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- Running time should be expressed by simple functions.



5/25

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Let f, g denote functions from  $\mathbb{N}$  to  $\mathbb{R}^+$ .

•  $\mathcal{O}(f) = \{g \mid \exists c > 0 \ \exists n_0 \in \mathbb{N}_0 \ \forall n \geq n_0 : [g(n) \leq c \cdot f(n)] \}$  (set of functions that asymptotically grow not faster than f)

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Note that  $\Theta(n)$  is on the right hand side, otw. this interpretation is wrong.

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Regardless of how we choose the anonymous function  $f(n) \in \mathcal{O}(n)$  there is an anonymous function  $g(n) \in \Theta(n^2)$  that makes the expression true.

10/25

How do we interpret an expression like:

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#### Careful!

"It is understood" that every occurrence of an  $\mathcal{O}$ -symbol (or  $\Theta, \Omega, o, \omega$ ) on the left represents one anonymous function.

Hence, the left side is not equal to

$$\Theta(1) + \Theta(2) + \cdots + \Theta(n-1) + \Theta(n)$$

11/25

We can view an expression containing asymptotic notation as generating a set:

$$n^2 \cdot \mathcal{O}(n) + \mathcal{O}(\log n)$$

### represents

$$\begin{split} \left\{ f: \mathbb{N} \to \mathbb{R}^+ \mid f(n) = n^2 \cdot g(n) + h(n) \\ & \text{with } g(n) \in \mathcal{O}(n) \text{ and } h(n) \in \mathcal{O}(\log n) \right\} \end{split}$$

12/25

Then an asymptotic equation can be interpreted as containement btw. two sets:

$$n^2 \cdot \mathcal{O}(n) + \mathcal{O}(\log n) = \Theta(n^2)$$

represents

$$n^2 \cdot \mathcal{O}(n) + \mathcal{O}(\log n) \subseteq \Theta(n^2)$$

#### Lemma 1

Let f, g be functions with the property

 $\exists n_0 > 0 \ \forall n \ge n_0 : f(n) > 0$  (the same for g). Then

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14/25

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The expressions also hold for  $\Omega$ . Note that this means that  $f(n) + g(n) \in \Theta(\max\{f(n), g(n)\})$ .

#### **Comments**

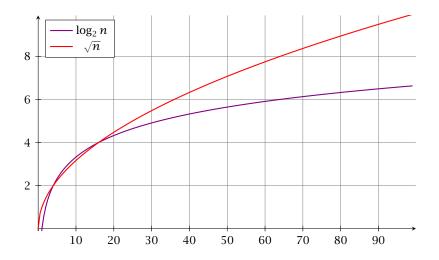
Do not use asymptotic notation within induction proofs.

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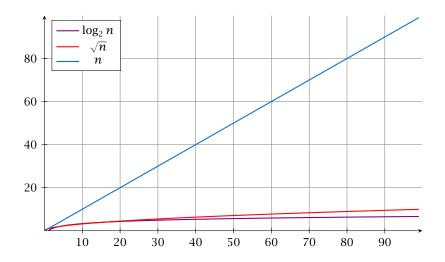
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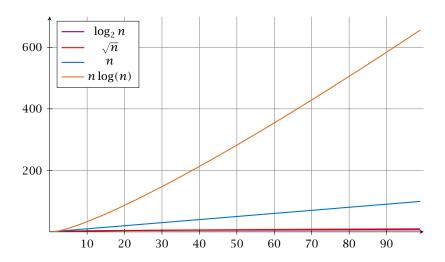
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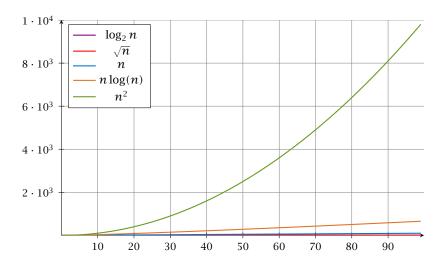
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- For any constants a, b we have  $\log_a n = \Theta(\log_b n)$ . Therefore, we will usually ignore the base of a logarithm within asymptotic notation.
- In general  $\log n = \log_2 n$ , i.e., we use 2 as the default base for the logarithm.



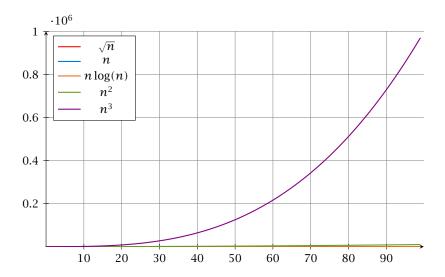




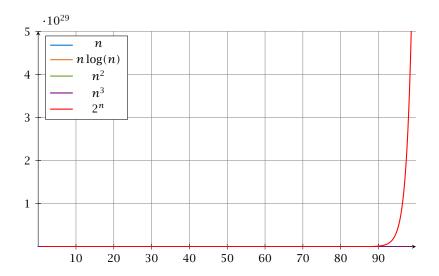












### Laufzeiten

Funktion	Eingabelänge n							
f(n)	10	$10^{2}$	$10^{3}$	$10^{4}$	$10^{5}$	$10^{6}$	10 <sup>7</sup>	$10^{8}$
$\log n$	33 <b>ns</b>	66ns	0.1µs	0.1µs	0.2µs	0.2µs	0.2µs	0.3µs
$\sqrt{n}$	32ns	$0.1 \mu s$	0.3µs	1µs	3.1 <b>µs</b>	10 <b>µs</b>	31 <b>µs</b>	$0.1  \mathrm{ms}$
n	100ns	1µs	10 <b>µs</b>	$0.1 \mathrm{ms}$	1ms	10ms	0.1s	1s
$n \log n$	0.3µs	6.6µs	0.1ms	1.3ms	16ms	0.2s	2.3s	27s
$n^{3/2}$	0.3µs	10µs	0.3ms	10ms	0.3s	10s	5.2min	2.7h
$n^2$	1µs	$0.1 \mathrm{ms}$	10ms	1s	1.7min	2.8h	11 <b>d</b>	3.2 <b>y</b>
$n^3$	10µs	10ms	10s	2.8h	115 <b>d</b>	317 <b>y</b>	3.2·10 <sup>5</sup> y	
$1.1^{n}$	26ns	0.1 ms	$7.8 \cdot 10^{25}$ y					
$2^n$	10µs	$4\cdot 10^{14}$ y						
n!	36ms	$3 \cdot 10^{142}$ y						

1 Operation = 10ns; 100MHz

Alter des Universums: ca.  $13.8 \cdot 10^9 \mathrm{y}$ 

In general asymptotic classification of running times is a good measure for comparing algorithms:

▶ If the running time analysis is tight and actually occurs in practise (i.e., the asymptotic bound is not a purely theoretical worst-case bound), then the algorithm that has better asymptotic running time will always outperform a weaker algorithm for large enough values of n.

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Clearly f = o(g). However, as long as  $\log n \le 1000$  Algorithm B will be more efficient.



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(set of functions that asymptotically grow not faster than f)

#### Example 2

 $f: \mathbb{N} \to \mathbb{R}_0^+, f(n,m) = 1 \text{ und } g: \mathbb{N} \to \mathbb{R}_0^+, g(n,m) = n-1$ 

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