## CSE 431/531: Algorithm Analysis and Design (Spring 2024) Dynamic Programming

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## Paradigms for Designing Algorithms

## Greedy algorithm

- Make a greedy choice
- Prove that the greedy choice is safe
- Reduce the problem to a sub-problem and solve it iteratively
- Usually for optimization problems


## Divide-and-conquer

- Break a problem into many independent sub-problems
- Solve each sub-problem separately
- Combine solutions for sub-problems to form a solution for the original one
- Usually used to design more efficient algorithms


## Paradigms for Designing Algorithms

Dynamic Programming

- Break up a problem into many overlapping sub-problems
- Build solutions for larger and larger sub-problems
- Use a table to store solutions for sub-problems for reuse


## Recall: Computing the $n$-th Fibonacci Number

- $F_{0}=0, F_{1}=1$
- $F_{n}=F_{n-1}+F_{n-2}, \forall n \geq 2$
- Fibonacci sequence: $0,1,1,2,3,5,8,13,21,34,55,89, \cdots$


## $\operatorname{Fib}(n)$

1: $F[0] \leftarrow 0$
2: $F[1] \leftarrow 1$
3: for $i \leftarrow 2$ to $n$ do
4: $\quad F[i] \leftarrow F[i-1]+F[i-2]$
5: return $F[n]$

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- Store each $F[i]$ for future use.


## Outline

(1) Weighted Interval Scheduling
(2) Subset Sum Problem
(3) Knapsack Problem
4. Longest Common Subsequence

- Longest Common Subsequence in Linear Space
(5) Shortest Paths in Directed Acyclic Graphs

6) Matrix Chain Multiplication
(7) Optimum Binary Search Tree
(8) Summary
(9) Summary of Studies Until Nov 1st

## Recall: Interval Schduling

Input: $n$ jobs, job $i$ with start time $s_{i}$ and finish time $f_{i}$
$i$ and $j$ are compatible if $\left[s_{i}, f_{i}\right)$ and $\left[s_{j}, f_{j}\right)$ are disjoint
Output: a maximum-size subset of mutually compatible jobs


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Input: $n$ jobs, job $i$ with start time $s_{i}$ and finish time $f_{i}$
each job has a weight (or value) $v_{i}>0$
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Output: a maximum-weight subset of mutually compatible jobs

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