

Recurrence Relations – Running Time for Recursive Functions

Chapter 2

Asymptotic Dominance in Action

	$O(\lg n)$	$O(n)$	$O(n \lg n)$	n^2	2^n	$n!$
10	0.003 μ s	0.01 μ s	0.033 μ s	0.1 μ s	1 μ s	3.63 ms
20	0.004 μ s	0.02 μ s	0.086 μ s	0.4 μ s	1 ms	77.1 years
30	0.005 μ s	0.03 μ s	0.147 μ s	0.9 μ s	1 sec	$8.4 \cdot 10^{15}$ yrs
40	0.005 μ s	0.04 μ s	0.213 μ s	1.6 μ s	18.3 min	
50	0.006 μ s	0.05 μ s	0.282 μ s	2.5 μ s	13 days	
100	0.007 μ s	0.1 μ s	0.644 μ s	10 μ s	$4 \cdot 10^{13}$ yrs	
1,000	0.010 μ s	1.00 μ s	9.966 μ s	1 ms		
10,000	0.013 μ s	10 μ s	130 μ s	100 ms		
100,000	0.017 μ s	0.10 ms	1.67 ms	10 sec		
1,000,000	0.020 μ s	1 ms	19.93 ms	16.7 min		
10,000,000	0.023 μ s	0.01 sec	0.23 sec	1.16 days		
100,000,000	0.027 μ s	0.10 sec	2.66 sec	115.7 days		
1,000,000,000	0.030 μ s	1 sec	29.90 sec	3.7 years		

Gnome Sort



<http://www.portlandoctopus.com/top-5-garden-gnomes/>

Divide and Conquer Algorithms

- Analysis of divide and conquer algorithms requires knowledge of:
 - Mathematical Induction
 - Substitution/Iterative Method
 - Recurrences

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4

Motivation

- The following structure and function exist:

```
struct Tree
{
    int info;
    Tree * left;
    Tree * right;
    Tree(int value, Tree * lchild, Tree * rchild) : info
        (value), left(lchild), right(rchild) { }
};

// return true if & only if all values in t are less than val
bool ValsLess(Tree * t, int val)
```

Thank you Owen Astrachan

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5

Motivation

```
// returns true if t represents a binary
// search tree containing no duplicate values;
bool IsBST(Tree * t)
{
    if (t == NULL) return true;
    return ValsLess(t->left, t->info) &&
        ValsGreater(t->right, t->info) &&
        IsBST(t->left) &&
        IsBST(t->right);
}
```

- What is the complexity or running time of the above function on a tree with n nodes?

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6

Another Example

- What is the asymptotic complexity of the function below? Assume Combine is $O(n)$

```
// precondition: a[left] <= ... <= a[right]
void DoStuff(vector<int> & a, int left, int right)
{
    int mid = (left + right)/2;
    if (left < right)
    {
        DoStuff(a, left, mid);
        DoStuff(a, mid + 1, right);
        Combine(a, left, mid, right);
    }
}
```

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7

Another Example

- What does the function below remind you of?

```
// precondition: a[left] <= ... <= a[right]
void DoStuff(vector<int> & a, int left, int right)
{
    int mid = (left + right)/2;
    if (left < right)
    {
        DoStuff(a, left, mid);
        DoStuff(a, mid + 1, right);
        Combine(a, left, mid, right);
    }
}
```

Merge Sort!

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8

Recurrence Relation

- A **recurrence relation** contains two equations
 - One for the general case
 - One for the base case

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Efficiency of Binary Search

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Merge Sort

- What was the running time of the Merge procedure in Merge Sort?

```
MERGE(A, p, q, r)
n1 ← q - p + 1
n2 ← r - q
create arrays L[1..n1+1] and R[1..n2+1]
for i ← 1 to n1
do L[i] ← A[p+i-1]
for j ← 1 to n2
do R[j] ← A[q+j]
L[n1+1] ← ∞
R[n2+1] ← ∞
i ← 1
j ← 1
for k ← p to r
do if L[i] ≤ R[j]
then A[k] ← L[i]
i ← i + 1
else A[k] ← R[j]
j ← j + 1
```

O(n)

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11

Merge Sort

```
MERGE-SORT(A, p, r)
∀ p & r are indices into the array (p < r)
if p < r          ∀Check for base case
then q ← ⌊(p + r) / 2⌋  ∀Divide
MERGE-SORT(A, p, q)    ∀Conquer
MERGE-SORT(A, q + 1, r) ∀Conquer
MERGE(A, p, q, r)     ∀Combine
```

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12

Recurrence Relation

- Let $T(n)$ be the time for Merge-Sort to execute on an n element array.
- The time to execute on a one element array is $O(1)$
- Then we have the following relationship:

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13

Merge Sort

- To solve the recurrence relation we'll write n instead of $O(n)$ as it makes the algebra simpler:
 - $T(n) = 2 T(n/2) + n$
 - $T(1) = 1$
- Solve the recurrence by iteration (substitution)
- Use induction to prove the solution is correct

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14

Recurrence Relations to Remember

$T(n) = T(n/2) + O(1)$		
$T(n) = T(n-1) + O(1)$		
$T(n) = 2 T(n/2) + O(1)$		
$T(n) = T(n-1) + O(n)$		
$T(n) = 2 T(n/2) + O(n)$		

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Your Turn

- Solve the following recurrence relation using the expansion (iteration) method
 - $T(n) = T(n-1) + 2n - 1$
 - $T(0) = 0$

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16

Approaches to Algorithm Design

- Incremental
 - Job is partly done – do a little more, repeat until done.
- Divide-and-Conquer (recursive)
 - Divide problem into sub-problems of the same kind.
 - For small subproblems, solve, else, solve them recursively.
 - Combine subproblem solutions to solve the whole thing.

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17

For Next Time

- So far we've covered chapters 1, 2, and 3.

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18
