

Algorithmen, KI und Data Science 1 (AKIDS 1): Complexity and Sorting

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Exercise 1.1

Recap: Difference between O, Ω, and Θ

O-Notation:

$f \in O(g)$: For some $C \in R^{>0}$, there exists a $n_0 \in N$, such that for all $n \in N$ with $n \geq n_0$ it holds :

$$0 \leq f(n) \leq C \times g(n)$$

Ω-Notation:

$f \in \Omega(g)$: For some $C \in R^{>0}$, there exists a $n_0 \in N$, such that for all $n \in N$ with $n \geq n_0$ it holds :

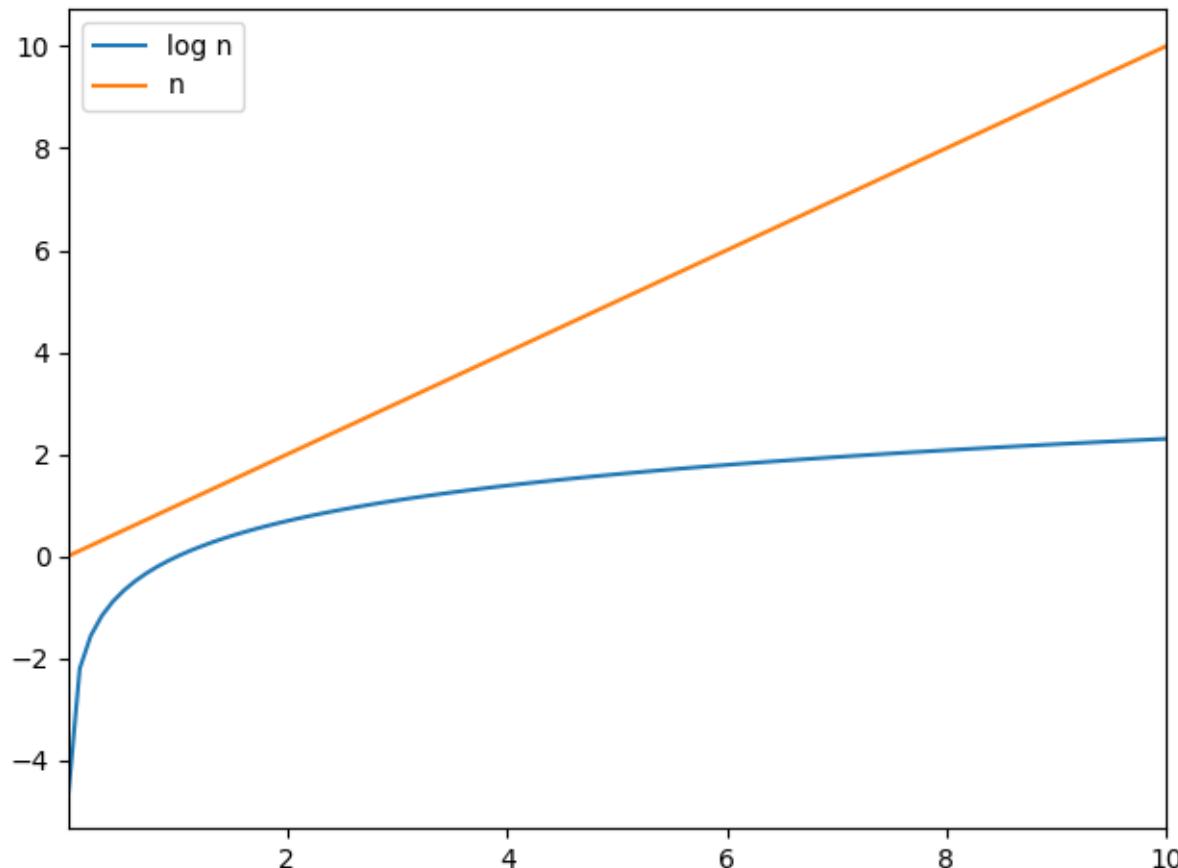
$$0 \leq C \times g(n) \leq f(n)$$

Θ-Notation:

$f \in \Theta(g)$: For some $C_1, C_2 \in R^{>0}$, there exists a $n_0 \in N$, such that for all $n \in N$ with $n \geq n_0$ it holds :

$$0 \leq C_1 \times g(n) \leq f(n) \leq C_2 \times g(n)$$

Example: $\log n \in O(n)$



Transitivity and Reflexivity

Transitivity

$f(n) \in \Theta(g(n))$ and $g(n) \in \Theta(h(n))$ imply $f(n) \in \Theta(h(n))$

$f(n) \in O(g(n))$ and $g(n) \in O(h(n))$ imply $f(n) \in O(h(n))$

$f(n) \in \Omega(g(n))$ and $g(n) \in \Omega(h(n))$ imply $f(n) \in \Omega(h(n))$

