

CS 320 Fall 2023
Solving recurrences
for Divide & Conquer

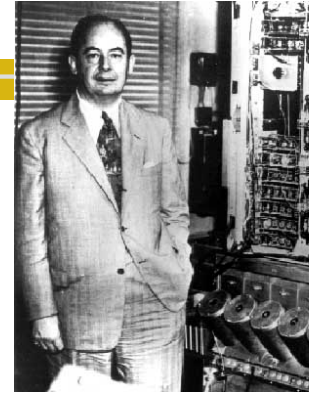
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Divide & Conquer

- Break up the problem into (multiple, smaller) parts
- Solve each of the parts recursively
- Combine the solution of each of the parts into a solution of the original problem

First example: Merge sort

- Divide the array into two halves
- Recursively sort each half
- Merge the two sorted halves



John von Neumann (1945)

Analysis

Divide $O(1)$

Merge $O(n)$

What about the recursive calls?

$$2T\left(\frac{n}{2}\right)$$

A L G O R I T H M S

A L G O R I T H M S

A G L O R H I M S T

A G H I L M O R S T

Complexity of merge

- Time: $O(n)$
- Space: $O(n)$
 - Often with two arrays of length n
 - Can you do (a constant factor) better?

Recurrence relations

- A recurrence relation for a sequence, $\{a_n\}$ is an equation that expresses a_n in terms of one or more of the previous elements of the sequence, a_1, a_2, \dots, a_{n-1}
- A special kind of recursive function
- There may be *base cases*, and the equation holds for $n \geq n_0$ for some constant n_0
 - Example: $a_n = 2a_{n-1} + 1$ and $a_1 = 1$
 - After setting up the recurrence relation, we *solve* it

Recurrence relation for Merge-sort

- *Define* the number of comparisons to sort an input of length n as: $T(n)$
- Use the *structure* of the D&C algorithm to define an equation/relation for $T(n)$

$$T(n) \leq \begin{cases} c & \text{if } n = 1 \\ T\left(\left\lfloor \frac{n}{2} \right\rfloor\right) + T\left(\left\lceil \frac{n}{2} \right\rceil\right) + cn & \text{otherwise} \end{cases}$$

Solving the Recurrence

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

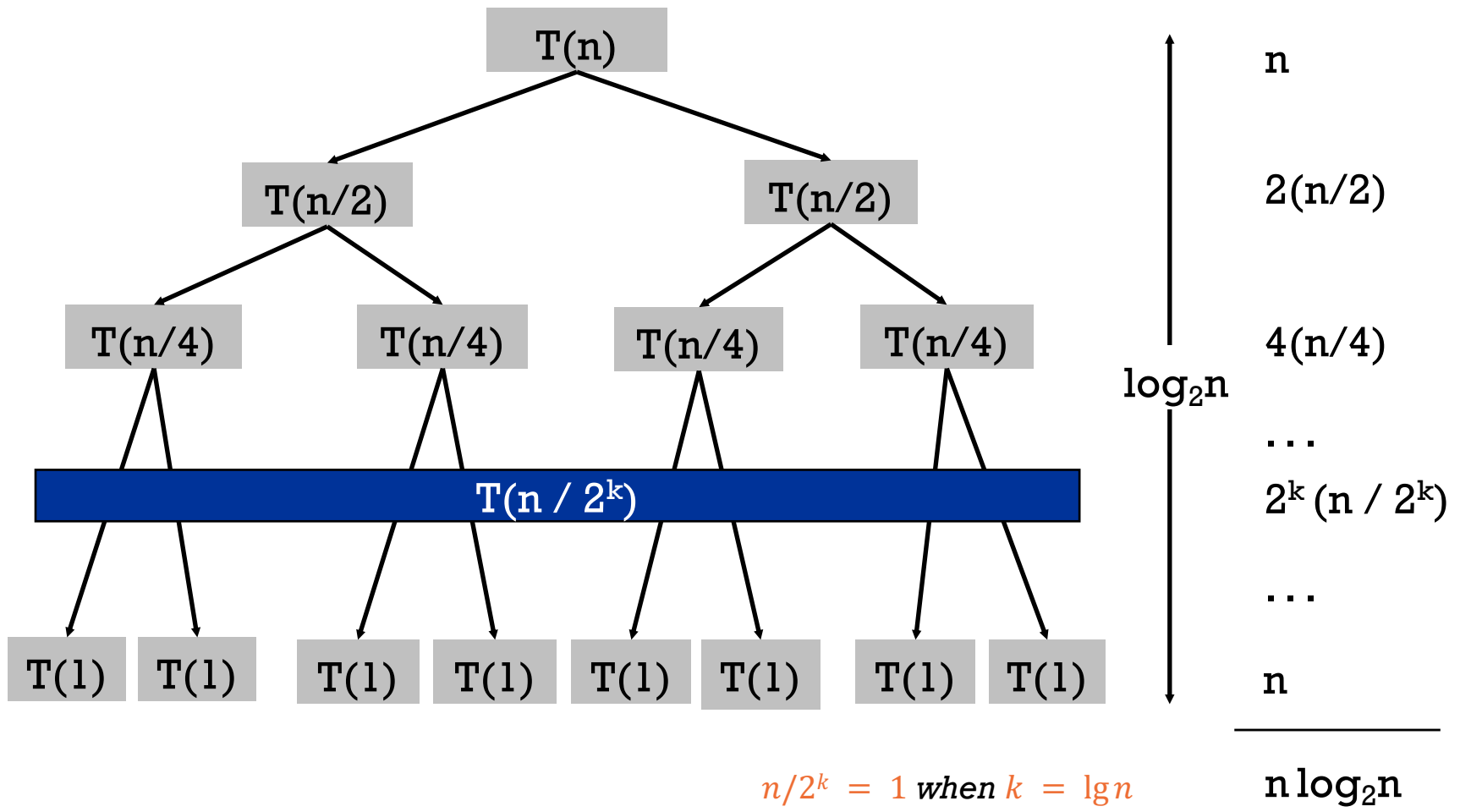
- Solution (*closed form*):

