

# SC2001/ CX2101: Algorithm Design and Analysis

**Week 11**

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Another example of the longest path problem not satisfying the principle of optimality

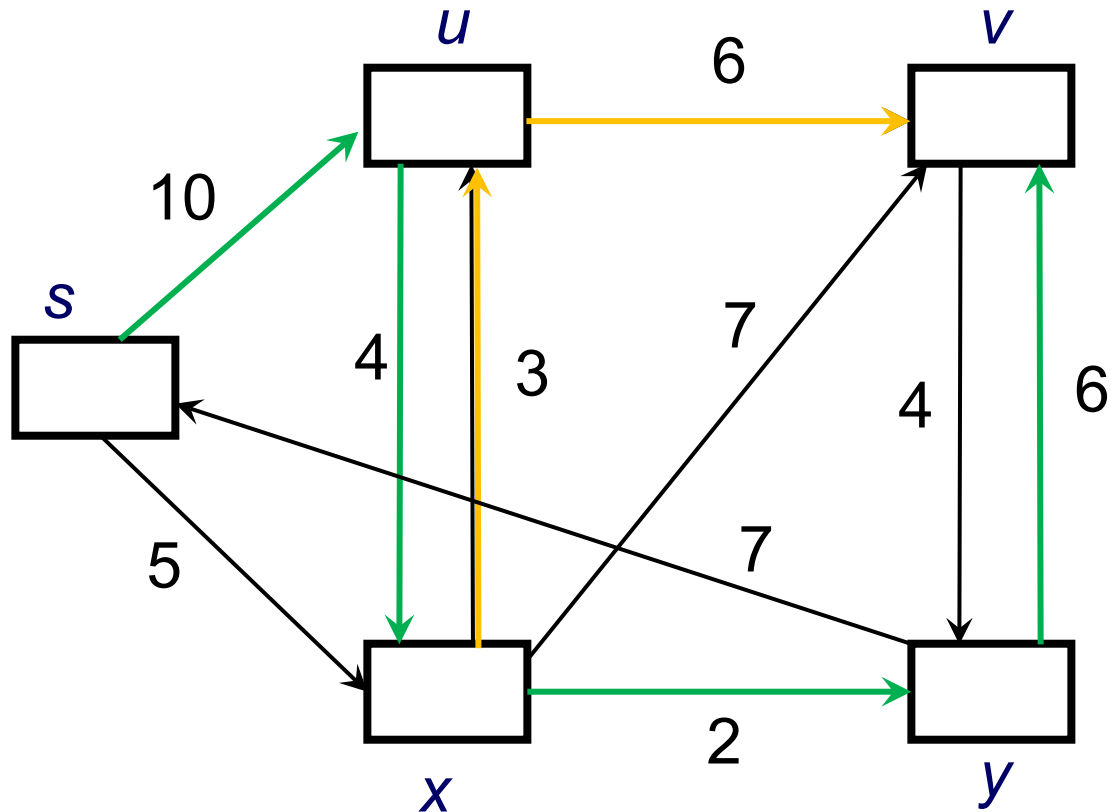
The longest path from  $s$  to  $v$  is

$s \rightarrow u \rightarrow x \rightarrow y \rightarrow v$

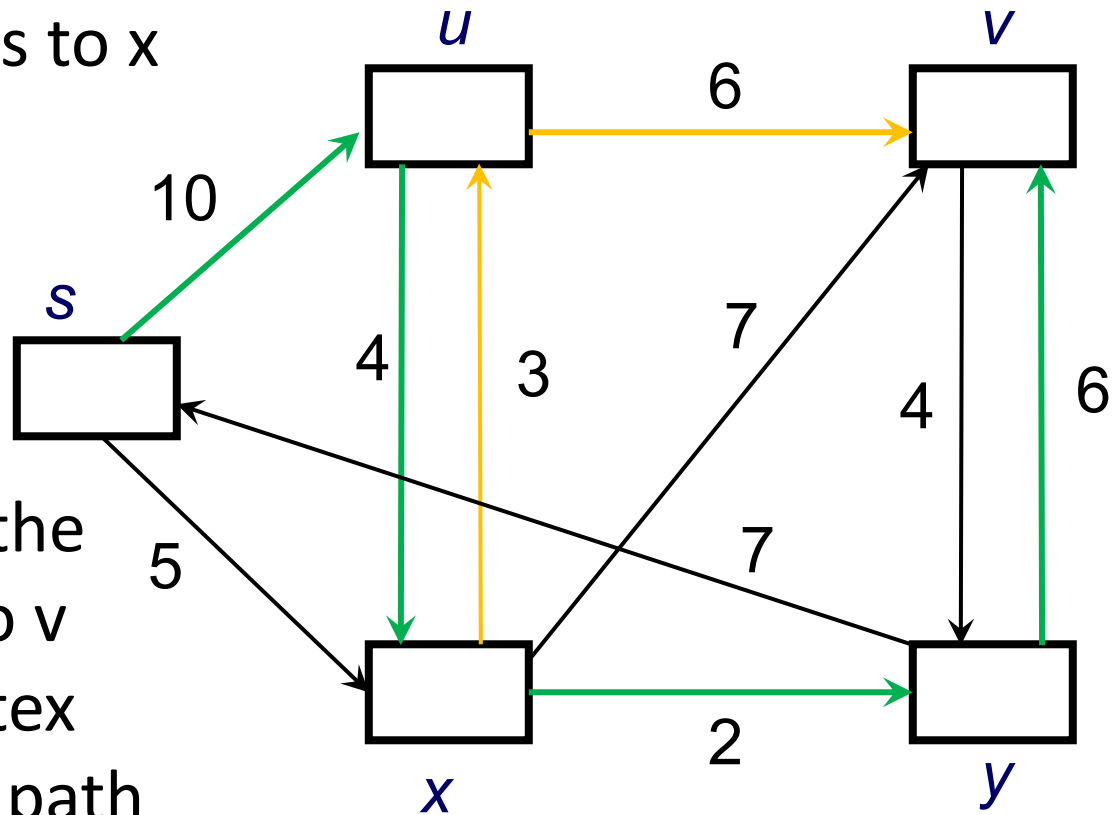
The longest path from  $x$  to  $v$ ?

