Lecture 22: Intro to Dynamic Programming COSC 311 *Algorithms,* Fall 2022

Announcements

- 1. '22E Honors Thesis talks Today!
 - 4-5pm in SCCE A131
- **2**, Courses Next Semester
 - 225 Algorithms and Visualization
 - 273 Parallel and Distributed Computing

Overview

- 1. Memoization 🗲
- 2. Profit Maximization, Revisited

- **So Far** Algorithmic Paradigms
- 1. Divide and Conquer

2. Greedy

And now...

Dynamic Programming

Features of Dynamic Programming

- 1. Break problem into smaller sub-problems
- 2. *Iterate* over sub-problems to produce solution

Compare to divide and conquer:

- 1. Break problem into smaller sub-problems
- 2. *Recursively* solve sub-problems
- 3. Combine solutions

Issues with Recursion

Recall the Fibonacci Sequence:

• 1, 1, 2, 3, 5, 8, 13, 21, 34, 55, ...

Issues with Recursion Recall the **Fibonacci Sequence**:

• 1, 1, 2, 3, 5, 8, 13, 21, 34, 55, ... Defined by: 1. f(1) = f(2) = 12. for n > 2, f(n) = f(n - 1) + f(n - 2) Issues with Recursion Recall the Fibonacci Sequence:

• 1, 1, 2, 3, 5, 8, 13, 21, 34, 55, ...

Defined by:

- 1. f(1) = f(2) = 1
- 2. for n > 2, f(n) = f(n 1) + f(n 2)

Recursive code:

Fib(n): if n <= 2 then return 1 return Fib(n-1) + Fib(n-2) Fib(n-2)

What is Running Time?



Redundant Recursive Calls

Recursive subproblems *overlap*

Memoization

Recursing without recursion

- store results of recursive calls
- iterate over results rather than making recursive calls
 - compute results "bottom up" rather than "top down"
 - iterate over stored values to compute "next" value

How to do for Fibonacci?

$$a[3] = aCr_{4} aEr_{5}$$

$$a[3] = aCr_{4} aEr_{5}$$

$$There over i = 3, 4, 5, ..., N$$

$$set aCift aEi-1J + aEi-2J$$

Memoized Fibonacci

MFib(n): a <- array of size n a[1] <- 1 a[2] <- 1 for each index i from 3 to n do 7n-2 iterations a[i] <- a[i-1] + a[i-2] endfor return a[n]

 $\partial(n)$ -

Running Time?

```
MFib(n):
  a <- array of size n
  a[1] <- 1
  a[2] <- 1
  for each index i from 3 to n do
     a[i] <- a[i-1] + a[i-2]
  endfor
  return a[n]
```

The Moral

With memoization, we converted

- recursive procedure
- many redundant recursive calls
- running time $\Omega(2^{n/2})$

into

- iterative procedure
- running time *O*(*n*)

Profit Maximization

Recall: Profit Maximization



Goal. Pick day *b* to buy and day *s* to sell to maximize profit.

Formalizing the Problem

Input. Array *a* of size *n*

• a[i] = price of Alphabet stock on day i

Output. Indices *b* (buy) and *s* (sell) with $1 \le b \le s \le n$ that maximize profit

•
$$p = a[s] - a[b]$$

Divide and Conquer Algorithm



Running time? Exercise: O(nlogn) from MT.

Another (Recursive) Procedure?

- consider last day, *n*
- two cases for optimal solution:
 - 1. max profit achieved by selling on day n
 - 2. max profit achieved by selling before day n

Questions.

 In case 1, how should we determine buy date?
 find min value in a[1..n], buy on that day
 In case 2, how should we compute max profit?
 Recursion: Solve on a[1..n-1].

Recursive Procedure



Questions

1. What makes MaxProfit inefficient?

Duplicated Work <u>computing</u> Min Value 2. What part(s) of the procedure can be memoized? <u>Mis</u>

Memoizing MaxProfit

Create two arrays:

- 1. min[i] stores minimum value in a[1..i]
- 2. max[i] stores maximum profit achievable by selling up to time i

Question. How to update these arrays?