

Introduction to julia

Presentation and Workshop

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Overview

What is Julia?

Installation & REPL

Main features

Packages

Pluto Notebooks

Workshop: Let's get you started with Julia!



What is Julia?



Goal: Scientific Computing & Fast Prototyping

In scientific computing we need

- high performance to tackle large scale problems
 - \Rightarrow compiled languages (C/C++, Rust)
 - ▶ all types are known at compile time
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 - ⇒ fast prototyping
 - types have to be inferred at runtime
 - code is interpreted (slow)



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Often: Fast code is written in C/C++ and is interfaced.

 \Rightarrow new users might have to compile the C/C++ (e.g. MEX files)



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A short history

2009 Adam Edelman starts the project with Jeff Bezanson, Stefan Karpinski, Viral B. Shah

2012 first public version

2018 Julia 1.0, i.e. no breaking releases since then

2024 Julia 1.11



Resources

```
Main homepage https://julialang.org
Documentation https://docs.julialang.org/en/v1/
Modern Julia Workflows https://modernjuliaworkflows.org/
Discourse https://discourse.julialang.org
Julia Hub webfrontend for the General Registry
  https://juliahub.com/ui/Packages
```

These slides

```
https://github.com/
Julia-Users-Trondheim/Intro-to-Julia/
blob/main/presentation/
introduction-to-julia.pdf
```





Installation & REPL



Installation

Windows Install Julia from the Microsoft Store by running this in the command prompt

```
winget install julia -s msstore
```

We can take a closer look at your individual installation after this presentation in the workshop.



Read-Eval-Print Loop (REPL)

The Julia command line is called REPL.

- for fast computations
- easily define variables & functions
- include("script.jl"); to run a script.



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Quick commands

^D Quit

L Clear console screen

Up Arrow last command



REPL modes

Starting with special characters on REPL enters specific modes

? help mode quick access to the documentation of a function

Example:

? sqrt displays the help for the sqrt function on REPL, see also the (HTML) documentation $\,$

```
https:
```

```
//docs.julialang.org/en/v1/base/math/#Base.sqrt-Tuple{Number}
```

- package mode quick access to manage packages
- ; shell mode quick access to shell without exiting Julia, e. g. to change folders



Main features



General philosophy & Code format

Philosophy

- Write functions not scripts
- Julia has data types, but not objects
- write generic code "acting" on data
- no need to write "vectorized code"
- avoid global variables



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Format

- blocks have an end
- ▶ Indentation with 4 spaces is recommended but not necessary
- ▶ functions that modify their data should be named with an !.



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► To install one for our demos use the package mode

] add Pluto

This has only to be done once.



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We will continue command demos in the Pluto notebook (similar to a Jupyter notebook, but with a persistent state)



```
lterate with for-loops
for i=1:4
    print(i," ")
end # prints "1 2 3 4"
```



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for i=1:4
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Combine several (and use ∈)
for i ∈ 1:3, j ∈ 1:2
    print(i,"×",j,", ")
end # prints 1×1, 1×2, ...
```



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Iterate with for-loops
for i=1:4
    print(i," ")
end # prints "1 2 3 4"
Combine several (and use \in)
for i \in 1:3, j \in 1:2
    print(i,"x",j,", ")
end # prints 1 \times 1, 1 \times 2, ...
Or through several of same length
for (i,j) \in zip(1:4, 5:8)
    print(i,"|",i," ")
end # prints 1/5 2/6 3/7 4/8
```



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```

```
or as a comprehension for vectors
x = [3*s for s \in 1:3]
creates [3, 6, 9]
Loops with "unknown end"
i = 1:
# do as long as i \le 4
while i \le 4
    print(i," ");
    i += 1
end # also prints "1 2 3 4"
```



Control flow II: Conditionals

Conditionals require an expression that evaluates to a Bool. Then if $(x > 3) \mid \mid (z < 2) \text{ # brackets } (x > 3) \text{ are optional print("x is at least 3")}$ else print("x is 3 or less") end



