

Julia



- 2012 developed by Jeff Bezanson, Alan Edelman, Stefan Karpinski and Viral B. Shah at MIT
- Used for numerical and scientific computing with high performance
 - Execution speed is similar to C and FORTRAN
 - Hierarchical and parameterized type system as well as method overloading ("multiple dispatching") as central concepts
 - Native calls from C-(compiled) code possible (without wrappers)
- Unicode is efficiently supported (e.g., UTF-8)
- Alongside C, C++ and FORTRAN, the only programming language that has entered the "PetaFlop Club"



Jeff Bezanson (1981–)



Alan Edelman (1963 –)



Stefan Karpsinski (1981–)



Viral Shah

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Features of Julia



Just-in-time (JIT) compilation implemented using LLVM

Approaching static compilation such as C!

Optional typing and type inference

- Every object has a type; type declaration can be made
- Types are run-time objects
- A rich language of types for constructing and describing objects

Multiple dispatch

- Functions are uniquely defined by their argument types
- Alternative to classes in object-oriented programming

Very simple core language that imposes very little

 Julia Base and the standard library are written in Julia itself, including primitive operations like integer arithmetic

Language support for meta-programming

Macros are a typed extensions to pre-processors that modify expression trees

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Overview



- 1. Installation
- 2. Variables and Functions
- 3. Types
- 4. Methods & Interfaces
- 5. Modules

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Command Line Installation



- All resources can be found on https://julialang.org/
- On command line, execute

```
curl -fsSL https://install.julialang.org | sh
```

This will install the command line interface that can be launched with julia

- The installation manager is called juliaup and you should regularly run juliaup update
- juliaup can also be used to install several (older) versions of Julia in parallel!

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Visual Studio Code Installation



VS Code has excellent support for Julia but you should install two packages





- Syntax highlighting
- Integrated Julia REPL
- Code completion
- Linter
- Debugger
- Plot gallery
- Grid viewer for tabular data

Automatic Julia code formatting

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Julia Variables



- Variables: A name associated (or bound) to a value
 - Variables names can contain any Unicode character (best typed by \<LaTeX name>-tab

```
● julia> δ = 0.00001
1.0e-5
● julia> 안녕하세요 = "Hello"
"Hello"
```

■ A variable assignment with = binds the value but does not change it

```
• julia> a = [1,2,3] # an array of 3 integers
3-element Vector{Int64}:
1
2
3
• julia> b = a # both b and a are names for the same array!
3-element Vector{Int64}:
1
2
3
• julia> a[1] = 42 # change the first element
42
• julia> a = 3.14159 # a is now the name of a different object
3.14159
• julia> b # b refers to the original array object, which has been mutated
3-element Vector{Int64}:
42
2
3
```

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Julia Functions



- Functions: An object that maps a tuple of argument values to a return value
 - Julia functions are not pure mathematical functions: they can alter and be affected by the global state of the program

Long syntax

Short syntax

```
• julia> f(x,y) = x + y
f (generic function with 1 method)
```

```
• julia> \Sigma(x,y) = x + y
 \Sigma (generic function with 1 method)
```

• Function call: A function is called using parantheses syntax

```
• julia> f(2,3) 5
```



Functions are objects: A function is an object that can be assigned

```
julia> g = f
f (generic function with 1 method)
julia> g(2,3)
5
```

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Julia Functions: Arguments and Returns



- Arguments: Values are not copied when they are passed to functions
 - Function arguments themselves act as new variable bindings
 - Modifications to mutable values made within a function will be visible to the caller

Function definition

```
• julia> function f(x, y)

x[1] = 42  # mutates x

y = 7 + y  # new binding for y, no mutation return y

end
f (generic function with 1 method)
```

- A common convention is to put an explanation mark in the function name if the function mutates an argument
- Return: The value returned by a function is the value of the last expression evaluated
 - The return keyword causes a function to return immediately

```
Function application
```

```
julia> a = [4,5,6]
3-element Vector{Int64}:
4
5
6

julia> b = 3
3

julia> f(a, b) # returns 7 + b == 10
10

julia> a # a[1] is changed to 42 by f
3-element Vector{Int64}:
42
5
6
julia> b # not changed
```

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```
• julia> function g(x,y)
return x * y
x + y
end
g (generic function with 1 method)
• julia> g(2,3)
6
```

Julia Functions: Operators & Anonymous Functions



- **Operators**: Operators are just functions with support for special syntax
 - Can also be applied using parenthesized argument lists
 - Can also be assigned to other variables