Why?

Note, the longest path means the longest simple path (path with no cycle)



The solution to the problem of the longest path from  $s$  to  $v$  includes the solutions of 2 subproblems: from  $s$  to  $x$  and from  $x$  to  $v$ .



The longest path for the subproblem from  $x$  to  $v$  passes through a vertex which appears in the path from  $s$  to  $x$

Summary: Greedy vs DP

# A greedy strategy for coin change problem

## Example: Making Change

**Problem:** A country has coins with denominations

$$1 = d_1 < d_2 < \cdots < d_k.$$

You want to make change for  $n$  cents, using the smallest number of coins.

**Example:** U.S. coins

$$d_1 = 1 \quad d_2 = 5 \quad d_3 = 10 \quad d_4 = 25$$

Change for 37 cents – 1 quarter, 1 dime, 2 pennies.

What is the algorithm?

# A greedy strategy for coin change problem

```
c = 0;  
While (n > 0) {  
    c++;  
    n = n - value of the highest valued coin that is ≤ n;  
}  
Return c;                                // O(n)
```

An even faster version of the greedy approach is

1.  $a_1 = \lfloor n/d_k \rfloor, \quad n = n - a_1 * d_k$
2.  $a_2 = \lfloor n/d_{k-1} \rfloor, \quad n = n - a_2 * d_{k-1}$
3. ...

// O(k), or O(1) if k is a known constant

# A greedy strategy for coin change problem

- Works for many cases, e.g. to change for 37 cents,

$$d_1 = 1 \quad d_2 = 5 \quad d_3 = 10 \quad d_4 = 25$$

The answer is optimal: one 25¢, one 10¢, two 1¢

- but fails for cases like  $d = \{1, 10, 25\}$  and we want to change 30 cents

# A Greedy Method for chain matrix multiplication

1. use an array to record the dimensions.

E. g

	$A_1$	$\times$	$A_2$	$\times$	$A_3$	$\times$	$A_4$
	0	1	2	3	4		
d	30	1	40	10	25		

2. choose the multiplication of two matrices whose cost is the minimum at each step:

First,  $\min(30 \times 1 \times 40, 1 \times 40 \times 10, 40 \times 10 \times 25) \Rightarrow 1 \times 40 \times 10$  is the minimum. So  $A_2 \times A_3$  first:

0	1	2	3	4
30	1	40	10	25



# A Greedy Method for chain matrix multiplication

0	1	2	3	4
30	1	40	10	25

Second,  $\min(30 \times 1 \times 10, 1 \times 10 \times 25) \Rightarrow 1 \times 10 \times 25$  is the minimum. So  $(A_2 \times A_3) \times A_4$

0	1	2	3	4
30	1	40	10	25

Last:  $30 \times 1 \times 25: A_1((A_2 \times A_3) \times A_4)$

Total :  $1 \times 40 \times 10 + 1 \times 10 \times 25 + 30 \times 1 \times 25 = 1400$

Works in most cases except some sequences of 3 matrices (i.e. 2 matrix multiplications) e.g.

$A_1 \times A_2 \times A_3 : 10 \times 1 \times 10 \times 15 \Rightarrow 10 \times 1 \times 10 + 10 \times 10 \times 15$

# Another Greedy Method for chain matrix multiplication

1. use an array to record the dimensions.

