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CSE 373 Analysis of Algorithms Fall 2016 Instructor: Prof. Sael Lee

# LEC16: INTRODUCTION TO DYNAMIC PROGRAMMING

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## DYNAMIC PROGRAMMING

- Dynamic programming is a very powerful, general tool for solving optimization problems on left-right-ordered items such as character strings.
- Once understood it is relatively easy to apply, it looks like magic until you have seen enough examples.
- Floyd's all-pairs shortest-path algorithm was an example of dynamic programming.

# **GREEDY VS. EXHAUSTIVE SEARCH**

- *Greedy* algorithms focus on making the best local choice at each decision point. In the absence of a correctness proof such greedy algorithms are very likely to fail.
- Dynamic programming gives us a way to design custom algorithms which systematically search all possibilities (thus guaranteeing correctness) while storing results to avoid recomputing (thus providing efficiency).

# **RECURRENCE RELATIONS**

- \* A recurrence relation is an equation which is defined in terms of itself.
- They are useful because many natural functions are easily expressed as recurrences:
  - + Polynomials:  $a_n = a_{n-1} + 1$ ;  $a_1 = 1 \rightarrow a_n = n$
  - + Exponentials:  $a_n = 2a_{n-1}$ ;  $a_1 = 2 \rightarrow a_n = 2^n$
  - + Weird:  $a_n = na_{n-1}; a_1 = 1 \rightarrow a_n = n!$
- Computer programs can easily evaluate the value of a given recurrence even without the existence of a nice closed form.

### **COMPUTING FIBONACCI NUMBERS**

$$F_n = F_{n-1} + F_{n-2}; F_0 = 0; F_1 = 1$$

 Implementing this as a recursive procedure is easy, but slow because we keep calculating the same value over and over.





$$\frac{F_{n+1}}{F_n} \approx \phi = \frac{1+\sqrt{5}}{2} \approx 1.61803$$

Thus  $F_n \approx 1.6^n$ .

Since our recursion tree has 0 and 1 as leaves, computing  $F_n$  requires  $\approx 1.6^n$  calls!

# WHAT ABOUT DYNAMIC PROGRAMMING?

We can calculate  $F_n$  in linear time by storing small values:

 $+F_0 = 0$ +  $F_1 = 1$ + for i = 1 to n

 $\times F_i = F_{i-1} + F_{i-2}$ 

Moral: we traded space for time.

# BENEFITS OF DYNAMIC PROGRAMMING

- Dynamic programming is a technique for efficiently computing recurrences by storing partial results.
- Once you understand dynamic programming, it is usually easier to reinvent certain algorithms than try to look them up!
- Dynamic programming to be one of the most useful algorithmic techniques in practice:
  - + Morphing in computer graphics.
  - + Data compression for high density bar codes.
  - + Designing genes to avoid or contain specified patterns.

#### **AVOIDING RECOMPUTATION BY STORING PARTIAL RESULTS**

- The trick to dynamic program is to see that the naive recursive algorithm repeatedly computes the same subproblems over and over and over again.
- If so, storing the answers to them in a table instead of recomputing can lead to an efficient algorithm.
- Thus we must first hunt for a correct recursive algorithm – later we can worry about speeding it up by using a results matrix.

## **BINOMIAL COEFFICIENTS**

- \* The most important class of counting numbers are the *binomial coefficients*, where  $\binom{n}{k}$  counts the number of ways to choose k things out of n possibilities.
  - + Committees How many ways are there to form a k-member committee from n people? By definition,  $\binom{n}{k}$ .
  - + Paths Across a Grid How many ways are there to travel from the upper-left corner of an n m grid to the lower-right corner by walking only down and to the right? Every path must consist of n + m steps, n downward and m to the right, so there are  $\binom{n+m}{n}$ such sets/paths.

## **COMPUTING BINOMIAL COEFFICIENTS**

- × Since  $\binom{n}{k} = n!/((n k)! k!)$ , in principle you can compute them straight from factorials.
- However, intermediate calculations can easily cause <u>arithmetic overflow</u> even when the final coefficient fits comfortably within an integer.



In Pascal's Triangle, each number is the sum of the two numbers directly above it:



### PASCAL'S RECURRENCE

- \* A more stable way to compute binomial coefficients is using the recurrence relation implicit in the construction of Pascal's triangle, namely, that  $\binom{n}{k} = \binom{n-1}{k-1} + \binom{n-1}{k}$
- \* It works because the nth element either appears or does not appear in one of the  $\binom{n}{k}$  subsets of k elements.



- × No recurrence is complete without basis cases.
- \* How many ways are there to choose 0 things from a set?
- × Exactly one, the empty set.
- \* The right term of the sum drives us up to  $\binom{k}{k}$ . How many ways are there to choose k things from a k-element set?
- × Exactly one, the complete set.

## **BINOMIAL COEFFICIENTS IMPLEMENTATION**

```
long binomial coefficient(n,m)
                       /* compute n choose m */
int n,m;
ł
   int i,j;
                       /* counters */
   long bc[MAXN][MAXN]; /* table of binomial coefficients */
   for (i=0; i \le n; i++) bc[i][0] = 1;
   for (j=0; j<=n; j++) bc[j][j] = 1;
   for (i=1; i<=n; i++)
   for (j=1; j<i; j++)
       bc[i][i] = bc[i-1][i-1] + bc[i-1][i];
   return( bc[n][m] );
```

# THREE STEPS TO DYNAMIC PROGRAMMING

- 1. Formulate the answer as a recurrence relation or recursive algorithm.
- 2. Show that the number of different instances of your recurrence is bounded by a polynomial.
- 3. Specify an order of evaluation for the recurrence so you always have what you need.