Reflexivity

- $f(n) \in \Theta(f(n))$
- $f(n) \in O(f(n))$
- $f(n) \in \Omega(f(n))$

Symmetry

Symmetry

- $f(n) \in \Theta(g(n))$ if and only if $g(n) \in \Theta(f(n))$

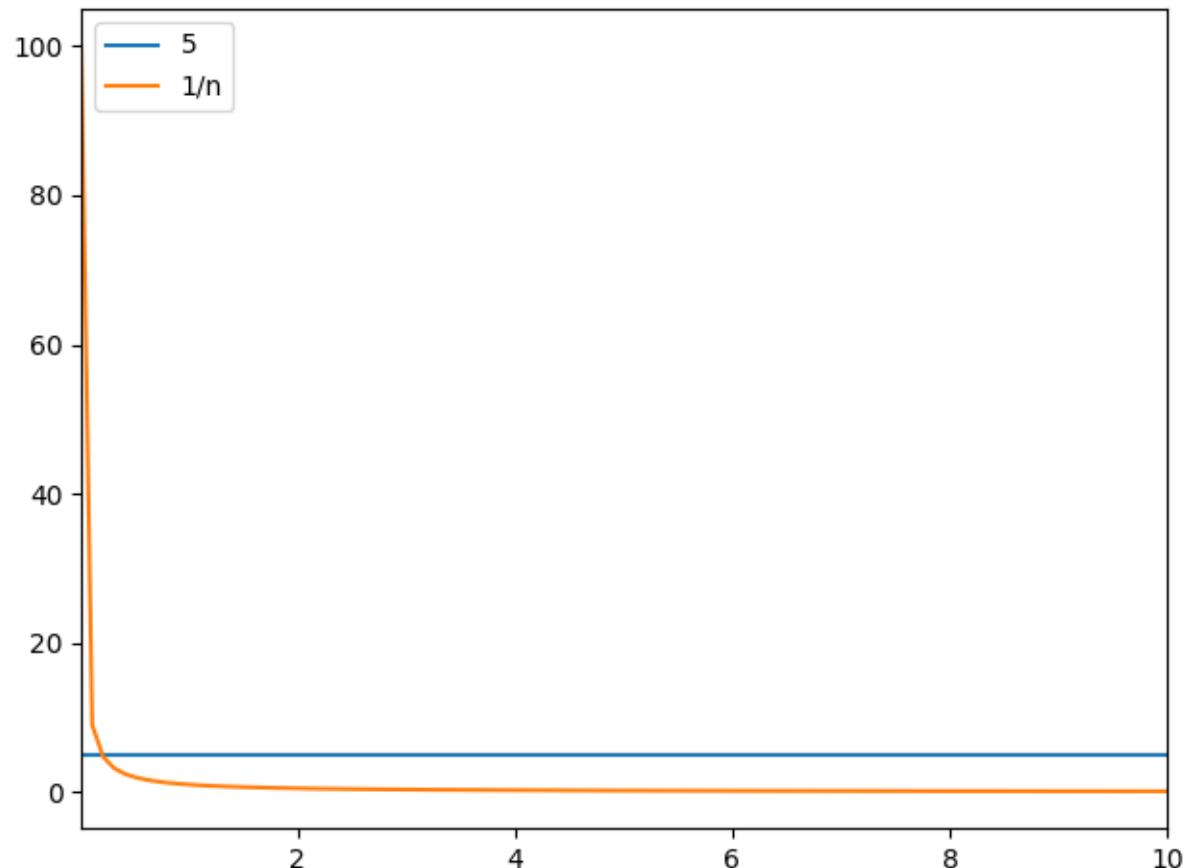
Transpose Symmetry

- $f(n) \in O(g(n))$ if and only if $g(n) \in \Omega(f(n))$

Complete the following table with symbols \circ , Ω , Θ .

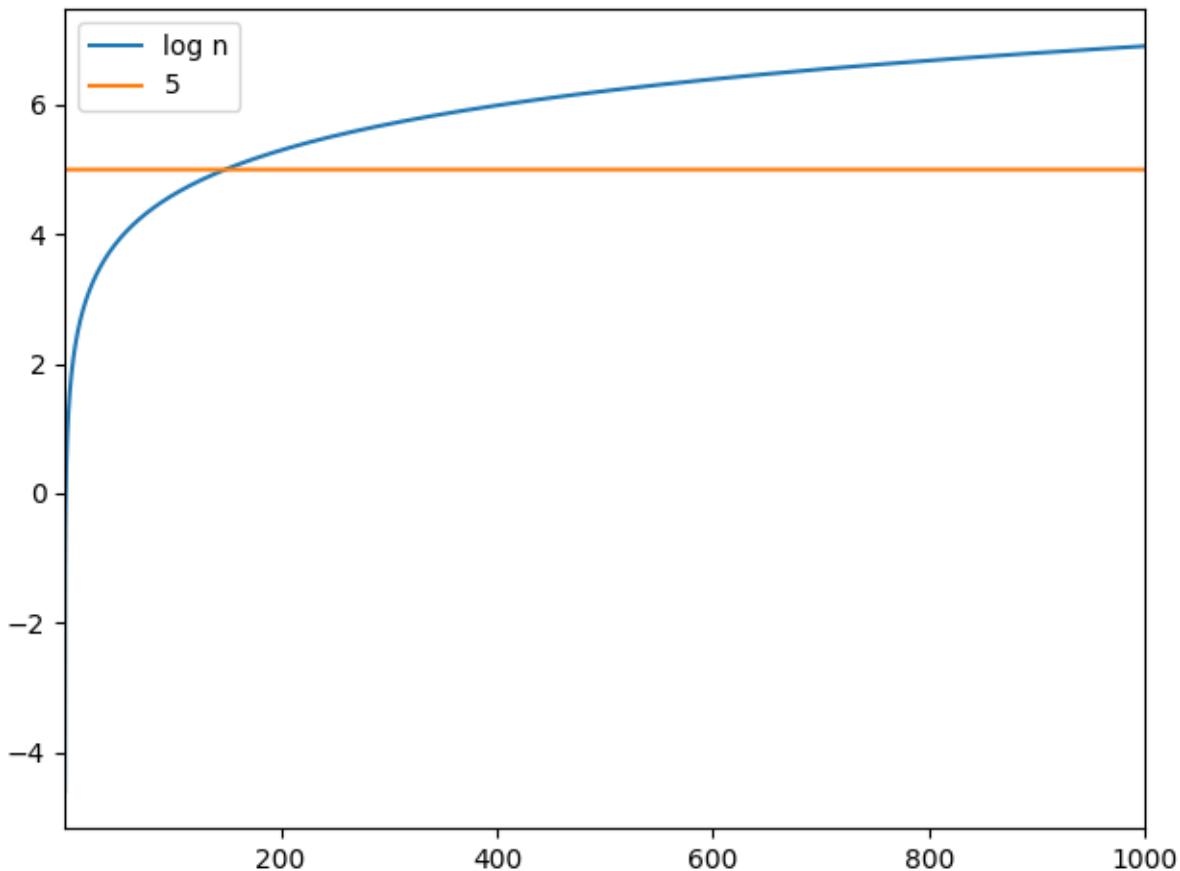
Using reflexivity

	$\log n$	$2^{n/2}$	\sqrt{n}	5	2^n	$1/n$	n	e^n	n^2
$\log n$	Θ						0		
$2^{n/2}$		Θ							
\sqrt{n}			Θ						
5				Θ					
2^n					Θ				
$1/n$						Θ			
n							Θ		
e^n								Θ	
n^2									Θ

