

Numerical Algorithms (MU4IN910)

Lecture 1: Introduction to MATLAB and to floating-point arithmetic

Stef Graillat

Sorbonne Université



Lecturer: Stef Graillat (office 26-00/313)

stef.graillat@sorbonne-universite.fr

Assignments, exams and grading

- 1 exam (50 %) + practicals (50 %)

Schedule

- 5 lectures: wednesday 8h30-10h30
- 5 tutorials: wednesday 10h45-12h45
- 5 practicals as a personal work in pairs
- Practical due one week after
→ report written in L^AT_EX and submitted in one pdf file on Moodle

Website: <http://www-pequan.lip6.fr/~graillat/teach/anum/>

Moodle: <https://moodle-sciences-22.sorbonne-universite.fr/course/view.php?id=4479>

Goals:

- **Mathematical concept**: mathematical definition of concepts and quantities
- **Algorithm**: how to efficiently calculate these quantities on a computer (via the use of MATLAB)?
- **Problem solving**: use concepts and algorithms to solve real-life problems

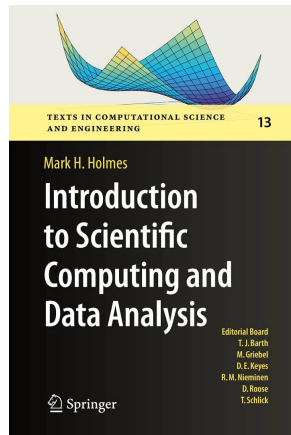
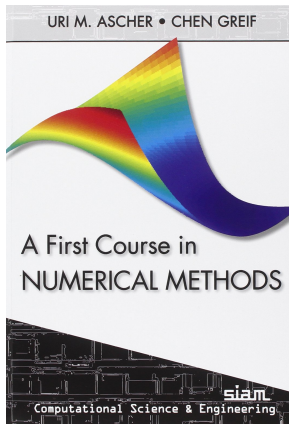
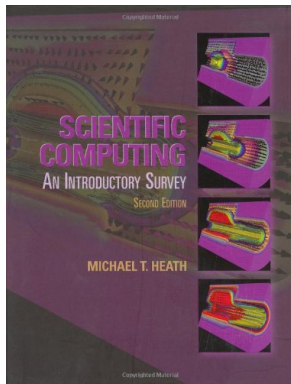
Numerical Algorithms

- 1 Introduction to MATLAB and floating-point arithmetic
- 2 Matrix computation
- 3 Introduction to numerical optimization
- 4 Nonlinear equations

Main references

- **A First Course in Numerical Methods**, Uri M. Ascher, Chen Greif, SIAM, 2011
- **Scientific Computing, An Introductory Survey**, Michael T. Heath, McGraw-Hill, 2002
- **Scientific Computing with Case Studies**, Dianne P. O'Leary, SIAM, 2009
- **Introduction to Scientific Computing and Data Analysis**, Mark H. Holmes, Springer, 2016
- **Mathématiques appliquées L3**, sous la direction de Jacques-Arthur Weil et Alain Yger, Pearson, 2009
- **Numerical Computing with MATLAB**, Cleve Moler, SIAM, 2004
- **MATLAB Guide**, Desmond J. Higham, Nicholas J. Higham, 3e édition, SIAM, 2017
- **Scientific Computing, An Introduction using Maple and MATLAB**, Walter Gander, Martin Gander, Felix Kwok, Springer, 2014
- **Numerical Recipes. The Art of Scientific Computing**, William Press, Saul Teukolsky, William Vetterling et Brian Flannery, 3rd Edition, Cambridge University Press, 2007

Main references



The concepts seen in this course can be applied to:

- robotics
- signal processing
- image processing
- finance
- biology
- etc.

1. Floating-point arithmetic

Can we count to 6 with a computer?

