Notes on the Julia Programming Language

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Abstract

This document summarizes my experience with the Julia language. Its main purpose is to document tips and tricks that are not covered in the official documentation.

1 My Setup $(1.1)^1$

Since things are in flux, I find it useful to use the offical JuliaPro installation. My startup file loads the packages OhMyREPL and Revise. Revise comes after packages from the standard libraries, so it does not track changes to those.

1.1 JuliaPro (1.1)

Sometimes JuliaPro gets slow and has trouble updating the REPL screen. Then restarting the computer is the only solution.

1.2 Julia + Editor (1.1)

It appears that the default editor is determined by the system wide file association. No need to set the JULIA_EDITOR environment variable.

¹ Each section is labeled with the Julia version for which it was last updated.

One drawback: Links in the terminal REPL are not clickable. A substantial drawback during debugging. So I end up using BBEdit as my main editor, but do some debugging in Juno. Not ideal.

2 Arrays (1.1)

2.1 Indexing

Extracting specific elements with indices given by vectors:

```
A = rand(4,3,5);
A[CartesianIndex.([1,2], [2,2]), 1] -> A[1,2,1] and A[2,2,1]
Similar to using sub2ind:
idxV = sub2ind(size(A), [1,2],[2,2],[1,1])
A[idxV]
```

To extract a "row" of a multidimensional matrix without hard-coding the dimensions, generate a view using selectdim.

3 Debugging

The Juno debugger stopped working in V.1.1 (invoking it hangs Julia). But the command line debugger may well be the better option. After using Debugger invoke @enter foo(x) to start a debugging session.

Particularly useful:

- bp on error
- bp add func: line with possible restrictions on particular argument types.

4 External Programs

One can execute bash commands with run.

Question: Trying to run a bash script using run('. myscript.sh') produces a permission denied error (even though permissions are set so that others can execute. Why?

5 Formatting

The Formatting package seems to be the best bet. It uses Python like syntax and can format multiple arguments simultaneously (not well documented). Example:

```
fs = FormatExpr("{1:.2f} and {2:.3f}")
format(fs, 1.123, 2)
yields "1.12 and 2.000".
```

6 Modules

6.1 LOAD_PATH

Only modules located somewhere along the LOAD_PATH can be loaded with using.

But: If a directory contains Project.toml, it becomes a project directory and only entries listed in Project.toml can be loaded (even if the directory is on the LOAD_PATH).

6.2 Extending a function in another module (1.1)

The problem:

• Module B defines type Tb and function foo(x :: Tb).

Module A contains a generic function bar(x) that calls foo(). It should use the foo() that matches the type of x. That is, when called as foo(x :: Tb), we want to call B.foo.

Solution:

- Module A:
 - Define the stub: function foo end
 - Call foo(x) from within bar.
- Module B:
 - Define function foo(x :: Tb)
 - import A.foo
- Now A.bar(x) knows about B.foo() and calls it when the type matches the signature.

See Duck typing when 'quack' is not in 'Base'.

7 Packages

7.1 Environments (1.1)

An environment is anything with a Manifest.toml. When you start Julia, you enter the version's environment (e.g. 1.1). When you add a package, you effectively edit Manifest.toml.

You can add additional environments using Pkg.activate() or pkg> activate . and then Pkg.add to initialize a Mainfest.toml in that directory.

7.1.1 Stacked environments

When you activate an environment, you do **not** deactivate previous environments. Instead, you now operate in a sort of union of all the environments that you activated during a session. This matters when both environments list the same packages in the Manifests.

Example: Start in environment 1.1 and Pkg.add(D). Pkg.activate(P) and Pkg.add(D) with a different version of D (or using the local path for D). Which version of D is used after using D?

I encountered a case where I could not covince Julia to update an unregistered package, even using Pkg.rm followed by Pkg.add. The reason was that 1.1 referenced the same package, pointing to a fixed github commit.

7.2 Creating a package (1.1)

The easiest way is PkgSkeleton.jl. You need to set your github info (user.name etc) using

git config --global user.name YourName

This must be done insite a git directory. Then generate generates the directory structure and the required files (Project.toml etc).

Note: The package name should not end in \blacksquare .jl \blacksquare – this is automatically appended when the package is registered (?).

7.3 Package workflow (1.1)

Your packages will generally be unregistered. Your workflow needs to account for the fact that Pkg does not track versions for unregistered packages. Here are the steps:

- 1. Initialize a package in a folder pDir; call the package P. This generates a directory structure with src, test, etc. If you plan on using this package as a dependency, it is best to place it in a sub-folder of JULIA_PKG_DEVDIR (~/.julia/dev by default). The reason is that Pkg.develop wants to download your code there.
- 2. While the code is being worked on: Pkg.activate(ps). This makes sure that changes are written to the package's environment (Project.toml).
- 3. To add registered dependencies, simply use Pkg.add(pkgName). No problem.
- 4. To add unregistered dependencies D that may change as you work on your project, use Pkg.develop instead.

