



Solving recurrences Master Theorem.

Slide credit: David Luebke (Virginia)

Review: Merge Sort

```
MergeSort(A, left, right) {  
    if (left < right) {  
        mid = floor((left + right) / 2);  
        MergeSort(A, left, mid);  
        MergeSort(A, mid+1, right);  
        Merge(A, left, mid, right);  
    }  
}  
// Merge() takes two sorted subarrays of A and  
// merges them into a single sorted subarray of A.  
// Code for this is in the book. It requires O(n)  
// time, and *does* require allocating O(n) space
```

Review: Analysis of Merge Sort

Statement Effort

```
MergeSort(A, left, right) {                                T(n)
    if (left < right) {
        mid = floor((left + right) / 2);                  Θ(1)
        MergeSort(A, left, mid);                          T(n/2)
        MergeSort(A, mid+1, right);                      T(n/2)
        Merge(A, left, mid, right);                     Θ(n)
    }
}
```

- So $T(n) = \Theta(1)$ when $n = 1$, and
 $2T(n/2) + \Theta(n)$ when $n > 1$
- This expression is a *recurrence*

Recurrences

- The expression:

$$T(n) = \begin{cases} c & n = 1 \\ 2T\left(\frac{n}{2}\right) + cn & n > 1 \end{cases}$$

is a *recurrence*.

- Recurrence: an equation that describes a function in terms of its value on smaller functions

Recurrence Examples

$$s(n) = \begin{cases} 0 & n = 0 \\ c + s(n-1) & n > 0 \end{cases}$$

$$s(n) = \begin{cases} 0 & n = 0 \\ n + s(n-1) & n > 0 \end{cases}$$

$$T(n) = \begin{cases} c & n = 1 \\ 2T\left(\frac{n}{2}\right) + c & n > 1 \end{cases}$$

$$T(n) = \begin{cases} c & n = 1 \\ aT\left(\frac{n}{b}\right) + cn & n > 1 \end{cases}$$

Solving Recurrences

- Substitution method
- Iteration method
- Master method

Solving Recurrences

- The substitution method (CLR 4.1)
 - A.k.a. the “making a good guess method”
 - Guess the form of the answer, then use induction to find the constants and show that solution works
 - Examples:
 - $T(n) = 2T(n/2) + \Theta(n)$ □ $T(n) = \Theta(n \lg n)$
 - $T(n) = 2T(\lfloor n/2 \rfloor) + n$ □ ???

Solving Recurrences

- The substitution method (CLR 4.1)
 - A.k.a. the “making a good guess method”
 - Guess the form of the answer, then use induction to find the constants and show that solution works
 - Examples:
 - $T(n) = 2T(n/2) + \Theta(n) \rightarrow T(n) = \Theta(n \lg n)$
 - $T(n) = 2T(\lfloor n/2 \rfloor) + n \rightarrow T(n) = \Theta(n \lg n)$
 - $T(n) = 2T(\lfloor n/2 \rfloor) + 17 + n \rightarrow ???$

Solving Recurrences

- The substitution method (CLR 4.1)
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 - Guess the form of the answer, then use induction to find the constants and show that solution works
 - Examples:
 - $T(n) = 2T(n/2) + \Theta(n) \rightarrow T(n) = \Theta(n \lg n)$
 - $T(n) = 2T(\lfloor n/2 \rfloor) + n \rightarrow T(n) = \Theta(n \lg n)$
 - $T(n) = 2T(\lfloor n/2 \rfloor + 17) + n \rightarrow \Theta(n \lg n)$

Solving Recurrences

- Another option is what the book calls the “iteration method”
 - Expand the recurrence
 - Work some algebra to express as a summation
 - Evaluate the summation
- We will show several examples

$$s(n) = \begin{cases} 0 & n = 0 \\ c + s(n-1) & n > 0 \end{cases}$$

- $s(n) =$
 $c + s(n-1)$
 $c + c + s(n-2)$
 $2c + s(n-2)$
 $2c + c + s(n-3)$
 $3c + s(n-3)$
 \dots
 $kc + s(n-k) = ck + s(n-k)$

$$s(n) = \begin{cases} 0 & n = 0 \\ c + s(n-1) & n > 0 \end{cases}$$

- So far for $n \geq k$ we have
 - $s(n) = ck + s(n-k)$
- What if $k = n$?
 - $s(n) = cn + s(0) = cn$