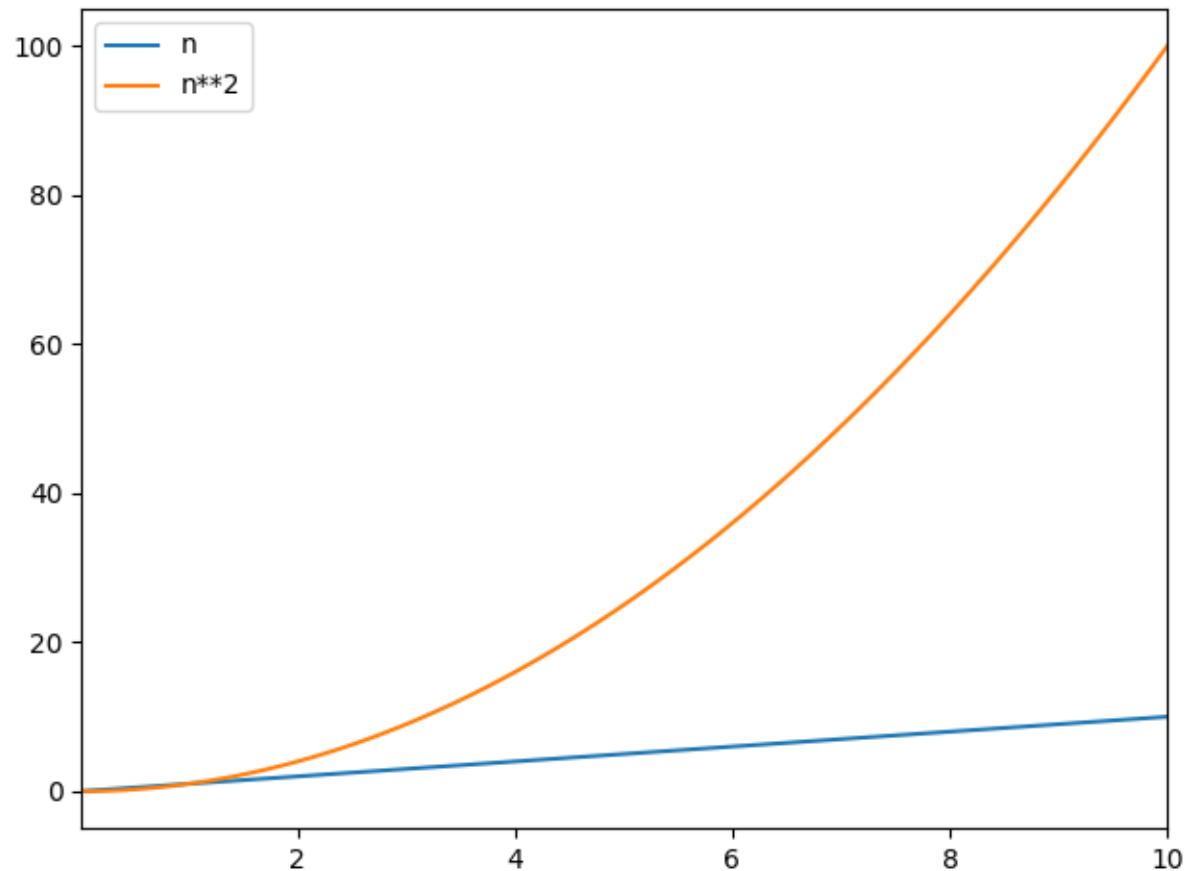
$5 \text{ vs. } \frac{1}{n}$ 

→ $5 \in \Omega(\frac{1}{n})$, $\frac{1}{n} \in O(5)$

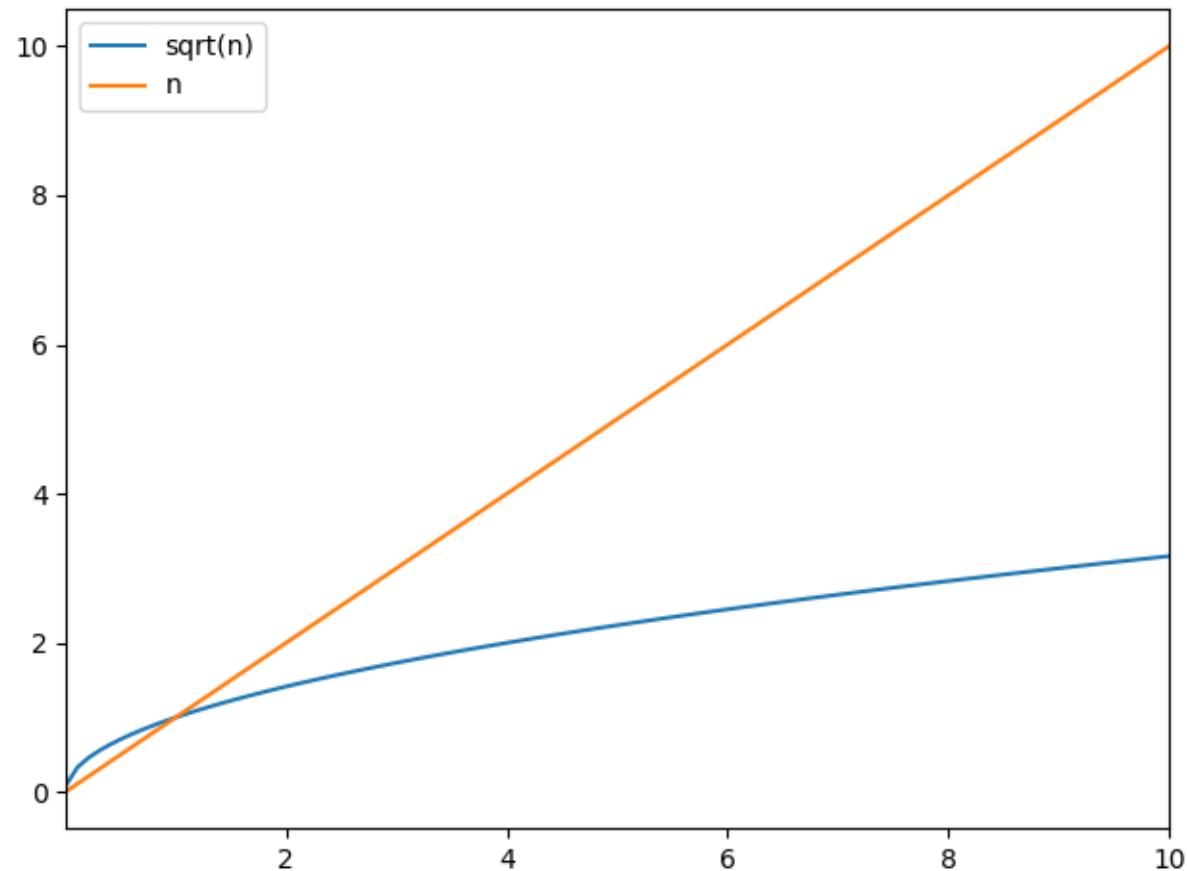
$\log n$ vs. 5