- (a) Write code that makes a PackageSpec for D. This simplifies managing the package. Call this ps. ps should point to D's local directory, not to a github url. Otherwise, you end up tracking what is on github rather than your local edits.
- (b) Pkg.develop(ps) simply changes the entry for D in Project.toml from pointing at the github repo to pointing at the local dir. Key point: This is only operative while the environment P is active.
- (c) Pkg.develop is an alternative to Pkg.add, which edits Project.toml to point at github.
- 5. To freeze the state of the code:
 - (a) push P and D to github.
 - (b) in the environment for P: Pkg.add(ps) where ps should now point at the github url for D.
 - (c) Even if you continue to push updates for unregistered dependencies to github, your package should track the fixed versions (identified by the sha key that defines the commit). Just don't run Pkg.update.

7.4 Unregistered packages as dependencies (1.1)

Important point: Unregistered packages need to be added as dependencies "by hand." Pkg cannot track when other packages depend on them. This is a known issue 810. That means:

- Suppose you are working in P with dependency D that depends on E.
- Pkg.add(D) does not add E to P's Project.toml.
- You need to explicitly Pkg.add(E).

Tracking changes in unregistered packages is difficult. While doing development work, adding dependencies with Pkg.develop seems to work. Once the code is copied to a remote computer, things get more complicated.

• The solution suggested on discourse suggests to always develop packages and to have relative paths in Manifest.toml. That would be relative paths of the form ../MyPackage. User directory expansion, as in ~/abc does not work.

Note that Pkg.update does nothing for unregistered dependencies.

It is easy enough to get the most recent version of the code on any computer. Forcing Julia to recompile once the code has changed is more difficult. On the local machine, Revise does the trick. But since you don't edit code on the remote machine before running it, I don't see how one can force a recompile. What works sometimes:

• Restart the REPL.

7.5 Creating a package registry (1.1)

Any registry that lives in ~/.julia/registries is automatically used by Pkg.

In principle, it is easy to create your own registry (see discourse for a guide). But there are problems:

- 1. Entries must be added to the registry by hand. Each package gets an entry line in the registry and a subdirectory in the registry directory with Versions.toml, Deps.toml, Compat.toml, Package.toml.
- 2. Any inconsistencies in the entries will cause the registry to be ignored by Pkg. So this approach is fragile.
- 3. Each time a package is changed or updated, the registry needs to be augmented by hand, including all dependencies.

Fredrik Ekre has an example in his github repo. At this time, the approach is not really workable.

7.6 Multiple Modules in one Package (1.1)

The cleanest approach is sub-modules. One can still import Foo.Bar to only use the sub-module (especially for testing). In the test function, non-exported functions can be called as Bar.f().

7.7 Testing a package (1.1)

Activate the package by issuing activate . in the package's directory (not in src). Then type test.

Placing test code inside a module:

- This can be useful when the test code defines structs that one would like to be able to modify without having to restart Julia all the time. Note that objects defined in tests are no longer visible once Pkg is exited.
- Place the module definition into test. Add push(LOAD_PATH, @__DIR__). This has to be done in each module. Not elegant.

8 Performance

The compiler does not optimize out if false statements. Hence, defining a constant that switches self-testing code on and off does not result in no-ops. Of course, the overhead is quite small.

8.1 Profiling

The output generated by the built-in profiler is hard to read. ProfileView does not compile (1.1).

StatProfilerHTML is a good alternative (1.1). It provides a flame graph with clickable links that show which lines in a function take up most time.

8.2 Type stability

One can automate checking for type stability using the code_warntype() function. Example:

- For function foo(x), call code_warntype(stdout, foo, (Int,1)).
- This can be written to a file by changing the IO argument.
- It generates output even if no issues are found.

• The amount of output generated is overwhelming. Signs of trouble are Union types, especially return types (at Body:).

9 Remote Clusters

9.1 Getting started with a test script

How to get your code to run on a typical Linux cluster?

- Get started by writing a simple test script (Test3.jl) so we can test running from the command line.
- Add the Julia binary to the PATH using (on Macos, editing ~/.bash_profile):

```
PATH="/Applications/Julia-1.1.app/Contents/Resources/julia/bin:$PATH"
```

• Then make sure you can run the test script with julia ■/full/path/to/Test3.jl■

Now copy Test3. jl to a directory on the cluster and repeat the same.

- You may need to add the Julia binary to the path. On Longleaf (editing ~/.bash_profile):
 export PATH="/nas/longleaf/apps/julia/1.1.0/bin:\$PATH"
- Then run julia "/full/path/to/Test3.jl"

Now run the test script via batch file:

```
sbatch -p general -N 1 -J "test_job" -t 3-00 --mem 16384 -n 1 --
mail-type=end --mail-user=lhendri@email.unc.edu -o "test1.out" -
-wrap="julia /full/path/to/Test3.jl"
```

9.2 Generate an ssh key

This allows log on without password. Instructions on the web.

Now you can use the terminal to log in with ssh user@longleaf.unc.edu.