E.g

	$A_1$	$\times$	$A_2$	$\times$	$A_3$	$\times$	$A_4$
	0	1	2	3	4		
d	30	1	40	10	25		

2. choose the multiplication of two matrices whose resulting matrix has minimum dimension at each step:

First,  $\min(30 \times 40, 1 \times 10, 40 \times 25) \Rightarrow 1 \times 10$  is the minimum.

So  $A_2 \times A_3$  first:

0	1	2	3	4
30	1	40	10	25

## Another Greedy Method for chain matrix multiplication

0	1	2	3	4
30	1	40	10	25

Second,  $\min(30 \times 10, 1 \times 25) \Rightarrow 1 \times 25$  is the minimum. So  $(A_2 \times A_3) \times A_4$

0	1	2	3	4
30	1	40	10	25

Last:  $30 \times 25$ :  $A_1((A_2 \times A_3) \times A_4)$

Total :  $1 \times 40 \times 10 + 1 \times 10 \times 25 + 30 \times 1 \times 25 = 1400$

Works in most cases but not for this (question in the tutorial T4Q5)

**d**

20	2	15	40	4
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$A_1 \times A_2 \times A_3 \times A_4 : A_1 \times (A_2 \times (A_3 \times A_4)) \Rightarrow 2680$  (1680 by DP)

# Summary: Greedy vs DP

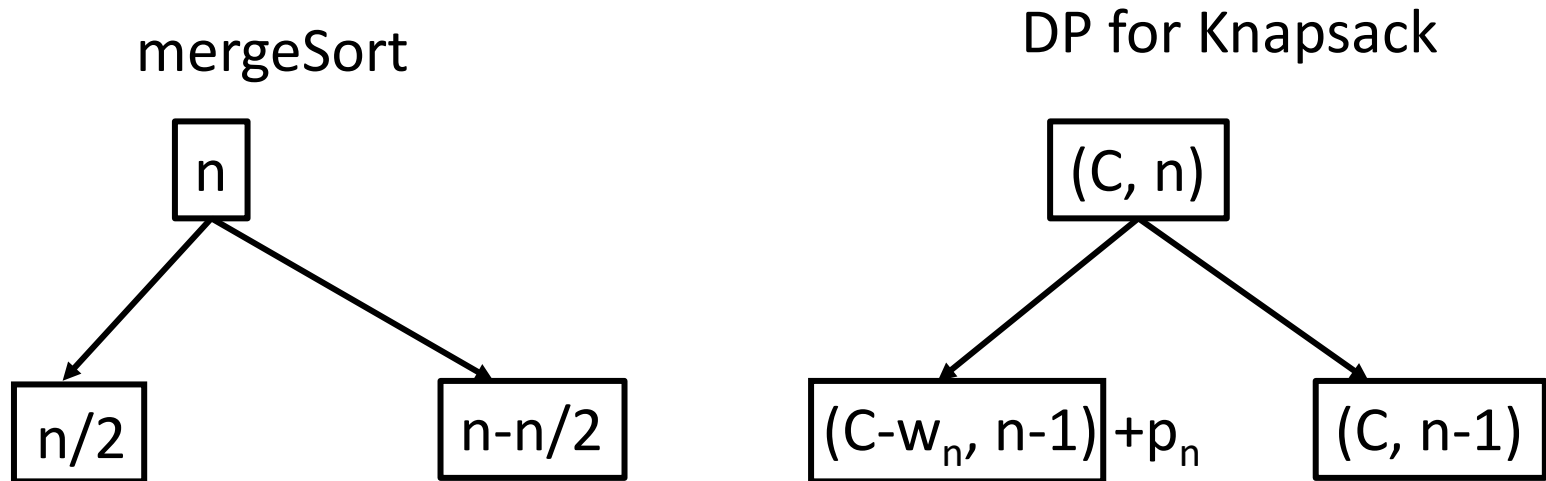
- DP is an **optimization** technique
- Both build solutions through a sequence of individual steps
- Both make a selection out of a collection of choices at each step
- At each step, the greedy method computes its **locally optimal** choice one after another and never revises any choices made – optimal solution not guaranteed, but many problems do get the optimal solution
- At each step, DP computes its solution by trying all choices before it arrives at the optimal choice

- Greedy heuristic algorithms are often used to solve problems because of its simplicity.
- Typically, DP algorithms are more expensive than greedy algorithms
- So DP is used only when no greedy strategy can be found to deliver the optimal solution

