

## CSSE 220 Day 22

Recursion, Efficiency, and the Time-Space Trade Off;
Selection Sort and Big-Oh

Checkout Recursion2 project from SVN

## Questions?



## Recursion

- What is a recursive method?
- Answer: A method that calls itself but on a "simpler" problem, so that it makes progress toward completion
- When to use recursive methods?
- Implementing a recursive definition

$$
n!=n \times(n-1)!
$$

- Implementing methods on a recursive data structure, e.g.: Size of tree to the right is the sum of sizes of subtrees B, C, D, E, plus 1
- Any situation where parts of the whole
 look like mini versions of the whole
- Folders within folders on computers
- Trees
- Pros: easy to implement, easy to understand code, easy to prove code correct
- Cons: takes more space than equivalent iteration (because of function calls)


## Key Rules to Using Recursion

- Always have a base case that doesn't recurse
- Make sure recursive case always makes progress, by solving a smaller problem
- You gotta believe
- Trust in the recursive solution
- Just consider one step at a time


## What the Fib?

- The nth Fibonacci number $F(n)$ is defined by:

$$
\begin{aligned}
& F(n)=F(n-1)+F(n-2) \text { for } n>1 \\
& F(1)=F(2)=1
\end{aligned}
$$

- Why does recursive Fibonacci take so long?!?
- Hint: How deep is the right-most branch of the tree below? Hence how big the tree? Hence how long does the computation take?
- How can we fix it?
- Use a memory table! Same idea as what some of you noticed about Ackermann, but more powerful with Fibonacci.



## Memory tables

- To speed up the recursive calculation of the nth Fibonacci number, just:

1. "Memorize" every solution we find to subproblems, and
2. Before you recursively compute a solution to a subproblem, look it up in the "memory table"

- So to compute the nth Fibonacci number, construct an array that has $n+1$ elements, all initialized to 0 . Then call Fib(n).
- The base case for $\operatorname{Fib}(\mathrm{k})$ remains the same as in the naive solution.
- At the beginning of the recursive step computing Fib(k), see if the $k^{\text {th }}$ entry in the array is 0 .

If it is NOT 0, return it.

- If it IS 0, compute Fib(k) recursively. Then store the computed value in the $\mathrm{k}^{\text {th }}$ spot of the array. Then return the computed value.

This is a classic time-space tradeoff

- A deep discovery of computer science
- Studied by "Complexity Theorists"
- Used everyday by software engineers

Tune the solution by varying the amount of storage space used and the amount of computation performed

## Two Mirrors



If you actually do this, what really happens is Douglas Hofstadter appears and talks to you for eight hours about strange loops.

## Mutual Recursion

Two or more methods that call each other repeatedly

- For example, Hofstadter Female and Male Sequences

$$
\begin{aligned}
& F(n)= \begin{cases}1 & \text { if } n=0 \\
n-M(F(n-1)) & \text { if } n>0\end{cases} \\
& M(n)= \begin{cases}0 & \text { if } n=0 \\
n-F(M(n-1)) & \text { if } n>0\end{cases}
\end{aligned}
$$

- Questions:
- How often are the sequences different in the first 50 positions? first 500? first 5,000? first 5,000,000?


## What is sorting? <br> 21 Let's see...

## Why study sorting? <br> 2) Shlemiel the Painter

## What makes a program "good"?

- Correct - meets specifications
- Easy to understand, modify, write
- Uses reasonable set of resources
- Time (runs fast)
- Space (main memory)
- Hard-drive space
- Peripherals
- Here we focus on "runs fast" - how much CPU time does the program / algorithm / problem take?
- Others are important too!


## Course Goals for Sorting: You should...

- Be able to describe basic sorting algorithms:
- Selection sort
- Insertion sort
- Merge sort
- Quicksort
- Know the run-time efficiency of each
- Know the best and worst case inputs for each


## Selection Sort

- Basic idea:
- Think of the list as having a sorted part (at the beginning) and an unsorted part (the rest)
- Find the smallest number in the unsorted part
- Move it to the end of the sorted part (making the sorted part bigger and the unsorted part smaller)

$$
\begin{aligned}
& \text { Repeat until } \\
& \text { unsorted part is } \\
& \text { empty }
\end{aligned}
$$

## Profiling Selection Sort

- Profiling: collecting data on the run-time behavior of an algorithm
- How long does selection sort take on:
- 10,000 elements?
- 20,000 elements?
- ...
- 100,000 elements?


## Big-Oh motivation: why profiling is not enough

- Results from profiling depend on:
- Power of machine you use
- CPU, RAM, etc
- Operating system of machine you use
- State of machine you use
- What else is running? How much RAM is available? ...
- What inputs do you choose to run?
- Size of input
- Specific input


## Big-Oh motivation: what it provides

- Big-Oh is a mathematical definition that allows us to:
- Determine how fast a program is (in big-Oh terms)
- Share results with others in terms that are universally understood
- Features of big-Oh
- Allows paper-and-pencil analysis
- Is much easier / faster than profiling
- Is a function of the size of the input
- Focuses our attention on big inputs
- Is machine independent


## Analyzing Selection Sort

- Analyzing: calculating the performance of an algorithm by studying how it works, typically mathematically
- Typically we want the relative performance as a function of input size
- Example: For an array of length $n$, how many times does selectionSort() call compareTo()?

| Handy Fact |
| :---: |
| $1+2+\ldots+(n-1)+n=\frac{n(n+1)}{2}$ |

## Asymptotic analysis

- We care most what happens when $n$ (the size of a problem) gets large
- Is the function basically linear, quadratic, exponential, etc. ?
- Consider: Why do we care most about large inputs?
- For example, when n is large (or even moderate):
- The difference between $n^{2}$ and $n^{2}-3$ is negligible.
- $\mathrm{n}^{3}$ is pretty large but $2^{\mathrm{n}}$ is REALLY large.
" We say, "selection sort takes on the order of $n^{2}$ steps"
- Big-Oh gives a formal definition for "on the order of"


## Definition of big-Oh

- Formal:
- We say that $f(n)$ is $O(g(n))$ if and only if
- there exist constants $c$ and $n_{0}$ such that
- for every $n \geq n_{0}$ we have
- $f(n) \leq c \times g(n)$
- Informal:
- $\mathrm{f}(\mathrm{n})$ is roughly proportional to $\mathrm{g}(\mathrm{n})$, for large n

- Example: $7 n^{3}+24 n^{2}+3000 n+45$ is $O\left(n^{3}\right)$
- Because it is $\leq 3,077 \times n^{3}$ for all $n \geq 1$

