

Recurrences

Recurrence

Recurrence

Towers of Hanoi

Merge sort

Time complexity of algorithms

Recall that we used induction to prove statements like

$$\sum_{k=0}^n k = 0 + 1 + 2 + 3 + \dots + n = \frac{n(n+1)}{2}$$

In problems like this, we used a common pattern:

$$\sum_{k=0}^0 k = 0$$

$$\sum_{k=0}^n k = \left(\sum_{k=0}^{n-1} k \right) + n, \quad \text{when } n > 0$$

That is, we can express the sum of natural numbers recursively in terms of a smaller sum.

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Let $S(n)$ be the sum of all natural numbers not greater than n :

$$S(n) = \sum_{k=0}^n k,$$

It can be convenient to redefine the sum $S(n)$ as a *recurrence*:

$$S(0) = 0$$

$$S(n) = S(n-1) + n \quad (\forall n > 0)$$

This is just another way to express the same function S .

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Exponentiation:

$$E(a, n) = a^n$$

Recursively:

$$E(a, 0) = 1$$

$$E(a, n) = E(a, n-1) \cdot a \quad (\forall n > 0)$$

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Factorial:

$$n! = 1 \cdot 2 \cdot \dots \cdot n$$

Recursively:

$$0! = 1$$

$$n! = (n-1)! \cdot n \quad (\forall n > 0)$$

The towers of Hanoi

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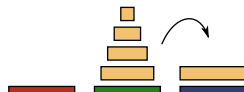
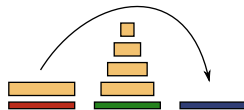
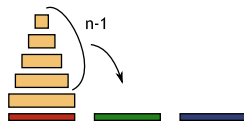


<http://www.mathsisfun.com/games/towerofhanoi.html>

The towers of Hanoi

Our recursive algorithm to move a tower of height n from #1 to #3:

1. Move an $(n-1)$ -tower from #1 to #2.
2. Move an 1-tower from #1 to #3.
3. Move an $(n-1)$ -tower from #2 to #3.



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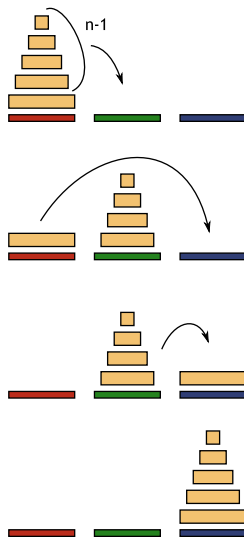
Time complexity of algorithms

The towers of Hanoi

Our recursive algorithm to move a tower of height n from #1 to #3:

1. Move an $(n-1)$ -tower from #1 to #2.
2. Move an 1-tower from #1 to #3.
3. Move an $(n-1)$ -tower from #2 to #3.

There is a way to find a recurrent formula for T_n , the total number of steps to move the tower from the peg 1 to the peg 3.



Recurrence

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Merge sort

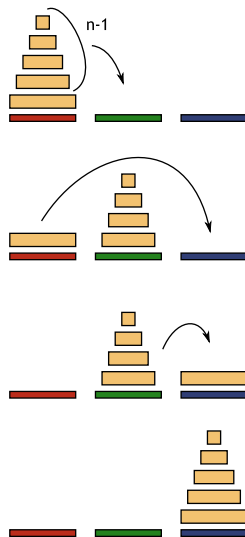
Time complexity of algorithms

The towers of Hanoi

T_n , the time to move a tower of height n :

$$T_1 = 1$$

$$T_n = T_{n-1} + 1 + T_{n-1} \quad (\forall n > 1)$$



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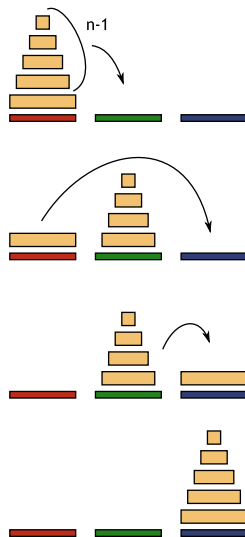
The towers of Hanoi

T_n , the time to move a tower of height n :

$$T_1 = 1$$

$$T_n = T_{n-1} + 1 + T_{n-1} \quad (\forall n > 1)$$

There is a proof by induction that this time is optimal for any algorithm.



