



Programming in Julia

language philosophy and design

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Part I

Julia versus other languages





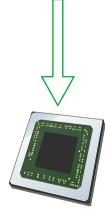
How does a computer work?

```
mutate!(traits, settings, locivar)
 6 Loop over an array of traits, mutating each value in place along a normal distribution.
 7 'locivar' can be used to scale the variance of the normal distribution used to draw new
 8 trait values (together with `settings[phylconstr]`).
10 function mutate!(traits::Array{Trait, 1}, settings::Dict{String, Any}, locivar::Float64 = 1.0)
       settings["phylconstr"] * locivar == 0 && return
      for trait in traits
          traitname = settings["traitnames"][trait.nameindex]
          occursin("segsimilarity", traitname) && settings["fixtol"] && continue
          occursin("tempopt", traitname) && (oldvalue -= 273)
          while oldvalue <= 0 # make sure sd of Normal dist != 0
              oldvalue = abs(rand(Normal(0,0.01)))
          newvalue = rand(Normal(oldvalue, oldvalue * settings["phylconstr"] * locivar))
          (newvalue > 1 && occursin("prob", traitname)) && (newvalue=1.0)
          while newvalue <= 0
              newvalue = rand(Normal(oldvalue, oldvalue * settings["phylconstr"] * locivar))
          occursin("tempopt", traitname) && (newvalue += 273)
          trait.value = newvalue
28 end
```

High-level language



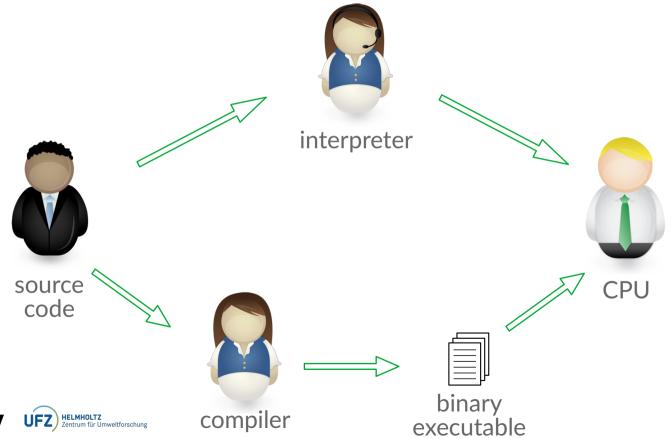








Compilation vs. interpretation





A closer look at: R

• **Purpose:** data analysis & visualisation

Context:

- Users are scientists, not software developers
- Most programs are small; apply pre-existing functions to new data
- Programs are often not specified ahead of time → exploratory programming

- Strongly interactive (REPL) → interpreted
- Huge library ecosystem for common functionality
- Dynamic typing (easier for users)





A closer look at: Python

• **Purpose:** general purpose, especially scripting

Context:

- Should be quick & easy to write
- Should be useful for as many use-cases as possible (both scripts and larger programs)
- Performance is generally not a priority (except in special libraries)

- Clean, minimalist syntax
- Portability, dynamic typing → interpreted
- Large standard library for common functionality ("batteries included")





A closer look at: C++

• Purpose: general purpose, especially application and systems programming

Context:

- Users are expert software developers
- Often needs to interface directly with hardware, performance is critical
- Efficiency is more important than ease of use
- Programs can be very large and complex → architectural programming

- Performance needed → compiled
- Close to hardware \rightarrow static typing, no automatic memory management
- 3rd-party libraries needed for most functionality



A closer look at: Julia

• **Purpose:** scientific computing

Context:

- Users are not all expert software engineers, but have to carry out complex and intensive calculations → should be easy to use, yet offer high performance
- Programs are usually novel and can be complex → either exploratory or architectural programming

- Performance + interactivity → just-in-time compilation
- Language designed to allow building complex scientific software
- __ Ideal: numeric utility of R + simplicity of Python + performance of C++

Pros and cons of Julia

Advantages

- Syntax is easy to learn and understand
- Performance is comparable to C/C++
- Designed by and for computational scientists → scientific libraries
- Good documentation, helpful community (Discourse forum)
- Flexible & expressive language (multiple dispatch, metaprogramming)

Disadvantages

- Not a typical object-oriented language, need time to learn concepts
- Annoying wait for compilation before every run (TTFX)
- Young language, small community
 → limited availability of highquality libraries
- Fast-moving development, future changes may break code





Question Time I





Part II

Julia's design: the type system





What are types?

primitive types (e.g. int, bool)





composite types (e.g. array, dict)









Static vs. dynamic typing

Static typing

- Variable types declared by programmer
- Easier for compiler/interpreter
- Rigid, types have to be known ahead of time and cannot change
- Allows compiler optimisation → better performance
- Mandatory in C/C++, optional in Julia

Dynamic typing

- Types inferred automatically by compiler/interpreter
- Easier for programmer
- Flexible, allows types to be determined or change at runtime
- Often slower
- Mandatory in R/Python, default in Julia





Type hierarchy Any Number AbstractFloat Real Float16 Integer Float32 Float64 Signed Unsigned Boo1 Int8 UInt8 Int16 UInt16 Int32 UInt32 Int64 UInt64

Int128

UInt128





Dispatching on type: methods

```
julia> function f(x,y)
            x + y
       end
f (generic function with 1 method)
julia> f(2,3)
julia> f("hello", "world")
ERROR: MethodError: no method matching +(::String, ::String)
julia > f(x::String, y::String) = x * y
f (generic function with 2 methods)
                                           Inheritance: A method declared for
                                           a supertype will automatically apply
julia> f("hello", "world")
"helloworld"
                                           to all subtypes, unless there are
                                           more specialised methods available!
```





Creating new types: structs

```
type declaration
julia> mutable struct Foo
            bar::String
            baz::Int
            qux
       end
julia> foo = Foo("Hello, world.", 23, 1.5)
                                                        object instantiation
Foo("Hello, world.", 23, 1.5)
julia> typeof(foo)
Foo
julia> foo.bar
"Hello, world."
julia > foo.baz = 20
                                                        only possible with
20
                                                        mutable structs!
```





Question Time II





Thank you for your attention!

Now let's get our hands dirty...