$2 - 1$	1.0000000000000000
$\left(\frac{1}{\cos(100\pi + \pi/4)} \right)^2$	2.0000000000000111
$3 \frac{\tan(\arctan(10000))}{10000}$	2.999999999997162
$\left(\left(\dots \left(\sqrt{\sqrt{\dots \sqrt{4}}} \right)^2 \dots \right)^2 \right)^2$ (20 times)	4.000000000629434
$5 \times \left\{ \frac{(1 + e^{-100}) - 1}{(1 + e^{-100}) - 1} \right\}$	NaN
$\frac{\log(e^{6000})}{1000}$	Inf

Floating-point numbers

A normalized floating-point number $x \in \mathbb{F}$ is a number which is written in the form

$$x = \pm \underbrace{x_0.x_1 \dots x_{p-1}}_{\text{mantissa}} \times b^e, \quad 0 \leq x_i \leq b-1, \quad x_0 \neq 0$$

b : the base, p : precision, e : exponent satisfying $e_{\min} \leq e \leq e_{\max}$

Machine precision $\epsilon = b^{1-p}$, $|1^+ - 1| = \epsilon$

Approximation of \mathbb{R} by \mathbb{F} , rounding $\text{fl} : \mathbb{R} \rightarrow \mathbb{F}$

Let $x \in \mathbb{R}$ then

$$\text{fl}(x) = x(1 + \delta), \quad |\delta| \leq \mathbf{u}.$$

Unit roundoff \mathbf{u} is equal to $\mathbf{u} = \epsilon/2$ for round to nearest

Standard model of floating-point arithmetic

IEEE 754 standard (1985,2008,2019)

- The arithmetic operations ops $(+, -, \times, /, \sqrt{})$ are performed as if they were calculated in infinite precision and then rounded off
- Default: rounded to nearest

Type	Size	Mantissa	Exponent	Unit roundoff	Interval
binary32 (simple)	32 bits	23+1 bits	8 bits	$\mathbf{u} = 2^{-24} \approx 5,96 \times 10^{-8}$	$\approx 10^{\pm 38}$
binary64 (double)	64 bits	52+1 bits	11 bits	$\mathbf{u} = 2^{-53} \approx 1,1 \times 10^{-16}$	$\approx 10^{\pm 308}$

Let $x, y \in \mathbb{F}$,

$$\text{fl}(x \circ y) = (x \circ y)(1 + \delta), \quad |\delta| \leq \mathbf{u}, \quad \circ \in \{+, -, \cdot, /\}$$

Exceptions

The arithmetic is “closed ”: each operation returns a result.

Exception	Results
Invalid operation	NaN (Not a Number)
Overflow	$\pm\infty$
Divide by zero	$\pm\infty$
Underflow	Denormalized numbers
Inexact	correctly rounded result

NaN is generated by operations such as $0/0$, $0 \times \infty$, ∞/∞ , $(+\infty) + (-\infty)$ and $\sqrt{-1}$.

Infinite symbols satisfy $\infty + \infty = \infty$, $(-1) \times \infty = -\infty$ and $(\text{fini})/\infty = 0$.

Accuracy of the computation

At each rounding, we lose a priori a bit of accuracy, we talk about **rounding error**.

Even if an isolated operation returns the best possible result (rounding of the exact result), a series of calculations can lead to large errors due to the accumulation of rounding errors.

The two main sources of rounding error during calculations are **cancellation** and **absorption**.

Example of cancellation

Let $f(x) = (1 - \cos(x))/x^2$, then $0 \leq f(x) < 1/2$ for all $x \neq 0$.

With $x = 1.2 \times 10^{-5}$, the cosine rounded to 10 significant digits is equal to

$$c = 0.9999\ 9999\ 99,$$

so

$$1 - c = 0.0000\ 0000\ 01$$

Therefore $(1 - c)/x^2 = 10^{-10}/1.44 \times 10^{-10} = 0.6944 \dots !!!!$

However, the subtraction $1 - c$ is exact.

