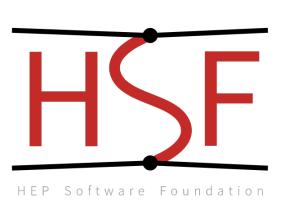
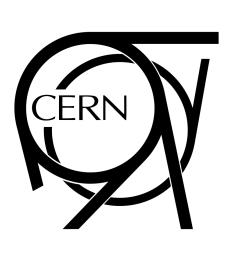


Julia in high-energy physics

A paradigm shift or just another tool?





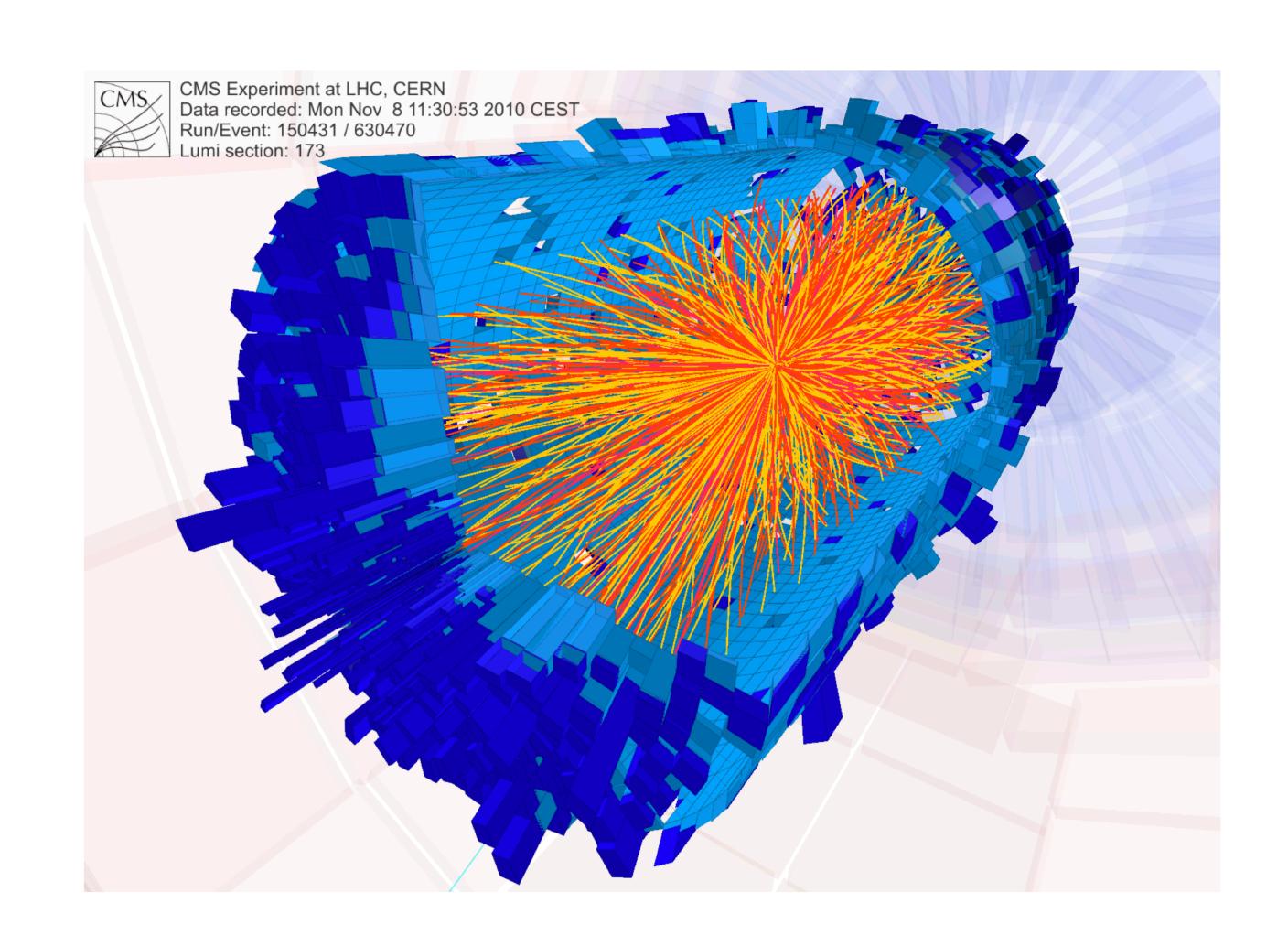


Introduction A new era for HEP-software?



Software requirements in HEP

- Efficiency
 - Fast execution
 - High data throughput
 - Scalability
- Developer-friendly
 - Quick bug fixes
 - Newest algorithms implemented
 - Good tooling
- User-friendly
 - Rapid development cycles
 - Low entry points
 - Interactivity

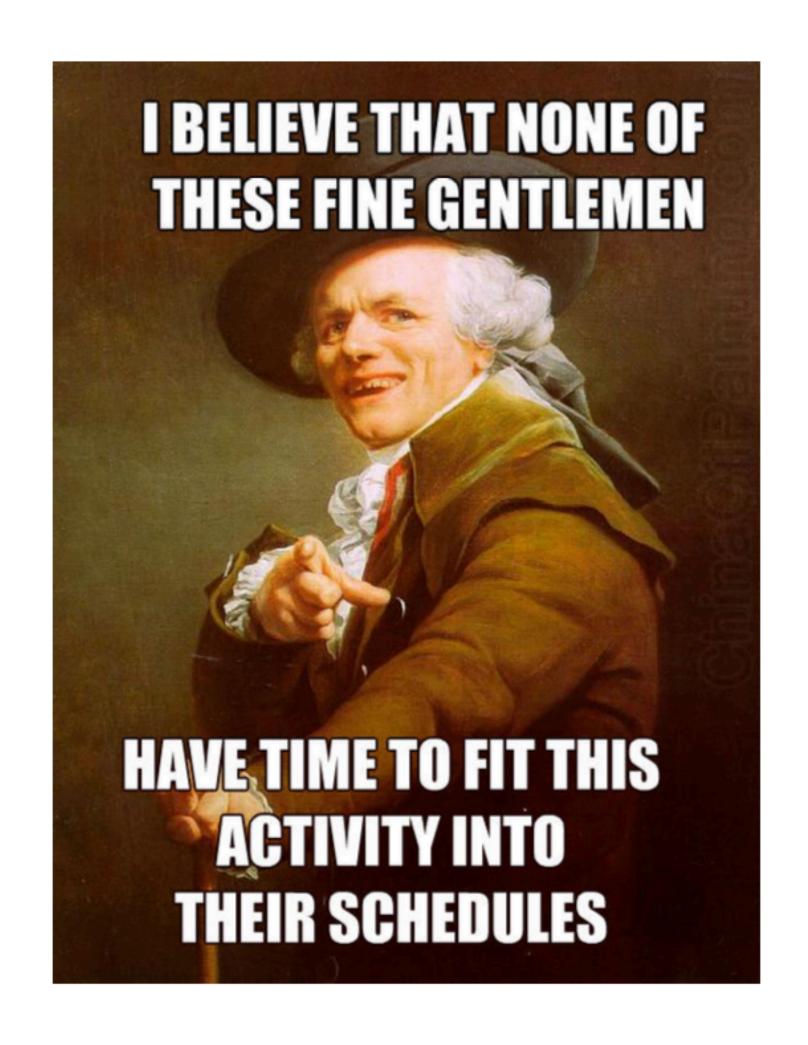


"[I propose that] you should use *two* languages for large software system: one, such as C or C++, for manipulating the complex internal data structures where performance is key and another, such as Tcl, for writing small-ish scripts that tie together the C pieces and are used for extensions."

Why is this problematic?

The two languages problem

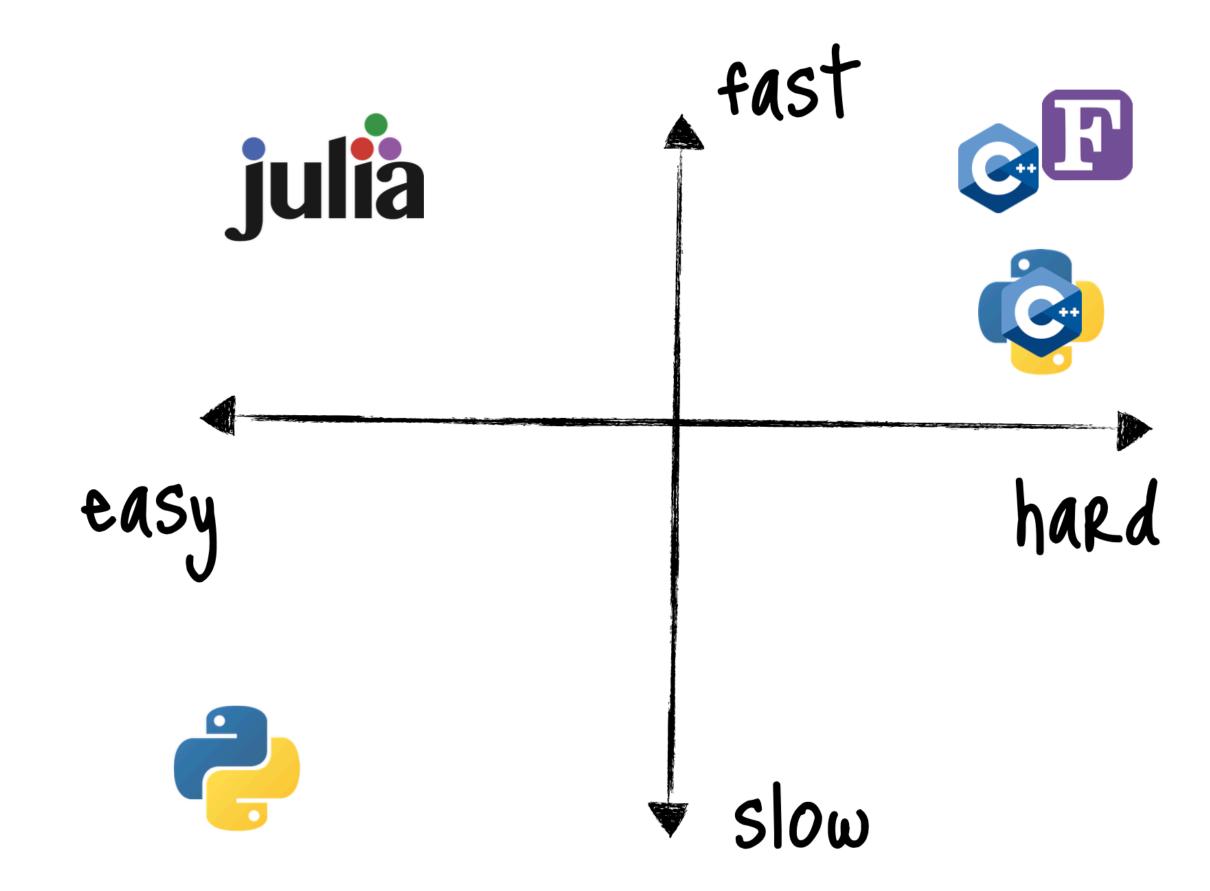
- Rewriting parts == refactoring
- Different languages == different logics
- Need for glue code
- Extending is a mess
- Debugging is a mess
- Scientists need to be polyglot
- Multithreading? Anyone?