→ $\log n \in \Omega(5)$, $5 \in O(\log n)$

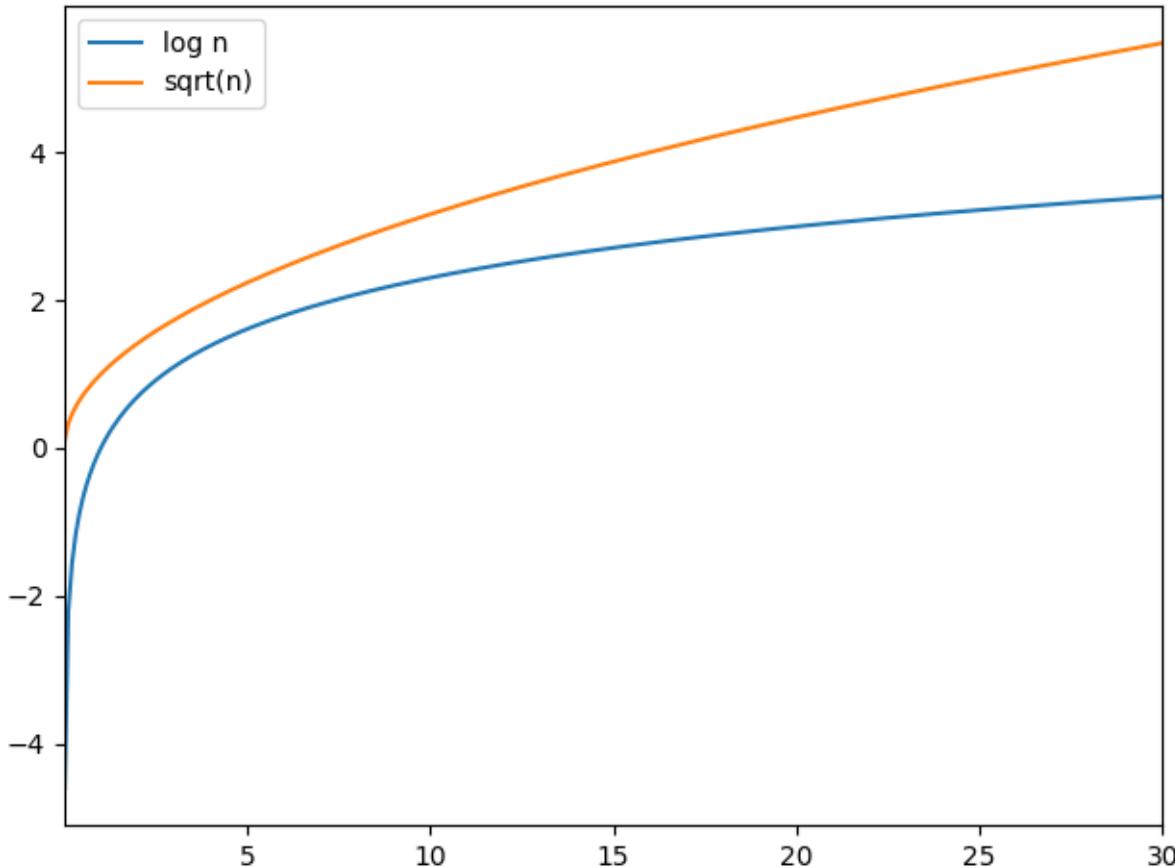
n vs. n^2 

→ $n \in O(n^2), n^2 \in \Omega(n)$

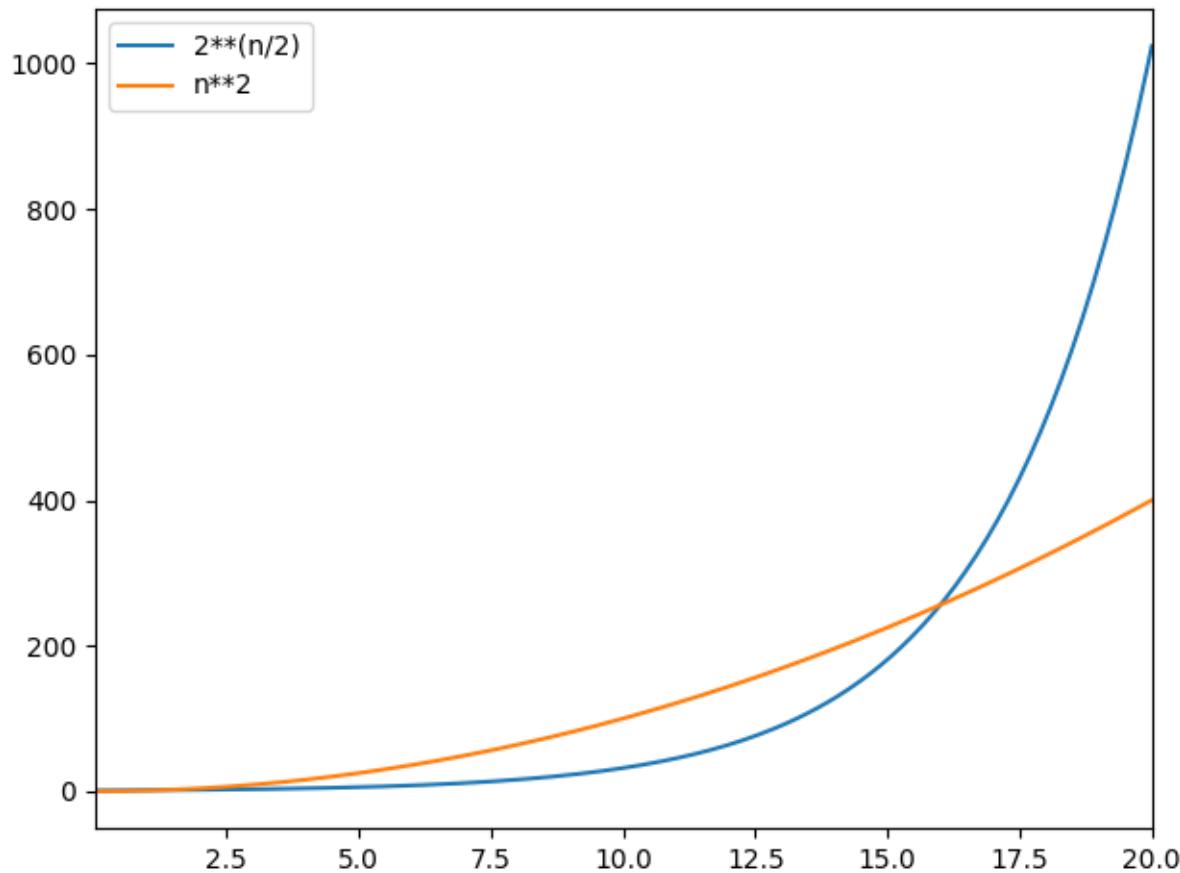
\sqrt{n} vs. n 