To avoid the cancellation, rewrite f in the form

$$f(x) = \frac{1}{2} \left(\frac{\sin(x/2)}{x/2} \right)^2$$

Example of absorption

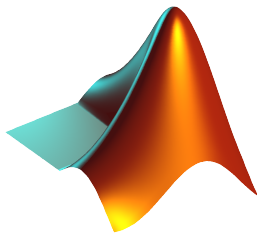
We calculate numerically, for large values of N , the sum:

$$\sum_{i=1}^N \frac{1}{i}$$

Results of a C program (floating-point single precision) on a Pentium 4 processor:

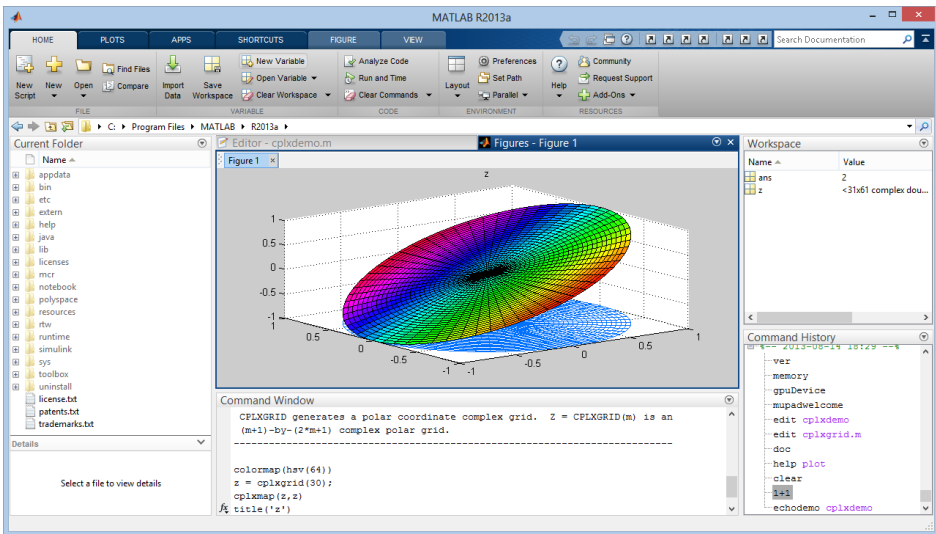
order	N			
	10^5	10^6	10^7	10^8
exact	1.209015e+01	1.439273e+01	1.669531e+01	1.899790e+01
$1 \rightarrow N$	1.209085e+01	1.435736e+01	1.540368e+01	1.540368e+01
$N \rightarrow 1$	1.209015e+01	1.439265e+01	1.668603e+01	1.880792e+01

2. Introduction to MATLAB



- MATLAB = MATrix LABoratory
- a programming language and a development environment
- MATLAB was designed by Cleve Moler in the late 1970s
- MATLAB is supplemented by multiple toolboxes
- MATLAB language supports OOP
- Interaction possible with C and Fortran languages
- for help on a command `command`, use `help command` or `doc command`
- to write comments `%`

Reference: MATLAB Guide, Desmond J. Higham, Nicholas J. Higham, 3rd edition, SIAM, 2017



- MATLAB was created in the 1970s by Cleve Moler, professor of mathematics at the University of New Mexico
- MATLAB was created from the Fortran LINPACK and EISPACK libraries
- MATLAB then evolved, integrating the LAPACK library in 2000
- There are free alternatives to MATLAB such as GNU Octave, FreeMat and Scilab
- The current version of MATLAB is MATLAB R2022b (version 9.13)

Vectors and matrices in MATLAB

MATLAB variables are mainly **vectors** and **matrices**

- Vectors:

- row vector

```
>> format compact
```

```
>> x = [1.1 10.1 100.1]
```

```
x =
```

```
    1.1000   10.1000   100.1000
```