Proposal of a solution

The Julia programming language

- Invented 2012 at MIT (mostly)
- Jeff Bezanson, Stefan Karpinski, Viral B. Shah, Alan Edelman
- Design goals
 - Open source
 - Speed like C, dynamic like Ruby
 - Obvious mathematical notation
 - General purpose like Python
 - As easy for statistics as R
 - Powerful linear algebra like in Matlab
 - Good for gluing programs together like the shell



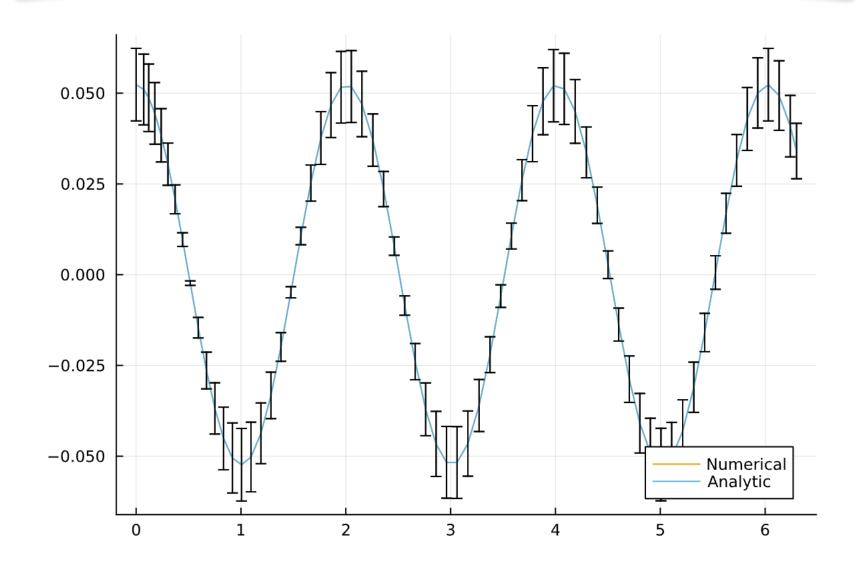
"Something that is dirt simple to learn, yet keeps the most serious hackers happy."

Julia is easy Ease of use

- Dynamically typed
- Powerful type system
- Garbage collection
- Extensive standard library
 - Mostly written in Julia
 - Math included
 - Performant
- Multiple dispatch for the win!

You can write Julia code as far away from the metal as you want!

```
using DifferentialEquations, Measurements, Plots
g = 9.79 \pm 0.02; # Gravitational constants
L = 1.00 \pm 0.01; # Length of the pendulum
#Initial Conditions
u_0 = [0 \pm 0, \pi / 60 \pm 0.01] # Initial speed and initial angle
tspan = (0.0, 6.3)
#Define the problem
function pendulum(du,u,p,t)
    \theta = u[1]
    d\theta = u[2]
    du[1] = d\theta
    du[2] = -(g/L)*\theta
end
#Pass to solvers
prob = ODEProblem(pendulum, uo, tspan)
sol = solve(prob, Tsit5(), reltol = 1e-6)
# Analytic solution
u = u_0[2] .* cos.(sqrt(g / L) .* sol.t)
plot(sol.t, getindex.(sol.u, 2), label = "Numerical")
plot!(sol.t, u, label = "Analytic")
```

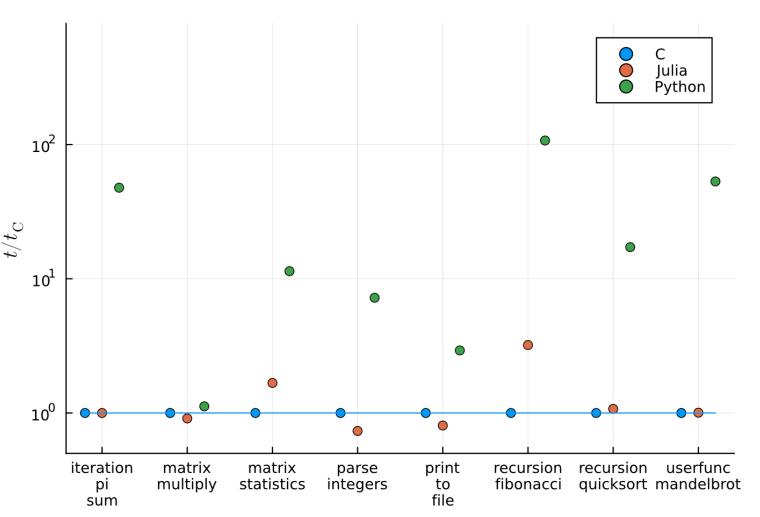


Julia is fast Not an interpreter

- Just-ahead-of-time compiler
- LLVM empowered
- Statically sizes arrays
- Built-in vector/matrix types
- Arbitrary optimization
- Compiler reflections available
- Native thread support

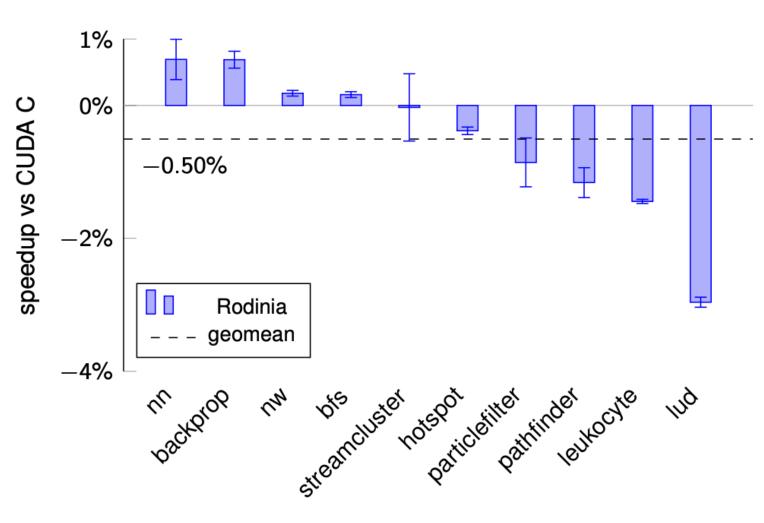
You can write Julia code as close to the metal as you want!

CPU performace



Data taken from [https://julialang.org/benchmarks/]

GPU performance



Taken from [Besard et al. IEEE Trans. Parallel Distrib. Syst. 30.4 (2018)]

Modern Language Development tooling

```
QEDcore.jl

— CHANGELOG.md

— LICENSE

— Manifest.toml

— Project.toml

— README.md

— docs
— src
— test
```

Packaging system

```
name = "QEDcore"
uuid = "35dc0263-cb5f-4c33-a114-1d7f54ab753e"
authors = [
    "Uwe Hernandez Acosta <u.hernandez@hzdr.de>",
    "Anton Reinhard <a.reinhard@hzdr.de>",
]
version = "0.1.1"
[deps]
DocStringExtensions = "ffbed154-4ef7-542d-bbb7-c09d3a79fcae"
QEDbase = "10e22c08-3ccb-4172-bfcf-7d7aa3d04d93"
Reexport = "189a3867-3050-52da-a836-e630ba90ab69"
SimpleTraits = "699a6c99-e7fa-54fc-8d76-47d257e15c1d"
StaticArrays = "90137ffa-7385-5640-81b9-e52037218182"
[compat]
DocStringExtensions = "^0.9"
QEDbase = "0.2.2"
Reexport = "^1.2"
SimpleTraits = "^0.9"
StaticArrays = "^1.9"
julia = "1.6"
```