```
• julia> +(1,2,3) 6
```

```
• julia> f = +
+ (generic function with 189 methods)
• julia> f(1,2,3)
6
```

- Operators with Special Names: A few special expressions correspond to calls to functions with special names
 - These operators can be made work for custom types by defining type-specific methods (see soon!)
- Anonymous Functions: A function object without a name
 - In other programming languages, they are called *lambda* functions

Long syntax

```
• julia> function (x)

x^2 + 2x - 1

end

#5 (generic function with 1 method)
```

Short syntax

```
• julia> x -> x^2 + 2x - 1
#7 (generic function with 1 method)
```

Expression	Calls
[A B C]	hcat
[A; B; C;]	vcat
[A B; C D;]	hvcat
[A; B;; C; D;;]	hvncat
Α'	adjoint
A[i]	getindex
A[i] = x	setindex!
A.n	getproperty
A.n = x	setproperty!

Julia Functions: Control Flow (Conditionals)



Compound expressions: Several sub-expressions that evaluate to the final expression

Long syntax

Short syntax

```
• julia> z = (x = 1; y = 2; x + y)
```

■ **Conditional evaluation**: Allows portions of code to be evaluated or not evaluated depending on the value of a *Boolean* expression

Function definition

Function application

```
• julia> test(1, 2)
x is less than y

• julia> test(2, 1)
x is greater than y

• julia> test(1, 1)
x is equal to y
```

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Ternary Operator: Shorthand for an if-then-else-end expression

```
• julia> test(x, y) = println(x < y ? "x is less than y" : x > y ? "x is greater than y" : "x is equal to y") test (generic function with 1 method)
```

```
• julia> test(1, 2)
  x is less than y
• julia> test(2, 1)
  x is greater than y
• julia> test(1, 1)
  x is equal to y
```

Julia Functions: Control Flow (Loops)



- while-loop: Evaluates the condition expression and as long it remains true, keeps also evaluating the body of the loop
- for-loop: Iterates through the container provided
 - A for-loop always introduces a new iteration variable in its body, regardless of whether a variable of the same name exists in the enclosing scope (a while-loop does not do that)

```
• julia> i = 1
1
• julia> while i <= 3
    println(i)
    i += 1
end
1
2
3</pre>
```

```
• julia> for i = 1:3 println(i) end 1 2 3
```

```
• julia> for i in [1,4,0] println(i) end 1 4 0
```

• break: Terminate the repetition of a while-loop before the test condition is false or stop iterating in a for-loop before the end of the iterable object is reached

```
• julia> i = 1
1
• julia> while true println(i) if i >= 3 break end i += 1 end
1
2
3
```

```
• julia> for j = 1:1000 println(j) if j >= 3 break end end 1 2 3
```

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Julia's Type System



- Julia's type system is dynamic (i.e. types are not known until run time)
 - There is no meaningful concept of a "compile-time type"!
 - It is possible to indicate that certain values are of specific types
 - Explicit types are used for method dispatching (see soon!)
- Julia's type system is also **nominative** (i.e., hierarchical relationships between types are explicitly declared)
 - Concrete types cannot subtype each other: all concrete types are final
 - Only abstract types can be supertypes
- Julia's type system is also parametric
 - Types can be parameterized by other types or by symbols (e.g. numbers)
 - Powerful for container types (e.g., lists)
- Note: Only values, not variables, have types!
 - Variables are simply names bound to values
 - "type of a variable" is shorthand for "type of the value to which a variable refers".

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Julia Types: Type Declarations



 : operator: Can be used to attach type annotations to expressions and variables in programs

```
o julia> (1+2)::AbstractFloat
ERROR: TypeError: in typeassert, expected AbstractFloat, got a value of type Int64
Stacktrace:
[I] top-level scope
@ REPL[27]:1

o julia> (1+2)::Int
3
```

- typeof function: Returns the type of the value that is passed to it
- **Function return types**: Can be specified at function definition

typeof function

Function return type

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Julia Types: Abstract Types



- Abstract Types: Cannot be instantiated, and serve only as nodes in the type graph.
 - Abstract types are declared using the abstract type keyword

```
abstract type «name» end
abstract type «name» <: «supertype» end
```

:< function: Allows to specify a supertype relationship or test for it

Type declarations

Supertype check

