

Announcements

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- Homework 1 is out!
- Due February 23 (one week) at midnight anywhere on Earth

Recurrence Relations

Recurrence Relations

- A **recurrence relation** is an equation that recursively defines a function's values in terms of earlier values
- Very useful for analyzing an algorithm's running time!

not pseudocode!

Recurrence Relations in Code

```
def betterPower(x, n):
```

```
    if n == 0:
```

```
        return 1
```

```
    else if n == 1:
```

```
        return x
```

```
    else if n % 2 == 0:
```

```
        return betterPower(x * x, n/2) if n > 1:
```

$T(n)$ = running time for
bP on value n
(for arbitrary x)

if $n=0$: $T(0) = c_1$
if $n=1$: $T(1) = c_2$

$T(n) = c_3 + T(n/2)$

(assume for simplicity that n is a power of 2)

How can we write the running time?

Recurrence Relations in Code

$$T(0) = c_1, \quad T(1) = c_2, \quad \dots, \quad T(n) = c_3 + T(n/2)$$

$$\begin{aligned} T(n) &= c_3 + T(n/2) \\ &= c_3 + c_3 + T(n/4) \\ &= c_3 + c_3 + c_3 + T(n/8) \end{aligned}$$

.....

$$= kc_3 + T(n/(2^k))$$

$$= c_3 * \log n$$

+ c₁

What should k be in order for us to get down to $T(1)$?

$$2^k = n, \text{ so } k = \log n$$

↗

Solving Recurrences

- Solving recurrence relations is like integrating an expression- there are tricks, but no techniques are guaranteed to work

Solving Recurrences

- Simplest method: Guess the solution, prove with induction
- Sometimes it helps to do a few expansions to get some intuition

Bounding Recurrences: the Master Method

Suppose $T(n) = aT(\lceil \frac{n}{b} \rceil) + O(n^d)$,
for $a > 0$, $b > 1$, $d \geq 0$.

"ceiling" symbol:
means to round
up

"other work"
besides recursion

$\Theta(\log_a x) =$
 $O(\log_b x)$
↑

- Case 1: If $d > \log_b a$, then $T(n) = O(n^d)$
- Case 2: If $d = \log_b a$, then $T(n) = O(n^d \log n)$
- Case 3: If $d < \log_b a$, then $T(n) = O(n^{\log_b a})$

Note that this doesn't actually SOLVE the recurrence!

Example

Bound the recurrence $T(n) = T\left(\left\lfloor \frac{n}{2} \right\rfloor\right) + 3n^3 + 2$.

Parameters:

$$a = 1$$

$$b = 2$$

$$d = 3$$

$$\text{Case} = \log_2 1 = 0 < 3$$

Solution =

Case 1

$$T(n) = O(n^3)$$

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

- Case 1: If $d > \log_b a$, then $T(n) = O(n^d)$
- Case 2: If $d = \log_b a$, then $T(n) = O(n^d \log n)$
- Case 3: If $d < \log_b a$, then $T(n) = O(n^{\log_b a})$

$T(1) = ?$
 $O(n^3)$
 d

Example: Binary search

$O(n^d)$ = "other work" ignoring recursion

a = number of rec. calls generated by each function call

Algorithm: Goal is to find the location of a value in a sorted array. Start by looking in the middle; then depending on the value there, look in either the first half or second half, and so on. Divide the array in half each time.

$k=3$

b = size of input in the rec. call

[1 3 4 6 8 9 10]

What is the running time of binary search?

$a=1$
 $b=2$

$T(n)$ = ^{worst case} running time of BinSearch on an array of size n

$$T(n) = 1 \cdot T\left(\frac{n}{2}\right) + O(n^0)$$

Example: Binary search

if other work = $O(n^d)$, then $d=1$

Write the running time as a recurrence relation:

What is the running time?

