

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:

$$\begin{aligned}T(n) &= (1) && \text{if } n = 1 \\T(n) &= 2 T(n/2) + (n) && \text{if } n > 1\end{aligned}$$

For arbitrary $n > 0$, the running time is

$$\begin{aligned}T(n) &= (1) && \text{if } n = 1 \\T(n) &= T(\lfloor n/2 \rfloor) + T(\lceil n/2 \rceil) + (n) && \text{if } n > 1\end{aligned}$$

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 ,
 - $\lfloor x \rfloor$ = greatest integer less than or equal to x
 - $\lceil x \rceil$ = least integer greater than or equal to x
- For any integer n ,
$$\lfloor n/2 \rfloor + \lceil n/2 \rceil = n$$
- For integers $a > 0$ and $b > 0$,
$$\lfloor n/a \rfloor / b = \lfloor n/(a b) \rfloor$$
$$\lceil n/a \rceil / b = \lceil n/(a b) \rceil$$

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Logarithms

- Definition: For any a, b, c :

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

- We use:

$$\lg n = \log_2 a \quad (\text{binary logarithm})$$

$$\ln n = \log_e a \quad (\text{natural logarithm})$$

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

$$a = b^{\log_b a}$$

$$\log(a \cdot b) = \log a + \log b$$

$$\log a^n = n \log a$$

$$\log_b a = (\log_c a) / (\log_c b) \quad (*)$$

$$\log(1/a) = -\log a$$

$$\log_b a = 1/\log_a b$$

$$a^{\log_b n} = n^{\log_b a}$$

- (*) 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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Forward 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:

$$T(n) \leq c n \lg n, \text{ for some } c \text{ and all } n > n_0$$

- Induction Hypothesis:

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

- Proof of Induction Goal:

$$\begin{aligned} T(n) &= 2 T(n/2) + n \\ &\leq 2(c (n/2) \lg (n/2)) + n \\ &= c n \lg (n/2) + n \\ &= c n \lg n - c n \lg 2 + n \\ &= c n \lg n - c n + n \\ &= c n \lg n \end{aligned}$$

provided $c \geq 1$

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... Forward Substitution Method

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

$$T(n_0) = c n \lg n$$

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

However, taking $n_0 = 2$ we have:

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

so

$$T(2) = c \cdot 2$$

holds provided $c \geq 2$.

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Summations...

- Linearity:

$$\begin{aligned} \sum_{k=1}^n (c a_k + b_k) \\ = c \sum_{k=1}^n a_k + \sum_{k=1}^n b_k \end{aligned}$$

Use for asymptotic notation:

$$\sum_{k=1}^n (f(k)) = \sum_{k=1}^n f(k)$$

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

- Arithmetic Series:

$$\begin{aligned} \sum_{k=1}^n k &= n(n+1)/2 \\ &= \Theta(n^2) \end{aligned}$$

- Geometric (or Exponential) Series: If $x \neq 1$ then

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

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... Summations

- Infinite Decreasing Geometric Series: If $|x| < 1$ then

$$\sum_{k=0}^{\infty} k \cdot x^k = 1 / (1 - x)$$

- Harmonic Series:

$$\begin{aligned} H_n &= 1 + 1/2 + 1/3 + \dots + 1/n \\ &= \sum_{k=1}^n 1/k \\ &= \ln n + O(1) \end{aligned}$$

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

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

$$\sum_{k=0}^{\infty} k \cdot x^k = x / (1 - x)^2$$

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Iteration (Backward Substitution) Method ...

Ü Express the recurrence as a summation of terms.

Ü Use techniques for summations.

- For example, we iterate

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

as follows:

$$\begin{aligned} T(n) &= n + 3 T(n/4) \\ &= n + 3 (n/4 + 3 T(n/16)) \\ &= n + 3 (n/4 + 3 (n/16 + 3 T(n/64))) \\ &= n + 3 n/4 + 9 n/16 + 27 T(n/64) \end{aligned}$$

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

We have to iterate until $n / 4^i = 1$, since $T(1) = \Theta(1)$,

or equivalently until $i > \log_4 n$.

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... Iteration (Backward Substitution) Method

- We continue:

$$\begin{aligned}
 T(n) &= n + 3n/4 + 9n/16 + 27T(n/64) \\
 &= n + 3n/4 + 9n/16 + 27n/64 + \dots + 3^{\log_4 n} \quad (1) \\
 &\quad \{\text{as } a^{\log_b n} = n^{\log_b a}\} \\
 &= n \sum_{i=0}^{\log_4 n} (3/4)^i + (n^{\log_4 3}) \\
 &\quad \{\text{decreasing geometric series:} \\
 &\quad (k | 0 \leq k < \infty) \cdot x^k = 1/(1-x)\} \\
 &= 4n + (n^{\log_4 3}) \\
 &\quad \{\log_4 3 < 1\} \\
 &= 4n + o(n) \\
 &= O(n)
 \end{aligned}$$

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The Master Theorem

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

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

where n/b stands for $\lfloor n/b \rfloor$ or $\lceil n/b \rceil$. Then

- $T(n) = \Theta(n^{\log_b a})$ if $f(n) = O(n^{\log_b a - \epsilon})$ for some $\epsilon > 0$,
 - $T(n) = \Theta(n^{\log_b a} \lg n)$ if $f(n) = \Theta(n^{\log_b a})$
 - $T(n) = \Theta(f(n))$ if $f(n) = \Omega(n^{\log_b a + \epsilon})$ for some $\epsilon > 0$ 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

$$T(n) = aT(n/b) + D(n) + C(n)$$
 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.

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Merge Sort with the Master Theorem

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

$$T(n) = \Theta(1) \quad \text{if } n = 1$$

$$T(n) = T(n/2) + T(n/2) + \Theta(n) \quad \text{if } n > 1$$

We can approximate this from below and above by

$$T(n) = 2T(n/2) + \Theta(n) \quad \text{if } n > 1$$

$$T(n) = 2T(n/2) + \Theta(n) \quad \text{if } n > 1$$

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

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

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

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

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Binary Search with the Master Theorem

- 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

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

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

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

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Towers of Hanoi with the Master Theorem (a bit odd application)

- The Towers of Hanoi algorithm 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 $n = \lg m$:

$$\begin{aligned} T(\lg m) &= 2 T(\lg m - 1) + 1 \\ &= 2 T(\lg m - \lg 2) + 1 \\ &= 2 T(\lg (m/2)) + 1 \end{aligned}$$

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 $\epsilon = 1$ and we get:

$$S(m) = \Theta(m)$$

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

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