

Recurrence relation for  $\overbrace{x_1, \dots, x_n, \dots}^{\text{sequence}}$

"an equation expressing

$x_n$ , for all  $n > n_0$

in terms of (one or more of)

$x_1, \dots, x_{n-1}$ "

EXAMPLE: Triangular numbers

•	• •	• • •	• • • •	• • • • •	• • • • • •			
1	,	3	,	6	,	10	,	?
$x_1$		$x_2$		$x_3$		$x_4$		$x_5$

$$x_n = \sum_{i=1}^n i = \overbrace{\frac{1}{2}(n^2+n)}^{\text{"closed form"}}$$

as you know it

$$\begin{aligned}x_n &= n + x_{n-1} \\x_1 &= 1\end{aligned}$$

L our  $n_0$

as a recurrence

EXAMPLE: Fibonacci numbers

•	•	• •	• • •	• • • •	• • • • •	• • • • • •	• • • • • • •	• • • • • • • •						
0	,	1	,	1	,	2	,	3	,	5	,	8	,	?
$x_1$		$x_2$		$x_3$		$x_4$		$x_5$		...				

$$\begin{aligned}x_n &= x_{n-1} + x_{n-2} \\x_1 &= 0 \\x_2 &= 1\end{aligned}$$

← Recurrence

$$L x_n = \frac{\varphi^n - (1-\varphi)^n}{\sqrt{5}} ; \varphi \approx 1.618$$

closed form

what is  
this number  
called?

## Divide & Conquer recurrence

"a recurrence relation on the form

$$x_n = a x_{\frac{n}{b}} + g(n)$$

for integers  $a, b$  and function  $g$ "

normally written

$$F(n) = a F\left(\frac{n}{b}\right) + g(n)$$

we will only consider cases where

$$g \in O(n)$$

that is, relations on the form

$$F(n) = a F\left(\frac{n}{b}\right) + cn$$

How to "solve" (bring to closed form) these?

- I will show 2 special cases.
- You want more : see "Master Theorem".

EXAMPLE: "Binary search"

$$B(n) = B\left(\frac{n}{2}\right) + 1$$

$$B(1) = 1$$

"Half-decrease" recurrence — unofficial name

Given

$T$  : running time of program  $P$  (function of  $n$ )

if

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

then  $T \in O(\log_2 n)$  ( $P$  runs in  $O(\log n)$  time)

Why: let's unroll the recurrence.

• Analyze first few levels:

level 1 :	problem size $n$ ,	contributes $c$
" 2 :	" $\frac{n}{2}$ ,	" $c$
" 3 :	" $\frac{n}{4}$ ,	" $c$

• identify Pattern

level  $j$  : "  $\frac{n}{2^j}$ , "

• summing over levels

$n$  reduces to 2 in  $\log_2 n$  levels.

Each level contributes  $c$ .

$\Rightarrow$  Total running time at most  $c \cdot \log_2 n \in O(\log n)$

In short, (assuming  $n=2^d$ )

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

$$= T\left(\frac{n}{2^2}\right) + 2c$$

= ...

$$= T\left(\frac{n}{2^{\log_2 n}}\right) + (\log_2 n)c$$

$$= T(1) + (\log_2 n)c \in O(\log n)$$

constant

TELESCOPING  
PROOF

Formal proof: induction in  $n$ .

 EXAMPLE: "Merge sort"

$$\begin{aligned} M(n) &\leq 2M\left(\frac{n}{2}\right) + cn \\ M(2) &\leq c \end{aligned}$$

