

## Complexity

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KTH

VT23

## run-time complexity of sum

Calculating the sum of all elements in a list:

sum/1

```
def sum([]) do 0 end
def sum([h|t]) do
  s = sum(t)
  h + s
end
```

sum/2

```
def sum([], s) do s end
def sum([h|t], s) do
  s1 = h+s
  sum(t, s1)
end
```

What are the run-time complexities of sum/1 and sum/2?

1 / 30

2 / 30

## run-time complexity of foo

foo/1

```
def foo([]) do [] end
def foo([h|t]) do
  z = foo(t)
  bar(z, h)
end
```

foo/2

```
def foo([], y) do y end
def foo([h|t], y) do
  z = zot(h, y)
  foo(t, z)
end
```

What are the run-time complexities of foo/1 and foo/2?

## run-time complexity of reverse

nreverse/1

```
def nreverse([]) do [] end
def nreverse([h|t]) do
  z = nreverse(t)
  append(z, [h])
end
```

reverse/2

```
def reverse([], y) do y end
def reverse([h|t], y) do
  z = [h | y]
  reverse(t, z)
end
```

What are the run-time complexities of nreverse/1 and reverse/2?

3 / 30

4 / 30

nreverse/1

```
def nreverse([]) do [] end
def nreverse([h|t]) do
  z = nreverse(t)
  append(z, [h])
end
```

Assume that append/2 takes  $kn$  ms to execute, where  $k$  is some constant time and  $n$  is the length of the list.

Describe the time  $T_n$  it takes to execute nreverse/1 of a list of length  $n$ :

$$T_0 = a \text{ ms}$$

$$T_n = T_{n-1} + k(n-1) + b \text{ ms}$$

$$\begin{aligned}
 T_n &= T_{n-1} + k(n-1) + b \\
 &= T_{n-2} + k(n-2) + k(n-1) + 2b \\
 &= T_{n-3} + k(n-3) + k(n-2) + k(n-1) + 3b \\
 &: \\
 &= T_{n-n} + k(n-n) + \dots + k(n-1) + nb \\
 &= a + 0 + k + 2k + 3k + \dots + (n-1)k + nb \\
 &= n \frac{(n-1)}{2} k + nb + a \\
 &= \left(\frac{k}{2}\right)n^2 - \frac{k}{2}n + bn + a \\
 &= \left(\frac{k}{2}\right)n^2 + \left(b - \frac{k}{2}\right)n + a
 \end{aligned} \tag{1}$$

5 / 30

6 / 30

We know:

$$T_n = \left(\frac{k}{2}\right)n^2 + \left(b - \frac{k}{2}\right)n + a$$

$$T_n \in O(n^2)$$

Do ordo calculations in your head without specifying the full  $T_n$  relation.

If we know that append/2 is in  $O(n)$  then:

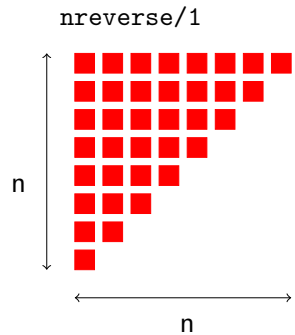
$$T_n \in n * O(n) + bn + a$$

Which means that:

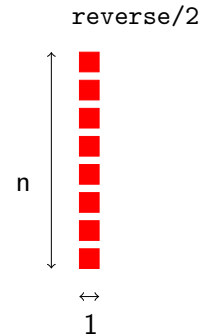
$$T_n \in O(n^2)$$

7 / 30

8 / 30



```
nreverse/1
def nreverse([]) do end
def nreverse([h|t]) do
  z = nreverse(t)
  append(z, [h])
end
```

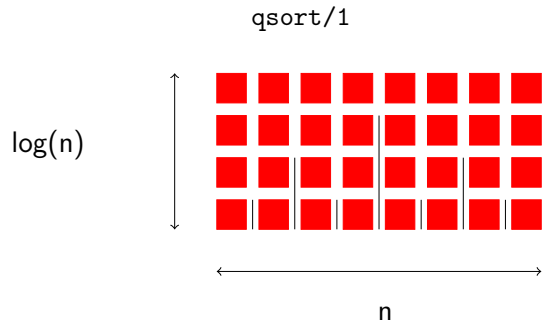


```
reverse/2
def reverse([], y) do y end
def reverse([h|t], y) do
  z = [h | y]
  reverse(t, z)
end
```

```
def qsort([]) do [] end
def qsort([h]) do [h] end
def qsort(all) do
  {low, high} = partition(all)
  lowS = qsort(low)
  highS = qsort(high)
  append(lowS, highS)
end
```

- What is done in each iteration?
- How many iterations do we have?

$$\begin{aligned}
 T_1 &= a \\
 T_n &= 2 \times T_{n/2} + nc \\
 &= 2 \times (2 \times T_{n/4} + (n/2)c) + nc \\
 &= 4 \times T_{n/4} + 2 \times nc \\
 &= 8 \times T_{n/8} + 3 \times nc \\
 &\vdots \\
 &= 2^k \times T_1 + k \times nc \\
 &= 2^{\lg(n)} \times a + \lg(n) \times nc \\
 &= n \times a + \lg(n)n \times c
 \end{aligned}
 \tag{2}$$



What if we run qsort on a already ordered list?

```
def msort([]) do [] end
def msort(l) do
  {a, b} = split(l)
  as = msort(a)
  bs = msort(b)
  merge(as, bs)
end
```

- What is done in each iteration?
- How many iterations do we have?
- What is the run-time complexity?
- Which is best qsort or msort?

