Julia Part II

Julia for Data Science

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UoA

Oct 31, 2017





A basic problem about any body of data is to make it more easily and effectively handleable by minds – our minds, her mind, his mind.

John W. Tukey, Exploratory Data Analysis, Addison-Wesley, 1977

Section 1

Get Started

Interface Cuteness

- Matlab uses help, Julia switches into help mode by typeing ?
 - lookfor in Matlab becomes apropos, e.g., apropos ("determinant")
- In Julia can access OS commands by typing;, e.g.,

```
; pwd
```

- Useful things to know
 - history with up and down keys
 - matches partial strings
 - auto-complete with TAB
- Standard shell-commands
 - ► Ctrl-c interrupt process
 - ▶ Ctrl-a start of the line
 - ▶ Ctrl-e end of the line
 - Ctrl-d exit
- Startup file ~/.juliarc.jl

Other useful bits and pieces

- Comments in shell-style #
- Functions that modify their arguments have a name like sort!
- Useful commands

```
whos()
@which sin(2)
versioninfo()
```

Numerical constants

```
pi
golden
e
im
eulergamma
```

- Long numbers: 1_000_000
- Others useful constants

```
JULIA_HOME # path to julia executable
nothing # function that returns void
```

Section 2

Plotting

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Plot packages

There are several plotting packages

- PyPlot: emulates Matlab, through Python's matplotlib
- Gadfly: emulates R's ggplot
- Plots: aims to become front end for all backends
- GR, UnicodePlots, Plotly, PlotlyJS, Vega, Winston, StatsPlots, PlotRecipes, GLVisualize, PGFPlots, Qwt, ...

PyPlot

https://github.com/JuliaPy/PyPlot.jl

- You should have it installed (see startup sheet)
 - ▶ it uses PyCall to call Python
 - uses Julia's multimedia backend to display using various Julia graphical backends (Qt, GTK, ...)

http://matplotlib.org/api/pyplot_api.html

- it should be fairly portable
- Syntax is intended to be similar to Matlab
 - ▶ as implemented in matplotlib

title ("A sinusoidally modulated sinusoid")

Main commands

You can get a listing of commands by typing PyPlot.TAB TAB Some examples

```
plot
qcf()
xlim
xlabel
xkcd
surf
bar
figure
fill
pie
t.ext.
scatter
```

When running in a script, you need to use show () to get the fig to display.

Example 1

```
using PyPlot
x = 0:0.1:2*pi;
y = 0:0.1:pi;
X = repmat(x, 1, length(y));
Y = repmat(y', length(x), 1);
S = [\cos(x[i]) + \sin(y[j]) \text{ for } i=1:\text{length}(x),
                                  j=1:length(y)
surf(X, Y , S, cmap=ColorMap("jet"), alpha=0.7)
xlabel("x")
ylabel("y")
```

Example 2

LaTeXString defined by L"...."

More Examples

```
https://gist.github.com/gizmaa/7214002
https://lectures.quantecon.org/jl/julia_plots.html
```

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Section 3

A Stupidly Short Tour of Packages

Installing Packages

Packages are a collection of code encapsulated into a set of **Modules**, and (usually) put on GitHub in a standard format

 Adding a package can be done in a few ways, but the most standard is

```
Pkg.add("PyPlot")
Pkg.update()
```

- takes care of dependencies
- installs code
- Get status, and see where code is

```
Pkg.status()
Pkg.Dir.path()
LOAD PATH
```

Using Packages

Packages are a collection of code encapsulated into a set of **Modules**, and (usually) put on GitHub in a standard format

Commands to use or import

```
using PyPlot import PyPlot
```

- using simple access to all exported functions
- ▶ import uses names space of module, e.g., PyPlot.plot
- Other ways to import code

```
include( "Code/my_code.jl" )
reload( "PyPlot" )
```

Lots of Packages

```
https://pkg.julialang.org/
```

- 1518 registered packages!
- Some trending packages

```
https://github.com/trending/julia
```