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

- Number of techniques

- Unrolling the recurrence
- Repeated substitution
- See a pattern, guess (i.e., make a hypothesis), and then, prove by induction

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



Seeing the pattern

- What is the “*label*” of each node?
- When does the label become “*small enough*” (base case)
- How many levels in the tree? [Hint: use the above two]
- How many nodes at each level?
- What is the “*contribution*” of each node?
- What is the contribution of *each level*?
- How many *leaves*?
- *Contribution of the leaves* (different from contribution of other levels)

Repeated substitution for $T(n) = \begin{cases} c & \text{if } n = 1 \\ 2T\left(\frac{n}{2}\right) + cn & \text{otherwise} \end{cases}$

■ **Claim:** $T(n) = cn \log_2 n$

$$\begin{aligned} T(n) &= 2T(n/2) + cn \\ &= 4T(n/4) + cn + 2cn/2 \\ &= 8T(n/8) + cn + cn + 4cn/4 \\ &\dots \\ &= 2^{\log_2 n} T(1) + \underbrace{cn + \dots + cn}_{\log_2 n} \\ &= O(n \log_2 n) \end{aligned}$$

← This reaches $T(1)$ when
 $n = 2^{\lg n}$
by definition of $\lg n$

Binary search

```
function BS(x, start, end)
  if (end <= start)
    return A[start]
  mid = (end + start)/2
  if A[mid] < x
    return BS(x, mid, end)
  return BS(x, start, mid-1)
```

- What is the recurrence?
- Apply repeated substitution (on doc cam or exercise)

Find max in an unsorted array

Algorithm:

- Base case $n=1$
- Otherwise: find the max of the two halves, and return the max of that

```
function FM(start, end)
  if (end = start)
    return A[start]
  mid = (end + start)/2
  return max( FM(start, mid-1), FM(mid, end) )
```

Find max in an unsorted array

Recurrence: base case: $T(1) = 0$

$$\begin{aligned}\text{Otherwise: } T(n) &= 2T\left(\frac{n}{2}\right) + 1 \\ &= 4T\left(\frac{n}{4}\right) + 2 + 1 \\ &= 8T\left(\frac{n}{8}\right) + 4 + 2 + 1 \\ &\vdots \\ &= 2^k T\left(\frac{n}{2^k}\right) + 2^{k-1} + 2^{k-2} + \dots + 2^0 \\ &= 2^k T\left(\frac{n}{2^k}\right) + 2 \cdot 2^{k-1} - 1 \\ &= 2^k T\left(\frac{n}{2^k}\right) + 2^k - 1\end{aligned}$$

Base case is reached when $2^k = n$, i.e., $k = \lg n$, So

$$T(n) = 0 + 2^{\lg n} - 1 = n - 1$$

Another example

```
function foo(A, B) // the size of A is n
  if (n == 1):
    return fuzz(A, B) // base case, fuzz is
    constant time

// Process A to build two parts, A0 and A1 of
// size n/2 each

C0 = foo (A0, B)
C1 = foo (A1, B)
return buzz(C0, C1) // buzz is O(n2)
```

General Divide & Conquer

```
function foo(parameters) // the size of A is n
  if (n <= b):           // base case
    return fuzz(A, B)   // constant time
// Divide input into a parts, each of size n/b
  divide()
// Make a calls to
  foo(new parameters) // size is n/b
  return combine(r1, ..., ra)
// Complexity of divide and combine is  $O(n^d)$ 
```