```
• julia> Integer <: Number true
• julia> Integer <: AbstractFloat false
```

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- No loss in performance when using function whose arguments are abstract types
 - They are recompiled for each tuple of concrete argument types with which it is invoked!

Julia Types: Primitive Types



- **Primitive Types**: Concrete type whose data consists of plain old bits
 - Primitive types are declared using the primitive type keyword

```
primitive type «name» «bits» end
primitive type «name» <: «supertype» «bits» end</pre>
```

The number of bits indicates how much storage the type requires

Type declarations

```
julia> primitive type F16
julia> primitive type F32
julia> primitive type F64
julia> primitive type F64
julia> primitive type Integer8
julia> primitive type UInteger8
Unsigned 8 end
```

Supertype check

```
    julia> Integer8 <: Integer true</li>
    julia> Integer8 <: Real true</li>
    julia> Integer8 <: AbstractFloat false</li>
```

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Unusual to redefine primitive types as Julia's primitive types are highly performant

Julia Types: Composite Types



- Composite Types: A collection of named fields
 - Composite types are declared using the struct keyword
 - New objects are created by applying the type name as a function (default constructor)
 - Fields can be accessed with the property notation <variable>.<fieldname>

```
• julia> struct Foo
bar
baz::Int
qux::Float64
end
```

```
• julia> foo = Foo("Hello, world.", 23, 1.5)
Foo("Hello, world.", 23, 1.5)
• julia> typeof(foo)
Foo
```

- julia> foo.bar
 "Hello, world."
 julia> foo.baz
 23
 julia> foo.qux
 1.5
- Composite objects declared with struct are immutable!
 - Efficient and easier to reason about for the (just-in-time) compiler (can use registers!)
- Composite objects declared with mutable struct are mutable!
 - Such objects are generally allocated on the heap and have stable memory addresses

```
• julia> mutable struct Bar
baz
qux::Float64
end

• julia> bar = Bar("Hello", 1.5)
Bar("Hello", 1.5)
```

```
• julia> bar.qux = 2.0
2.0
• julia> bar.baz = 1//2
1//2
```

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Julia Types: Parametric Types



 Parametric Types: Types that have parameters that introduce a whole family of new types – one for each possible combination of parameter values

Type parameters are introduced after the type name, surrounded by curly braces

The type parameter can be any type at all (or a value of any primitive type!)

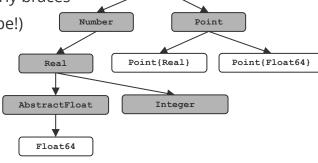
```
• julia> struct Point{T}
x::T
y::T
end
```

```
• julia> p1 = Point{Float64}(1.0, 2.0)
Point{Float64}(1.0, 2.0)

• julia> p2 = Point(1,2)
Point{Int64}(1, 2)

• julia> typeof(p1)
Point{Float64}

• julia> typeof(p2)
Point{Int64}
```



Any

The non-parametric type is a super-type of any specialized parametric type!

• julia> Point{Float64} <: Point
true
• julia> Point{Float64} <: Point{Real}
false
• julia> Point{Float64} <: Point{<:Real}
true</pre>

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Any type that is a subtype of Real

Julia Types: Tuple Types



- Tuple Types: Parameterized immutable type where each parameter is the type of one field
 - Tuple types may have any number of type parameters
 - Tuples do not have field names; fields are only accessed by index
 - Tuple types are covariant in their parameters

```
• julia> struct Tuple2{A,B}
a::A
b::B
end
```

```
• julia> x = (1,"foo",2.5)
(1, "foo", 2.5)
```

```
julia> typeof(x)
Tuple{Int64, String, Float64}

julia> x[1]
1
```

```
• julia> Tuple{Int,AbstractString} <: Tuple{Real,Any}
true
• julia> Tuple{Int,AbstractString} <: Tuple{Real,Real}
false</pre>
```

- Named Tuple Types: Tuple types with field names
 - Field names can be used to access elements

```
julia> x = (a=1,b="hello")
(a = 1, b = "hello")

julia> y = NamedTuple{(:a, :b)}((1, ""))
(a = 1, b = "")
```

```
    julia> typeof(x)
        @NamedTuple{a::Int64, b::String}
    julia> typeof(y)
        @NamedTuple{a::Int64, b::String}
```