"Half-divide" recurrence

unofficial name

if

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

then  $T \in \Theta(n \log n)$

Why:

$$\begin{aligned} T(n) &= T\left(\frac{n}{2^0}\right) = 2T\left(\frac{n}{2^1}\right) + cn \\ &= 2\left(2T\left(\frac{n}{2^2}\right) + \frac{1}{2}cn\right) + cn \\ &= 2\left(2T\left(\frac{n}{2^2}\right)\right) + 2 \cdot \frac{1}{2}cn + cn \\ &= 2\left(2T\left(\frac{n}{2^2}\right)\right) + cn + cn \\ &= 2\left(2T\left(\frac{n}{2^2}\right)\right) + 2cn \\ &= 2^{②}T\left(\frac{n}{2^2}\right) + ②cn \\ &= \dots = 2^{\log_2 n} \textcircled{T(1)} + \log_2 n \cdot cn \\ &= n \textcircled{T(1)}^{\text{constant}} + cn \log_2 n \in \Theta(n \log n) \end{aligned}$$

## Solved Exercise 1

Given:

$A$ : array of  $n$  numbers

$p$ : peak index (not given) we just know it exists)

Values in  $A[1], \dots, A[p]$  are increasing,  
Values in  $A[p], \dots, A[n]$  are decreasing.

Goal: Show how  $p$  can be found reading at most  $\mathcal{O}(\log n)$  entries of  $A$ .

Solution: We obtain  $\mathcal{O}(\log n)$  by

- i) Doing constant work,
  - ii) Dropping half of input,
  - iii) Continue recursively on remainder.
- } "half decrease", if done right

We should use i) to determine which half to drop in ii).  
So choose wisely.

and neighbor

We look at centre of  $A$ , that is,  $A[c]$ , with  $c = \lfloor \frac{n}{2} \rfloor + 1$ .  
Three cases can occur.

- $A[c-1] < A[c] < A[c+1]$ : We are in "ascending half" of  $A$ .  $p > c$ , so we drop  $A[1], \dots, A[c]$ , and proceed recursively on rest.
- $A[c-1] > A[c] > A[c+1]$ : We are in "descending half" of  $A$ .  $p < c$ , so we drop  $A[c], \dots, A[n]$ , and proceed recursively on rest.
- $A[c-1] < A[c] > A[c+1]$ : We found  $p$  (!) ; -  $p := c$ .  $\Sigma$  (why?)

A straightforward algorithm implementing this runs in  $\mathcal{O}(\log n)$ .  
More specifically, this idea has recurrence

$$T(n) \leq T\left(\frac{n}{2}\right) + 3 \quad \# \text{ reads from array}$$

$$T(3) \leq 3$$

## Solved Exercise 2

Given

$n$  : # days

$i, j$ : day index ;  $1 \leq i, j \leq n$

$p(i)$  : price of stock on day  $i$

Buy 1000 shares on  $i$ ,  
sell the shares on  $j$ ;  $i < j$

Goal: Find  $i, j$  maximising profit (if profit is possible). Algorithm must run in  $O(n \log n)$ .

Solution: Apply D&C:

- Split problem in two,
- Find optimal solution for each half,
- Combine these solutions into solution for the whole.

If we can find an algorithm implementing this idea, it will have recurrence  $T(n) \leq 2T(\frac{n}{2}) + cn$ , and thus run in  $O(n \log n)$ .

Assume wlg. that initial input size  $n = 2^k$  for some  $k \in \mathbb{Z}$

Step i) and ii) are easy; the recursion does this for us. The tricky part is iii).

In iii), there are three cases:

- best to buy and sell in first half,  $s$ ,
- best to buy in  $s$  and sell in  $s'$ ,
- best to buy and sell in second half,  $s'$ ,

$(i, j)$  are found by taking maximum of

$$\max_{\substack{1 \leq i \leq \frac{n}{2}; \frac{n}{2}+1 \leq j \leq n}} (p(j) - p(i)) = \max_{\frac{n}{2}+1 \leq j \leq n} (p(j)) - \min_{1 \leq i \leq \frac{n}{2}} (p(i))$$

$T(\frac{n}{2})$   
 $\Gamma(\frac{n}{2})$   
 $\Theta(n)$

$(i, j)$  for that value is our result.

An algorithm implementing this has recurrence

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

For some  $c$ , and thus  $O(n \log n)$ .