Master Theorem

- Let $a \geq 1, b > 1, n = b^k$ and $T(n)$ be given by

$$T(n) = aT\left(\frac{n}{b}\right) + cn^d$$

- The solution of the recurrence is

$$T(n) = \begin{cases} O(n^d) & \text{if } a < b^d \\ O(n^d \log n) & \text{if } a = b^d \\ O(n^{\log_b a}) & \text{if } a > b^d \end{cases}$$

Merge-sort by master theorem

- $a = 2, b = 2, d = 1$

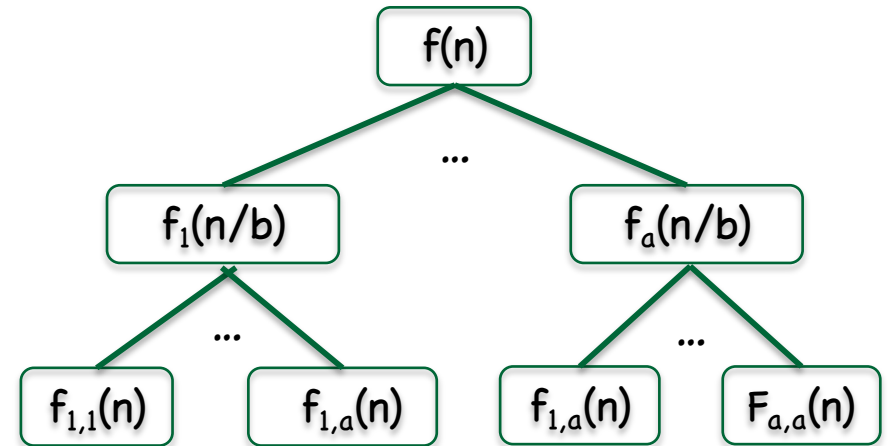
- So, $a = 2$, and $b^d = 2$

... and the solution is

$$T(n) = O(n^d \log n) = O(n \log n)$$

Divide & Conquer call tree

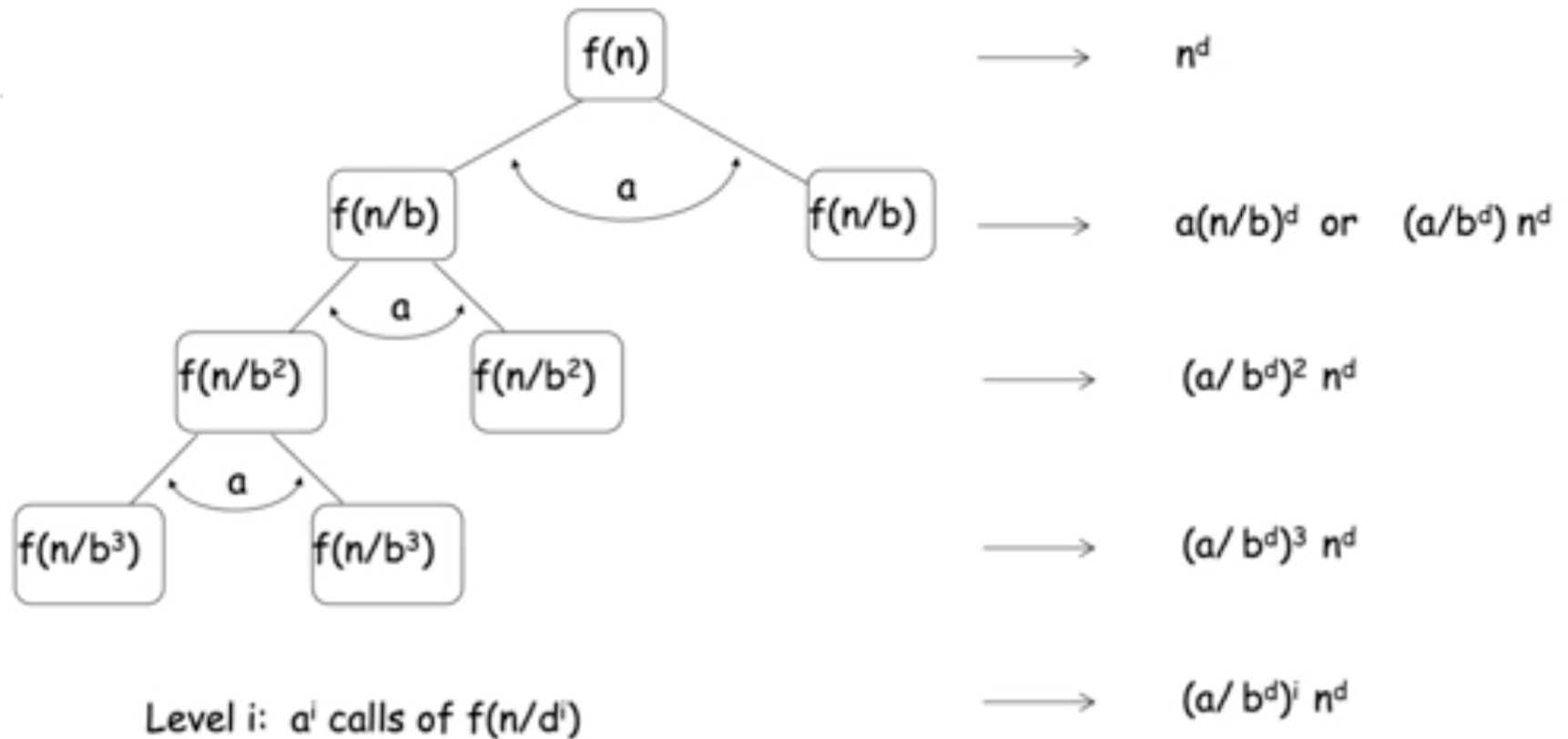
```
Function foo(A) //size n
  if (n <= b) return (base(A))
  A1 ... Aa = divide() // size n/b
  // Recurse
  C1 = foo(A1)
  ⋮
  Ca = foo(Aa)
  return combine(C1, ..., Ca)
```