$$s(n) = \begin{cases} 0 & n = 0 \\ c + s(n-1) & n > 0 \end{cases}$$

- So far for $n \geq k$ we have
 - $s(n) = ck + s(n-k)$
- What if $k = n$?
 - $s(n) = cn + s(0) = cn$
- So $s(n) = \begin{cases} 0 & n = 0 \\ c + s(n-1) & n > 0 \end{cases}$
- Thus in general
 - $s(n) = cn$

$$s(n) = \begin{cases} 0 & n = 0 \\ n + s(n-1) & n > 0 \end{cases}$$

- $s(n)$

$$= n + s(n-1)$$

$$= n + n-1 + s(n-2)$$

$$= n + n-1 + n-2 + s(n-3)$$

$$= n + n-1 + n-2 + n-3 + s(n-4)$$

$$= \dots$$

$$= n + n-1 + n-2 + n-3 + \dots + n-(k-1) + s(n-k)$$

$$s(n) = \begin{cases} 0 & n = 0 \\ n + s(n-1) & n > 0 \end{cases}$$

- $s(n)$

$$= n + s(n-1)$$

$$= n + n-1 + s(n-2)$$

$$= n + n-1 + n-2 + s(n-3)$$

$$= n + n-1 + n-2 + n-3 + s(n-4)$$

$$= \dots$$

$$= n + n-1 + n-2 + n-3 + \dots + n-(k-1) + s(n-k)$$

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

$$s(n) = \begin{cases} 0 & n = 0 \\ n + s(n-1) & n > 0 \end{cases}$$

- So far for $n \geq k$ we have

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

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- So far for $n \geq k$ we have

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

- What if $k = n$?

$$\sum_{i=1}^n i + s(0) = \sum_{i=1}^n i + 0 = n \frac{n+1}{2}$$

$$s(n) = \begin{cases} 0 & n = 0 \\ n + s(n-1) & n > 0 \end{cases}$$

- So far for $n \geq k$ we have

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

- What if $k = n$?

$$\sum_{i=1}^n i + s(0) = \sum_{i=1}^n i + 0 = n \frac{n+1}{2}$$

- Thus in general

$$s(n) = n \frac{n+1}{2}$$

$$T(n) = \begin{cases} c & n = 1 \\ 2T\left(\frac{n}{2}\right) + c & n > 1 \end{cases}$$

- $T(n) =$
 $2T(n/2) + c$
 $2(2T(n/2/2) + c) + c$
 $2^2T(n/2^2) + 3c$
 $2^2(2T(n/2^2/2) + c) + 3c$
 $2^3T(n/2^3) + 4c + 3c$
 $2^3(2T(n/2^3/2) + c) + 7c$
 $2^4T(n/2^4) + 15c$
 \dots
 $2^kT(n/2^k) + (2^k - 1)c$

$$T(n) = \begin{cases} c & n = 1 \\ 2T\left(\frac{n}{2}\right) + c & n > 1 \end{cases}$$

- So far for $n > 2k$ we have

☞ $T(n) = 2^k T(n/2^k) + (2^k - 1)c$

- What if $k = \lg n$?

☞ $T(n) = 2^{\lg n} T(n/2^{\lg n}) + (2^{\lg n} - 1)c$

$$= n T(n/n) + (n - 1)c$$

$$= n T(1) + (n-1)c$$

$$= nc + (n-1)c = (2n - 1)c$$

Generalize previous result

$$T(n) = \begin{cases} c & n = 1 \\ aT\left(\frac{n}{b}\right) + cn & n > 1 \end{cases}$$

- More general recurrence than last one
- Previous example: $a=b=2$.
- New feature: *result depends on whether $a < b$, $a=b$ or $a > b$!*

$$T(n) = \begin{cases} c & n = 1 \\ aT\left(\frac{n}{b}\right) + cn & n > 1 \end{cases}$$