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There is lazy evaluation: the second parts of
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Conditionals can be used inline with

$$y = (x > 4) ? 1 : 3*x$$



```
0.00
    phase(z)
Compute the phase of a complex number z
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function phase(z)
    return atan(imag(z), real(z))
end
 naming convention snake case
```



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- ► (multiline) "String" upfront: doc-string, may use Markdown



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(last evaluated expression returned)



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Shorter form

```
magnitude(z) = sqrt(imag(z)^2+real(z)^2)
```



positional optional parameters are defined by providing defaults

```
f(a, b=2, c=3) = a*exp(b/c)
f(1) #equals f(1,2,3)
f(1,3) #equals f(1,3,3)
f(1,3,5) #equals f(1,3,5)
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- \blacktriangleright in g(1; b=4, b=3) the last one "wins", so b is 3.
- You can "collect and pass on":
 - ▶ h1(args...) = f(1, args...)
 - ► h2(; kwargs...) = g(1; kwargs...)
 - ▶ or combine both as h3(args...; kwargs...) = #def here



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X = [0.1, 0.2, 0.3]; Y = [1.0, 2.0, 3.0]
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▶ functions can modify their input

```
function add_scalar!(X, v)
    X .+= v # X an array, v a scalar: add to every entry
```

end

Convention: such a functions name ends in !, it returns the modified

return X # the X we got passed is now changed



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Variant I. default: immutable

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struct TimeSeries <: ExperimentData
   name::String
   data::Vector</pre>
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end # default constructor:
ts = TimeSeries("A", [1,2,3])

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                                         (modified in-place)
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                                         more efficient
Variant II. mutable – reassign fields:
mutable struct Measurement <: ExperimentData</pre>
    name::String
    value::Float64
end # same constructor
m = Measurement("B", 3.1415)
```



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- m.name="B"; m.value=4.5 both work (if same type)
- slightly less efficient



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```
mutable struct TimeSeries2{T} <: ExperimentData
    param::T  # maybe some concentration
    data::Vector{T} # actually parametrized by element-type
end # Constructor now maybe a bit clumsy:
ts2 = TimeSeries2{Float64}(3.1415, [1.2, 1.3])</pre>
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- ▶ makes the previous (implicit) Vector{Any} to a concrete type
- ▶ nicer constructor: Define a parametric function

```
function TimeSeries2(c::T, v::Vector{T}) where {T}
    return TimeSeries2{T}(c, v)
end # Then we have back
ts2 = TimeSeries2(3.1415, [1.2, 1.3])
```



Dispatch: "finding" the "best fitting version" of a function.



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function. For
f(x) = "A"
f(x::Number) = "B"
f(x::Float64) = "C"
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f(x) = "A"
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We get that
f.(["a", 1, 1.0im, 2.0])
is ["A", "B", "B", "C"]
```



method of a function

```
function g(a::Number, t::TimeSeries)
Dispatch: "finding" the
"best fitting version" of a
                             TimeSeries(t.name, a .* t.data)
function. For
                           end
                           function g(a::String, t::TimeSeries)
f(x) = "A"
                             TimeSeries("$(a) $(t.name)", t.data)
f(x::Number) = "B"
                           end
f(x::Float64) = "C"
                           function g(a::Number, ts::TimeSeries2)
                             TimeSeries2(a*t.param, a .* t)
We get that
                           end
f.(["a", 1, 1.0im, 2.0])
is ["A", "B", "B", "C"]
\Rightarrow dispatch to
"most fitting"
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f.(["a", 1, 1.0im, 2.0])
                           Avoid ambiguities. Defining
is ["A", "B", "B", "C"]
                           g(a::Float64, b) = 2*a+b
                           g(a, b::Float64) = a+2*b
\Rightarrow dispatch to
                           makes g(1.0,2.0) ambiguous.
"most fitting"
method of a function
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\Rightarrow dispatch to
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"most fitting"
                           g(a::Float64, b::Float64) = 2*a + 2*b
method of a function
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Operators are Functions

```
Operators like +, *, ^ are functions. Add a method to + via
function Base.:+(t::TimeSeries, s::TimeSeries)
    if !(length(t.data)==length(s.data))
        error("Time series not of same length")
    end
    return TimeSeries(
        "$(t.name) and $(s.name)",
        t.data .+ s.data
end
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        t.data .+ s.data
end
Then
u = TimeSeries("A", [1,2]) + TimeSeries("B", [3,4])
returns TimeSeries ("A and B", [4, 6]).
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    end
    return TimeSeries(
        "$(t.name) and $(s.name)",
        t.data .+ s.data
end
Then
u = TimeSeries("A", [1,2]) + TimeSeries("B", [3,4])
returns TimeSeries ("A and B", [4, 6]).
To ensure same type parameter, define a function with
Base::+(t::TimeSeries2{T}, s::TimeSeries2{T}) where {T}
```