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The towers of Hanoi

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$$T_1 = 1$$

$$T_n = 2T_{n-1} + 1 \quad (\forall n > 1)$$

Our goal is to find a closed form expression for T_n as a function of n , without any recurrence.

The towers of Hanoi

Before we get a closed form formula for T_n ,
what are the numbers?

$$T_1 = 1$$

$$T_n = 2T_{n-1} + 1 \quad (\forall n > 1)$$

We can compute a list like this:

$$T_1 = 1$$

$$T_2 = 3$$

$$T_3 = 7$$

$$T_4 = 15$$

$$T_5 = 31$$

$$T_6 = 63$$

...

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$$T_1 = 1$$

$$T_n = 2T_{n-1} + 1 \quad (\forall n > 1)$$

$$T_1 = 1$$

$$T_2 = 3$$

$$T_3 = 7$$

$$T_4 = 15$$

$$T_5 = 31$$

$$T_6 = 63$$

...

Guess and verify method... Let's try $T_n = 2^n - 1$?

The towers of Hanoi

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$$T_1 = 1$$

$$T_n = 2T_{n-1} + 1 \quad (\forall n > 1)$$

Guess and verify method... Let's try $T_n = 2^n - 1$? We can show by induction that this formula is correct.

The base case, $n = 1$:

$$T_1 = 2^1 - 1 = 1.$$

Ok, the base case is true.

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$$\begin{aligned}T_1 &= 1 \\T_n &= 2T_{n-1} + 1 \quad (\forall n > 1)\end{aligned}$$

We want to prove the closed form formula $T_n = 2^n - 1$.

The inductive step, $n > 1$:

Assume that $T_n = 2^n - 1$, and show that then $T_{n+1} = 2^{n+1} - 1$.

Proof. From the recurrence:

$$T_{n+1} = 2T_n + 1$$

By the inductive hypothesis:

$$2T_n + 1 = 2(2^n - 1) + 1 = 2^{n+1} - 2 + 1 = 2^{n+1} - 1$$

The towers of Hanoi

If it is difficult to guess the closed form expression for the recurrence, there is another technique:

$$T_1 = 1$$

$$T_n = 2T_{n-1} + 1$$

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The towers of Hanoi

If it is difficult to guess the closed form expression for the recurrence, there is another technique:

$$T_1 = 1$$

$$T_n = 2T_{n-1} + 1$$

$$= 2(2T_{n-2} + 1) + 1$$

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$$T_1 = 1$$

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$$= 2(2T_{n-2} + 1) + 1$$

$$= 2^2T_{n-2} + 2 + 1$$

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If it is difficult to guess the closed form expression for the recurrence, there is another technique:

$$T_1 = 1$$

$$T_n = 2T_{n-1} + 1$$

$$= 2(2T_{n-2} + 1) + 1$$

$$= 2^2 T_{n-2} + 2 + 1$$

$$= 2^2 (2T_{n-3} + 1) + 2 + 1$$

$$= 2^3 T_{n-3} + 4 + 2 + 1$$

...

$$= 2^{n-1} \underbrace{T(n - (n-1))}_{=1} + \dots + 4 + 2 + 1$$

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The towers of Hanoi

If it is difficult to guess the closed form expression for the recurrence, there is another technique:

$$T_1 = 1$$

$$T_n = 2T_{n-1} + 1$$

$$= 2(2T_{n-2} + 1) + 1$$

$$= 2^2 T_{n-2} + 2 + 1$$

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$$= 2^3 T_{n-3} + 4 + 2 + 1$$

...

$$= 2^{n-1} \underbrace{T(n - (n-1))}_{=1} + \dots + 4 + 2 + 1$$

$$= 2^{n-1} + \dots + 4 + 2 + 1 = \sum_{k=0}^{n-1} 2^k =$$

Recurrence

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The towers of Hanoi

If it is difficult to guess the closed form expression for the recurrence, there is another technique:

$$T_1 = 1$$

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$$= 2(2T_{n-2} + 1) + 1$$

$$= 2^2 T_{n-2} + 2 + 1$$

$$= 2^2 (2T_{n-3} + 1) + 2 + 1$$

$$= 2^3 T_{n-3} + 4 + 2 + 1$$

...

$$= 2^{n-1} \underbrace{T(n - (n-1))}_{=1} + \dots + 4 + 2 + 1$$

$$= 2^{n-1} + \dots + 4 + 2 + 1 = \sum_{k=0}^{n-1} 2^k = \frac{1-2^n}{1-2} = 2^n - 1.$$

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The towers of Hanoi

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Why is it useful to know that the recurrence

$$T_1 = 1$$

$$T_n = 2T_{n-1} + 1 \quad (\forall n > 1)$$

is equivalent to the closed form formula $T_n = 2^n - 1$?