- $T(n) =$
- $aT(n/b) + cn$
- $a(aT(n/b/b) + cn/b) + cn$
- $a^2T(n/b^2) + cna/b + cn$
- $a^2T(n/b^2) + cn(a/b + 1)$
- $a^2(aT(n/b^2/b) + cn/b^2) + cn(a/b + 1)$
- $a^3T(n/b^3) + cn(a^2/b^2) + cn(a/b + 1)$
- $a^3T(n/b^3) + cn(a^2/b^2 + a/b + 1)$
- ...
- $a^kT(n/b^k) + cn(a^{k-1}/b^{k-1} + a^{k-2}/b^{k-2} + \dots + a^2/b^2 + a/b + 1)$

$$T(n) = \begin{cases} c & n = 1 \\ aT\left(\frac{n}{b}\right) + cn & n > 1 \end{cases}$$

- So we have
 - $T(n) = a^k T(n/b^k) + cn(a^{k-1}/b^{k-1} + \dots + a^2/b^2 + a/b + 1)$
- For $k = \log_b n$
 - $n = b^k$
 - $T(n) = a^k T(1) + cn(a^{k-1}/b^{k-1} + \dots + a^2/b^2 + a/b + 1)$
 $= a^k c + cn(a^{k-1}/b^{k-1} + \dots + a^2/b^2 + a/b + 1)$
 $= ca^k + cn(a^{k-1}/b^{k-1} + \dots + a^2/b^2 + a/b + 1)$
 $= cna^k/b^k + cn(a^{k-1}/b^{k-1} + \dots + a^2/b^2 + a/b + 1)$
 $= cn(a^k/b^k + \dots + a^2/b^2 + a/b + 1)$

$$T(n) = \begin{cases} c & n = 1 \\ aT\left(\frac{n}{b}\right) + cn & n > 1 \end{cases}$$

- So with $k = \log_b n$
 - $T(n) = cn(a^k/b^k + \dots + a^2/b^2 + a/b + 1)$
- What if $a = b$?
 - $T(n) = cn(k + 1)$
 $= cn(\log_b n + 1)$
 $= \Theta(n \log n)$

$$T(n) = \begin{cases} c & n = 1 \\ aT\left(\frac{n}{b}\right) + cn & n > 1 \end{cases}$$

- So with $k = \log_b n$
 - $T(n) = cn(a^{k/b} + \dots + a^2/b^2 + a/b + 1)$
- What if $a < b$?

$$T(n) = \begin{cases} c & n = 1 \\ aT\left(\frac{n}{b}\right) + cn & n > 1 \end{cases}$$

- So with $k = \log_b n$
 - $T(n) = cn(a^{k/b} + \dots + a^{2/b} + a/b + 1)$
- What if $a < b$?
 - Recall that $\sum(x^k + x^{k-1} + \dots + x + 1) = (x^{k+1} - 1)/(x - 1)$

$$T(n) = \begin{cases} c & n=1 \\ aT\left(\frac{n}{b}\right) + cn & n>1 \end{cases}$$

- So with $k = \log_b n$
 - $T(n) = cn(a^{k/b} + \dots + a^{2/b^2} + a/b + 1)$
- What if $a < b$?
 - Recall that $(x^k + x^{k-1} + \dots + x + 1) = (x^{k+1} - 1)/(x - 1)$
 - So:

$$\frac{a^k}{b^k} + \frac{a^{k-1}}{b^{k-1}} + \dots + \frac{a}{b} + 1 = \frac{(a/b)^{k+1} - 1}{(a/b) - 1} = \frac{1 - (a/b)^{k+1}}{1 - (a/b)} < \frac{1}{1 - a/b}$$

$$T(n) = \begin{cases} c & n=1 \\ aT\left(\frac{n}{b}\right) + cn & n>1 \end{cases}$$

- So with $k = \log_b n$
 - $T(n) = cn(a^{k/b} + \dots + a^{2/b^2} + a/b + 1)$
- What if $a < b$?
 - Recall that $\sum(x^k + x^{k-1} + \dots + x + 1) = (x^{k+1} - 1)/(x - 1)$

■ So:

$$\frac{a^k}{b^k} + \frac{a^{k-1}}{b^{k-1}} + \dots + \frac{a}{b} + 1 = \frac{(a/b)^{k+1} - 1}{(a/b) - 1} = \frac{1 - (a/b)^{k+1}}{1 - (a/b)} < \frac{1}{1 - a/b}$$