- column vector

```
>> x = [1.1; 10.1; 100.1]
```

```
x =
```

```
    1.1000
```

```
   10.1000
```

```
  100.1000
```

Vectors in MATLAB

- Display and assignment

```
>> x = [1.1; 10.1; 100.1];
```

```
>> x
```

```
x =
```

```
    1.1000
```

```
   10.1000
```

```
  100.1000
```

```
>> x(3) = -1.1
```

```
x =
```

```
    1.1000
```

```
   10.1000
```

```
   -1.1000
```

The vectors are indexed from 1 (and not 0 as in C)

Vectors in MATLAB

- Transpose of a vector

```
>> x = [1.1 10.1 100.1]'
```

```
x =
```

```
    1.1000
```

```
   10.1000
```

```
  100.1000
```

- To enter a vector or a command occupying more than one line

```
>> x = [0 .05 .10 .15 .20 .25 .30 .35 .40 .45 .50 ...  
       .55 .60 .65 .70 .75 .80 .85 .90 .95 1];
```

- Length of a vector

```
>> length(x)
```

```
ans =
```

```
    21
```

Vectors in MATLAB

- Vector of any size (by default vectors are row vectors)

```
n = 20;  
h = 1/n;  
for k=1:n  
    x(k) = k*h;  
end
```

- Initialization of a column vector

```
x = zeros(n,1);
```

- colons. The notation $a:b$ denotes the row vector going from a to b in steps of 1 while for $a:s:b$ the step is s

```
>> x = 1:5
```

```
x =
```

```
1 2 3 4 5
```

```
>> x = 4:-1:0
```

```
x =
```

```
4 3 2 1 0
```

Vectors in MATLAB

- `>> x = 0:0.05:1; % vector with 21 components.`
`>> x = 0.05*(0:20) % another way to generate the same vector`
- Access parts of a vector
`x(1:4)` extract the first 4 elements of `x`
- `linspace(a,b,n)` produces a row vector with n components which divides $[a, b]$ in $n - 1$ equal intervals.
`x = linspace(0,1,21);`

Matrices in MATLAB

The power of MATLAB comes from its matrix operations

- fast
- and accurate

2 ways to enter matrices

```
>> A = [1 2 3  
        2 4 7  
        -1 0 5]
```

```
A =  
    1 2 3  
    2 4 7  
   -1 0 5
```

```
>> a = [1 2 3; 1 4 8; 3 -1 0]
```

```
a =  
    1 2 3  
    1 4 8  
    3 -1 0
```

Special functions for creating matrices

- `zeros(m,n)` returns a matrix of 0 of size $m \times n$

```
>> A = zeros(2,4)
```

```
A =
```

```
0 0 0 0
```

```
0 0 0 0
```

- `ones(m,n)` returns a matrix of 1 of size $m \times n$

```
>> A = ones(3)
```

```
A =
```

```
1 1 1
```

```
1 1 1
```

```
1 1 1
```

- `eye(n)` returns the identity matrix of size n

```
>> A = eye(3)
```

```
A =
```

```
1 0 0
```

```
0 1 0
```

```
0 0 1
```

- Solving a linear system $Ax = b$

```
>> A = [1 2 3; 2 4 7; -1 0 5];
```

```
>> b = [1 1 1]';
```

```
>> x = A\b
```

```
x =
```

```
-6
```

```
5
```

```
-1
```

- Another use of colons:

```
>> C = A([1 3], :)
```

```
C =
```

```
    1    2    3  
   -1    0    5
```

- Vectorized functions

We want to evaluate a function on a vector of value x_i :

$a = x_1 < x_2 < \dots < x_n = b$.

