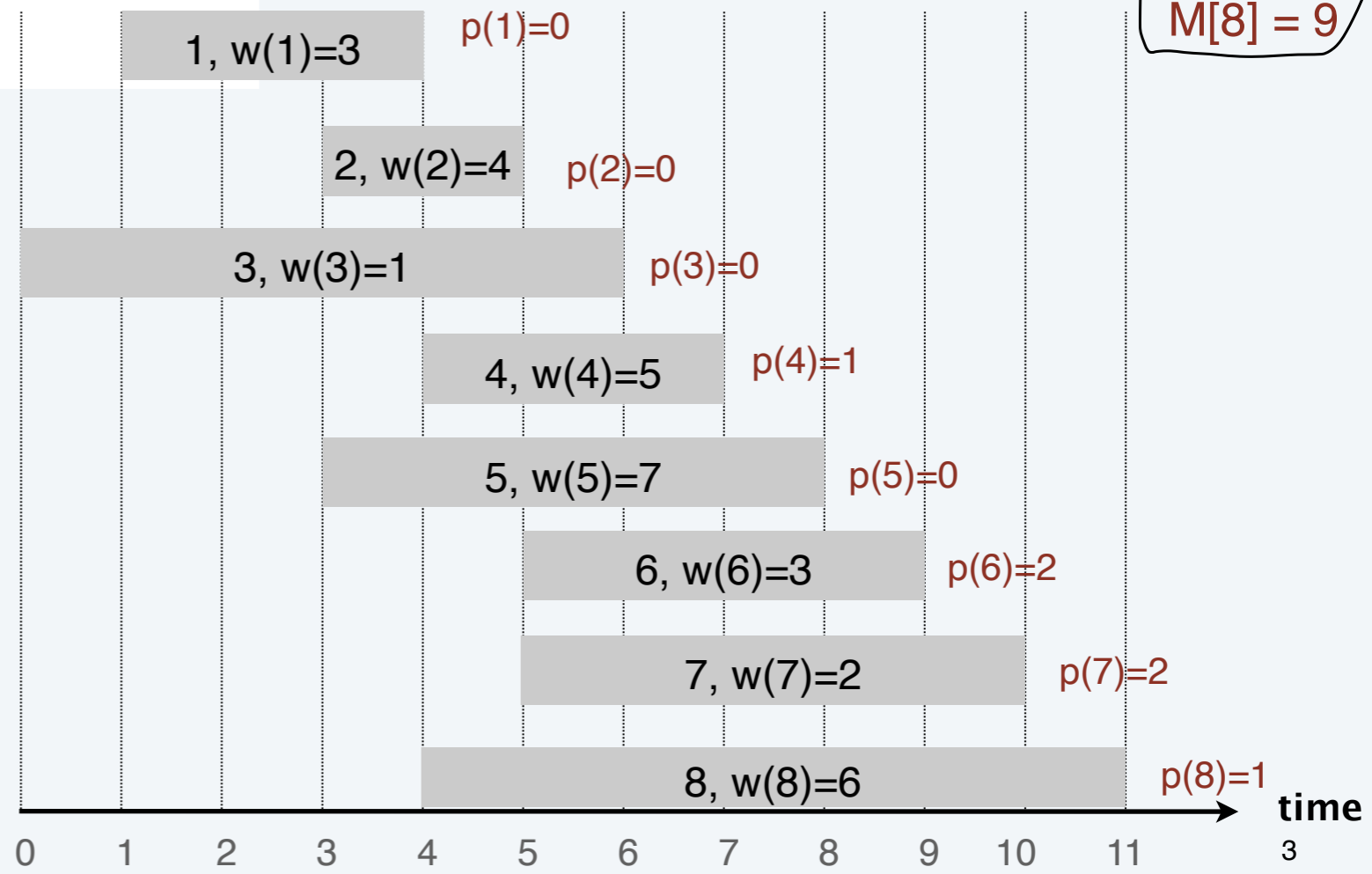
ELSE

    RETURN FIND-SOLUTION( $j-1$ ).

~~[9, 1]~~

$M[0] = 0$   
 $M[1] = 3$   
 $M[2] = 4$   
 $M[3] = 4$   
 $M[4] = 8$   
 $M[5] = 8$   
 $M[6] = 8$   
 $M[7] = 8$   
 $M[8] = 9$

find-solution(n)



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## Can there be more than one optimal set of intervals?

FIND-SOLUTION( $j$ )

---

IF ( $j = 0$ )

    RETURN  $\emptyset$ .

ELSE IF ( $w_j + M[p[j]] > M[j-1]$ )

    RETURN  $\{j\} \cup \text{FIND-SOLUTION}(p[j])$ .

ELSE

    RETURN FIND-SOLUTION( $j-1$ ).

1. Yes

2. No

---

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FIND-SOLUTION( $j$ )

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ELSE

    RETURN FIND-SOLUTION( $j-1$ ).

1. Yes

2. No

With table: which one does this algorithm find?