■ $T(n) = cn \cdot \Theta(1) = \Theta(n)$

$$T(n) = \begin{cases} c & n = 1 \\ aT\left(\frac{n}{b}\right) + cn & n > 1 \end{cases}$$

- So with $k = \log_b n$
 - $T(n) = cn(a^{k/b} + \dots + a^2/b^2 + a/b + 1)$
- What if $a > b$?

$$T(n) = \begin{cases} c & n = 1 \\ aT\left(\frac{n}{b}\right) + cn & n > 1 \end{cases}$$

- So with $k = \log_b n$
 - $T(n) = cn(a^{k/b} + \dots + a^2/b^2 + a/b + 1)$

- What if $a > b$?

$$\frac{a^k}{b^k} + \frac{a^{k-1}}{b^{k-1}} + \dots + \frac{a}{b} + 1 = \frac{(a/b)^{k+1} - 1}{(a/b) - 1} = \Theta((a/b)^k)$$

$$T(n) = \begin{cases} c & n = 1 \\ aT\left(\frac{n}{b}\right) + cn & n > 1 \end{cases}$$

- So with $k = \log_b n$
 - $T(n) = cn(a^{k/b} + \dots + a^2/b^2 + a/b + 1)$

- What if $a > b$?

$$\frac{a^k}{b^k} + \frac{a^{k-1}}{b^{k-1}} + \dots + \frac{a}{b} + 1 = \frac{(a/b)^{k+1} - 1}{(a/b) - 1} = \Theta((a/b)^k)$$

- $T(n) = cn \cdot \Theta(a^k / b^k)$

$$T(n) = \begin{cases} c & n = 1 \\ aT\left(\frac{n}{b}\right) + cn & n > 1 \end{cases}$$

- So with $k = \log_b n$
 - $T(n) = cn(a^{k/b} + \dots + a^2/b^2 + a/b + 1)$

- What if $a > b$?

$$\frac{a^k}{b^k} + \frac{a^{k-1}}{b^{k-1}} + \dots + \frac{a}{b} + 1 = \frac{(a/b)^{k+1} - 1}{(a/b) - 1} = \Theta((a/b)^k)$$

- $T(n) = cn \cdot \Theta(a^k / b^k)$
 $= cn \cdot \Theta(a^{\log n} / b^{\log n}) = cn \cdot \Theta(a^{\log n} / n)$

$$T(n) = \begin{cases} c & n = 1 \\ aT\left(\frac{n}{b}\right) + cn & n > 1 \end{cases}$$

- So with $k = \log_b n$
 - $T(n) = cn(a^k/b^k + \dots + a^2/b^2 + a/b + 1)$

- What if $a > b$?

$$\frac{a^k}{b^k} + \frac{a^{k-1}}{b^{k-1}} + \dots + \frac{a}{b} + 1 = \frac{(a/b)^{k+1} - 1}{(a/b) - 1} = \Theta((a/b)^k)$$

- $T(n) = cn \cdot \Theta(a^k / b^k)$
 $= cn \cdot \Theta(a^{\log n} / b^{\log n}) = cn \cdot \Theta(a^{\log n} / n)$

recall logarithm fact: $a^{\log n} = n^{\log a}$

$$T(n) = \begin{cases} c & n = 1 \\ aT\left(\frac{n}{b}\right) + cn & n > 1 \end{cases}$$

- So with $k = \log_b n$
 - $T(n) = cn(a^k/b^k + \dots + a^2/b^2 + a/b + 1)$

- What if $a > b$?

$$\begin{aligned} \frac{a^k}{b^k} + \frac{a^{k-1}}{b^{k-1}} + \dots + \frac{a}{b} + 1 &= \frac{(a/b)^{k+1} - 1}{(a/b) - 1} = \Theta((a/b)^k) \\ \text{■ } T(n) &= cn \cdot \Theta(a^k / b^k) \\ &= cn \cdot \Theta(a^{\log n} / b^{\log n}) = cn \cdot \Theta(a^{\log n} / n) \\ &\quad \text{recall logarithm fact: } a^{\log n} = n^{\log a} \\ &= cn \cdot \Theta(n^{\log a} / n) = \Theta(cn \cdot n^{\log a} / n) \end{aligned}$$

$$T(n) = \begin{cases} c & n = 1 \\ aT\left(\frac{n}{b}\right) + cn & n > 1 \end{cases}$$

