### Reading: Chapter 2 & 3.

#### **Announcements:**

• discussion on Piazza

• grading:

• homework: 30%

• peer review: 30%

• midterms: 20% (10/27)

• final: 15% (12/3)

• participation: 5%

- sections, Mondays, various times.
- homework partners
- Homework plan:
  - assigned thursday, due thursday, work in pairs, graded for accuracy and quality.
  - peer review.
- TAs: Sam Taggart, Aleck Johnsen, Yiding Feng
- office hours

# Algorithms

- algorithms are everywhere. examples:
  - digital computers,
  - parlementary procedure,
  - scientific method,
  - biological processes.
- algorithms design and analysis governs everything.
- good algorithms are closest things to magic.
- course philosophy: no particular algorithm is important.
- course goals: how to design, analize, and think about algorithms.
- we will not cover anything you could figure out on your own.

# Algorithms for Fibonacci Remembering Redundant Computa-Numbers

"0, 1, 1, 2, 3, 5, 8, 13, 21, ..."

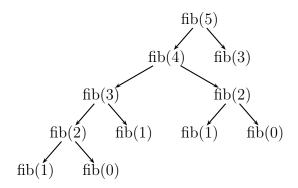
**Question:** recursive alg?

**Algorithm:** Recursive Fibonacci

fib(k):

- 1. if  $k \leq 1$  return k
- 2. (else) return fib(k-1) + fib(k-1)

#### Example:



#### **Analysis**

"what is runtime?"

Let 
$$T(k)$$
 = number of calls to fib
$$T(0) = T(1) = 1$$

$$T(k) = T(k-1) + T(k-2)$$

$$\geq 2T(k-2)$$

$$\geq 2 \times 2T(k-4)$$

$$\geq \underbrace{2 \times 2 \times \cdots \times 2}_{k/2 \text{ times}} \times 1$$

$$= 2^{k/2}$$

Conclusion: at least "exponential time"!

# tion (memoization)

remember redundant computation Idea: (memoize)

Algorithm: Memoized Recursive Fibonacci fib-helper(k)

- 1. if  $memo[k] \ge 0$  return memo[k]
- 2. (else) return fib-helper(k-1) + fibhelper(k-2)

fib(k)

- 1. memo = new int[k];
- 2. memo[0] = 0, memo[1] = 1, memo[2,...,k]= -1;
- 3. return fib-helper(k)

#### Example:

	MAAA						
ſ	0	1	1	2	3	5	

#### **Analysis**

- cost to fill in each entry: 1 additions.
- $\bullet$  number of entries: k
- total cost: T(k) = k additions.

Conclusion: "linear time".

**Note:** memoizing redundant computation is essential part of "dynamic programming".

#### Iterative Algorithm

Algorithm: Iterative Memoized Fibonacci fib(k):

- 1. memo = new int[k];
- 2. memo[0] = 0, memo[1] = 1
- 3. for i = 2..k

$$memo[i] = memo[i-1] + memo[i-2]$$

4. return memo[k]

**Question:** Can we compute fib with less memory (space)?

Algorithm: Iterative Fibonacci

fib(k):

- 1. last[0] = 0, last[1] = 1;
- 2. for i = 2..k
  - (a) tmp = last[1]
  - (b) last[1] = last[0] + last[1]
  - (c) last[0] = tmp
- 3. return last[1]

Question: faster alg?

## Fast Fibonacci

Note: algorithm operates on last like a matrix multiply

fib(k):

1. 
$$z = [0 \ 1]; A = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix}$$

2. multiply 
$$z \times \underbrace{A \times A \cdots \times A}_{k-1 \text{ times}}$$

3. return z[1]

**Note:** just need to compute  $z \times A^{k-1}$ 

# Exponentiation

"compute  $A^k$ "

**Note:** If  $k = k_1 + k_2$  then  $A^k = A^{k_1}A^{k_2}$ 

- compute  $A^{k_1}$  and  $A^{k_2}$  and multiply.
- if  $k_1 = k_2$  then redundant computation

**Idea:** factor  $A^k = (A^k / 2)^2 \times A^k / 2$ 

Algorithm: Repeated Squaring

- 1. if k = 1 return A
- $2. \ k' = \lfloor k/2 \rfloor.$
- 3. B = repeated-square(A, k').
- 4. if k odd

return  $B \times B \times A$ 

5. else

return  $B \times B$ 

#### Analysis

Let T(k) = number of multiplies.

$$T(1) = 0$$

$$T(k) = T(k/2) + 2$$

$$= T(k/4) + 2 + 2$$

$$= \underbrace{2 + 2 + 2 \cdots 2}_{\log k \text{ times}}$$

$$= 2 \log k$$

**Note:** finding subproblems is important part of "divide and conquer"

**Algorithm:** Fibonacci numbers via repeated squaring

fib(k):

$$1. \ A = \left[ \begin{array}{cc} 0 & 1 \\ 1 & 1 \end{array} \right].$$

- 2.  $z = [0 \ 1] \times \text{repeated-square}(A, k 1)$ .
- 3. return z[1].

## Analysis

 $2 \log k$  2x2 matrix multiplies.

## Conclusions

- runtime analysis
- memoization
- divide and conquer