```
• julia> x.b
"hello"
• julia> y.a
1
```

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Tuple types are an abstraction of the arguments of a function

Julia Types: Numbers



- Julia provides a broad range of primitive numeric types
 - □ Integers: Hexadecimal, octal and binary literals are input via 0x, 0o and 0b

```
• julia> x = 0x1
0x01
```

Floating Point Numbers: 64-bit and 32-bit literals are input via <leading digits>e<exponent> or <leading digits>f<exponent>

```
• julia> 2.5e-4
0.00025
• julia> 2.5f-4
0.00025f0
```

Machine epsilon: Distance between two adjacent floating-point numbers



Arbitrary Precision Arithmetic: Julia wraps the GMP and GMPFR library



Туре	Signed?	Number of bits	Smallest value	Largest value
Int8	✓	8	-2^7	2^7 - 1
UInt8		8	0	2^8 - 1
Int16	✓	16	-2^15	2^15 - 1
UInt16		16	0	2^16 - 1
Int32	✓	32	-2^31	2^31 - 1
UInt32		32	0	2^32 - 1
Int64	✓	64	-2^63	2^63 - 1
UInt64		64	0	2^64 - 1
Int128	✓	128	-2^127	2^127 - 1
UInt128		128	0	2^128 - 1
Bool	N/A	8	false (O)	true (1)

Туре	Precision	Number of bits
Float16	half	16
Float32	single	32
Float64	double	64

Julia Types: Strings



- Strings are finite sequences of characters
 - The built-in concrete type used for strings (and string literals) in Julia is String
 - It supports the full range of Unicode characters via the UTF-8 encoding
 - String literals are delimited by double quotes or triple double quotes
 - Strings are arrays of characters with the special indexers begin and end

```
• julia> s2 = "\u2200 x \u2203 y"
"∀ x ∃ y"
```

```
• julia> str = "Hello, world.\n"
  "Hello, world.\n"

• julia> s1 = """Contains "quote" characters"""
  "Contains \"quote\" characters"
```

```
• julia> str[begin]
    'H': ASCII/Unicode U+0048 (category Lu: Letter, uppercase)
• julia> str[1]
    'H': ASCII/Unicode U+0048 (category Lu: Letter, uppercase)
• julia> str[end]
    '\n': ASCII/Unicode U+000A (category Cc: Other, control)
• julia> str[end-1]
    '.': ASCII/Unicode U+002E (category Po: Punctuation, other)
```

- Strings are concatenated with * and interpolated with \$
 - The shortest complete expression after the **\$** is taken as the expression whose value is to be interpolated into the string

```
• julia> greet * ", " * whom * ".\n"
"Hello, world.\n"
• julia> "$greet, $whom.\n"
"Hello, world.\n"
```

```
• julia> greet = "Hello"
"Hello"

• julia> whom = "world"
"world"
```

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Julia supports regular expression with the regex:

```
• julia> m = match(r"^\s*(?:#\s*(.*?)\s*$|$)", "# a comment ")
RegexMatch("# a comment ", 1="a comment")
```

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Julia Methods



- A function is an object that maps a tuple of arguments to a return value
 - It is common for the same conceptual function or operation to be implemented quite differently for different types of arguments
- **Method**: A definition of one possible behavior for a function

```
• julia> f(x::Float64, y::Float64) = 2x + y f (generic function with 1 method)
```

```
• julia> f(x::Number, y::Number) = 2x - y
f (generic function with 2 methods)
```

```
• julia> methods(f)
# 2 methods for generic function "f" from Main:
[1] f(x::Float64, y::Float64)
@ REPL[100]:1
[2] f(x::Number, y::Number)
@ REPL[101]:1
```

- Method dispatch: Choice of method to execute when a function is applied
 - Multiple dispatch: Using all of a function's arguments to choose which method

```
• julia> f(2.0, 3.0)
7.0
• julia> f(2.0, 3)
1.0
```

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```
② julia> f("foo", 3)
    ERROR: MethodError: no method matching f(::String, ::Int64)

Closest candidates are:
    f(::Number, ::Number)
    @ Main REPL[101]:1
```

```
    julia> f(x,y) = println("Whoa there, Nelly.")
    f (generic function with 3 methods)
    julia> f("foo", 1)
    Whoa there, Nelly.
```