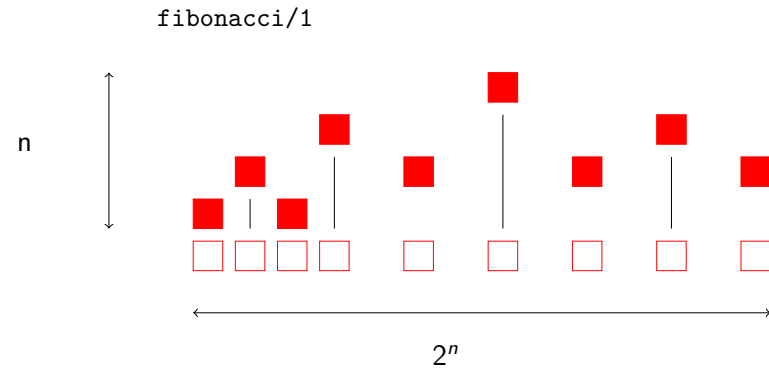
```
def fib(0) do 0 end
def fib(1) do 1 end
def fib(n) do
  fib(n-1) + fib(n-2)
end
```

- What is done in each iteration?
- How many iterations do we have?

Let's cheat a bit to make it simpler:

$$\begin{aligned}
 T_0 &= a \\
 T_n &= 2 \times T_{n-1} + c \\
 &= 2 \times (2 \times T_{n-2} + c) + c \\
 &= 4 \times T_{n-2} + 3 \times c \\
 &= 8 \times T_{n-3} + 7 \times c \\
 &\vdots \\
 &= 2^n \times T_0 + (2^n - 1) \times c \\
 &= 2^n \times a + 2^n \times c - c
 \end{aligned}
 \tag{3}$$

The more precise answer is  $O(1.6^n)$



The smarter implementation is  $O(n)$   
 ... an even smart solution is  $O(\log(n))$

What is the difference between a smart programmer and a not so smart programmer?

3 billion years?

Let's represent trees as:

```

:nil
{:node, key, value, left, right}
    
```

- new: create a empty tree
- insert: add an element to the three
- lookup: search for an element
- modify: modify an element

Why use trees, why not use lists?

Operations on a tree.

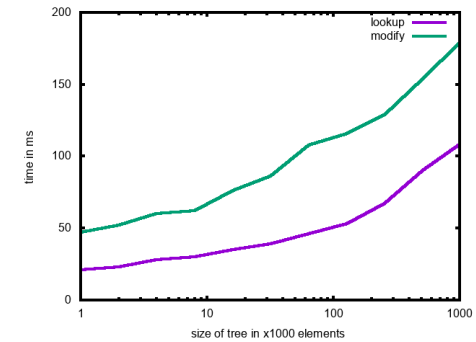


Figure: Execution time in ms of 100.000 calls

Why use trees, why not use tuples?

```
def new([a,b,c]) do {a,b,c} end

def lookup({a,_,_}, 1) do a end
def lookup({_, b,_}, 2) do b end
:

def modify({_,b,c}, 1, v) do {v, b, c} end
def modify({a,_,c}, 2, v) do {a, v, c} end
:
```

```
def new(list) do List.to_tuple(list) end
def lookup(tuple, k) do elem(tuple, k) end
def modify(tuple, k, v) do put_elem(tuple, k, v) end
```

The functions `put_elem/3` will create a copy of the original tuple!

Operations on a tuple.

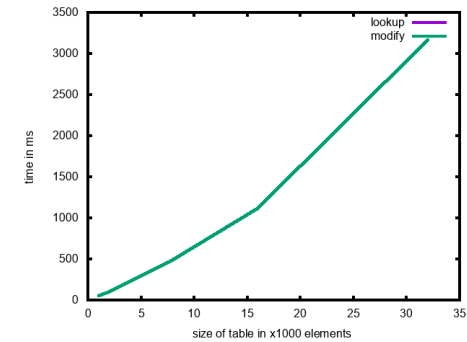


Figure: Execution time in ms of 100,000 calls

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Tuple vs tree.

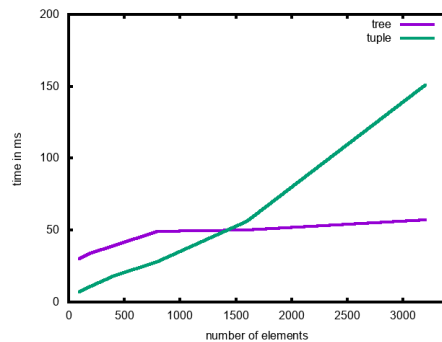


Figure: Modify operations, execution time in ms of 100,000 calls

*Programmers waste enormous amounts of time thinking about, or worrying about, the speed of noncritical parts of their programs, and these attempts at efficiency actually have a strong negative impact when debugging and maintenance are considered. We should forget about small efficiencies, say about 97 percent of the time: premature optimization is the root of all evil. Yet we should not pass up our opportunities in that critical 3 percent.*

Donald Knuth

27 / 30

28 / 30

code size



execution time



- understand the problem before starting coding
- write well structured code that is easy to understand
- use abstractions to separate functionality from implementation
- think about complexity
- benchmark your program
- if needed, optimize