Functors: function-like structures

```
Consider (actually taken from the Julia documentation)  \begin{tabular}{ll} struct & Polynomial\{R\} \\ & coeffs:: Vector\{R\} \\ end \\ \end \\ \en
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Consider (actually taken from the Julia documentation)
struct Polynomial{R}
    coeffs::Vector(R)
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We can turn a Polynomial into a function as well defining
function (p::Polynomial)(x)
    v = p.coeffs[end] # Horner Schema, (a 2x + a 1)x + a 0
    for i = (length(p.coeffs)-1):-1:1
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end
For p = Polynomial([1, 10, 100]); p(3) we get
100 \cdot 3^2 + 10 \cdot 3 + 1 = 931
```



TLDR: Main differences to Python

- ▶ for, if, while etc. blocks are terminated by end
- indentation is nice, but not mandatory
- ► Julia is 1-indexed
- ► Strings have single "quotation marks", multiline strings three



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- the imaginary unit is im (not j)
- ► Matrix multiplication is A * B, element wise multiplication A .* B
- ► Julia has no objects/classes



TLDR: Main differences to R

- 'single' quotation marks are for characters
- vectors are constructed with square brackets v = [1,2,3]
- operations on vectors of different length are not allowed
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- ▶ function arguments are not copied when calling a function
- ▶ 1:5 is an AbstractRange, use collect(1:5) to create the vector
- you do not need vectorization for performance
- ▶ logical indexing: in R x [x>3] has two alternatives in Julia
 - x[x .> 3] (uses a temporary vector memory)
 - ▶ filter(z->z>3, x) might be nicer to read
 - filter!(z->z>3, x) updates x inplace (avoids the temporary memory)



TLDR: Main differences to Matlab

- array indexing uses square brackets A[i,j]
- ► Arrays are not copied by default A=B references the same, do A=copy(B) for an actual copy
- similarly function arguments are references, input variables can be modified
- ▶ 1-dimensional vectors exist and are not Nx1 matrices
- ▶ 42 is an integer, not a float, use 42.0 for the float.
- ► A == B does not return a matrix of booleans but true or false use A .== B to get such a matrix
- dimensions are not "constant-broadcasted":
 - ightharpoonup [1:10] + [1:10] ' creates a 10×10 matrix in Matlab
 - ► [1:10] + [1:10] ' is a dimension mismatch, because a column vector can not be added to a row vector



Packages



```
module MyModule #Same naming convention as types: CamelCase
   f(x) = x^2 # is exported
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- ► Default packages are among others Base (loaded on start) LinearAlgebra, Random, Statistics, ...



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After a package is installed, it can be used with using PackageName, PackageA, PackageB,



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Overall: Reproducible environment / setup to run your experiments in



Pluto Notebooks



Pluto.jl – Motivation



Similarities & differentes to Jupyter



Live Demo



Thanks for your attention!

Are there (further) questions?



Workshop: Let's get you started with Julia!