- ▶ Deep Learning https://github.com/denizyuret/Knet.jl
- ► IJulia is a Jupyter interactive environment
 https://github.com/JuliaLang/IJulia.jl
- Gadfly is ggplot-like plotting https://github.com/GiovineItalia/Gadfly.jl
- PyCall lets you call Python https://github.com/JuliaPy/PyCall.jl
- Convex programming
 https://github.com/JuliaOpt/Convex.jl

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I will talk about a couple of direct use in Data Science

DataFrames

- Concept comes from R (as fas as I know)
- Like a 2D array except
 - can have missing values
 - multiple data types
 - ★ quantitative
 - categorical (strings)
 - labelled columns
- Nice mapping from Frame to CSV (or similar)

```
https:
```

//en.wikibooks.org/wiki/Introducing_Julia/DataFrames

DataFrames

Download the following dataset, and put in a local folder called Data

https://raw.githubusercontent.com/vincentarelbundock/Rdatasets/master/csv/datasets/Titanic.csv

```
using DataFrames
data = readtable("Data/Titanic.csv",
         nastrings=["NA", "na", "n/a", "missing"])
head (data)
size(data)
showcols (data)
data[:Name]
temp = deepcopy(data)
push! ( temp, @data([1314, "my bit", "nth", NA, "male
tail(temp)
deleterows! (temp, 3:5)
data[ data[:,:Sex] .=="female", : ]
data[ :height ] = @data( rand(size(data,1)) )
sort!(data, cols = [order(:Sex), order(:Age)])
                                            4 = → = 900
```

JSON

- JSON = JavaScript Object Notation
- Data exchange format
 - increasingly popular
 - lightweight
 - portable
- Stores name/value pairs
 - so it maps to a Dictionary well
 - but lots of other data can be stored as JSON

http://www.json.org/



JSON

Download the following dataset, and put in a local folder called Data

https://raw.githubusercontent.com/corysimmons/colors.json/master/colors.json

```
import JSON
c = JSON.parsefile("Data/colors.json")
c["purple"]
JSON.print(c)
```

Distributions

- Package for probability distributions and associate facilities
 - moments
 - pdf, cdf, logpdf, mgf
 - samples
 - Estimation: MLE, MAP
- Included here because
 - its useful
 - its a nice example of a Julia package
 - ★ type hierarchy used to provide structure to RVs e.g., Distributions → Univariate → Continuous → Normal
 - multiple dispatch used to call correct version of generically named functions
 - easy to add a new one

https:

//juliastats.github.io/Distributions.jl/latest/

Distributions

```
using Distributions
srand (123)
d = Normal(0.0, 1.0)
x = rand(d, 10)
quantile.( d, [ 0.5, 0.975] )
params (d)
minimum(d)
location(d)
scale(d)
x = rand(d, 100)
fit mle(Normal, x)
```

Section 4

Parallel Processing

Julia Macros

Macros look a bit like functions, but begin with @, e.g.,

```
@printf("Hello %s\n", "World!")
@printf "Hello %s\n" "World!"
```

Why?

- Macros are parsed at compile time, to construct custom code for run time
 - e.g., for @printf, we want to interpret the format string at compile time,
 - ★ In C, the printf function re-parses the format string each time it is called, which is inefficient
 - ★ Also means that C compilers need to be very smart to avoid many hard-to-debug mistakes of the wrong types of arguments being passed to printf

Julia Macros

Julia uses quite a few macros, and you can define your own

```
@time [sin(i) for i in 1:100000];
@which sin(1)
@show 2 + 2
macroexpand(quote @time sin(i) end)
```

Macros can be MUCH faster ways of implementing code

```
https://statcompute.wordpress.com/2014/10/10/julia-function-vs-macro/
```

- Macros can be used to automate annoying bits of replicated code,
 e.g., @time
- It's part of the meta-programming paradigm of Julia
 - ideas from Lisp
 - Julia code is represented (internally) as Julia data
 - so you can change the "data"

What Julia Does

- Raw Julia code is parsed
 - converted into an Abstract Syntax Tree (AST), held in Julia
 - syntax errors are found
- Create a deeper AST
 - Macros play here they can create and modify unevaluated code
- Parsed code is run
 - hopefully really fast

So what does that have to do with Parallel Programming?