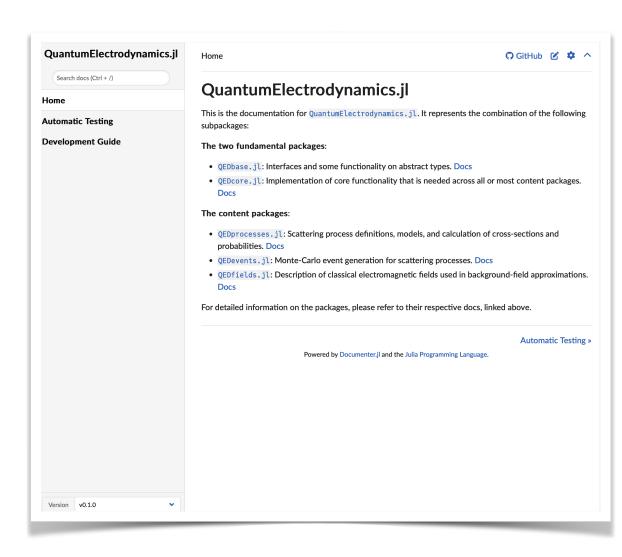
Project.toml

```
(@v1.11) pkg> add QEDcore
   Resolving package versions...
   Installed QEDcore - v0.1.1
   Updating `~/.julia/environments/v1.11/Project.toml`
  [35dc0263] + QEDcore v0.1.1
   Updating `~/.julia/environments/v1.11/Manifest.toml
  [7d9f7c33] + Accessors v0.1.38
  [dce04be8] + ArgCheck v2.3.0
  [49dc2e85] + Calculus v0.5.1
  [38540f10] + CommonSolve v0.2.4
  [a33af91c] + CompositionsBase v0.1.2
  [187b0558] + ConstructionBase v1.5.8
  [3587e190] + InverseFunctions v0.1.17
  [eff96d63] + Measurements v2.11.0
  [5ad8b20f] + PhysicalConstants v0.2.3
  [10e22c08] + QEDbase v0.2.2
  [35dc0263] + QEDcore v0.1.1
  [f2b01f46] + Roots v2.2.1
  [699a6c99] + SimpleTraits v0.9.4
  [90137ffa] + StaticArrays v1.9.7
  [1e83bf80] + StaticArraysCore v1.4.3
Precompiling project...
 4 dependencies successfully precompiled in 6 seconds.
```

Package manager (Pkg.jl)

```
Test Summary: | Pass Total Time
phase spaces
Test Summary: | Pass Total Time
four momentum |
Test Summary:
               Pass Total Time
gamma matrices
Test Summary:
                        69 1.4s
Lorentz vector
Test Summary: | Pass Total Time
Test Summary: | Pass Total Time
particle types |
Test Summary:
particle states | 4367
Test Summary:
                | Pass Total Time
particle spinors |
Test Summary:
                    | Pass Total Time
particle base states | 4367
Test Summary:
                     Pass Total Time
particle propagators |
              | Pass Total Time
Test Summary:
process interface | 148 1.3s
    Testing QEDcore tests passed
```

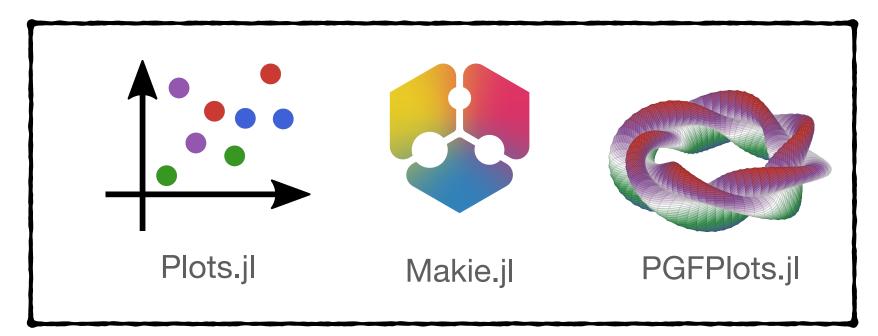
Testing (integrates with Pkg.jl)



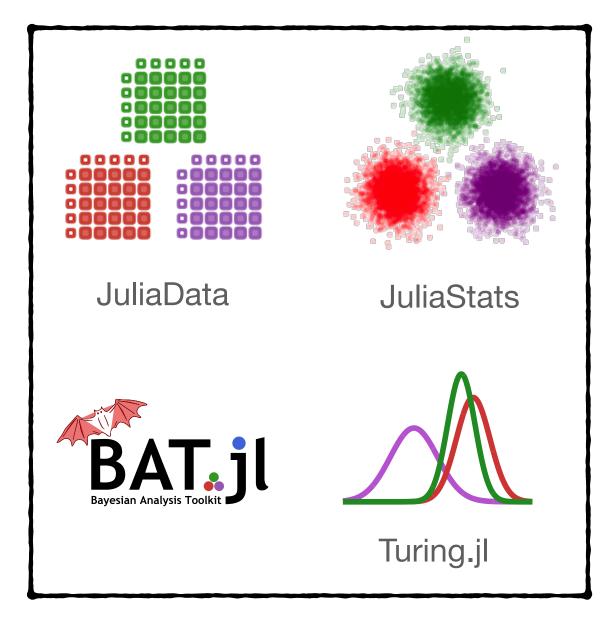
Documenter.jl

Rich eco-system > 10k packages

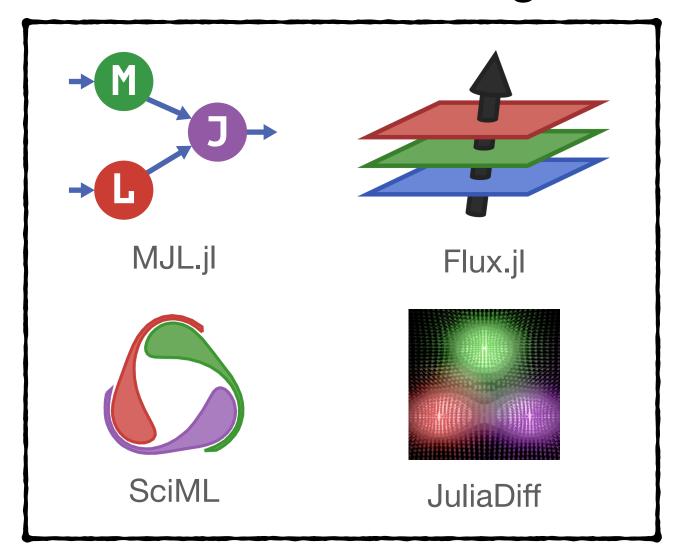
Visualization



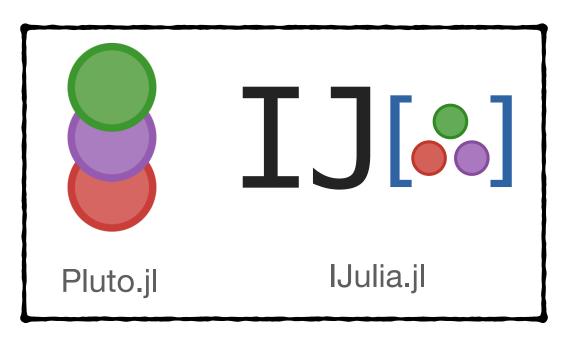
Data and Statistics



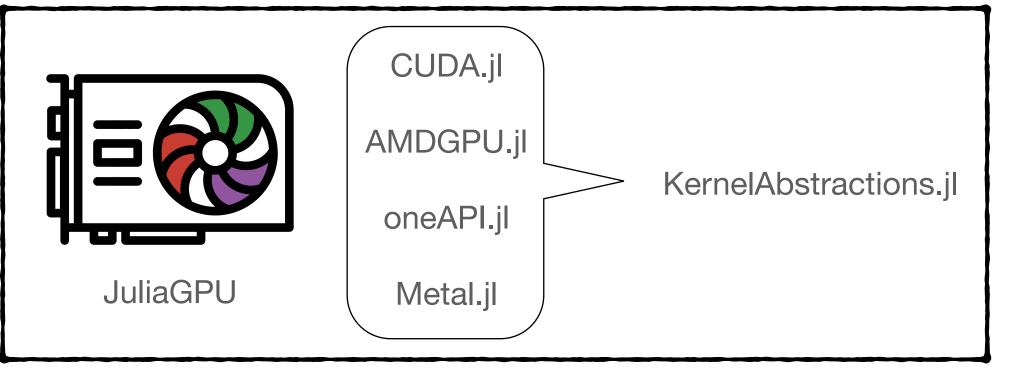
Machine learning



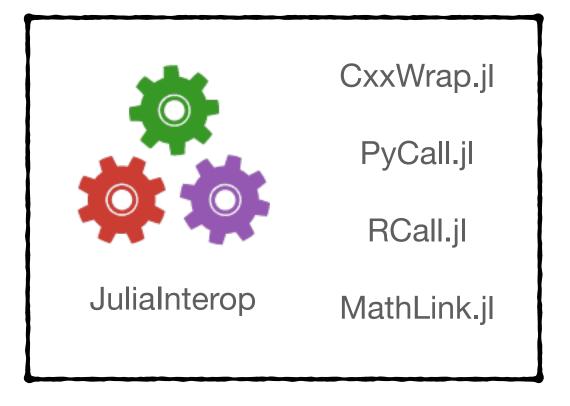
Notebooks



GPU support



Interoperability



Drawbacks of using Julia?

Julia should be better or shouldn't it?

- Formatter/Linter/LSP could be better
- Little scripts*
- Startup time*
- Vendor lock
 - Only LLVM and Clang
 - Only one reference implementation

- Building binaries*
- Calling Juila from other Languages*
- Context-based programming*
- Cumbersome static performance prediction
- Cumbersome static analysis/ checking*

Does it fit the HEP needs?