→ $\sqrt{n} \in O(n)$, $n \in \Omega(\sqrt{n})$

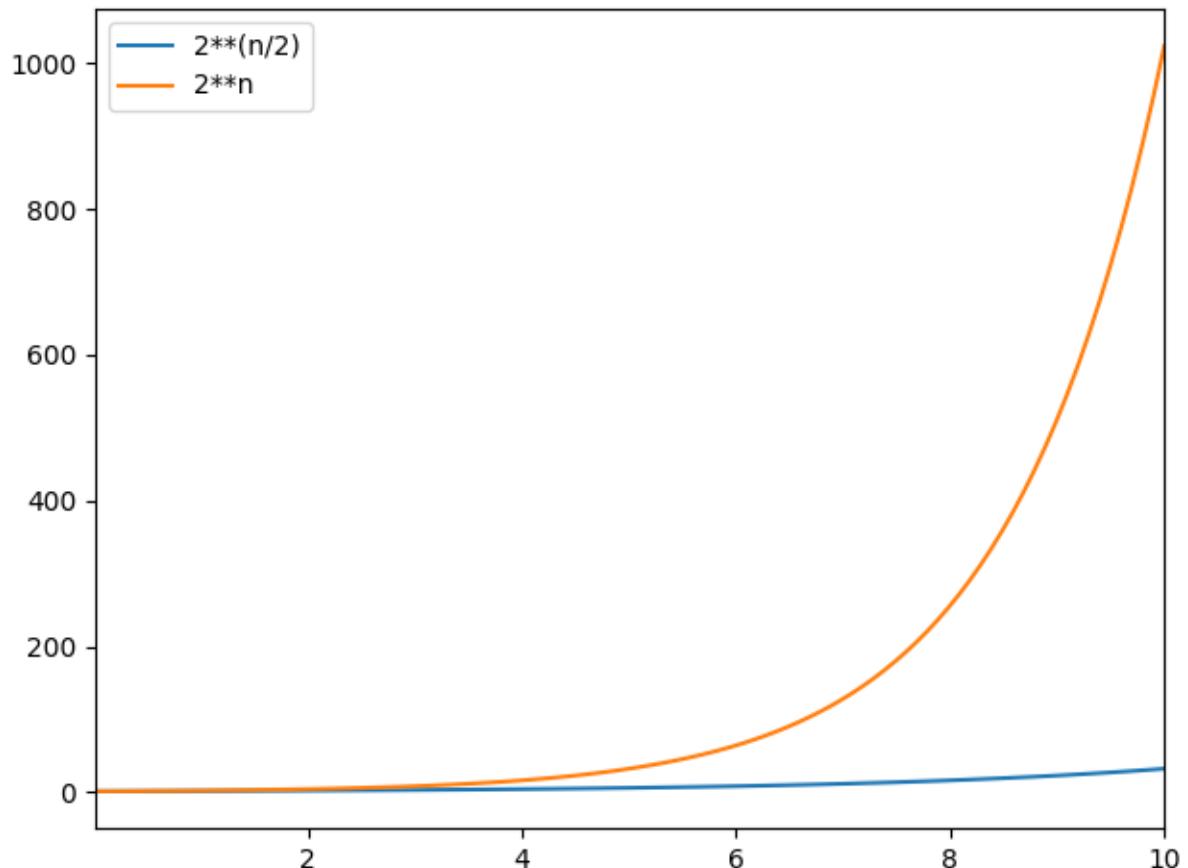
$\log n$ vs. \sqrt{n}



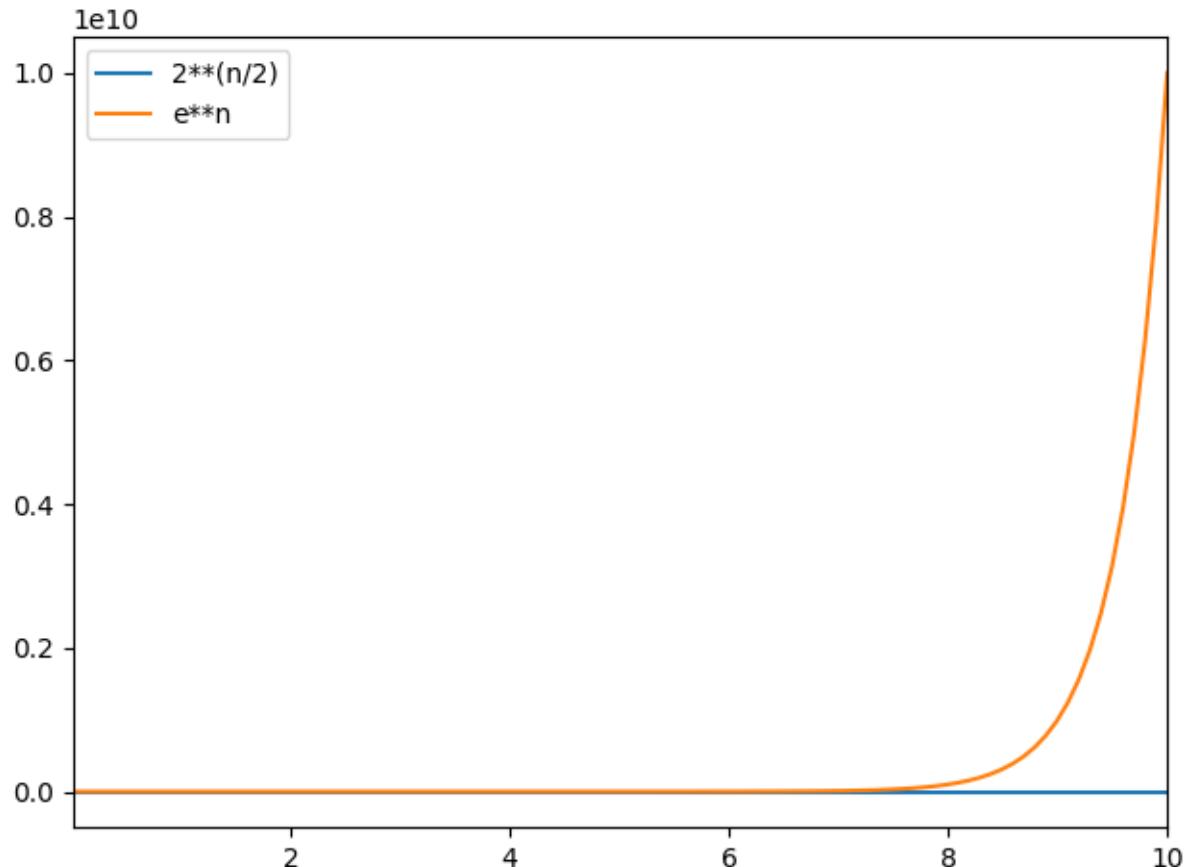
→ $\log n \in O(\sqrt{n}), \sqrt{n} \in \Omega(\log n)$