Standard functions take vectors as arguments and return a vector.

```
n=21;
```

```
x = linspace(0,2*pi,n);
```

```
y = cos(x);
```

- In general, one write the list of commands in a text file (via the `edit` editor integrated into MATLAB) or via a favorite text editor
- To save variables, use the `save` function. Saving the variables `A` and `b` in the file `svar.mat` can be done by:

```
save svar A b
```

One reload the variables by

```
load svar
```

- type `help` command.

```
>> help length
```

```
LENGTH    Length of vector.
```

```
    LENGTH(X) returns the length of vector X.  It is  
    equivalent to MAX(SIZE(X)) for non-empty arrays  
    and 0 for empty ones.
```

- to have it in graphic mode, type `doc`

```
>> doc length
```

2 types of file (which have the extension .m):

- **script M-files**: neither input nor output and uses workspace variables
- **function M-files**: contains a function that accepts input arguments and returns output arguments and internal variables are local to the function

Example (to save in a file `sumprod.m`):

```
function [s,p]= sumprod(x)
    n = length(x);
    s=0;
    p=1;
    for i=1:n
        s = s + x(i);
        p = p*x(i);
    end
```

Structure of an M-files function

- 1 the keyword `function`
- 2 list of output arguments (in brackets `[]` if there are several ones)
- 3 the symbol `=`
- 4 the name of the function (which must be the same as the name of the `.m` file)
- 5 the list in parenthesis of the entries
- 6 the body of the function

To edit the files, type `edit`

Useful commands: `dir`, `ls`, `cd`, `type`, `lookfor`, `path`

- Numeric format: command format to display fixed or floating-point numbers

```
>> format short, pi^4 % fixed, 5 digits  
ans =  
    97.4091
```

```
>> format shortE, pi^4 % float, 5 digits  
ans =  
    9.7409e+001
```

```
>> format long, pi^4 % fixed, 15 digits  
ans =  
    97.40909103400242
```

```
>> format longE, pi^4 % float, 15 digits  
ans =  
    9.740909103400242e+001
```

- Display of strings

- the command `disp` allows to display a string or a variable

```
>> x=3;
```

```
>> disp(x);
```

```
3
```

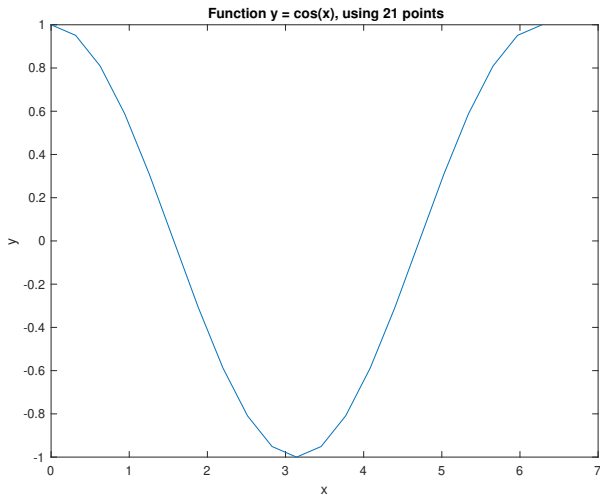
```
>> disp('test');
```

```
test
```

- the command `fprintf` allows to display strings and numbers (similar to the `printf` command in C)

Plot of a function $y = f(x)$ over an interval $[a, b]$

```
n = 21;  
x = linspace(0,2*pi,n);  
y = cos(x);  
plot(x,y)  
title('Function y = cos(x), using 21 points')  
xlabel('x')  
ylabel('y')
```



Control structures and tests

- loop for
 for variable = expression
 instructions
 end
- loop while
 while expression
 instructions
 end
- test if

```
if expression
    instructions
end
```

```
if expression
    instructions
else
    instructions
end
```

- relational operations

==	equal
~=	different
<	strictly lower
>	strictly greater
<=	less or equal
>=	greater or equal

- logical operations

&	and
	or
~	not

- The toolbox contains the kernel of MuPAD and MATLAB functions which communicate with this kernel
- New type: objects `sym` created by the `sym` and `syms` commands
- Create a symbolic variable `x`

```
>> syms x
```

