

Time complexity

Week 13

The running time

- Typically, the running time of an algorithm depends on the size of the input.
- When the input grows, also the running time grows.
- Let us denote by n the size of the input. For instance, if an algorithm has a list as its input, the size n is the number of the elements in the list.
- We denote by $T(n)$ the running time of the algorithm with size n input.

Running time of a loop

For instance, in the loop

```
for i in range(n):  
    k = k + i
```

the size of the "input" is n . This is because the loop runs n times, and the variable k is updated in each iteration.

We can think that each addition $k = k + i$ takes a constant time c . Therefore, the running time of the algorithm is

$$T(n) = cn$$

Loop inside a loop

Similarly, the running time of

```
for i in range(n):  
    for j in range(n):  
        k = k + i + j
```

is $T(n) = c \cdot n^2$

A practical test

```
import time

n = 1000
k = 0

start_time = time.time()    # Current time in seconds since the epoch

for i in range(n):
    for j in range(n):
        k = k + i + j

end_time = time.time()

running_time = end_time - start_time
print(f"With the size of n = {n}, the running time is {running_time:.3f} seconds")
```

File: two_loops.py

Running times

- With the size of $n = 1\,000$, the running time is 0.070 seconds
- With the size of $n = 10\,000$, the running time is 8.211 seconds
- With the size of $n = 50\,000$, the running time is 192.962 seconds
- The running time starts to grow very fast. This means that an algorithm which contains loops inside a loop will be non-usable very fast.
- With three nested loops the situation is much worse.

Grow rate of functions

n	$n \lg(n)$	n^2	n^3
10	33.22	10^2	10^3
100	664.39	10^4	10^6
1000	9965.78	10^6	10^9
10000	132877.12	10^8	10^{12}
100000	1660964.39	10^{10}	10^{15}

In the above table, $\lg n$ denotes the base-2 logarithm of n .