Computational challenges in HEP

- Large data volumes
 - PBs of experimental data
 - Extensive processing pipelines
- High computational cost
 - Event generation
 - Detector modelling
- Large-scale heterogeneous environments
 - Multi-architecture machines
 - Scalability
- Legacy and maintenance
 - Old codebases
 - Interoperability

HEP computing collaborations for the challenges of the next decade

Contacts: Simone Campana (<u>Simone.Campana@cern.ch</u>), Zach Marshall (<u>ZLMarshall@lbl.gov</u>) Alessandro Di Girolamo (<u>Alessandro Di Girolamo@cern.ch</u>) Heidi

Schellman (H Stewart (grae

A Roadmap for HEP Software and Computing R&D for the 2020s

The HEP Software Foundation⁵ · Johannes Albrecht⁶⁹ · Antonio Augusto Alves Jr⁸¹ · Guilherme Amadio⁵ · Giuseppe Andronico²⁷ · Nguyen Anh-Ky¹²² · Laurent Aphecetche⁶⁶ · John Apostolakis⁵ · Makoto Asai⁶³ · Luca Atzori⁵ · Marian Babik⁵ · Giuseppe Bagliesi³² · Marilena Bandieramonte⁵ · Sunanda Banerjee¹⁶ · Martin Barisits⁵ · Lothar A. T. Bauerdick¹⁶ · Stefano Belforte³⁵ · Douglas Benjamin⁸² · Catrin Bernius⁶³ · Wahid Bhimji⁴⁶ · Riccardo Maria Rianchi¹⁰⁵ · Ian Rird⁵ · Catherine Riscarat⁵² · Jakoh Rlomer⁵ · Kenneth Rloom⁹⁷ ·

Tommaso Boccali³² · Ma Concezio Bozzi²⁸ · Ma

Challenges in Monte Carlo Event Generator Software for High-Luminosity LHC

The HSF Physics Event Generator WG · Andrea Valassi¹ • · Efe Yazgan² • · Josh McFayden¹,¾ • · Simone Amoroso⁵ · Joshua Bendavid¹ · Andy Buckley⁶ · Matteo Cacciari⊓,8 · Taylor Childers⁰ · Vitaliano Ciulli¹ · Rikkert Frederix¹¹ · Stefano Frixione¹² · Francesco Giuli¹³ · Alexander Grohsjean⁵ · Christian Gütschow¹⁴ · Stefan Höche¹⁵ · Walter Hopkins⁰ · Philip Ilten¹⁶,¹७ · Dmitri Konstantinov¹8 · Frank Krauss¹⁰ · Qiang Li²⁰ · Leif Lönnblad¹¹ · Fabio Maltoni²¹,²² · Michelangelo Mangano¹ · Zach Marshall³ · Olivier Mattelaer²² · Javier Fernandez Menendez²³ · Stephen Mrenna¹⁵ · Servesh Muralidharan¹,⁰ · Tobias Neumann¹⁴,²⁴ · Simon Plätzer²⁵ · Stefan Prestel¹¹ · Stefan Roiser¹ · Marek Schönherr¹⁰ · Holger Schulz¹ · Markus Schulz¹ · Elizabeth Sexton-Kennedy¹⁵ · Frank Siegert²⁶ · Andrzej Siódmok²⊓ · Graeme A. Stewart¹

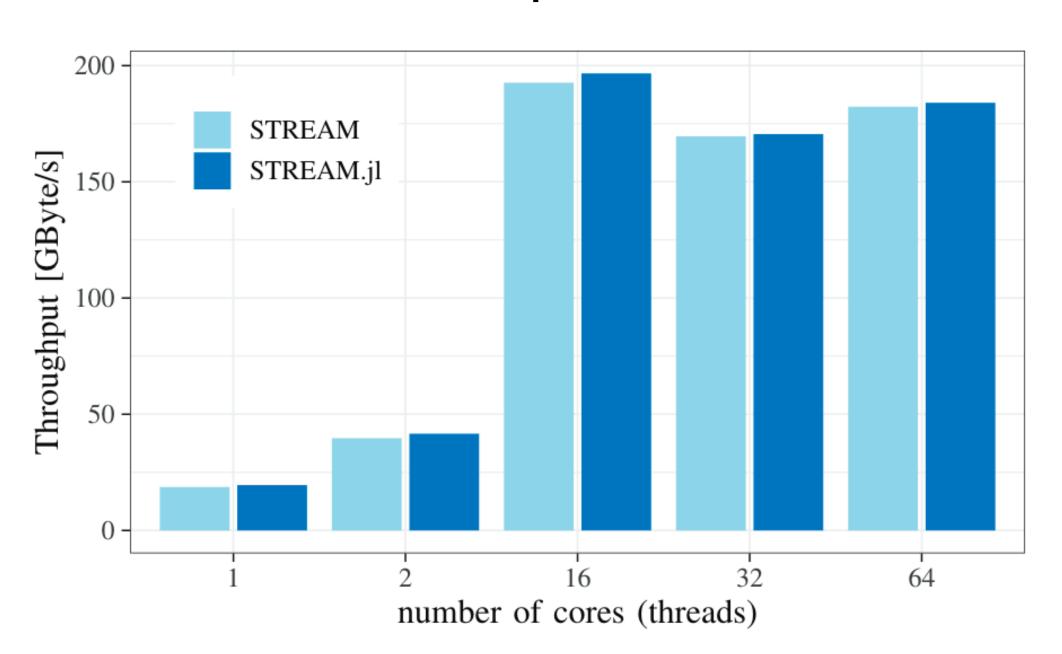
Data throughput

Memory bandwidth benchmarks

Benchmarking Julia's Communication Performance: Is Julia HPC ready or Full HPC?

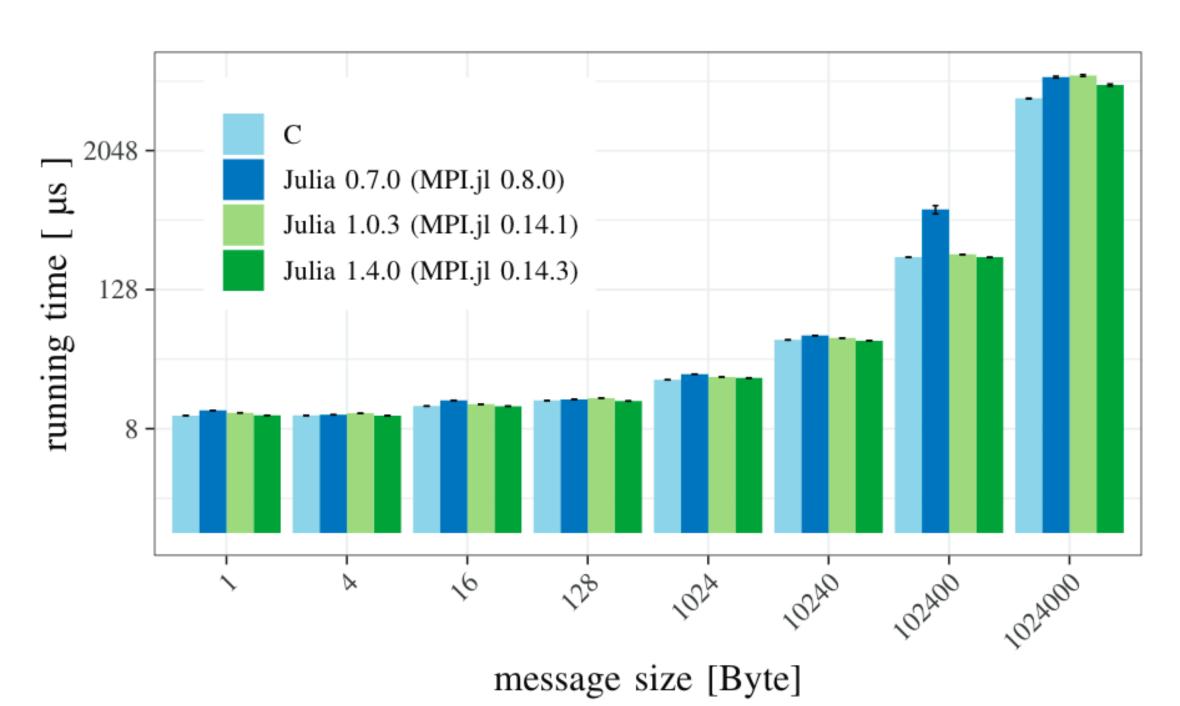
Sascha Hunold TU Wien, Faculty of Informatics Vienna, Austria Sebastian Steiner
TU Wien, Faculty of Informatics
Vienna, Austria

Intra-node performance



STREAM benchmark up to 64 AMD CPU cores LoC: 378 (C) vs 156 (Julia)

Inter-node performance



MPI broadcasting benchmark: 36 × 32 processes

Single-node performace

Single-thread axpy benchmarks on Fugaku (A64FX)

```
function axpy!(a::T, x::Vector{T}, y::Vector{T}) where {T<:Number}</pre>
    @simd for i in eachindex(x, y)
        @inbounds y[i] = muladd(a, x[i], y[i])
    end
    return y
end
```

Single precision 15.0 Julia Fujitsu BLAS 12.5 OpenBLAS ARMPL 10.0 GFLOPS 5.0 2.5 10² 10¹ 10³ 10⁵ 10^{6} 10⁷ 10^{4} Vector size