Symbolic Math Toolbox

- Manipulation of symbolic expression in x

```
>> f = 1/(1+x^2)
```

```
f =
```

```
1/(1+x^2)
```

```
>> g = int(f) % integration
```

```
g =
```

```
atan(x)
```

```
>> diff(g) % differentiation
```

```
ans =
```

```
1/(1+x^2)
```

```
>> syms a
```

```
>> y = solve(f-a) % solves f(x)-a=0
```

```
y = [ 1/a*(-a*(-1+a))^(1/2)]
```

```
[-1/a*(-a*(-1+a))^(1/2)]
```


- Multiprecision arithmetic: function `vpa`

```
>> digits % by default
```

```
Digits = 32
```

```
>> vpa('sqrt(2)')
```

```
ans =
```

```
1.4142135623730950488016887242097
```

```
>> digits(50)
```

```
>> vpa('sqrt(2)')
```

```
ans =
```

```
1.4142135623730950488016887242096980785696718753769
```

- Exact arithmetic: one manipulate symbolic expressions

```
>> z = sym('sqrt(2)')
```

```
z =
```

```
sqrt(2)
```

```
>> z^2-2
```

```
ans =
```

```
0
```

Matrix storage

For **programming efficiently** algorithms on matrices, it is very important to know how matrices are **stored** in **memory**!

Example: **matrix-vector product**

Given a matrix A of size $m \times n$ and a vector x of length n , one wants to compute $y = Ax$

- The vector y is defined by dot products between the lines of A and x

$$y_i = A(i,:)x$$

```
[m,n]=size(A);  
y = zeros(m,1);  
for i=1:m,  
    for j=1:n,  
        y(i) = y(i) + A(i,j)*x(j);  
    end  
end
```

- We can express Ax using the columns of A instead

$$Ax = x_1 A(:, 1) + x_2 A(:, 2) + \cdots + x_n A(:, n)$$

which is equivalent to compute the transpose of $(Ax)^T$

```
[m, n] = size(A);  
y = zeros(m, 1);  
for j = 1:n,  
    for i = 1:m,  
        y(i) = y(i) + A(i, j)*x(j);  
    end  
end
```

Both algorithms perform mn multiplications and mn additions!

- The computer stores the information in **memory pages**
- Part of the information is stored in **caches**
- What does not fit in the cache is stored in the **main memory** (slower to access)
- Finally, what does not fit in main memory is stored on the **hard disk**

To use data, **the page containing the data must be moved to cache** (therefore, other data must be cleared from the cache), these pages being at best in the RAM (most of the time when the calculations are reasonable) or at worst in the hard disk (for example for the multiplication of square matrices of several million lines).

For an algorithm to be effective, it is necessary to limit the number times a page is moved to cache!

Question to ask when one wants to multiply a matrix and a vector:

What is the algorithm to use to make the algorithm efficient?

The right question to ask is in fact:

Are the matrices ordered by **row** or by **column**?

Language	Storage scheme
C, C++	by row
Fortran	by column
Java	by row
MATLAB	by column

Basic tools for matrix manipulations: the BLAS

There are tasks that are used in almost all matrix problems!

Libraries have been developed to prevent programmers do not have to reprogram them each time

The **Basic Linear Algebra Subroutines** or BLAS are now available for almost all programming languages

The BLAS are divided into **levels**. Level- k BLAS need $\mathcal{O}(n^k)$ operations (here n is the size of the vectors or matrices)

Basic tools for matrix manipulations: the BLAS

Example 1

- *Level-1 BLAS: vector operations*
 - ① *sscal* computes ax where a is a scalar and x is a vector
 - ② *saxpy* computes $ax + y$
 - ③ *sdot* computes $x^* y$
- *Level-2 BLAS: matrix-vector operations*
 - ① *matrix-vector product*
 - ② *solution of linear systems involving triangular matrix*
- *Level-3 BLAS: matrix-matrix operations*
 - ① *matrix-matrix product*
 - ② *solution of multiple linear systems involving triangular matrix*

When a BLAS exist for a task that is needed, it is a good idea to use it on the algorithm

MATLAB automatically uses the BLAS. In other languages, one need to call specific subroutines to do it.

