

CSE 326: Data Structures

Lecture #2

Analysis of Algorithms I

(And A Little More About Linked Lists)

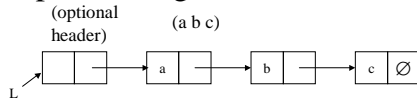
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Winter 2002

Assignment #1

Goals of this assignment:

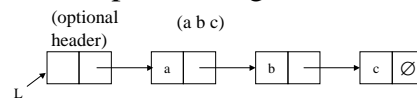
- Introduce the ADTs (abstract data types) for lists and sparse vectors, motivated by an application to information retrieval.
- Show the connection between the empirical runtime scaling of an algorithm and formal asymptotic complexity
- Gain experience with the Unix tools g++, make, gnuplot, csh, and awk.
- Learn how to use templates in C++.
 - We will use new g++ version 3.0 compiler - does templates right!

Implementing Linked Lists in C



```
struct node{
    Object element;
    struct node * next; }
Everything else is a pointer to a node!
typedef struct node * List;
typedef struct node * Position;
```

Implementing in C++



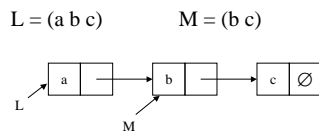
Create separate classes for

- Node
- List (contains a pointer to the first node)
- List Iterator (specifies a position in a list; basically, just a pointer to a node)

Pro: syntactically distinguishes uses of node pointers

Con: a lot of verbage! Also, is a position in a list really distinct from a list?

Structure Sharing



- Important technique for conserving memory usage in large lists with repeated structure
- Used in many recursive algorithms on lists

Implementing Linked Lists Using Arrays

	1	2	3	4	5	6	7	8	9	10
Data		F	O	A	R	N		R		T
Next		3	8	6	4	-1		10		5

First = 2

“Cursor implementation” Ch 3.2.8

Can use same array to manage a second list of unused cells

List ADT
↑
Polynomial ADT

A_i is the coefficient of the x^{i-1} term:

$5 + 2x + 3x^2$ (5 2 3)

$7 + 8x$ (7 8)

$3 + x^2$ (3 0 2)

Problem?

$4 + 3x^{2001}$

Sparse List Data Structure:
 $4 + 3x^{2001}$

(<4 0> <2001 3>)

Addition of Two Polynomials?

$15 + 10x^{50} + 3x^{1200}$

p →

$5 + 30x^{50} + 4x^{100}$

q →

Addition of Two Polynomials

- Similar to merging two sorted lists – one pass!

$15 + 10x^{50} + 3x^{1200}$

p →

$5 + 30x^{50} + 4x^{100}$

q →

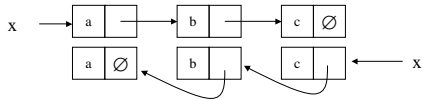
r →

To ADT or NOT to ADT?

- Issue: when to bypass / expand List ADT?
- Using general list operations:

```
reverse(x) {
  y = new list;
  while (! x.empty()){
    /* remove 1st element from x, insert in y */
    y.insert_after_kth( x.kth(1), 0);
    x.delete_kth(1);
  }
  return y; }
Disadvantages?
```

Destructive LL Version



```
reverse(node * x) {
    last = NULL;
    while (x->next != NULL){
        tmp = x->next;
        x->next = last;
        last = x;
        x = tmp;
    }
    return x; }
```

Faster in practice?
Asymptotically faster?