$a = 1$

$b = 2$

$d = 0$ $O(1)$

Case = $\log_2 1 = 0 = d$

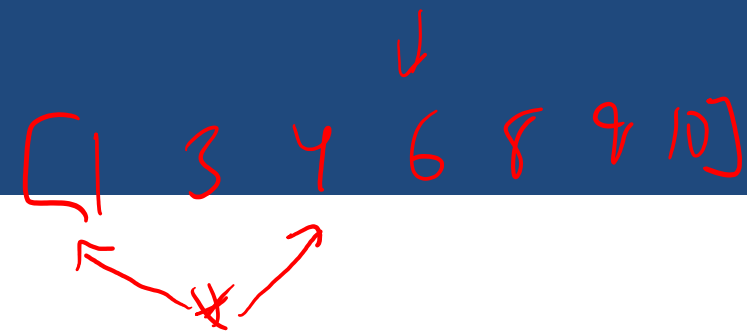
Solution = $O(n^0 \log n)$

case 2 = $O(\log n)$



Bin Search (A, k)
find mid
compare mid to k
recurse on either
left or right
half

Example: Binary search



BinSearch(A, k)

if $\text{len}(A) == 1$: if $A[0] == k$, return that index

else:
branch → get midpoint ← constant time because array
compare mid to k ← constant time
decide which half to branch on
create subarray representing that half
→ recurse on that subarray

In-Class Exercise

$O(n^0) = O(1) = \text{constant}$
 $\log n = O(n)$

$T(n) = 3T(\frac{n}{2}) + \log n$

$a=3$ $d=1$
 $b=2$

Bound the following recurrences:

$T(n) = T(\lceil \frac{n}{3} \rceil) + 2n^2 + 2$

$a=1$ $b=3$ $d=2$ $T(n) = O(n^2)$

$T(n) = 5T(\lceil \frac{n}{2} \rceil) + 5n^4$

$a=5$ $b=2$ $d=4$ $T(n) = O(n^4)$

$T(n) = 2T(\lceil \frac{n}{2} \rceil) + 3n^3 + 2n$

$a=2$ $b=2$ $d=3$ $O(n^3)$

$T(n) = 3T(\lceil \frac{n}{2} \rceil) + 2^n$

$a=3$ $b=2$ $d=\infty$?

$T(n) = aT(\lceil \frac{n}{b} \rceil) + O(n^d)$
<ul style="list-style-type: none"> Case 1: If $d > \log_b a$, then $T(n) = O(n^d)$ Case 2: If $d = \log_b a$, then $T(n) = O(n^d \log n)$ Case 3: If $d < \log_b a$, then $T(n) = O(n^{\log_b a})$

$2^n \neq O(n^d)$ for constant d

MM does not apply

Proof of the Master Method

For simplicity, we assume that n is a power of b. This does not affect the final result, because n is at most a constant factor of b away from a power of b.

$$T(n) = a T\left(\left\lceil \frac{n}{b} \right\rceil\right) + O(n^d)$$

↑

Recursion Tree

#rec. calls

other work

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

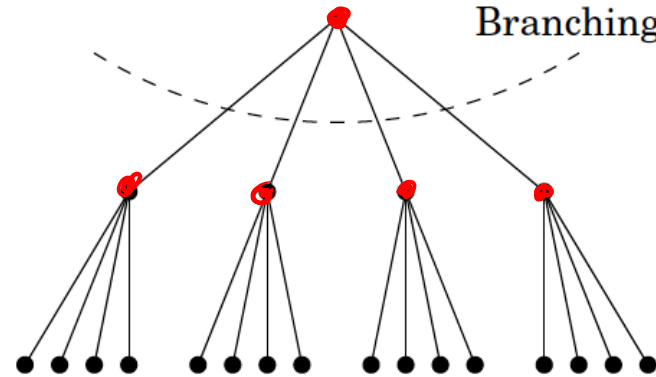
size of each call

Branching factor $a = 4$

Size n

Size n/b

Size n/b^2



⋮

Size 1



Recursion Tree

function(A)
→ some work
→ some recursion

work assoc.
w/ each
call

Levels?

