

Operations Research

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- Welcome to USEEN3 : Operations Research !
- **Academic Team** : Safia Kedad-Sidhoum and Daniel Porumbel.
- Objectives and program :
 - ▶ Modeling combinatorial optimization problems
 - ▶ Linear programming
 - ▶ Integer linear programming
 - ▶ Shortest path algorithms
 - ▶ Minimum spanning tree algorithms
 - ▶ Network flow algorithms
 - ▶ **Applications** : Routing and traffic, Network design, sudo snap install typora Network connectivity and reliability, Energy consumption

Quick Overview and Organization (2/2)

- Lesson 1 - Tue. 07/10
Introduction to networks models / computational complexity
- Lesson 2-3 - Tue. 14/10 & Tue. 21/10
Graph search algorithms - Application : Network accessibility
- Lesson 4 - Tue. 28/10
Shortest paths (Dijkstra) - Application : Routing tables
- Lesson 5 - Tue. 4/11
Shortest paths (Bellman/Ford)
- Lesson 6 - Tue. 18/11
Minimum spanning trees - Application : Network design
- Lesson 7 - Tue. 25/11
Linear Programming
- Lesson 8-9 - Tue. 02/12 & Tue. 09/12
Integer Programming
- Lesson 10 - Tue. 16/12 :
Maximum flow problem - Min-cost flow - Application : Network flow routing

Exam Tue. 6/01/2026

References

Books...

- The Algorithms Illuminated Book Series, Tim Roughgarden, 2017.
- Algorithm Design, Eva Tardos and Jon Klein, 2005.
- Network Flows : Theory, Algorithms, and Applications, Ravindra Ahuja, Thomas Magnanti, James Orlin, 1993

- Important for all other branches of computer science.
 - ▶ Routing protocols in communication networks piggyback on classical shortest path algorithms.
 - ▶ Public-key cryptography relies on efficient number-theoretic algorithms.
 - ▶ Computer graphics requires the computational primitives supplied by geometric algorithms.
 - ▶ Database indices rely on balanced search tree data structures.
 - ▶ Computational biology uses dynamic programming algorithms to measure genome similarity.
- Driver of technological innovation (PageRank algorithm for computing the relevance of various Web pages to a given search query).
- Lens on other sciences.

Algorithm efficiency?

What is a **Fast** Algorithm?

A **fast algorithm** is an algorithm whose worst-case running time grows slowly with the input size

Asymptotic notation provides the basic vocabulary for discussing the design and analysis of algorithms.

- Worst-Case Analysis : gives a running time bound that is valid even for the “worst” inputs
- “Big-picture” analysis : balance predictive power with mathematical tractability by ignoring constant factors and lower-order terms
- Asymptotic Analysis : focus on the rate of growth of an algorithm’s running time, as the input size n grows large

Asymptotic notation

- **High level idea** : Suppress constant factors (too system-dependent) and lower-order terms (irrelevant for large inputs)

Examples (Simple)

- Searching an array
- Searching two arrays
- Checking for a Common Element
- Checking for Duplicates

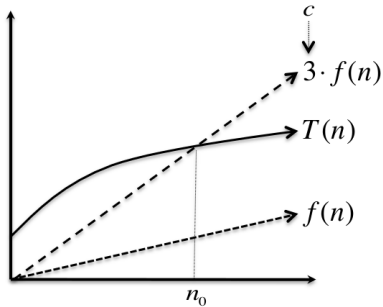
Codes + Quizz + Discussions

O notation (1/2)

It concerns functions $T(n)$ defined on the positive integers $n = 1, 2, \dots$.

Definition (Intuitive)

$T(n) = O(f(n))$ if and only if $T(n)$ is eventually bounded above by a constant multiple of $f(n)$.



$T(n)$ will denote a bound on the worst-case running time of an algorithm, as a function of the size n of the input.

Navigation icons: back, forward, search, etc.

Ω and Θ notations (1/2)

Definition (Ω)

$T(n) = \Omega(f(n))$ if and only if there exist positive constants c and n_0 such that $T(n) \geq cf(n)$ for all $n \geq n_0$.

Definition (Θ)

$T(n) = \Theta(f(n))$ if and only if there exist positive constants c_1 , c_2 and n_0 such that $c_1 f(n) \leq T(n) \leq c_2 f(n)$ for all $n \geq n_0$.

Algorithm designers often use O notation even when Θ notation would be more accurate. This course will follow that line as we generally focus on upper bounds about how long our algorithms could possibly run.

Example (Quizz)

Navigation icons: back, forward, search, etc.

O notation (2/2)

Definition (Mathematical)

$T(n) = O(f(n))$ if and only if there exist positive constants c and n_0 such that $T(n) \leq cf(n)$ for all $n \geq n_0$.

For all $n \geq n_0$ expresses that the inequality only needs to hold eventually, once n is sufficiently large (with the constant n_0 specifying how large).

Remark : c and n_0 are constants, they cannot depend on n .

Property

Suppose $T(n) = a_k n^k + \dots + a_1 n + a_0$, where $k \geq 0$ is a nonnegative integer and the a_i 's are real numbers (positive or negative). Then $T(n) = O(n^k)$.

Proof.

Navigation icons: back, forward, search, etc.

Ω and Θ notations (2/2)

Example (Additional examples)

- Adding a Constant to an Exponent
- Multiplying an Exponent by a Constant
- Maximum vs. Sum

Navigation icons: back, forward, search, etc.

Running Time

The running times of different algorithms on inputs of increasing size, for a processor performing a million high-level instructions per second.

Very long : exceeds 10^{25} years

	n	$n \log_2 n$	n^2	n^3	1.5^n	2^n	$n!$
$n = 10$	< 1 sec	< 1 sec	< 1 sec	< 1 sec	< 1 sec	< 1 sec	4 sec
$n = 30$	< 1 sec	< 1 sec	< 1 sec	< 1 sec	< 1 sec	18 min	10^{25} years
$n = 50$	< 1 sec	< 1 sec	< 1 sec	< 1 sec	11 min	36 years	very long
$n = 100$	< 1 sec	< 1 sec	< 1 sec	1 sec	12,892 years	10^{17} years	very long
$n = 1,000$	< 1 sec	< 1 sec	1 sec	18 min	very long	very long	very long
$n = 10,000$	< 1 sec	< 1 sec	2 min	12 days	very long	very long	very long
$n = 100,000$	< 1 sec	2 sec	3 hours	32 years	very long	very long	very long
$n = 1,000,000$	1 sec	20 sec	12 days	31,710 years	very long	very long	very long