# Weighted interval scheduling: bottom-up dynamic programming

**BOTTOM-UP**( $n, s_1, \dots, s_n, f_1, \dots, f_n, w_1, \dots, w_n$ )

Sort jobs by finish time and renumber so that  $f_1 \leq f_2 \leq \dots \leq f_n$ .

Compute  $p[1], p[2], \dots, p[n]$ .

$M[0] \leftarrow 0$ .

**FOR**  $j = 1$  **TO**  $n$

$M[j] \leftarrow \max \{ M[j-1], w_j + M[p[j]] \}$ .

$M[0] = 0$

$M[1] =$

$M[2] =$

$M[3] =$

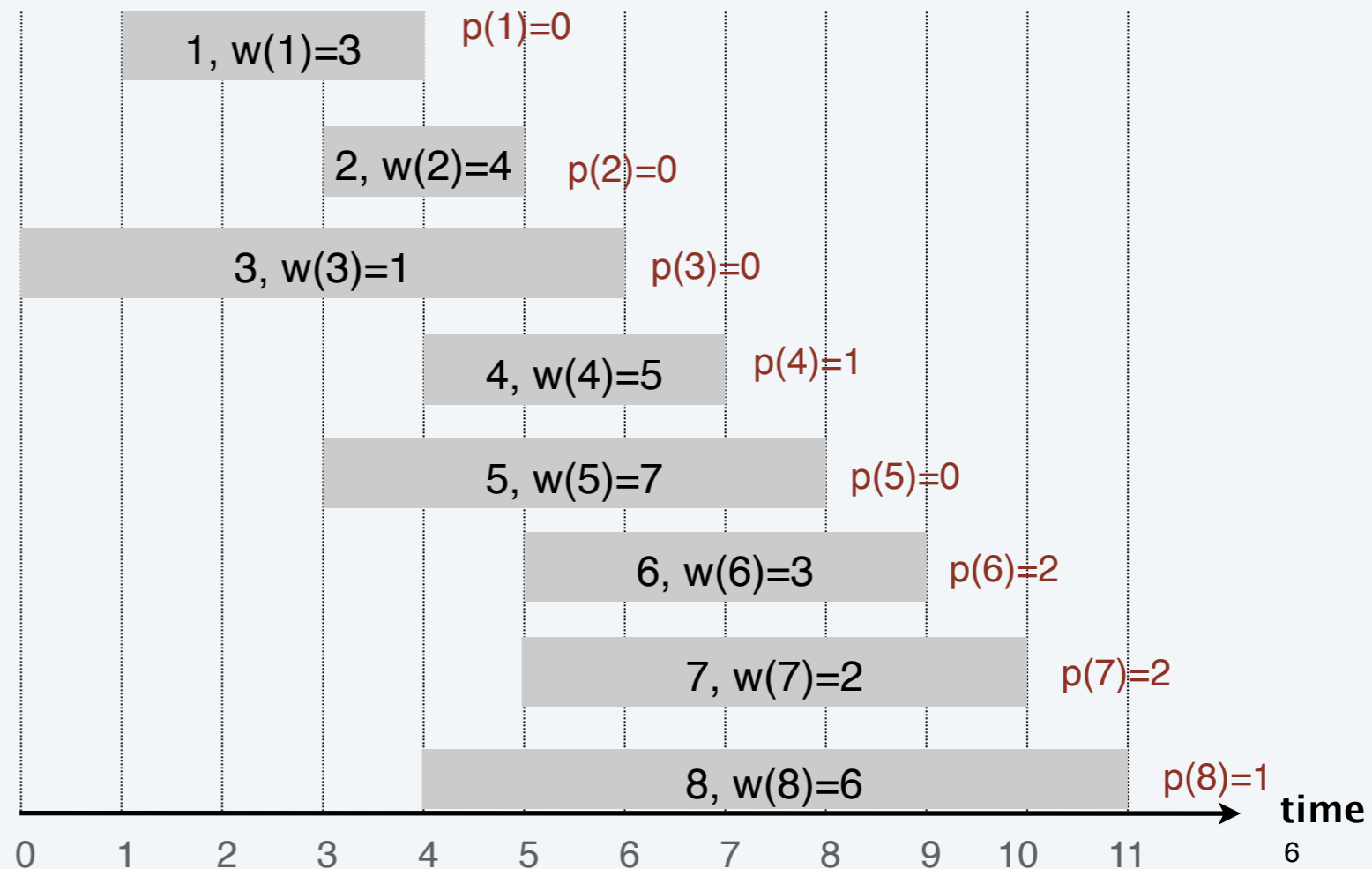
$M[4] =$

$M[5] =$

$M[6] =$

$M[7] =$

$M[8] =$



# Memoization allowed us to go from $O(2^n)$ to $O(n)$ ...

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Can we memoize merge sort?

`mergesort(L):`

`$L_1$  = first half of  $L$`

`$L_2$  = second half of  $L$`

`$sorted\_L_1$  = mergesort( $L_1$ )`

`$sorted\_L_2$  = mergesort( $L_2$ )`

`return merged  $L_1$  and  $L_2$`

# Memoization allowed us to go from $O(2^n)$ to $O(n)$ ...

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return merged  $L_1$  and  $L_2$

No-the key was overlapping subproblems