Analysis of Algorithms

- Analysis of an algorithm gives insight into how long the program runs and how much memory it uses
 - time complexity
 - space complexity
- Why useful?
- Input size is indicated by a number n
 - sometimes have multiple inputs, e.g. m and n
- Running time is a function of n
 - n , n^2 , $n \log n$, $18 + 3n(\log n^2) + 5n^3$

Simplifying the Analysis

- Eliminate low order terms
 - $4n + 5 \Rightarrow 4n$
 - $0.5 n \log n - 2n + 7 \Rightarrow 0.5 n \log n$
 - $2^n + n^3 + 3n \Rightarrow 2^n$
- Eliminate constant coefficients
 - $4n \Rightarrow n$
 - $0.5 n \log n \Rightarrow n \log n$
 - $\log n^2 = 2 \log n \Rightarrow \log n$
 - $\log_3 n = (\log_3 2) \log n \Rightarrow \log n$

Order Notation

- BIG-O $T(n) = O(f(n))$
Upper bound
Exist constants c and n' such that
 $T(n) \leq c f(n)$ for all $n \geq n'$
- OMEGA $T(n) = \Omega(f(n))$
Lower bound
Exist constants c and n' such that
 $T(n) \geq c f(n)$ for all $n \geq n_0$
- THETA $T(n) = \theta(f(n))$
Tight bound
 $\theta(n) = O(n) = \Omega(n)$

Examples

$n^2 + 100n = O(n^2)$ because
 $(n^2 + 100n) \leq 2n^2$ for $n \geq 10$

$n^2 + 100n = \Omega(n^2)$ because
 $(n^2 + 100n) \geq 1n^2$ for $n \geq 0$

Therefore:
 $n^2 + 100n = \theta(n^2)$

Notation Gotcha

- Order notation is not symmetric; write
 $2n^2 + 4n = O(n^2)$
but never
 $O(n^2) = 2n^2 + 4n$
right hand side is a crudification of the left
- Likewise
 $O(n^2) = O(n^3)$
 $\Omega(n^3) = \Omega(n^2)$

Mini-Quiz

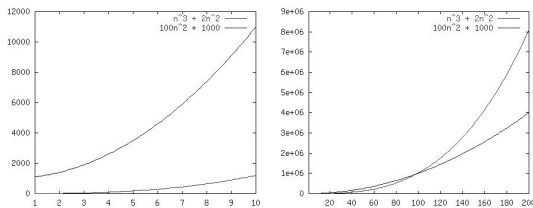
1. $5n \log n = O(n^2)$
2. $5n \log n = \Omega(n^2)$
3. $5n \log n = O(n)$
4. $5n \log n = \Omega(n)$
5. $5n \log n = \theta(n)$
6. $5n \log n = \theta(n \log n)$

Silicon Downs

Post #1	Post #2
$n^3 + 2n^2$	$100n^2 + 1000$
$n^{0.1}$	$\log n$
$n + 100n^{0.1}$	$2n + 10 \log n$
$5n^5$	$n!$
$n^{-1.5}2^n/100$	$1000n^{1.5}$
$8^{2 \log n}$	$3n^7 + 7n$