The 7-disk puzzle will require $T_7 = 2^7 - 1 = 127$ moves to complete.

And the 100-disk puzzle will require

$$T_{100} = 2^{100} - 1 = 1267650600228229401496703205375 \text{ moves.}$$

Merge sort

Recurrence

Towers of Hanoi

Merge sort

Time complexity of
algorithms

function Merge:

Given two sorted lists, combine them into a single sorted list:

$$[1, 2, 4, 5] + [3, 4, 5, 6] \mapsto [1, 2, 3, 4, 4, 5, 5, 6]$$

function Sort:

Given a list: if it contains a single element, return it. Otherwise, split it in two halves sort them separately and merge the results:

$$S[5] \mapsto [5]$$

$$S[6, 7, 1, 8, 9, 7, 4, 3] \mapsto S[6, 7, 1, 8] + S[9, 7, 4, 3]$$

Merge sort

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$S[1, 8, 3, 6, 5, 4, 7, 2] \mapsto$

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$$\begin{aligned} &S[1, 8, 3, 6, 5, 4, 7, 2] \mapsto \\ &S[1, 8, 3, 6] + S[5, 4, 7, 2] \mapsto \end{aligned}$$

Merge sort

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algorithms

$$S[1, 8, 3, 6, 5, 4, 7, 2] \mapsto$$

$$S[1, 8, 3, 6] + S[5, 4, 7, 2] \mapsto$$

$$(S[1, 8] + S[3, 6]) + (S[5, 4] + S[7, 2]) \mapsto$$

Merge sort

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Time complexity of algorithms

$$S[1, 8, 3, 6, 5, 4, 7, 2] \mapsto$$

$$S[1, 8, 3, 6] + S[5, 4, 7, 2] \mapsto$$

$$(S[1, 8] + S[3, 6]) + (S[5, 4] + S[7, 2]) \mapsto$$

$$((S[1] + S[8]) + (S[3] + S[6])) + ((S[5] + S[4]) + (S[7] + S[2])) \mapsto$$

Merge sort

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$$S[1, 8, 3, 6, 5, 4, 7, 2] \mapsto$$

$$S[1, 8, 3, 6] + S[5, 4, 7, 2] \mapsto$$

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Merge sort

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$$S[1, 8, 3, 6, 5, 4, 7, 2] \mapsto$$

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$$[1, 8] + [3, 6] + [4, 5] + [2, 7] \mapsto$$

Merge sort

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$$S[1, 8, 3, 6, 5, 4, 7, 2] \mapsto$$

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$$[1, 3, 6, 8] + [2, 4, 5, 7] \mapsto$$

Merge sort

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$$S[1, 8, 3, 6, 5, 4, 7, 2] \mapsto$$

$$S[1, 8, 3, 6] + S[5, 4, 7, 2] \mapsto$$

$$(S[1, 8] + S[3, 6]) + (S[5, 4] + S[7, 2]) \mapsto$$

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$$([1, 8] + [3, 6]) + ([4, 5] + [2, 7]) \mapsto$$

$$[1, 3, 6, 8] + [2, 4, 5, 7] \mapsto$$

$$[1, 2, 3, 4, 5, 6, 7, 8]$$

Merge sort

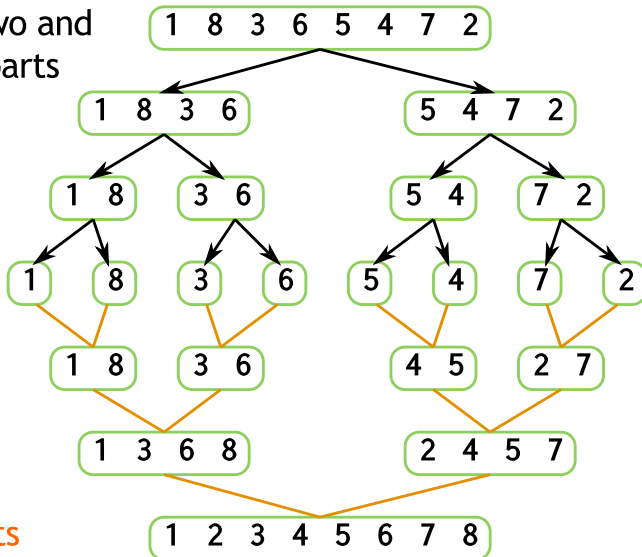
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Towers of Hanoi