$2^{\frac{n}{2}}$ vs. n^2 

→ $2^{\frac{n}{2}} \in \Omega(n^2), n^2 \in O(2^{\frac{n}{2}})$

$2^{\frac{n}{2}}$ vs. 2^n 

→ $2^{\frac{n}{2}} \in O(2^n)$, $2^n \in \Omega(2^{\frac{n}{2}})$

$2^{\frac{n}{2}}$ vs. e^n 

→ $2^{\frac{n}{2}} \in O(e^n), e^n \in \Omega(2^{\frac{n}{2}})$

Summary of observations

- $\frac{1}{n} \in O(5)$
- $5 \in O(\log n)$
- $\log n \in O(\sqrt{n})$
- $\sqrt{n} \in O(n)$
- $n \in O(n^2)$
- $n^2 \in O(2^{\frac{n}{2}})$
- $2^{\frac{n}{2}} \in O(2^n)$
- $2^n \in O(e^n)$



transitivity and transpose symmetry

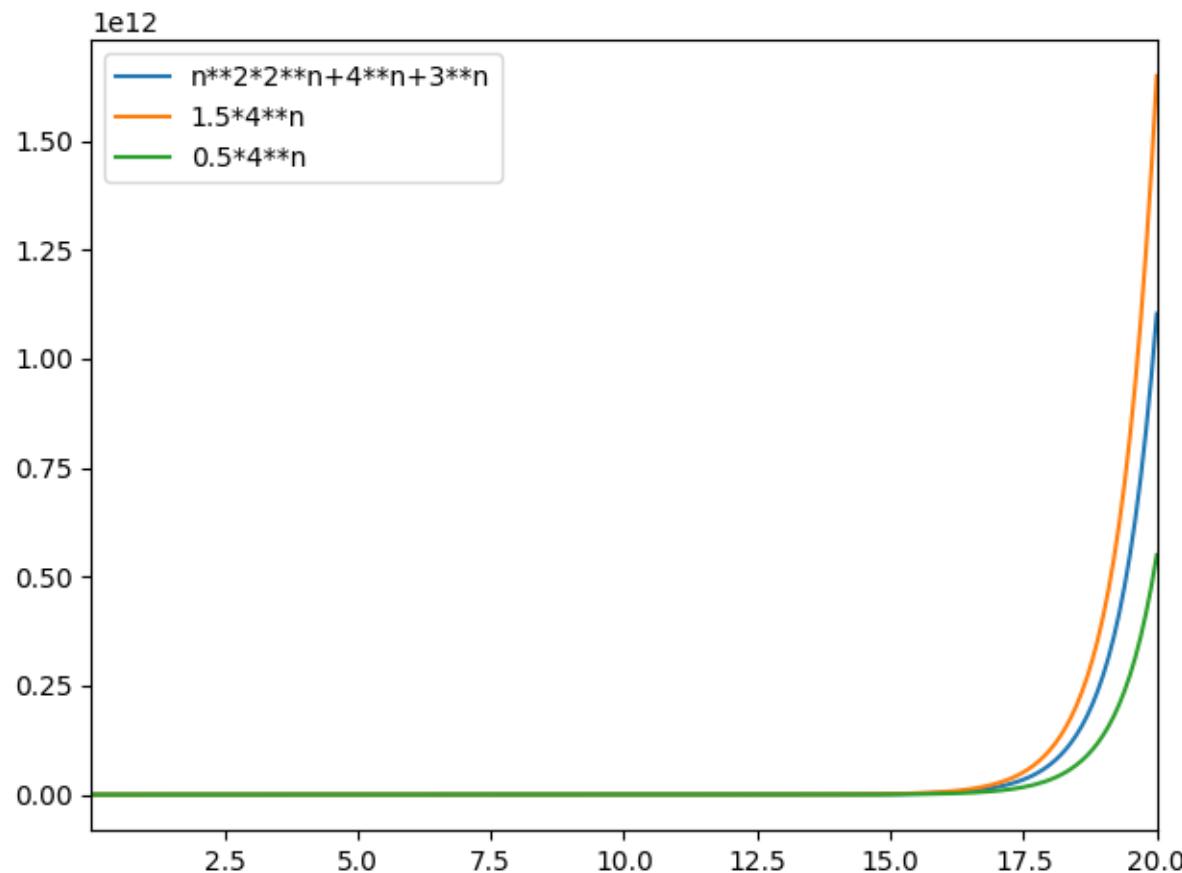
Solution

	$\log n$	$2^{n/2}$	\sqrt{n}	5	2^n	$1/n$	n	e^n	n^2
$\log n$	Θ	0	0	Ω	0	Ω	0	0	0
$2^{n/2}$	Ω	0	Ω	Ω	0	Ω	Ω	0	Ω
\sqrt{n}	Ω	0	0	Ω	0	Ω	0	0	0
5	0	0	0	0	0	Ω	0	0	0
2^n	Ω	Ω	Ω	Ω	0	Ω	Ω	0	Ω
$1/n$	0	0	0	0	0	Θ	0	0	0
n	Ω	0	Ω	Ω	0	Ω	0	0	0
e^n	Ω	Ω	Ω	Ω	Ω	Ω	Ω	0	Ω
n^2	Ω	0	Ω	Ω	0	Ω	Ω	0	0

For each of the following functions f_i , provide a function g_i having as few terms as possible and satisfying $f_i \in \Theta(g_i)$.