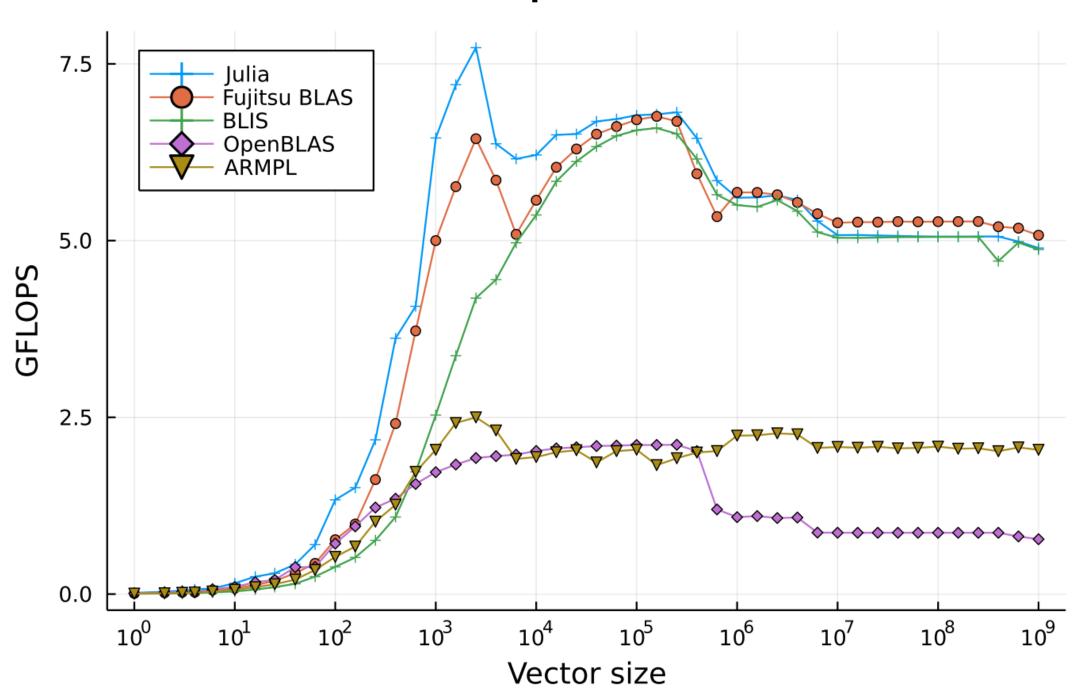
Productivity meets Performance: Julia on A64FX

1st Mosè Giordano London, United Kingdom m.giordano@ucl.ac.uk

2nd Milan Klöwer Advanced Research Computing Atmospheric, Oceanic and Planetary Physics University of Oxford Oxford, United Kingdom milan.kloewer@physics.ox.ac.uk

3rd Valentin Churavy CSAIL, EECS Massachusetts Institute of Technology Cambridge, United States of America vchuravy@mit.edu

Double precision



Julia on scale Celeste.jl project

- 2017 at NERSC (Berkley)
 - ^o Analysis of 178 TB telescope data
 - $^{\rm o}$ Inferred parameters of 1.88×10^8 stars
 - O Done in 14.6 min
 - $^{\rm o}~1.3\times10^6$ threads on 650.000 Intel Xeon Phi cores
 - ° 1.54 PFLOPS peak performance

Cataloging the Visible Universe through Bayesian Inference at Petascale

Jeffrey Regier*, Kiran Pamnany[†], Keno Fischer[‡], Andreas Noack[§], Maximilian Lam*, Jarrett Revels[§], Steve Howard[¶], Ryan Giordano[¶], David Schlegel^{||}, Jon McAuliffe[¶], Rollin Thomas^{||}, Prabhat^{||}

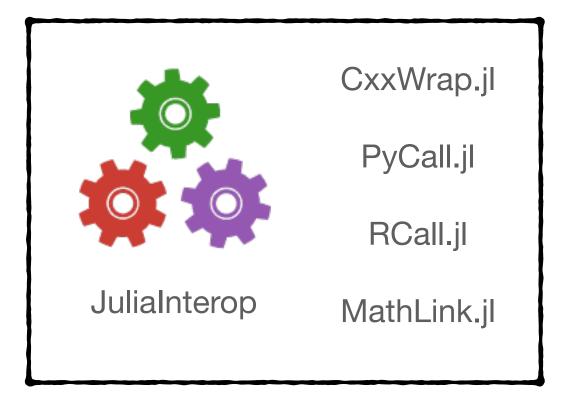


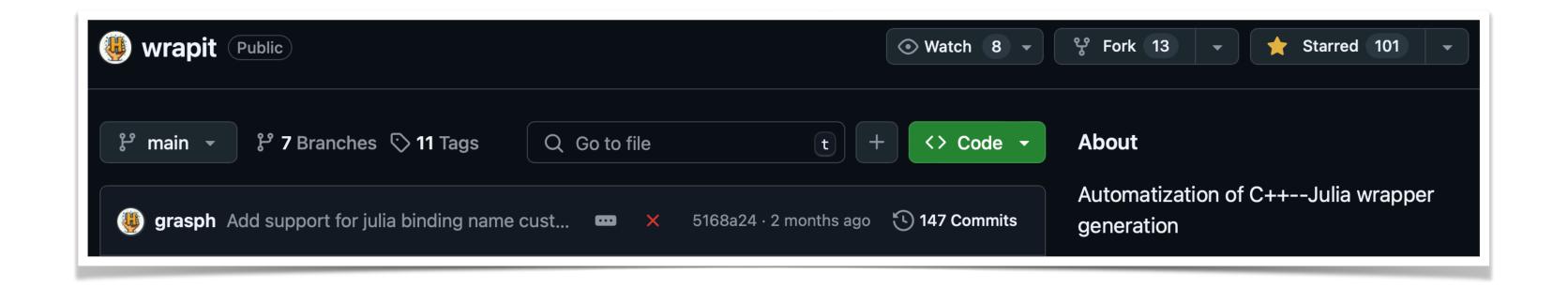
Interoperability and Legacy code

Everything is wrapped

- Use foreign code from Julia
- Wrapit and CxxWrap.jl for (semi-) automatic building of bindings
- non-exhausted list of wrapped libraries
 - Geant4.jl
 - ROOT.jl
 - XRootD.jl
 - Pythia8.jl
 - FastJet.jl
 - UpROOT.jl
 - Etc.

Interoperability





Julia on the HEP workbench

HEP paper using Julia

Study of the doubly charmed tetraquark T_{cc}^+

LHCb Collaboration*



PHYSICAL REVIEW D **104**, L091102 (2021) **Observation of excited \Omega_c^0 baryons in \Omega_b^- \to \Xi_c^+ K^- \pi^- decays**R. Aaij $et~al.^*$ (LHCb Collaboration)

PHYSICAL REVIEW D **98**, 096021 (2018)

Pole position of the a₁(1260) from τ-decay

M. Mikhasenko,^{1,*} A. Pilloni,^{2,3} A. Jackura,^{4,5} M. Albaladejo,^{2,6} C. Fernández-Ramírez,⁷ V. Mathieu,² J. Nys,⁸ A. Rodas,⁹ B. Ketzer,¹ and A. P. Szczepaniak^{4,5,2}

(Joint Physics Analysis Center Collaboration)

Note on Klein-Nishina effect in strong-field QED: the case of nonlinear Compton scattering

U. Hernandez Acosta^{1,2}, B. Kämpfer^{1,3}

more are about to be published...

