3. Recurrences

- A recurrence is an equation defining a function f(n) recursively in terms of smaller values of n.
- E.g., the running time of Merge-Sort, if n is a power of 2, is:

$$T(n) = (1)$$
 if $n = 1$
 $T(n) = 2 T(n/2) + (n)$ if $n > 1$

For arbitrary n > 0, the running time is

$$T(n) = (1)$$
 if $n = 1$
 $T(n) = T(n/2) + T(n/2) + (n)$ if $n > 1$ Why?

- · We use 3 methods for solving recurrences
 - Substitution Method
 - Iteration Method
 - Master Method

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Floors and Ceilings

For any real number x,

x = greatest integer less than or equal to x

x = least integer greater than or equal to x

· For any integer n,

$$n/2 + n/2 = n$$

For integers a 0 and b 0,

Logarithms

• Definition: For any a, b , c:

$$\log_b a = c$$
 $b^c = a$

We use:

lg n = log₂ a (binary logarithm) ln n = log_e a (natural logarithm)

• Properties (writing log for a logarithm with arbitrary base):

$$\begin{array}{lll} a & = & b^{\log_b a} \\ \log (a \ b) = & \log a + \log b \\ \log a^n & = & n \log a \\ \log_b a & = & (\log_c a) / (\log_c b) & (*) \\ \log (1/a) = & -\log a \\ \log_b a & = & 1/\log_a b \\ a^{\log_b n} & = & n^{\log_b a} \end{array}$$

• (*) implies that e.g. (lg n) = ($log_c n$) for any c.

The base of the logarithm is irrelevant for asymptotic analysis!

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Substitution Method ...

- Ü Guess a solution.
- Ü Verify by induction.
- · For example, for

$$T(n) = 2 T(n/2) + n \text{ and } T(1) = 1$$

we guess
$$T(n) = O(n \lg n)$$

Induction Goal:

$$T(n)$$
 c n lg n, for some c and all n > n_0

Induction Hypothesis:

$$T(n/2)$$
 c $n/2$ lg $n/2$

Proof of Induction Goal:

... Substitution Method

- So far the restrictions on c, n_0 are only c 1
- Base Case:

$$T(n_0)$$
 cn $\lg n$

Here, $n_0 = 1$ does not work, since T(1) = 1 but $c 1 \lg 1 = 0$.

However, taking $n_0 = 2$ we have:

$$T(2) = 4$$
 2 lg 2 = 2

SO

holds provided c 2.

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Summations...

· Linearity:

$$(k \mid 1 \mid k \mid n \cdot ca_k + b_k)$$

= $c(k \mid 1 \mid k \mid n \cdot a_k) + (k \mid 1 \mid k \mid n \cdot b_k)$

Use for asymptotic notation:

$$(k | 1 k n \cdot (f(k))) = (k | 1 k n \cdot f(k))$$

In this equation, the -notation on the left hand side applies to variable k, but on the right-hand side, it applies to n.

· Arithmetic Series:

$$(k | 1 k n \cdot k) = n(n+1)/2$$

= (n^2)

• Geometric (or Exponential) Series: If x = 1 then

$$(k \mid 0 \quad k \quad n \cdot x^{k}) = (x^{n+1} - 1) / (x - 1)$$

... Summations

• Infinite Decreasing Geometric Series: If |x| < 1 then

$$(k \mid 0 \quad k \leftarrow x^k) = 1/(1-x)$$

· Harmonic Series:

$$H_n = 1+1/2+1/3+...+1/n$$

= (k | 1 k n · 1 / k)
= ln n + O(1)

• Further series obtained by integrating or differentiating the formulas above.

For example, by differentiating the infinite decreasing geometric series and multiplying with \boldsymbol{x} we get:

$$(k \mid 0 \quad k \leftarrow k \times^k) = x / (1 - x)^2$$

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Iteration Method ...

- Ü Express the recurrence as a summation of terms.
- $\ddot{\text{U}}$ Use techniques for summations.

T(n) = 3 T(n/4) + n

· For example, we iterate

- The i-th term in the series is $3^i \ \ n \ / \ 4^i$.

We have to iterate until $n / 4^i = 1$, since T(1) = (1), or equivalently until $i > \log_4 n$.

... Iteration Method

· We continue:

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T(n) = n+3 \ n/4 + 9 \ n/16 + 27 \ T(n/64) 
n+3 \ n/4 + 9 \ n/16 + 27 \ n/64 + ... + 3^{\log_4 n} \quad (1)
\{as\ a^{\log_b n} = n^{\log_b a}\}
n(\ i \mid 0 \ i < \cdot (3/4)^i) + (n^{\log_4 3})
\{decreasing\ geometric\ series: (k \mid 0 \ k < \cdot x^k) = 1/(1-x)\}
4 \ n+ (n^{\log_4 3})
\{\log_4 3 < 1\}
= 4 \ n+o(n)
= O(n)
```

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The Master Method

- Let a 1 and b > 1 be constants and f(n) be a function. Assume

$$T(n) = a T(n/b) + f(n)$$

where n/b stands for n/b or n/b. Then

- $T(n) = (n^{\log_b a})$ if $f(n) = O(n^{\log_b a})$ for some > 0,
- $T(n) = (n^{\log_b a} \lg n) \text{ if } f(n) = (n^{\log_b a})$
- T(n) = (f(n)) if $f(n) = (n^{\log_b a^+})$ for some > 0 and if a f(n/b) c f(n) for some c < 1 and sufficiently large n.

 Note 1: This theorem can be applied to divide-and-conquer algorithms, which are all of the form

$$T(n) = a T(n/b) + D(n) + C(n)$$

where D(n) is the cost of dividing and C(n) the cost of combining.

Note 2: Not all possible cases are covered by the theorem.

Merge Sort with the Master Method

• For arbitrary n > 0, the running time of Merge-Sort is

$$T(n) = (1)$$

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

We can approximate this from below and above by

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

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

respectively. According to the Master Theorem, both have the same solution which we get by taking

$$a = 2, b = 2, f(n) = (n)$$
.

Since $n = n^{\log_2 2}$, the second case applies and we get:

$$T(n) = (n \lg n)$$

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Binary Search with the Master Method

- The Master Theorem allows us to ignore the floor or ceiling function around n/b in T(n/b) in general.
- · Binary Search has for any n > 0 a running time of

$$T(n) = T(n/2) + (1)$$
.

Hence a = 1, b = 2, f(n) = (1). Since 1 = $n^{\log_2 1}$ the second case applies and we get:

$$T(n) = (lg n)$$

Hanoi with the Master Theorem

· Hanoi has for any n > 0 a running time of

$$T(n) = 2 T(n-1) + 1$$
.

In order to bring this into a form such that the Master Theorem is applicable, we rename $n = \lg m$:

$$T(lg m) = 2 T(lg m - 1) + 1$$

= 2 T(lg m - lg 2) + 1
= 2 T(lg (m/2)) + 1

Defining S(m) = T(lg m) we get the new recurrence:

$$S(m) = 2 S(m/2) + 1$$

Hence a = 2, b = 2, f(m) = 1. Since 1 = $m^{log_2 2-1}$ the first case applies with = 1 and we get:

$$S(m) = (m)$$

With S(m) = T(lg m) and n = lg m we finally get:

$$T(n) = (2^n)$$