$$\log_b n$$

Problems in level k ?

$$(root = L0) / k$$

Size of problem in level k ?

$$n / b^k$$

Work done in each problem?

$$O\left(\left(\frac{n}{b^k}\right)^d\right)$$

Work done in level k ?

$$O\left(a^k \left(\frac{n}{b^k}\right)^d\right)$$

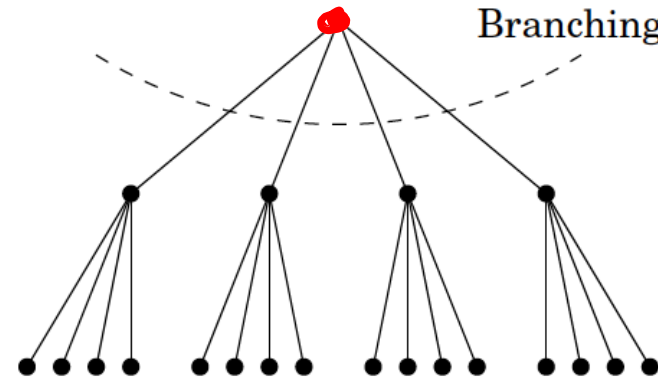
$$T(n) = aT\left(\left\lceil \frac{n}{b} \right\rceil\right) + O(n^d)$$

not same
for all
nodes

Size n

Size n/b

Size n/b^2



⋮

Size 1



Proof of the Master Method

$$\sum_{k=0}^{\log_b n} O\left(a^k \left(\frac{n}{b^k}\right)^d\right) = \text{total running time}$$

$$\sum O\left(a^k \cdot \frac{n^d}{b^{kd}}\right) = n^d \sum_{k=0}^{\log_b n} O\left(\left(\frac{a}{b^d}\right)^k\right)$$

$$r = \left(\frac{a}{b^d}\right)$$

$$= n^d \sum_{k=0}^{\log_b n} O(r^k)$$

Proof of the Master Method

Case 1: $r < 1$, ratio decreases as k goes up
sum is given by $O(n^d \cdot r^0) = O(n^d)$

Case 2: $r = 1$, then $r^k = 1$, so the sum
is given by $O(n^d \cdot \# \text{ terms}) =$
 $O(n^d \log_b n) = O(n^d \log n)$

Proof of the Master Method

Case 3: $r > 1$, r^k increases as k goes up,
sum dominated by last term

$$\begin{aligned} O(n^d \cdot r^{\text{highest } k}) &= O(n^d \cdot r^{\log_b n}) \\ &= O(n^d \cdot \left(\frac{a}{b^d}\right)^{\log_b n}) = O(n^d \cdot \frac{a^{\log_b n}}{(b^{\log_b n})^d}) = \\ &= O(a^{\log_b n}) = O(n^{\log_b a}). \end{aligned}$$

Proof of the Master Method

MM: d vs. $\log_b a$
proof: compared $r = \frac{a}{b^d}$ vs. 1

In-Class Exercise

$T(n)$ = running time on input of size n

Analyze the running time of the following function:

$$T(n) = 3T\left(\frac{n}{3}\right) + O(n)$$

```
function SplitMax(array A):
```

```
    if length(A) == 1:
```

```
        return A[0]
```

```
    else:
```

```
        * A1 = A[0 : length(A)/3]
```

```
        A2 = A[(length(A)/3) + 1 : 2*(length(A)/3)]
```

```
        A3 = A[2*(length(A)/3) + 1 : length(A)]
```

```
        return max(SplitMax(A1), SplitMax(A2), SplitMax(A3))
```

$O(n \log n)$

$x = 0$
* for $i = 1 : \text{len}(A)$

$x += 1$
 $O(n)$

$O(1)$ { $\frac{n}{3}$

$\uparrow O(1)$