Julia Parametric Methods



Method definitions can have type parameters qualifying the signature

```
    julia> same_type(x::T, y::T) where {T} = true same_type (generic function with 1 method)
    julia> same_type(x,y) = false same_type (generic function with 2 methods)
```

```
• julia> same_type(1, 2)
true
• julia> same_type(1.0, 2.0)
true
```

```
• julia> same_type(1, 2.0)
false
• julia> same_type(Int32(1), Int64(2))
false
```

 Similar to subtype constraints on type parameters in type declarations one can also constrain type parameters of methods using where:

```
    julia> same_type_numeric(x::T, y::T) where {T<:Number} = true same_type_numeric (generic function with 1 method)</li>
    julia> same_type_numeric(x::Number, y::Number) = false same_type_numeric (generic function with 2 methods)
```

```
• julia> same_type_numeric(1, 2)
true
• julia> same_type_numeric(1.0, 2.0)
true
```

```
• julia> same_type_numeric(1, 2.0)
false

• julia> same_type_numeric("foo", 2.0)
ERROR: MethodError: no method matching same_type_numeric(::String, ::Float64)
```

- Constructor methods: Constructor is just like any other function in Julia
 - Defining a method with the same name as composite type is called an *outer constructor*

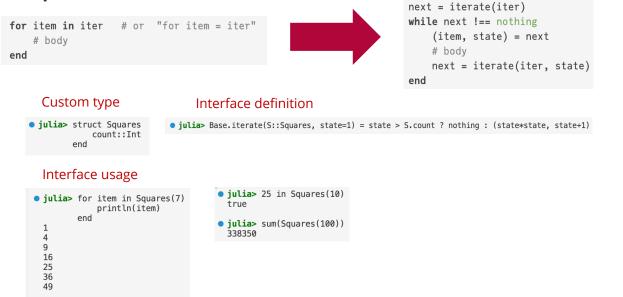
```
    julia> OrderedPair(a) = OrderedPair(a,a)
    OrderedPair
    julia> OrderedPair(10)
    OrderedPair(10, 10)
```

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Julia Interfaces



- Interface: By extending a few specific methods to work for a custom type, objects of that type can be used in other methods that are written to generically build upon those behaviors.
- **Example**: Iterator interface



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Julia Interfaces (ctd)



- Indexing interface allows any composite type to behave like an array
 - This is true even if the type never explicitly stores an array!
- **Example**: Array of square numbers

```
• julia> Squares(100)[23]
529
• julia> Squares(23)[end]
529
```

Methods to implement	Brief description
<pre>getindex(X, i)</pre>	X[i], indexed element access
setindex!(X, v, i)	X[i] = v, indexed assignment
firstindex(X)	The first index, used in X[begin]
lastindex(X)	The last index, used in X[end]

The interface even allows to support new indexing modes!

```
• julia> Base.getindex(S::Squares, i::Number) = S[convert(Int, i)]
• julia> Base.getindex(S::Squares, I) = [S[i] for i in I]
```

```
• julia> Squares(10)[[3,4.,5]]
3-element Vector{Int64}:
9
16
25
```

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Julia Modules



- Modules: Help organize code into coherent units
 - Modules are separate namespaces, each introducing a new global scope
 - Same name can be used for different functions or global variables without conflict
 - Can be precompiled for faster loading, and may contain code for runtime initialization
- Module Definition: Using the keywords module ... end
 - Explicit exports of symbols are done with export

```
• julia> module NiceStuff
export nice, DOG
struct Dog end # singleton type, not exported
const DOG = Dog() # named instance, exported
nice(x) = "nice $x" # function, exported
end
Main.NiceStuff
```

- Module Usage: Using the keywords using
 - Local defined modules start with a dot

```
• julia> using .NiceStuff
• julia> nice("house %")
"nice house %"
```

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Summary



Julia Concepts

- Runs programs via JIT compilation implemented using LLVM (nearly as fast as C!)
- Julia is 13 years old and has reached wide-spread adoption in scientific computing
- Extensibility comes from using method dispatch and unbundling functions and data
- Julia provides garbage collection and requires no explicit memory management

Julia Type System

- Julia's type system is dynamic, nominative and parametric
- It only has abstract, primitive and composite types (and the latter ones are all final)
- Every object has a type; type declaration can be made

Julia Methods

- A method is the concrete definition of a function's behavior
- Every function can have many methods; Julia run-time picks the right method based on the types of the arguments (method dispatch)
- We only scratched the surface; keep using the language to get confident!

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See you next week!