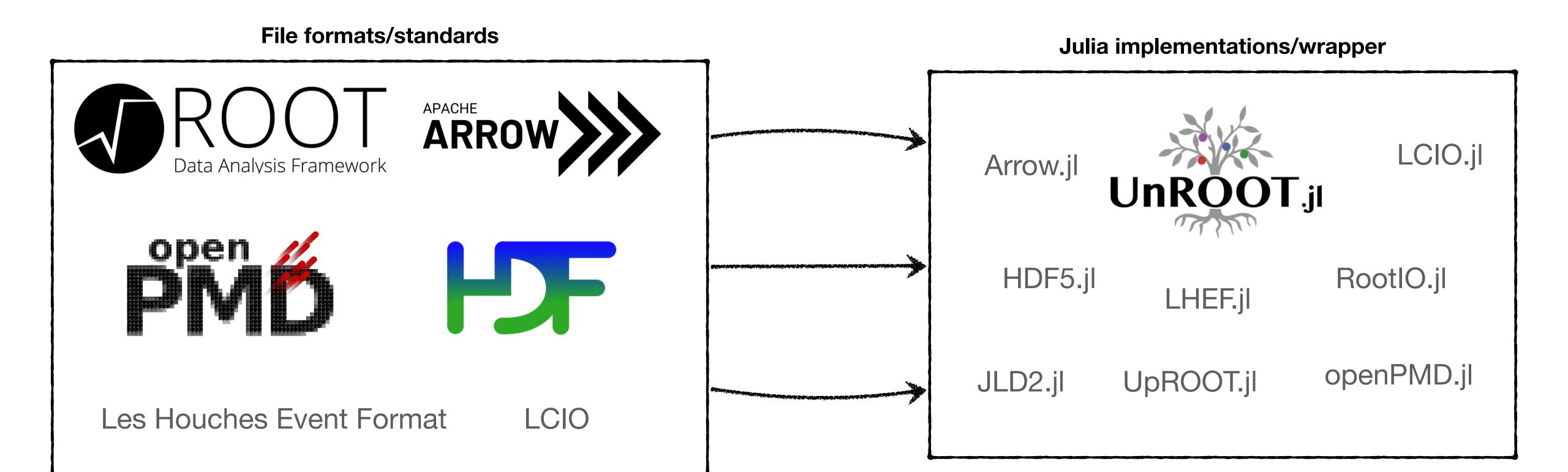
Loading data HEP data formats

UnROOT: an I/O library for the CERN ROOT file format written in Julia

Tamás Gál 1,2, Jerry (Jiahong) Ling 3, and Nick Amin 4

High-performance end-user analysis in pure Julia programming language

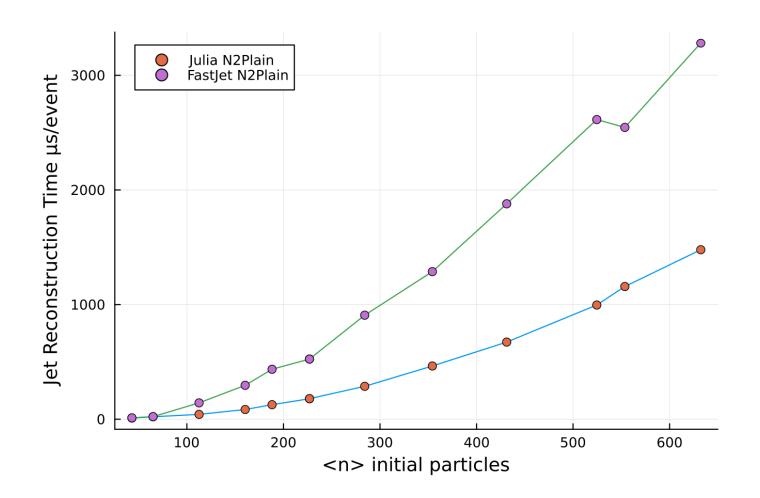
Jerry Ling^{1,*} and Tamás Gál^{2,**}

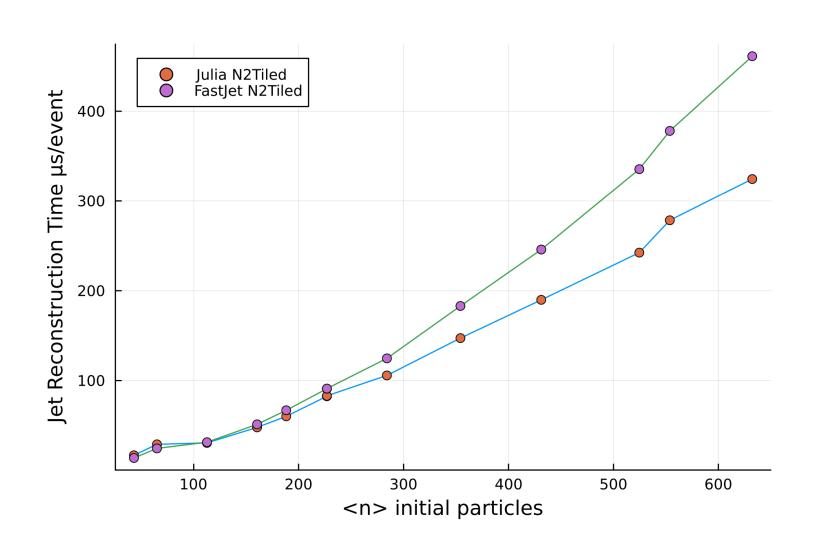


JetReconstruction.jl

Example for rewriting

- Sequential jet clustering
 - Algorithms from FastJet
 - Fully written in Julia
 - Visualization included
- Lesson learned
 - Better ergonomics
 - Better tooling
 - Neat visualization
 - More flexible usage

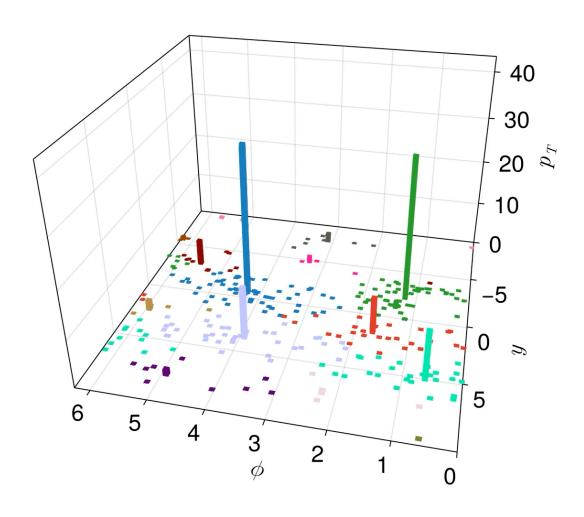




Polyglot Jet Finding

Graeme Andrew Stewart^{1,*}, Philippe Gras^{2,}, Benedikt Hegner^{1,}, and Atell Krasnopolski³

 $Anti - k_T$ Jet Reconstruction, 13TeV pp collision

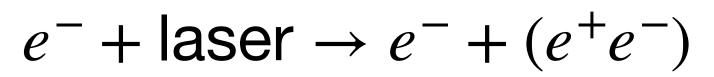


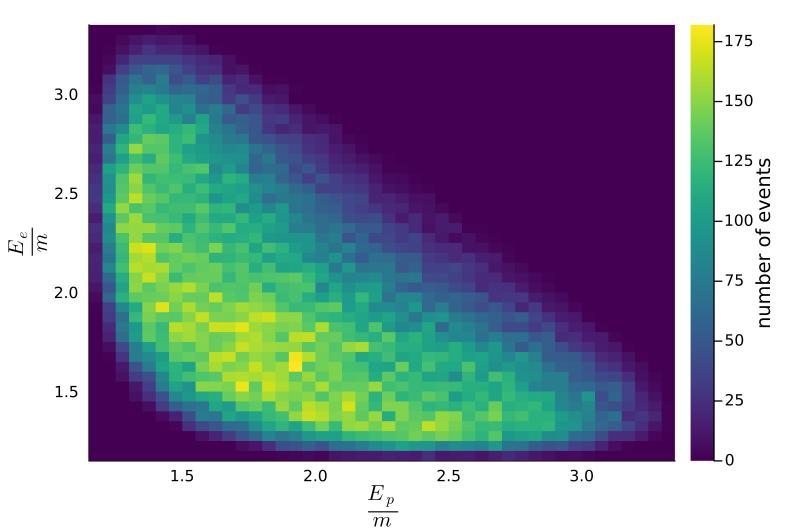


Quantum Electrodynamics.jl

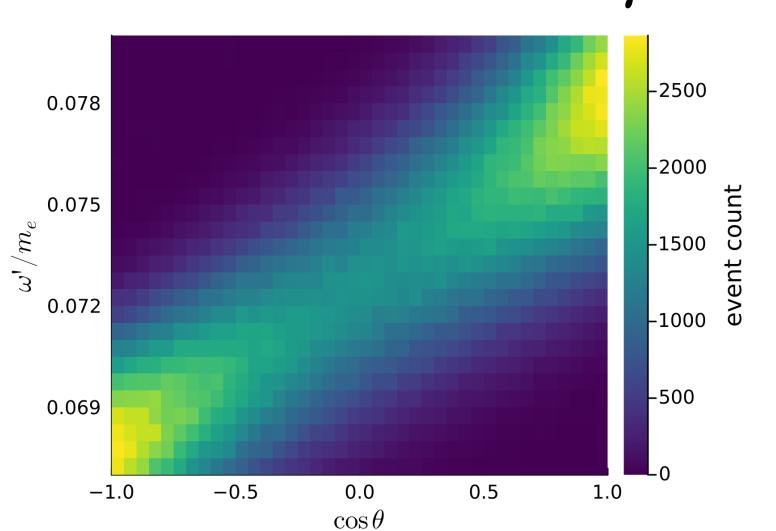
Interfaces and tools available

- Particles
- Lorentz Vectors
- Phase space points
- Computational models
- Scattering processes
- Particle distributions
- Laser fields
- Event generation