MATLAB and Maple at PPTI

- MATLAB is available at PPTI. It is MATLAB R2019b (version 9.7) with the Symbolic Math Toolbox.

To run MATLAB, type in a terminal

```
/usr/local/Matlab-2019/R2019b/bin/matlab
```

- Maple is also available at PPTI. It is Maple 2020.

To run Maple, type `/usr/local/maple2020/bin/xmaple`

Agreement with MATLAB and Maple at SU

- SU now makes available to you (for staff and students) MATLAB
http://logiciels.upmc.fr/fr/marches_conclus_par_l_upmc/matlab.html
- Whether you are a teacher, researcher, student, you can have a Maple license on your personal commuter
http://logiciels.upmc.fr/fr/marches_conclus_par_l_upmc/maple.html

Conclusion

To get an overview of MATLAB functionality, type `demo`

Representation of real numbers

Examples

Integer part - fractional part In base β a number can be represented by

$$A = \sum_{-\infty}^{+\infty} a_i \beta^i$$

Rational numbers The fractional part $A_f = \sum_{-\infty}^{-1} a_i \beta^i$ is periodic from a certain rank.

$$\frac{237}{315} = 0.7523809523809\overline{523809}$$

Real numbers The fractional part can be aperiodic.

$$\sqrt{2} = 1.414213562373095048801688724209 \dots$$

$$\sqrt{2} = 1 + \frac{1}{2} - \frac{1}{8} + \frac{1}{16} - \frac{5}{128} \dots + (-1)^{n+1} \frac{\binom{2n}{n}}{(2n-1)2^{2n}}$$

$$e = 1 + \sum_{n=1}^{\infty} \frac{1}{n!} = 2.71828182845904523536 \dots$$

$$\pi = 4 \sum_{n=0}^{\infty} \frac{(-1)^n}{2n+1} = 3.141592653589793238 \dots$$

- Representation of normalized numbers

$$X = (-1)^s \times x_0, x_1 x_2 \dots x_n \times \beta^e = (-1)^s \times \sum_{i=0}^n x_i \times \beta^{-i+e}$$

with $x_0 \neq 0$ (except if $X = 0$).

- The number of digits is fixed: notion of approximate value.
- $\frac{-237}{315} = (-1)^1 \times 7.523809523 \times 10^{-1}$
 $\sqrt{2} = (-1)^0 \times 1.414213562 \times 10^0$

- A number is represented in single precision (32 bits) or in double precision (64 bits) as follows:

s(1)	e = exp biased (8 or 11)	... mantissa (23 or 52)
------	--------------------------	-------------------------

- If $e \neq 0$ and $e \neq 255(11111111)$ or $e \neq 2047(1111111111)$

$$(-1)^s \times 2^{(e)-b} \times \textcolor{violet}{1}, \dots \text{mantissa} \dots$$

The bias is $b = 127$ (01111111) in single precision, and $b = 1023$ (01111111111) in double precision.

- if, exp biased = 0:

$$(-1)^s \times 2^{1-\text{bias}} \times \textcolor{violet}{0}, \dots \text{mantissa} \dots$$

- If $e = 255(11111111)$ (or $e = 2047(1111111111)$):
then the number represents infinity if the mantissa is 0 or a NaN (not a number) if not.