Summary: Divide-and-conquer vs DP

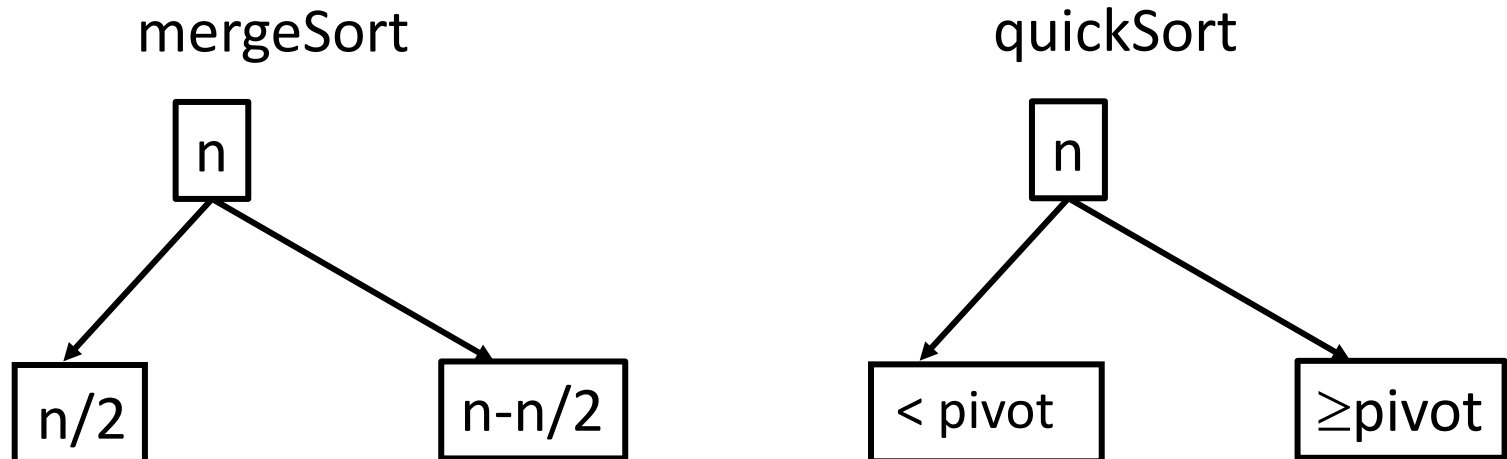
# Summary: Divide-and-conquer vs DP

- Both techniques divide their problems into subproblems, find subsolutions to the subproblems, and synthesize larger solutions from smaller ones.



# Summary: Divide-and-conquer vs DP

- Divide and Conquer divides a problem at prespecified deterministic points (e.g., always in the middle), combine the subsolutions to obtain the solution to a larger problem

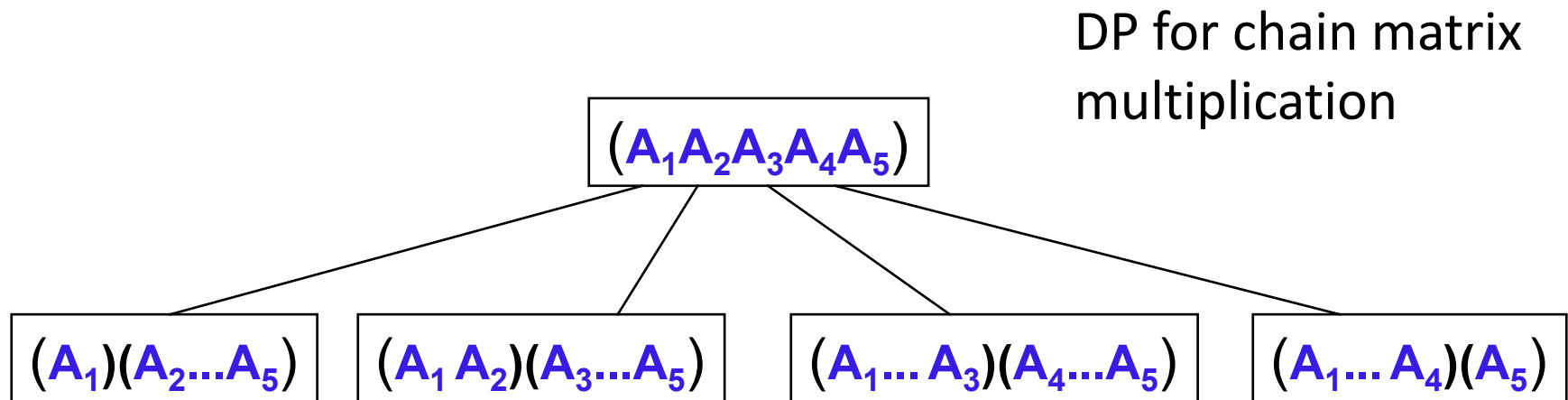


Divide at deterministic point



# Summary: Divide-and-conquer vs DP

- For optimization problems, DP divides a problem at every possible split points rather than at a pre-specified points. After trying all split points, it determines which split point is optimal. The subsolution at the split point is part of the solution to a larger problem



Divide at all possible split point