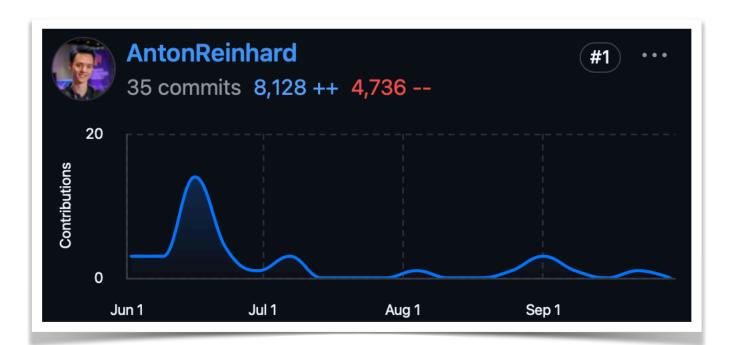


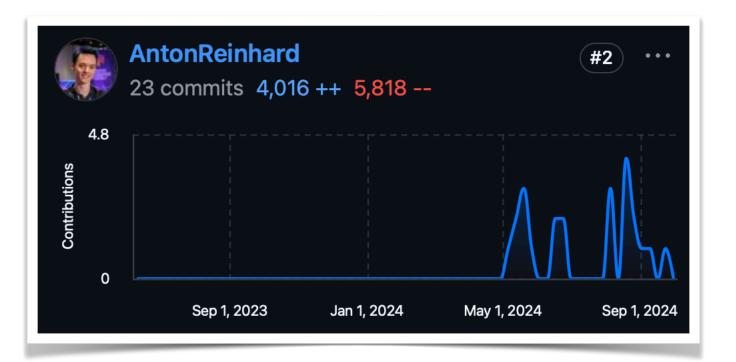


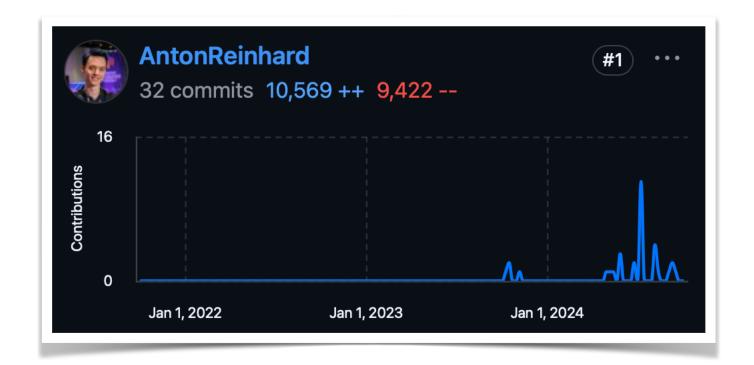
Software development and training

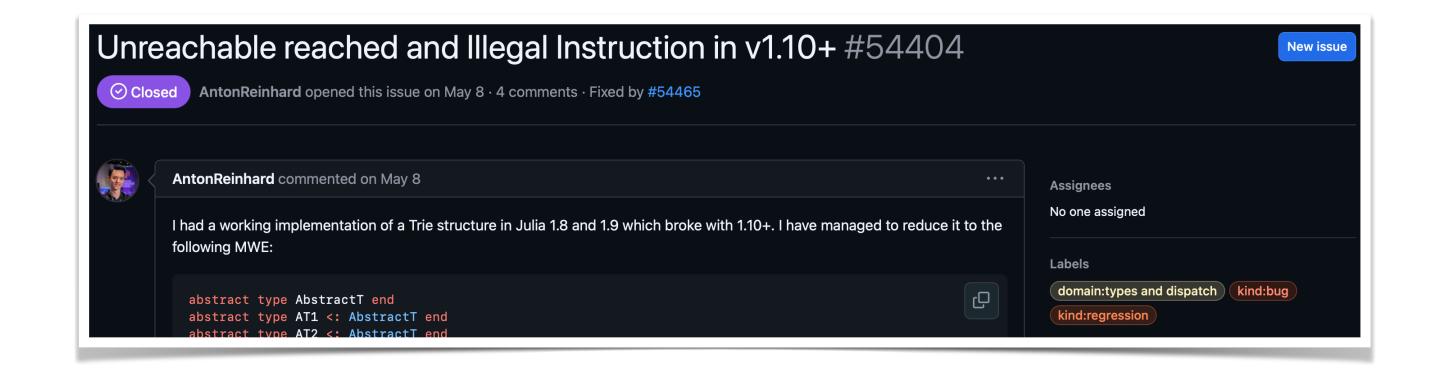
Anton Reinhard

- 1.5 years experience in Julia (coming from C++)
- Two packages (~5k LoC)
 - ComputableDAGs.jl
 - QEDFeynmanDiagrams.jl
 - Stressing the compiler to the max
 - For CPU and GPU
- Main contributor to QuantumElectrodynamics.jl (~20k LoC)









Easy access

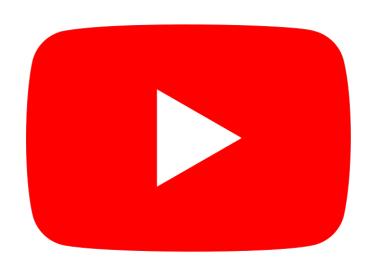
New people are very welcome!

- Availability: GitHub
- Open-source nature
- Friendly community
- Many communication channels



https://github.com/JuliaLang

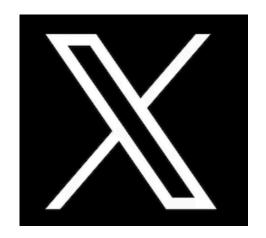






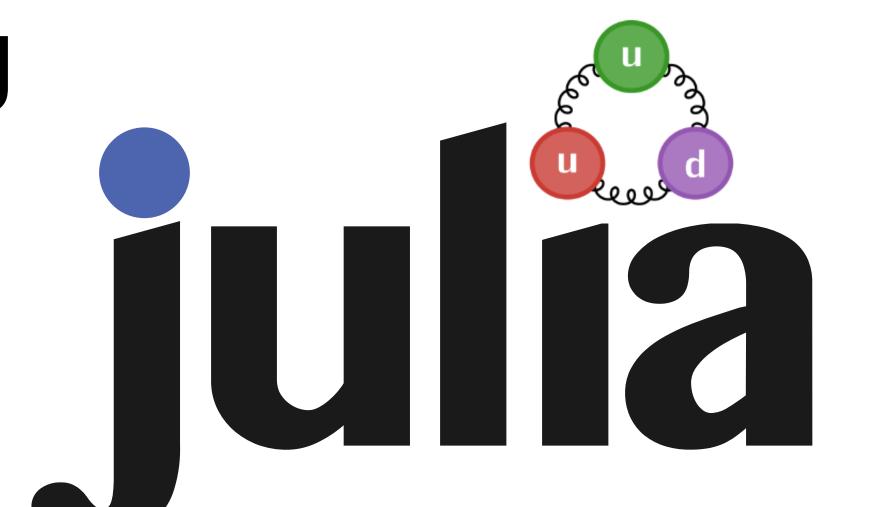






Community building JuliaHEP @ HSF

- JuliaHEP working group (2022)
- JuliaHEP annual workshop
 - 2023: ECAP in Erlangen
 - 2024: CERN
- Monthly community calls
- Monitoring/Supporting development: https://github.com/JuliaHEP
- Tutorial material + example project





Potential of the Julia Programming Language for High Energy Physics Computing

A paradigm shift or just another tool?

Balanced perspective

- Julia is a competitive contender in the HEP software game
- Consider using Julia-wrapped versions of existing code in your next little side project (or allowing your student to do so)
- Making use of the Julia infrastructure when adding new features
- Incrementally rewriting the existing code to benefit even more
- did I mention it runs on GPU as well ;-)

Balanced perspective

- Julia is a competitive contender in the HEP software game
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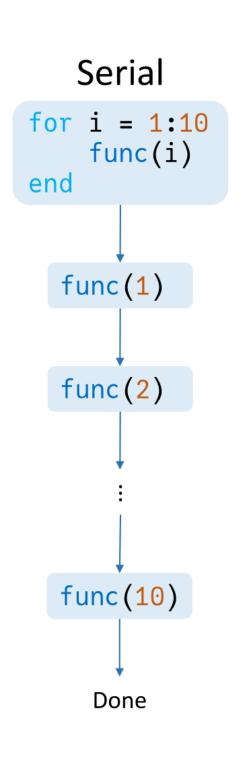
Thank you for your attention!

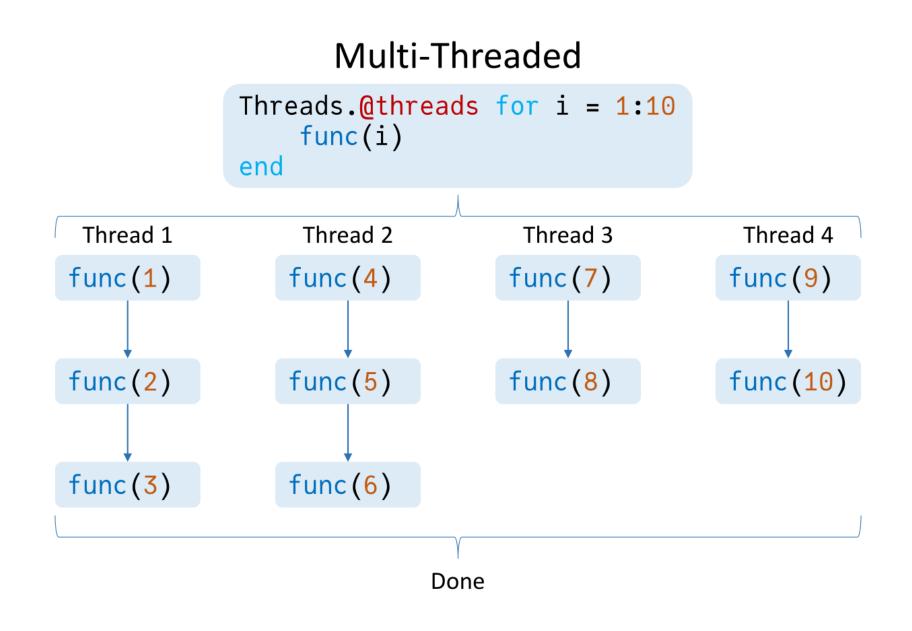
Backup

Parallel computing

Native Threading support

- Support for OpenMP-like models
 - Parallelization of loops
- Support for M:N threading
 - M user threads are mapped onto N kernel threads
- Support for task migration
 - Tasks can be started, suspended, and resumed again





