Julia for research

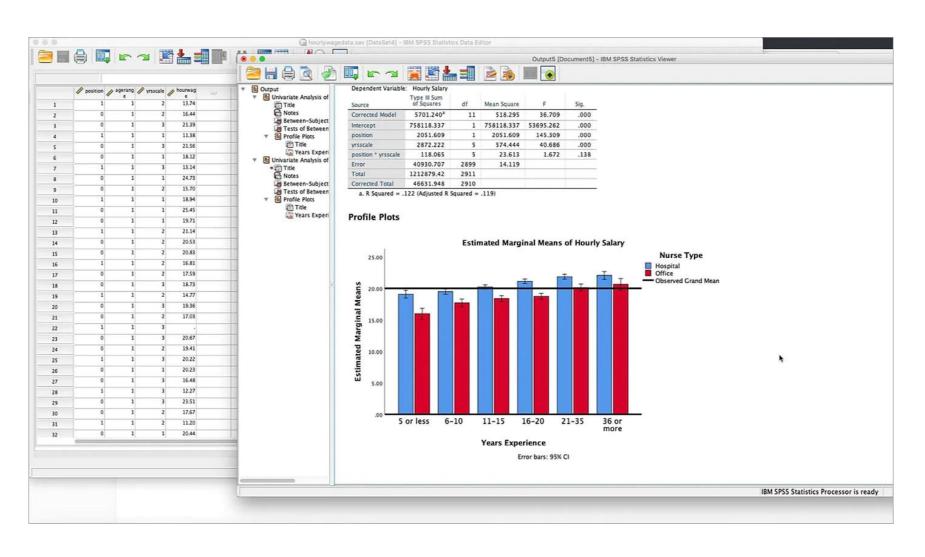
Tomas Fiers



Lunchtime data club at the School of Psychology

Dec 7, 2022

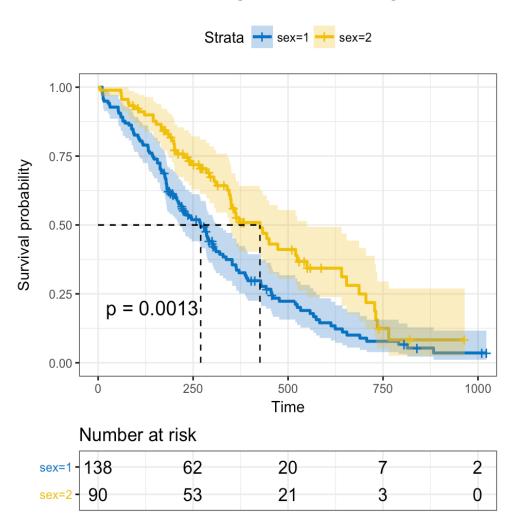
Why learn a programming language?



Why program?

- Automate analyses
 - → less error-prone
- Customize
 - Special plots
 - Tweak analyses
- Run simulations!
- For fun

An example custom plot

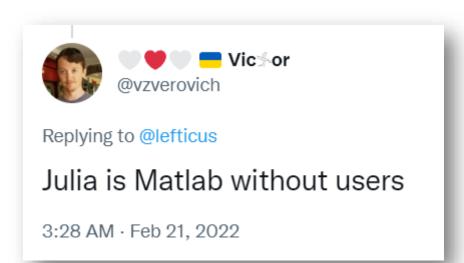


Choosing a programming language

as a researcher

	First release	Free & open?	Online community
		(hackable, "own your code running environment")	(→ Learning resources & documentation polish)
Main choices:			
R	1995 / 1976 (S)	Yes	Huge
Python	1991	Yes	Huge
Others			
Julia	2012	Yes	Medium
Matlab	1979	No	Large

..is also choosing a community





Julia syntax

```
# Simulate a simple neuron for input current 'I'
# and return when the neuron fires its first spike
function first_spike(I, Δt, τ)
    N = length(I)
    v = 0
    for i in 1:N
        v += Δt * (I[i] - v) / τ # Euler integration
        if v > 2 # Spike!
        return time = i * Δt
        end
    end
    return nothing
end
```

In Python / R / Matlab: "Avoid for-loops"
"Write vectorized code"

Compilation. Your code whe CPU

- If one line of Julia code corresponds to just a few CPU instructions
- ..then the same line in base Python / R / Matlab will often correspond to an order of magnitude more instructions
 - ..That's why the code that does the 'real' numeric work in these languages is actually written in C / C++ NumPy, PyTorch, Tensorflow, dplyr, ...: all have their core written in a different language
 - ..That's why, to have your code run fast, you're discouraged from writing for-loops for numeric code ..
 - .. and instead use the provided library functions
 e.g. np.where(...)
 - Python is often used as "glue-code" (see next slide)
 - If you want a custom numeric algorithm that's not provided by the libraries, you need to learn C / C++ The "two languages-problem"