Merge sort

Time complexity of algorithms

Split in two and
sort the parts



Merge
sorted lists

Merge sort

Recurrence

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Merge sort

Time complexity of
algorithms

How much time does it take to sort a list of n elements?

To estimate the time complexity, we are going to *count the number of comparisons* between the elements.

We assume that the size of the given list is a power of 2. It makes the analysis easier, but does not affect the result.

Merge sort

Recurrence

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Merge sort

Time complexity of
algorithms

- (a) To merge two lists of size $n/2$, we need to do at most $n - 1$ comparisons.
- (b) To sort a list, we have to split it in two, sort both halves, and merge them.

Therefore,

$$T(1) = 0$$

$$T(n) = 2T(n/2) + n - 1 \quad (\forall n > 1)$$

Merge sort

Given

$$T(n) = 2T(n/2) + n - 1 \quad (\forall n > 1)$$

Since we assume that $n = 2^k$,

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

Recurrence

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Merge sort

Time complexity of
algorithms

Merge sort

Given

$$T(n) = 2T(n/2) + n - 1 \quad (\forall n > 1)$$

Since we assume that $n = 2^k$,

$$T(n) = T(2^k) = 2T(2^k/2) + 2^k - 1 = 2T(2^{k-1}) + (2^k - 1)$$

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Merge sort

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Merge sort

Time complexity of
algorithms

Given

$$T(n) = 2T(n/2) + n - 1 \quad (\forall n > 1)$$

Since we assume that $n = 2^k$,

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

Merge sort

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Merge sort

Time complexity of algorithms

Given

$$T(n) = 2T(n/2) + n - 1 \quad (\forall n > 1)$$

Since we assume that $n = 2^k$,

$$\begin{aligned} T(n) &= T(2^k) = 2T(2^k/2) + 2^k - 1 = 2T(2^{k-1}) + (2^k - 1) \\ &= 2(2T(2^{k-2}) + 2^{k-1} - 1) + (2^k - 1) \\ &= 2^2T(2^{k-2}) + (2^k - 2) + (2^k - 1) \\ &= 2^2(2T(2^{k-3}) + 2^{k-2} - 1) + (2^k - 2) + (2^k - 1) \\ &= 2^3T(2^{k-3}) + (2^k - 4) + (2^k - 2) + (2^k - 1) \\ &= 2^3(2T(2^{k-4}) + 2^{k-3} - 1) + (2^k - 4) + (2^k - 2) + (2^k - 1) \\ &= 2^4T(2^{k-4}) + (2^k - 8) + (2^k - 4) + (2^k - 2) + (2^k - 1) \\ &= \dots = 2^k \underbrace{T(2^{k-k})}_{T(1)=0} + \sum_{i=0}^{k-1} (2^k - 2^i) = \sum_{i=0}^{k-1} (2^k - 2^i). \end{aligned}$$

Merge sort

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Merge sort

Time complexity of algorithms

$$T(n) = T(2^k) = \sum_{i=0}^{k-1} (2^k - 2^i) = \sum_{i=0}^{k-1} (n - 2^i) = n \cdot k - \sum_{i=0}^{k-1} 2^i.$$