Multiple dispatch

Function and methods

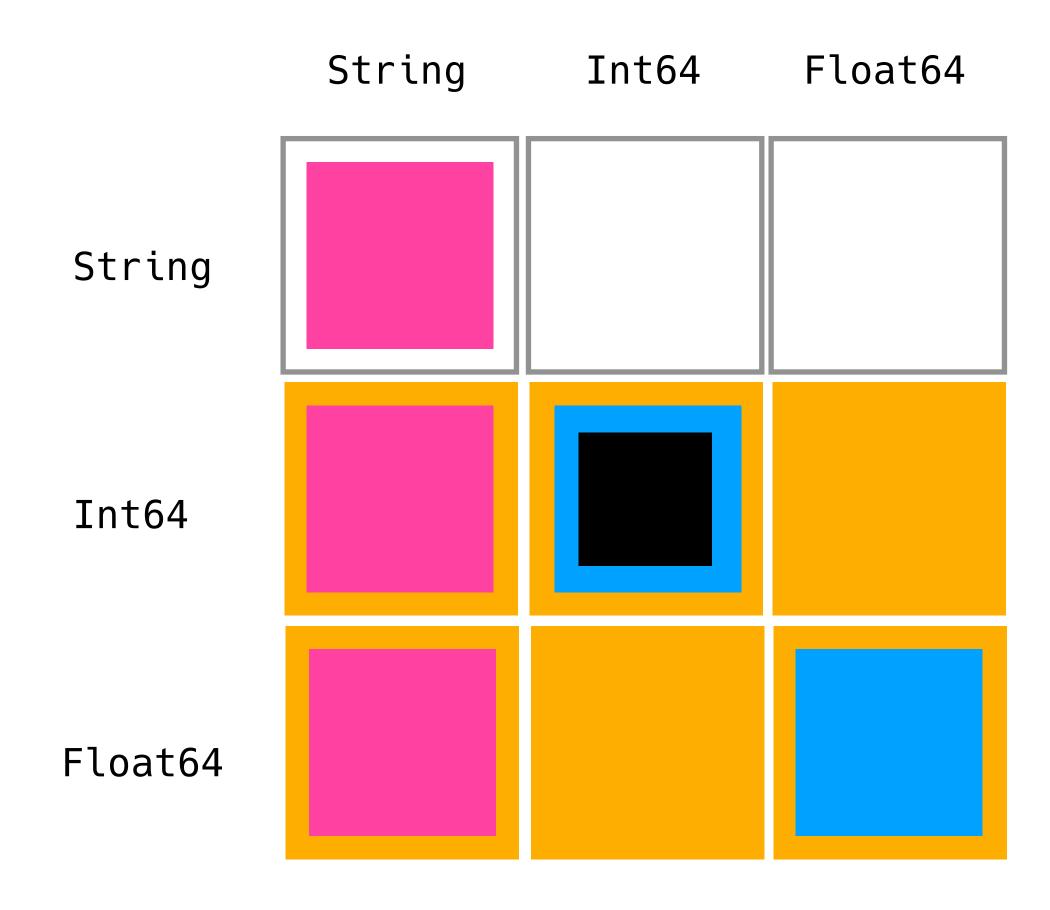
```
f(::Any, ::Number)

f(::T, ::T) where {T<:Number}

f(::Int64, ::Int64)

f(::String, ::Any)</pre>
```

Float64<:AbstractFloat<:Real<:Number<:Any</pre>



Multiple dispatch II

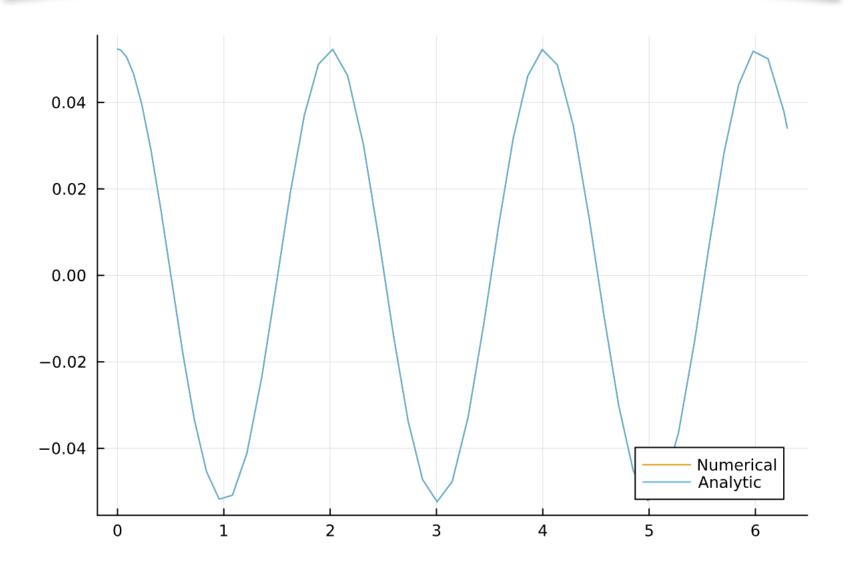
Expressiveness

Dispatch degree	Syntax	Dispatched on	Selection power
None	f(x,y,z)	{ }	1
Single	x.f(y,z)	{x}	X
Multiple	f(x::X,y::Y,z::Z)	{x,y,z}	X · Y · Z

Multiple dispatch III Unreasonable effectiveness

- Allows generic code based on abstract types
- Allows arbitrary optimization
- Orthogonal development
- Solves the expression problem

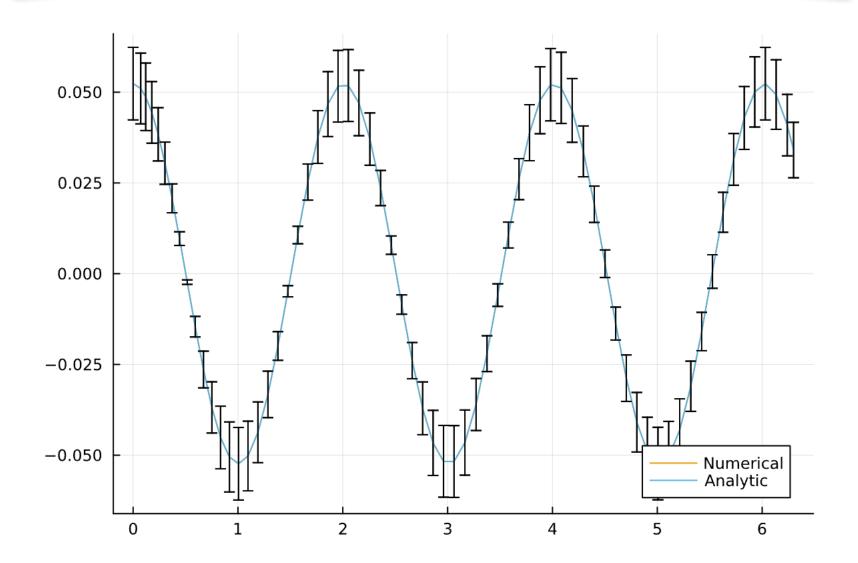
```
using DifferentialEquations, Plots
q = 9.79
                  # Gravitational constants
L = 1.00
                  # Length of the pendulum
#Initial Conditions
                             # Initial speed and initial angle
u_0 = [0, \pi / 60]
tspan = (0.0, 6.3)
#Define the problem
function pendulum(du,u,p,t)
    \theta = u[1]
    d\theta = u[2]
    du[1] = d\theta
    du[2] = -(g/L) * \theta
end
#Pass to solvers
prob = ODEProblem(pendulum, u₀, tspan)
sol = solve(prob, Tsit5(), reltol = 1e-6)
# Analytic solution
u = u_0[2] .* cos.(sqrt(g / L) .* sol.t)
plot(sol.t, getindex.(sol.u, 2), label = "Numerical")
plot!(sol.t, u, label = "Analytic")
```



Multiple dispatch III Unreasonable effectiveness

- Allows generic code based on abstract types
- Allows arbitrary optimization
- Orthogonal development
- Solves the expression problem

```
using DifferentialEquations, Measurements, Plots
g = 9.79 \pm 0.02; # Gravitational constants
L = 1.00 \pm 0.01; # Length of the pendulum
#Initial Conditions
u_0 = [0 \pm 0, \pi / 60 \pm 0.01] # Initial speed and initial angle
tspan = (0.0, 6.3)
#Define the problem
function pendulum(du,u,p,t)
    \theta = u[1]
    d\theta = u[2]
    du[1] = d\theta
    du[2] = -(g/L)*\theta
end
#Pass to solvers
prob = ODEProblem(pendulum, u₀, tspan)
sol = solve(prob, Tsit5(), reltol = 1e-6)
# Analytic solution
u = u_0[2] .* cos.(sqrt(g / L) .* sol.t)
plot(sol.t, getindex.(sol.u, 2), label = "Numerical")
plot!(sol.t, u, label = "Analytic")
```



Alternatives*?

Why not use ...

... only low-level languages?

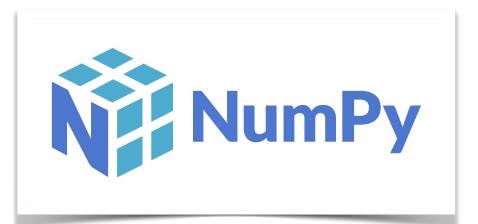
- Take years to learn...
- ...decades to master
- Boilerplate code
- Hardware specific
- Mostly non-interactive
- Missing tools/libraries

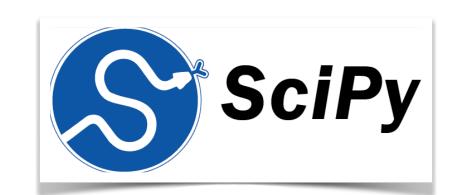


Why not use ...

... third-party libraries?

- "Use C/C++ under the hood"
- Valid in their scope
- Hard to do something outside the box
- Interoperability? Anyone?
- The vendor decides what is performance-critical

















Why not use ...

... Numba, PyPy, Pythran, etc?

- Sufficient for small code pieces
- These are second languages
 - Support only a subset of the host language(s) ...
 - ... and/or add new commands/ logic/concepts
- Usually not a low-level language
 - e.g Numba is neither Python nor C