- So with $k = \log_b n$
 - $T(n) = cn(a^k/b^k + \dots + a^2/b^2 + a/b + 1)$

- What if $a > b$?

$$\begin{aligned} \frac{a^k}{b^k} + \frac{a^{k-1}}{b^{k-1}} + \dots + \frac{a}{b} + 1 &= \frac{(a/b)^{k+1} - 1}{(a/b) - 1} = \Theta((a/b)^k) \\ \text{■ } T(n) &= cn \cdot \Theta(a^k / b^k) \\ &= cn \cdot \Theta(a^{\log n} / b^{\log n}) = cn \cdot \Theta(a^{\log n} / n) \\ &\quad \text{recall logarithm fact: } a^{\log n} = n^{\log a} \\ &= cn \cdot \Theta(n^{\log a} / n) = \Theta(cn \cdot n^{\log a} / n) \\ &= \Theta(n^{\log a}) \end{aligned}$$

$$T(n) = \begin{cases} c & n=1 \\ aT\left(\frac{n}{b}\right) + cn & n>1 \end{cases}$$

- So...

$$T(n) = \begin{cases} \Theta(n) & a < b \\ \Theta(n \log_b n) & a = b \\ \Theta(n^{\log_b a}) & a > b \end{cases}$$

The Master Theorem

- Given: a *divide and conquer* algorithm
 - An algorithm that divides the problem of size n into a subproblems, each of size n/b
 - Let the cost of each stage (i.e., the work to divide the problem + combine solved subproblems) be described by the function $f(n)$
- Then, the Master Theorem gives us a cookbook for the algorithm's running time:

The Master Theorem

- if $T(n) = aT(n/b) + f(n)$ then

$$T(n) = \begin{cases} \Theta(n^{\log_b a}) & f(n) = O(n^{\log_b a - \varepsilon}) \\ \Theta(n^{\log_b a} \log n) & f(n) = \Theta(n^{\log_b a}) \\ \Theta(f(n)) & f(n) = \Omega(n^{\log_b a + \varepsilon}) \text{ AND } af(n/b) < cf(n) \text{ for large } n \end{cases} \quad \left. \begin{array}{l} \varepsilon > 0 \\ c < 1 \end{array} \right\}$$

Using The Master Method

- $T(n) = 9T(n/3) + n$
 - $a=9, b=3, f(n) = n$
 - $n^{\log_b a} = n^{\log_3 9} = \Theta(n^2)$
 - Since $f(n) = O(n^{\log_3 9 - \varepsilon})$, where $\varepsilon=1$, case 1 applies:

$$T(n) = \Theta(n^{\log_b a}) \text{ when } f(n) = O(n^{\log_b a - \varepsilon})$$

- Thus the solution is $T(n) = \Theta(n^2)$

Sorting revisited

- We have seen algorithms for sorting: INSERTION-SORT, MERGESORT
- More generally:
- We are given a sequence of **items**
- Each item has a characteristic called **sorting key**. The values of the sorting key belong to a set on which there exists a total order relationship
- **Sorting the sequence** = arrange its elements such that the sorting keys are in increasing (or decreasing) order

Sorting revisited



Other assumptions:

- We shall consider that the sequence is stored in a random access memory (e.g. in an array)
- This means that we will discuss about **internal sorting**
- We shall analyze only sorting methods which are **in place** (the additional space needed for sorting has at most the size of an element/few elements).
- Stability: preserves **ordering of elements with identical keys**

Stability

Example:

- Initial configuration:

((Adam,9), (John, 10), (Peter,9), (Victor,8))

- Stable sorting :

((John,10),(Adam,9),(Peter,9),(Victor,8))

- Unstable sorting :

((John,10), (Peter,9),(Adam,9), (Victor,8))

Insertion sort – stability

8 **5** 9 8' 1 → 5 8 9 8' 1

5 8 **9** 8' 1 → 5 8 9 8' 1

5 8 9 **8'** 1 → 5 8 8' 9 1

5 8 8' 9 1 → 1 5 8 8' 9

The insertion method is **stable**

Coming up: Sorting methods

- (done) INSERTION-SORT, MERGESORT
- SELECTION-SORT, QUICKSORT
- HEAPSORT
- Sorting in linear time
- Order statistics: median, etc.