- Julia has several functions and macros to aid in parallel processing
- I think the coolest is the "Map/Reduce" functionality introduced by @parallel macro
 - maybe you can see why it is a macro?

Setting up for Multi-Processor Ops

There are two approaches for a single, multicore machine

```
> julia -p 4
julia > addprocs(3)
julia > procs()
julia > nprocs()
```

I'm not going to get into how to build a cluster

Map Reduce

- Many simple processes can be massively parallelised easily by decomposing them into Map-Reduce operations
- Map: apply an (independent) function or mapping to a small piece of data
- Reduce: combine the results of all the mappings into a summary
- It's a particularly good framework for multiple simulations run in parallel



First make sure that all processes have the required environment

```
@everywhere cd("/home/mroughan/Presentation/Julia/C
@everywhere include("my_code.jl")
```

Now run parallelised loop, aggregating results with operator +

```
nheads = @parallel (+) for i = 1:200_000_000
    Int(rand(Bool))
end
```

But take care – data is not automatically shared!!!!!!!

Section 5

Tips and tricks

Type stability

Use @time\$ to compare the speed of these two functions for large n

```
function t1(n)
    s = 0
    for i in 1:n
        s += s/i
    end
end
```

```
function t2(n)
    s = 0.0
    for i in 1:n
        s += s/i
    end
end
```

Don't avoid loops

Use @time\$ to compare the speed of these two functions for large n

```
function t1(n)
    x = zeros(n)
    for i in 1:n
       x[i] = i^2
    end
    return x
end
```

```
function t2(n)
  x = collect(1:n).^2
end
```

Avoid global variables

- Apart from the usual arguments
- Hard for compiler to optimise around, because type may change
 - ▶ if you need them, and they don't change, define them as constants

```
const DEFAULT_VAL = 0
```

- Note variables defined in the REPL are global
- Execute code in functions, not global scope
 - write functions, not scripts

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Pre-allocate outputs

Use @time\$ to compare the speed of these two functions for large n

Access arrays in memory order, along columns

- 2D arrays stored in column order (as in Fortran)
 - C and Python numpy are in row order
- Accessing in this order avoids jumping around in memory
 - get the best value out of pipeline and cache

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Lots more tips

- https://docs.julialang.org/en/latest/manual/ performance-tips/
- https://github.com/Gnimuc/JuliaSO
- http://blog.translusion.com/posts/julia-tricks/
- https://julialang.org/blog/2017/01/moredots

Standard Tools

- Debugging https://github.com/Keno/Gallium.jl
- BenchmarkTools package https://github.com/JuliaCI/BenchmarkTools.jl
- Profiler https: //docs.julialang.org/en/latest/manual/profile/
- Lint package https://github.com/tonyhffong/Lint.jl
- Unit testing https: //docs.julialang.org/en/stable/stdlib/test/
- Literate programming (aka Knitr, ...) https://github.com/mpastell/Weave.jl, and iJulia
- •

Standard Tools

- There is a lot more to learn
 - function definition
 - creating modules
 - types
 - interfaces to other languages
 - **...**
- I tried to concentrate on things where I think it is hard to get started learning yourself

Final Comment

Julia is v.shiny, but it's not all roses

- Current version is 0.6
 - each 0.1 increment has introduced "breaking" changes
 - the core is still evolving
 - it's getting better, but change is painful
- Some libraries aren't all there
 - stagnation, ...
- Plotting
 - argggh!

Conclusion

I don't like endings, so here are some quotes to go on with.

We – or the Black Chamber – have a little agreement with [Knuth]; he doesn't publish the real Volume 4 of the Art of Computer Programming, and they don't render him metabolically challenged.

Charles Stross, The Atrocity Archive, 2001

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Some more useful references

- https://github.com/trending/julia
- https://docs.julialang.org/en/latest/manual/ performance-tips/

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Bonus frames