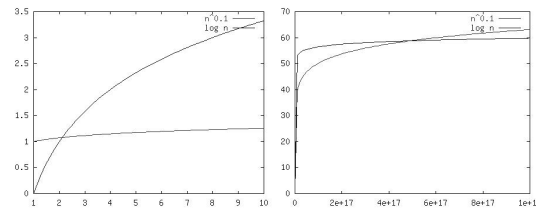
Race I

$n^3 + 2n^2$ vs. $100n^2 + 1000$



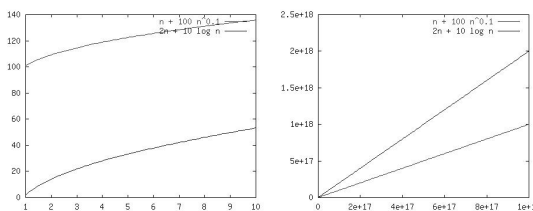
Race II

$n^{0.1}$ vs. $\log n$



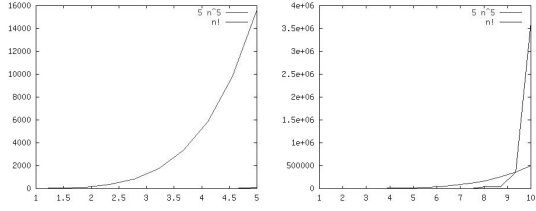
Race III

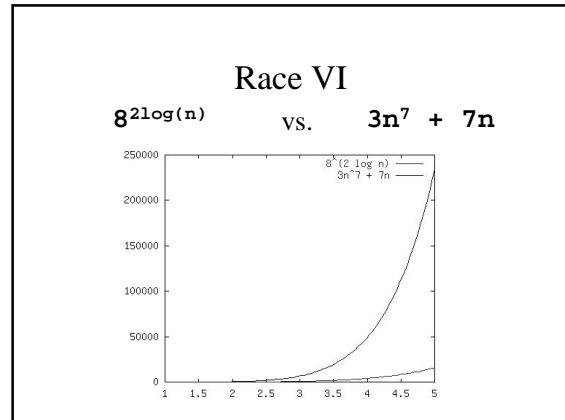
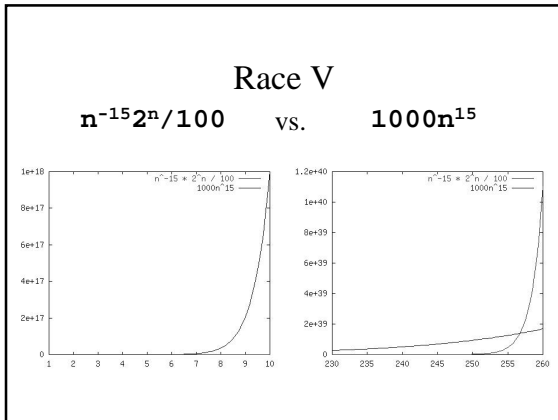
$n + 100n^{0.1}$ vs. $2n + 10 \log n$



Race IV

$5n^5$ vs. $n!$





The Losers Win

Post #1	Post #2	Better algorithm!
$n^3 + 2n^2$	$100n^2 + 1000$	$O(n^2)$
$n^{0.1}$	$\log n$	$O(\log n)$
$n + 100n^{0.1}$	$2n + 10 \log n$	TIE $O(n)$
$5n^5$	$n!$	$O(n^5)$
$n^{-15}2^n/100$	$1000n^{15}$	$O(n^{15})$
$8^{2\log n}$	$3n^7 + 7n$	$O(n^6)$

Common Names

constant:	$O(1)$	
logarithmic:	$O(\log n)$	
linear:	$O(n)$	
log-linear:	$O(n \log n)$	
superlinear:	$O(n^{1+c})$	(c is a constant > 0)
quadratic:	$O(n^2)$	
polynomial:	$O(n^k)$	(k is a constant)
exponential:	$O(c^n)$	(c is a constant > 1)

- ### Analyzing Code
- C++ operations - constant time
 - consecutive stmts - sum of times
 - conditionals - sum of branches, condition
 - loops - sum of iterations
 - function calls - cost of function body
 - recursive functions - solve recursive equation
- Above all, use your head!*

Conditionals

- Conditional
`if C then S1 else S2`

time \leq time(C) + MAX(time(S₁), time(S₂))

Nested Loops

```
for i = 1 to n do
  for j = 1 to n do
    sum = sum + 1
```

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```

$$\sum_{i=1}^n \sum_{j=1}^n 1 = \sum_{i=1}^n n = n^2$$

Nested Dependent Loops

```
for i = 1 to n do
  for j = i to n do
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Nested Dependent Loops

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for i = 1 to n do
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```

$$\sum_{i=1}^n \sum_{j=i}^n 1 = \sum_{i=1}^n (n-i+1) = \sum_{i=1}^n (n+1) - \sum_{i=1}^n i =$$

?

Nested Dependent Loops

```
for i = 1 to n do
  for j = i to n do
    sum = sum + 1
```

$$\sum_{i=1}^n \sum_{j=i}^n 1 = \sum_{i=1}^n (n-i+1) = \sum_{i=1}^n (n+1) - \sum_{i=1}^n i =$$

$$n(n+1) - \frac{n(n+1)}{2} = \frac{n(n+1)}{2} = O(n^2)$$

To Do

- Finish reading Ch 1 and 2
- Start reading Ch 3
- Get started on assignment #1!