- If $e \neq 0$ and $e \neq 255(11111111)$ or $e \neq 2047(1111111111)$ then the number is said to be normalized and the significand has an implicit 1 at the beginning
- Otherwise, the number is denormalized. This makes it possible to represent very small numbers
- Remark: 0 has a sign

Examples

- $1 = 0\ 01111111\ 000000000000000000000000$
- $2 = 0\ 10000000\ 000000000000000000000000$
- $3 = 0\ 10000000\ 100000000000000000000000$
- $9 = 0\ 10000010\ 001000000000000000000000$
- $1/2 = 0\ 01111110\ 000000000000000000000000$
- $1/3 = 0\ 01111101\ 010101010101010101010101$ _{101010101...}
- $1/10 = 0\ 01111011\ 10110011001100110011001$ _{100110011...}

Special values

Number	Representation in double precision (hexadecimale)	Decimal value
+0	00000000 00000000	0.0
-0	80000000 00000000	-0.0
1	3FF00000 00000000	1.0
2	40000000 00000000	2.0
Normalized maximum	7FEFFFFFF FFFFFFFF	$1.7976931348623157e + 308$
Normalized positive minimum	00100000 00000000	$2.2250738585072014e - 308$
Denormalized maximum	000FFFFFF FFFFFFFF	$2.2250738585072009e - 308$
Denormalized positive minimum	00000000 00000001	$4.9406564584124654e - 324$
$-\infty$	FFF00000 00000000	$-\infty$
$+\infty$	7FF00000 00000000	$+\infty$
NaN	7FF xxxx...xxxxxxxx	Not A Number

IEEE 754 standard: rounding

- If x and y are 2 representable numbers, then the result of an operation $res = x \odot y$ is not, in general, a representable number.
- For example, in base $B = 10$, the number $1/3$ is not representable with a finite number of digits.
- It is necessary to **round** the result, that is to say to return one of the closest representable numbers.

IEEE 754 standard: rounding modes



The norm proposes 4 rounding modes:

- rounding **toward $+\infty$** denoted $\Delta(x)$: return the smallest floating-point number greater or equal the exact result x
- rounding **toward $-\infty$** denoted $\nabla(x)$: return the largest floating-point number less or equal the exact result x
- rounding **toward 0**, denoted $\mathcal{Z}(x)$: return $\Delta(x)$ for negative numbers and $\nabla(x)$ for positive numbers
- rounding **to the nearest**, denoted $\circ(x)$: return the nearest floating-point number of the exact result x (breaks ties by rounding to the nearest even floating-point number)

The 3 first rounding modes are called **directed** rounding modes.

IEEE 754 standard: correct rounding

Let x and y be two representable number, \odot be on operations $+$, $-$, \times , $/$ and \diamond a rounding mode.

The standard requires that the result of the computation $x \odot y$ be equal to $\diamond(x \odot_{exact} y)$. The result must be similar to the one obtain by computing with infinite precision and then rounding this result.

Similar for square root.

This property is called **correctly rounding**.

The standard describes an algorithm for addition, subtraction, multiplication, division and square root and requires that the implementation produces the same result as those algorithms.

IEEE 754 standard: comparisons

The standard requires the comparison to be exact and not to overflow.

The specified comparison in the standard are:

- equality
- greater than
- less than

The sign of zero is not taken into account.

In case of a comparison with a NaN, the comparison returns False.

More precisely in case of equality: if $x = \text{NaN}$ then $x = x$ returns False and $x \neq x$ returns True (equality is not reflexive but it is a way to detect a NaN)

IEEE 754 standard: flags

No calculation should hinder the proper functioning of the machine. A mechanism with 5 flags makes it possible to inform the system about the behavior of operations:

INVALID operation the result by default is a NaN

DIVIDE by ZERO the result is $\pm\infty$

OVERFLOW overflow toward ∞ : the result is either $\pm\infty$ or the greatest floating-point number (in absolute value) depending on the sign of the exact result and the rounding mode

UNDERFLOW overflow toward 0: the result is either ± 0 or a subnormal

INEXACT inexact result: raised when the exact result is not representable exactly. Returns the correctly rounded result by default.

A flag, when raised, stay raised until there is a reset by the user (**sticky flags**). They can be read and write by the user.