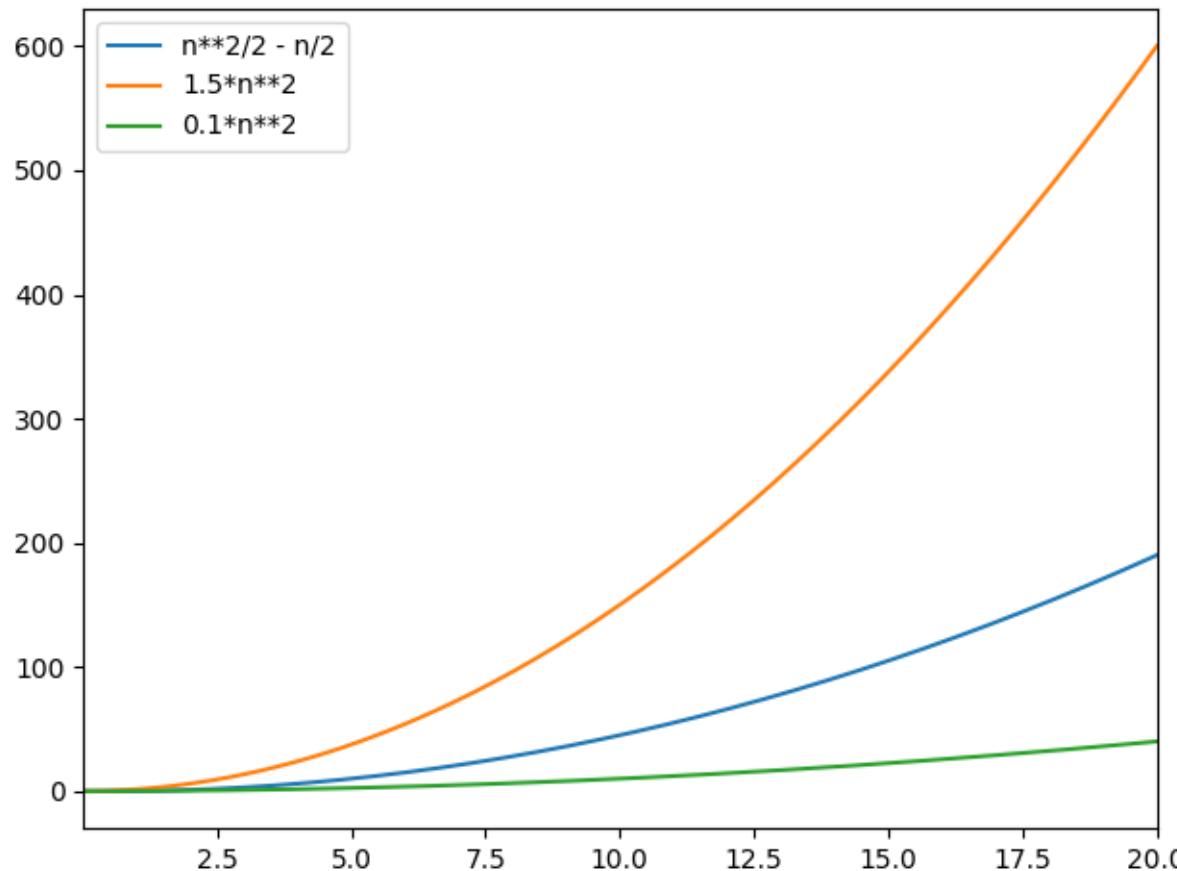
- *Example:* $f_0(n) = 3n^2 + 3 \in \Theta(n^2)$
- $f_1(n) = n^2 2^n + 4^n + 3^n$
- $f_2(n) = \frac{n(n-1)}{2}$
- $f_3(n) = \log(n^{70})$
- $f_4(n) = 9n\log(n) + 30n(\log(n))^2 + n$