- Base is constant time
- Divide and combine takes $O(n^d)$

$$f(n) = af(n/b) + n^d$$

$f(1) = c$ ← does not play a role, as we only care about O



Stops when $n/b^i = 1$ i.e. $i = \log_b n$

$$n^d \sum_{i=0}^{\log_b n} (a/b^d)^i$$

Three Cases for $r = (a/b^d)$

Geometric series: $\sum_{i=0}^k r^i = \frac{r^{k+1}-1}{r-1}$ Here $r = (a/b^d)$

1. $r < 1$ e.g. $r = \frac{1}{2}$ $1 + 1/2 + 1/4 + \dots < 2$ for any k
2. $r = 1$ $\sum_{i=0}^k 1^i = k+1 = O(k)$
3. $r > 1$ e.g. $r = 2$ $1 + 2 + 4 + \dots + 2^k = 2^{k+1}-1 = O(2^k)$

The three cases in practice

$$T(n) = 2T(n/2) + n \quad // \text{ mergesort}$$

$$r = 1 \quad a=2, b=2, d=1 \quad r = a/b^d=1 \quad n^1 \sum_{i=0}^{\log n} 1^i = n (\log n + 1) \\ T(n) = O(n \log n)$$

$$T(n) = 2T(n/2) + 1 \quad // \text{ e.g. recursive max in array size } n:$$

if $n=1$, then the element is the max.

$r > 1$ else divide array in 2 halves, find max of each and choose max of the two

$$a=2, b=2, d=0 \quad r = a/b^d=2 \quad n^0 \sum_{i=0}^{\log n} 2^i = (2^{\log n + 1} - 1)/(2 - 1) = (2n - 1)/1 \\ T(n) = O(n)$$

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

$$r < 1 \quad a=2, b=2, d=2 \quad r = a/b^d=1/2 \quad n^2 \sum_{i=0}^{\log n} \left(\frac{1}{2}\right)^i = n^2 (1 + 1/2 + 1/4 + \dots) < 2 n^2 \\ T(n) = O(n^2)$$

Draw trees for these and do the analysis, as in slides 9, 10, 11

Towers of Hanoi

- Move all disks to third peg, without ever placing a larger disk on a smaller one.
- What's the recurrence relation? $a_n = 2a_{n-1} + 1$ with the base case that $a_1 = 1$
- Let's solve by repeated substitution
 - Plug in the definition
 - Do the algebra to collect all the non-recursive expressions together
 - Identify a pattern
 - Determine how many times the pattern occurs until we hit the base case

Hanoi by repeated substitution

$$\begin{aligned}T(n) &= 2T(n - 1) + 1 \\&= 2(2T(n - 2) + 1) + 1 \\&= 4T(n - 2) + 2 + 1 \\&= 4(2T(n - 3) + 1) + 2 + 1 \\&= 8T(n - 3) + 4 + 2 + 1\end{aligned}$$

- What is the label and how is it changing?
- What about the other terms?
- When do we hit the base case?

Hanoi by repeated substitution

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

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

$$= 4T(n - 2) + 2 + 1$$

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

$$= 8T(n - 3) + 4 + 2 + 1$$

⋮

$$= 2^i T(n - i) + \sum_{j=0}^{i-1} 2^j$$

- When does the label become 1?
- When $i = n - 1$ So our solution is

Hanoi by repeated substitution

$$\begin{aligned} T(n) &= 2^{n-1}T(1) + \sum_{j=0}^{n-2} 2^j \\ &= \sum_{j=0}^{n-1} 2^j = 2^n - 1 = \Theta(2^n) \end{aligned}$$

- This is a geometric series
- The Master Theorem does not apply