The sum of the geometric progression is

$$\sum_{i=0}^{k-1} 2^i = \frac{2^k - 1}{2 - 1} = 2^k - 1 = n - 1.$$

Thus $T(n) = n \cdot k - n + 1$. And since $n = 2^k$, $k = \log_2 n$, so

$$T(n) = n \log_2 n - n + 1.$$

Merge sort

Recurrence

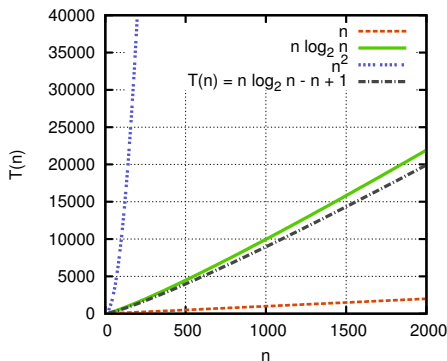
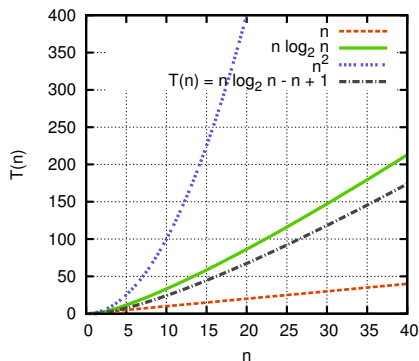
Towers of Hanoi

Merge sort

Time complexity of algorithms

To sort a list of length n , takes time (the number of comparisons)

$$T(n) = n \log_2 n - n + 1 \approx n \log_2 n.$$



Merge sort

Recurrence

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Merge sort

Time complexity of
algorithms

In merge sort, we had a recurrence:

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

In general, if the time complexity of an algorithm is expressed by a recurrence:

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

To solve such recurrences, there is a so called *Master theorem*:

https://en.wikipedia.org/wiki/Master_theorem

It covers different forms of the function f , as well as difference values of the constants a and b .

Time complexity, big-O

Recurrence

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Merge sort

Time complexity of
algorithms

Let's say that we've got this function as an estimation of the time complexity of an algorithm:

$$T(n) = 6n \log_2 n + 100n + \log_2 n + 50$$

Informally:

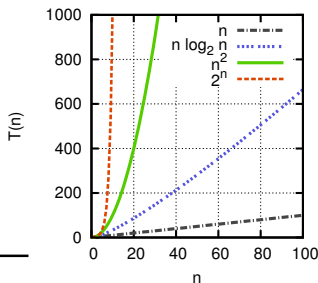
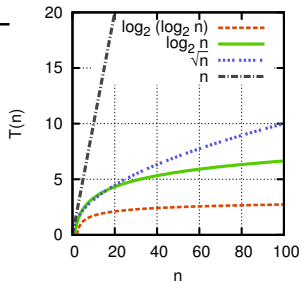
- (a) If $T(n)$ is a sum, we take the fastest growing term only.
- (b) We don't really care about constant factors.

$$T(n) = O(n \log_2 n)$$

Time complexity, big-O

Some common time complexities, from the slowest to the fastest:

Running time	Name
$O(1)$	constant
$O(\log(\log n))$	log-logarithmic
$O(\log n)$	logarithmic
$O(\sqrt{n})$	square root (sub-linear)
$O(n)$	linear
$O(n \log n)$	n-log-n
$O(n^2)$	quadratic
$O(2^n)$	exponential
$O(n!)$	factorial
$O(2^{(2^n)})$	double exponential



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Time complexity, big-O

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Faster than linear algorithms, for example $O(\log n)$, cannot go through the whole input. They are cherry-picking in some sense, knowing where to search for the answer. Usually the input is structured in some way.

Example: *Binary search in a sorted array.*

Linear time algorithms, $O(n)$, usually have to read the whole input.

Example: *Search for the largest element in an unsorted array.*

Time complexity, big-O

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Algorithms *slower than linear*, $O(n \log n)$ or $O(n^2)$, $O(n^7)$. Not only read the whole input, but also perform some extra work, but in a reasonably efficient way.

Example: *Sorting algorithms.*

Algorithms *much slower than linear*, exponential, for example, $O(2^n)$, are doing some non-trivial work.

Example: *Satisfiability of a statement in propositional logic.*

Time complexity, big-O

Recurrence

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Formally:

We say that

$$T(n) = O(f(n))$$

if there are constants C and k such that

$$|T(n)| \leq C|f(n)| \quad \text{for all } n > k$$

This definition says that after $n > k$, all slowly-growing terms don't really matter, and $T(n)$ behaves similarly to $f(n)$. To be more exact, $T(n)$ never exceeds $C \cdot f(n)$ when n is large enough.

$$T(n) = 6n \log_2 n + 100n + \log_2 n + 50$$

$$T(n) = O(n \log_2 n)$$