$$f_1(n) = n^2 2^n + 4^n + 3^n$$



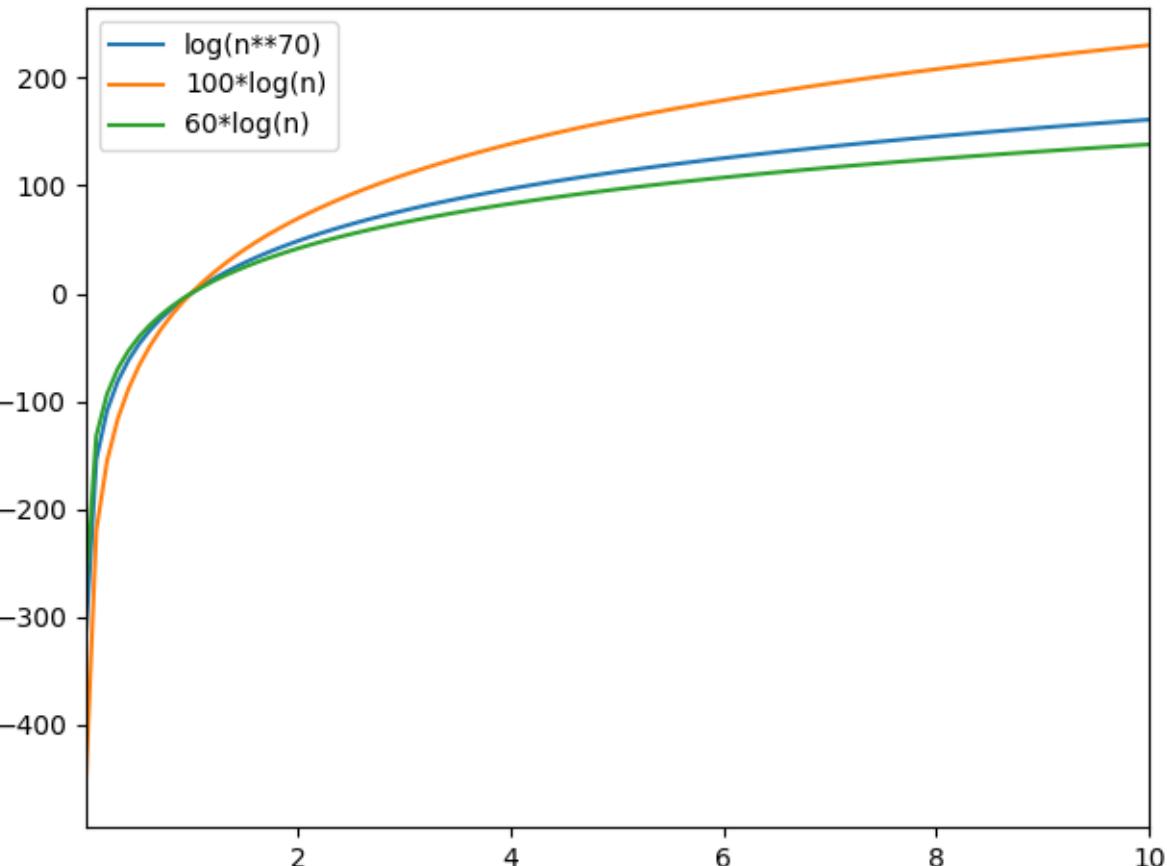
→ $f_1 \in \Theta(4^n)$

$$f_2(n) = \frac{n(n - 1)}{2} = \frac{n^2}{2} - \frac{n}{2}$$



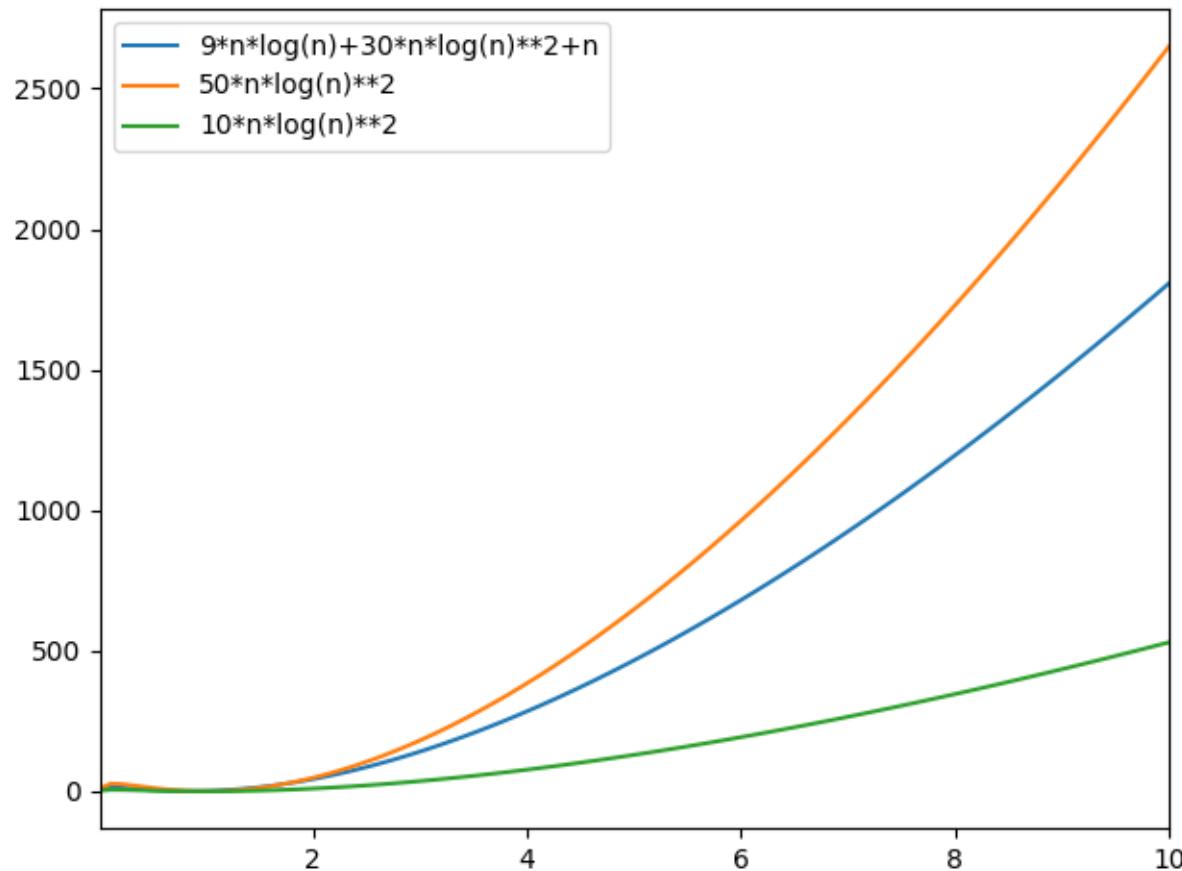
→ $f_2 \in \Theta(n^2)$

$$f_3(n) = \log(n^{70}) = 70 \log(n)$$



→ $f_3 \in \Theta(\log(n))$

$$f_4(n) = 9n\log(n) + 30n(\log(n))^2 + n$$



→ $f_3 \in \Theta(n\log(n)^2)$

Exercise 1.3

Exercise 1.3

- Given Algorithm 1, explain in your own words, what the algorithm does.
- Determine its worst-case time complexity.
- Assuming that nums is sorted. Implement the algorithm $\text{algo2}(\text{nums}, v)$ in the provided .ipynb that solves the problem in $O(\log(n))$, where n is the length of nums .

Solution 1.3

Algorithm 1 algo1(nums, v)

```
for i = 0 to nums.length – 1 do
    if nums[i] == v then
        return i
    end if
end for
return NIL
```

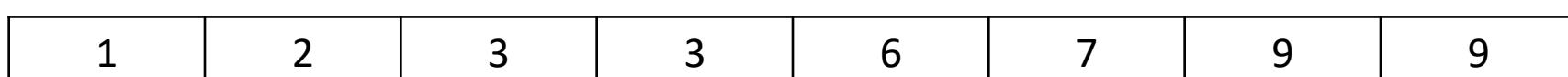
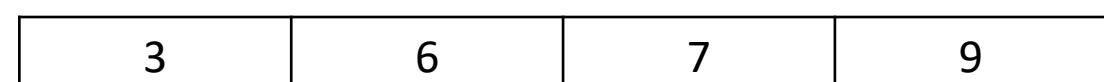
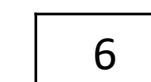
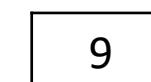
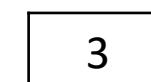
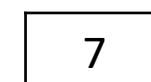
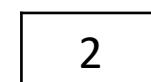
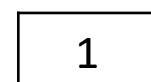
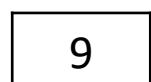
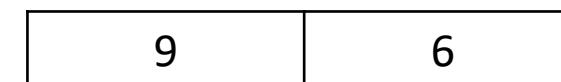
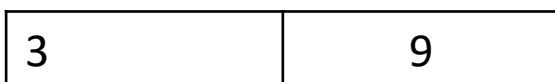
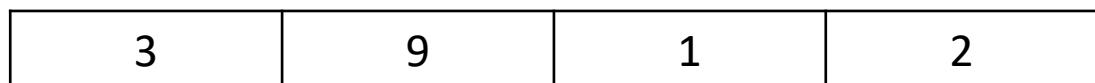
- Search for *v* in *nums*
- Return index of *v*, if found, otherwise NIL
- Complexity: $algo1 \in O(\text{nums})$

Solution 1.3

- Algorithm we are looking for: „binary search“
- Idea (divide and conquer):
 - Compute median of the array (half the array)
 - Check whether v is found?
 - Check whether v is lower or higher than median
 - Proceed with the corresponding half

Exercise 2.1

Illustrate each step of merge sort on the following sequence: <3,9,1,2,7,3,9,6>.



Exercise 2.2

What value does partition return when all elements in the subarray $A[p:r]$ have the same value?

```
partition(A, p, r)
    pivot = A[r]
    s = p - 1 # index of the
    for i = p to r - 1:
        if A[i] ≤ pivot
            s = s + 1
            exchange(A[i], A[s])
    exchange(A[s+1], A[r])
    return s + 1
```

- $A[i] \leq pivot$ is always true
- → method returns r

```
quick_sort(A, p, r)
    q = partition(A, p, r)
    quick_sort(A, p, q - 1)
    quick_sort(A, q + 1, r)
```

Exercise 2.3

Give a brief argument that the running time of partition on a subarray of size n is $O(n)$

```
partition(A, p, r)
    pivot = A[r]
    s = p - 1 # index of the
    for i = p to r - 1:
        if A[i] ≤ pivot
            s = s + 1
            exchange(A[i], A[s])
    exchange(A[s+1], A[r])
    return s + 1
```

- $p - (r - 1)$ iterations in the loop
 - Each takes constant time

$$\rightarrow p - r = n$$

Exercise 2.4

Recap: insersort

```
insert_sort(L) # L is a list of numbers
    for i = 1 to L.length - 1 # 0-indexing,
        key = L[i]
        j = i-1
        while j > -1 and L[j] > key
            L[j+1] = L[j]
            j = j - 1
        L[j+1] = key
```