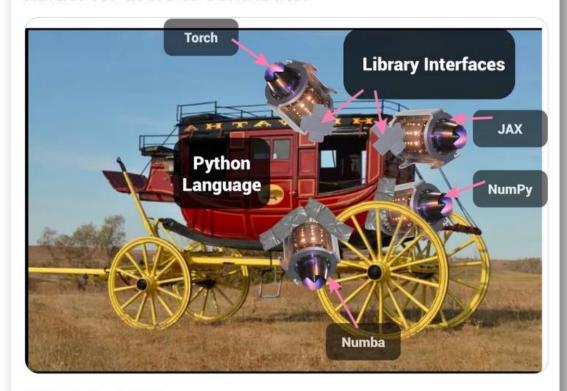
- * Matlab added JIT compilation in 2015 (but it's rather opaque)
 - Python can have JIT compilation via the fantastic **Numba** package. (But you can only use base

 Python with Numba, not arbitrary other packages).



The more I use Julia, the more Python and its numeric libraries look like a Victorian-era stagecoach with jet engines duct-taped to it, each pointing a different direction (=mutually incompatible).

It's such a weird ecosystem, and makes it so much harder for users to contribute.



5:50 PM · Nov 7, 2022

JIT compilation

- If one line of Julia code corresponds to just a few CPU instructions
- ..then the same line in base Python / R / Matlab* will often correspond to an order of magnitude more CPU instructions
- Why is this?
 - The same line of code (say, z = x + y) does different things, based on the **type** of X and y
 - If they're integers (8 + 3), use the `leaq` CPU instruction
 - If one is a float (8 + 3.3), call `convert` and use the floating point processor unit
 - If they're both plots, call subroutines, to compose the plots together into a bigger figure
 - •
 - Python, R, and Matlab need to check the types of X and Y every time the line is run, and then call the appropriate subroutines
 - Hence all these extra CPU instructions
 - Julia will *infer* the types of x and y
 - When? The first time that the function which contains our code is called
 - It does this (i.e. does 'type inference') based on the arguments that the function was called with (more specifically, their types), and by analyzing the function source code
 - I then compiles a fast version of the function This is just-in-time (JIT) compilation

Data analysis in Julia

- DataFrames.jl
 - Tidyverse's dplyr & Python's Pandas equivalent
 - Better API than Pandas, imho
 - In the very capable hands of Bogumił Kamiński
 - Check out his tutorials: github.com/bkamins/Julia-DataFrames-Tutorial
- <missing> datatype is built-in in Julia
 - distinct from <nothing>
- I plot using Python's matplotlib 😜
 - Via PyPlot.jl
 - There's also Makie.jl
 - ..and Gadfly.jl, which is ggplot-inspired

Julia likes

Unicode variable names & operators

```
izh() = begin
    # Conductance-based synaptic current
    I_syn = g_e*(v-E_e) + g_i*(v-E_i)
    # Izhikevich 2D system
    \Delta \cdot v = (k*(v-v_1)*(v-v_t) - u - I_syn) / C # Membrane potential
    \Delta \cdot u = a * (b * (v - v_r) - u)
                                                # Adaptation current
    # Synaptic conductance decay
    # (g<sub>e</sub> is sum over all exc synapses)
    \Delta \cdot g_e = -g_e / \tau
    \Delta \cdot g_i = -g_i / \tau
end
has\_spiked() = (v \ge v_s)
on_self_spike() = begin
    V = V_r
    u += \Delta u
end
```

Julia likes

Community

- Discourse forum & Slack
- Scientists
- Contribute to ecosystem (open source, build upon others)
- As close-to-the-metal as you like
 - Look under the hood
 - Understand why something is slow / fast
- "structs and functions" design style
 - Versus when you're designing software in Python, you tend to use more OOP (inheritance)

Inspectability

- `@edit` to jump to source code of anything... amazing
- `@code_native` to see cpu instructions
- `?` for documentation
- •

Dependency management

- Project.toml
- Manifest.toml

• Macro's

Lisp-like. 'Code as data'

Julia annoyances

- Package startup time ("time-to-first-plot")
 - Language developers are working hard this year to improve this
- No winning plotting package yet
- `name.<tab>` completion (API discovery) not as good as Python
- Getting floats to print with lower precision is way more difficult than it should be for new users
- Traits / interfaces (lack of)
- Error handling is underdeveloped / under-practiced ("→ silent fails & crashes")
- See also:
 - yuri.is/not-julia
 - danluu.com/julialang
 - <u>viralinstruction.com/posts/badjulia</u>

"Julia has a correctness problem"

- (i.e. there's nasty hidden bugs everywhere)
- Not true for Base Julia:
 - every line there is pored over by many language developers
 - automatic test coverage is very comprehensive
- For other people's packages:
 - Not a problem in my experience.
 - But you have to inspect the packages that you use, if they're not in Julia Base;
 and make a value judgement about their quality
 - A lot of Julia packages are of very high quality in my experience
 - Except for the lack of error checking (of inputs and outputs)
 - Julia doesn't hold your hand: you gotta know what you're doing mathematically / numerically / statistically

Why did I switch to Julia?

• Advent of Code:) (2021)

• Physical units in neuron simulations:

- I could just keep using:
 - my Jupyter notebook workflow
 - my Matplotlib experience

```
parameters = (
    # Izhikevich neuron
    C = 100 * pF
    k = 0.7 * (nS/mV)
    V_1 = -60 \times mV
    v_t = -40 * mV
    a = 0.03 / ms
    b = -2 * nS
    V_r = -50 \times mV
    \Delta u = 100 \times pA
    # Synapses
    E_e = 0 * mV
    E_i = -80 * mV
    \tau = 7 * ms
    # Inputs
    N_e = 40
    N_i = 10
    N = N_e + N_i
    \Delta g_e = 60 \text{nS} / N_e
    \Delta g_i = 60 \text{nS} / N_i
    # Integration
    \Delta t = 0.1 ms
    T = 10seconds
```

Julia tips

- Code must be type-inferable ("type-stable")
 - Put everything in (small) functions
 - If using globals: `const`, or typed
- Read the manual
 - Especially the "Performance tips" section, if you're wondering why your code is not as fast as promised. Also:
- Ask questions on the forum
 - discourse.julialang.org
 - People are very eager to help, and the community managers do a great job
- Use Revise.jl (Use all of Tim Holy's packages actually).
 - This minizes nr. of times you have to restart the Julia session (re: time-to-first-X problem)
 - Plus:
 - If using VS Code, there's a plugin for Julia. Also: the JuliaMono font:)
 - On Windows, use the Julia REPL in the Windows Terminal
 - Checkout startup.jl
- Don't load unnecessary packages
 - Julia Base has no real latency (time-to-first-X) problem. It's loading many packages that gets you
 - Especially packages that have many dependencies themselves (looking at you SciML ecosystem :P)
 - Do you really need this package? Can you just implement it yourself / copy the relevant part?
- Learn by doing
 - Like by doing some Advent of Code puzzles!

```
# Code excerpt from the
# JuliaMono homepage.
# Original by Zygmunt Szpak
\otimes = kron
N = length(\mathcal{D}[1])
\mathcal{M}, \mathcal{M}' = \mathcal{D}
\Lambda_1, \Lambda_2 = C
e_1 = (0SMatrix [1.0; 0.0; 0.0])
e_2 = @SMatrix [0.0; 1.0; 0.0]
for n = 1:N
       index = SVector(1,2)
       \Lambda_{n}[1:2,1:2] = \Lambda_{1}[n][index,i]
       \Lambda_n[3:4,3:4] = \Lambda_2[n][index,i]
                  = hom(\mathcal{M}[n])
                  = hom(\mathcal{M}'[n])
                 = (m \otimes m')
       \partial_{\mathsf{x}} \mathbf{u}_{\mathsf{n}} = \left[ (\mathbf{e}_{\mathsf{1}} \otimes \mathbf{m}') \right] (\mathbf{e}_{\mathsf{2}} \otimes \mathbf{m}')
                = \partial_x \mathbf{u}_n * \Lambda_n * \partial_x \mathbf{u}_n
                = \theta' * B_n * \theta
       \Sigma_n^{-1} = inv(\Sigma_n)
end
```

Should you use **julia**?

- Do you 'just' need data analysis, automation, and pretty, customized plots?
 - Then, no

- Or do you also write custom numeric algorithms / simulations?
 - Then, yes :)
 - ..Unless you already know Matlab and don't have the time
 - ..Plus, Python and R have huge ecosystems of packages that might already do your custom thing
 - Also, Python has **Numba** for JIT-optimization of hot inner loops (<u>numba.pydata.org</u>). That might be enough

Code sharing, git, GitHub

- tfiers.github.io/phd
 - made with <u>JupyterBook</u>
 - auto-built and -published with GitHub Actions on GitHub Pages
- github.com/schluppeck/ng-data-club
 - /presentations subdirectory, by ISO8601 date+slug