9.3 Rsync File Transfer

A reliable command line transfer option is rsync. The command would be something like

rsync -atuzv "/someDirectory/sourceDir/" "username@longleaf.unc.edu:someDirectorySNotes:

- The source dir should end in "/"; the target dir should not.
- Exluding .git speeds up the transfer.
- --delete ensures that no old files remain on the server.

9.4 Git File Transfer

- 1. Change into the package directory (which is already a git repo).
- git remote add longleaf ssh://lhendri@longleaf.unc.edu/nas/longleaf/home/lhen
 3. Initialize the remote directory with a bare repo: git init --bare.
- 3. Initialize the remote directory with a bare repo: git init --bare. Bare means that the actual files are not copied there. It needs to be bare so push does not produce errors later.
- 4. Verify the remote: git remote show longleaf

When files have changed:

1. Change into the package directory

2. Add a remote destination (once):

- 2. git commit -am ■commit message■
- 3. git push longleaf master

Note that this does not upload any files! So this only works for packages, not for code that should be run outside of packages.

9.5 Running code on the cluster

Steps:

- 1. Copy your code and all of its dependencies to the cluster (see Section 9.3).
- 2. Write a Julia script that contains the startup code for the project and then runs the actual computation (call this batch.jl).
- 3. Write a batch file that submits julia batch.jl as a job to the cluster's job scheduler. For UNC's longleaf cluster, this would be slurm. So you need to write job.sl that will be submitted using sbatch job.sl.

9.5.1 The Julia script

Submitting a job is (almost) equivalent to julia batch.jl from the terminal.

• Note: cd() does not work in these command files. To include a file, provide a full path.

If you only use registered packages, life is easy. Your code would simply say:

```
using Pkg
# This needs to be run only once
Pkg.add(MyPackage)
# If you want the latest version each time
Pkg.update()
using MyPackage
MyPackage.run()
```

If MyPackage contains unregistered dependencies, things get more difficult. Now batch.jl must:

- 1. Activate the package's environment.
- 2. develop all unregistered dependencies. This replaces the invalid paths to directories on the local machine (e.g. /Users/lutz/julia/...) with the corresponding paths on the cluster (e.g. /nas/longleaf/...). Note: I verified that one cannot replace homedir() with ~ in Manifest.toml.

- 3. using MyPackage
- 4. MyPackage.run()

This approach requires you to keep track of all unregistered dependencies and where they are located on the remote machine. My way of doing this is contained in PackageTools.jl in the shared repo (this is not a package b/c its very purpose is to facilitate loading of unregistered packages).

For an example implementation of the entire process, see batch_commands.jl in TestPkg2LH.

- This uses PackageToolsLH to handle directories on different computers and file transfer.
- write_command_file() writes the julia file that is to be executed remotely (command_file.jl).
- write_sbatch writes the sbatch file that will be submitted to slurm.
- project_upload() uses rsync to copy the code of the project, its dependencies, and some general purpose code that is required at startup (mainly PackageToolsLH itself) to the remote machine.

9.5.2 The sbatch file

How this works can be looked up online. The only trick is that the Julia command requires a full path (or a relative path, but that's a little risky) on the remote machine.

PackageToolsLH keeps track of where things are on each machine. It is used to build the full paths.

10 Types (1.1)

I find it easiest to write model specific code NOT using parametric types. Instead, I define type aliases for the types used in custom types (e.g., Double=Float64). Then I hardwire the use of Double everywhere. This removes two problems:

- 1. Possible type instability as the compiler tries to figure out the types of the custom type fields.
- 2. It becomes possible to call constructors with, say, integers of all kinds without raising method errors.

10.1 Constructors (1.1)

Constructing objects with many fields:

• Define an inner constructor that leaves the object (partially) uninitialized. It is legal to have new(x) even if the object contains additional fields.

10.2 Loading and saving (1.1)

using FileI0 and extension .jld2 automatically saves in jld2 format. This can save used defined types.

Loading user defined types is more complicated. All modules needed to construct the loaded types need to be known in the loading module and in Main. See Issue 134. It is not possible to use Core.eval(Main, :(using Module)) for unclear reasons.

Implications:

- 1. Each user defined type needs its own load function.
- 2. All dependencies need to imported into Main ■by hand■ for each loaded object.

An alternative is BSON. jl. It has the same limitation.

One could save the ParamVectors in each object and reconstruct the object from those (recursively). This, of course, only works for objects that can be constructed from ParamVectors. Each ParamVector could be stored as a Dict{Symbol, Any}. But even easier: store the ParamVectors directly. Constructing them after loading only requires modelLH. The approach would then be:

1. Collect the ParamVectors from all model objects into a Dict{Symbol, ParamVector}. The symbol identifies the associated model object.

- 2. Save the Dict.
- 3. In Main: using modelLH, so that loading works.
- 4. Function that loads the model:
 - (a) Construct the model object with arbitrary default values.
 - (b) Load the ParamVectors.
 - (c) Sync each ParamVector's parameters into the correct model object. Essentially, the model object needs a constructor that accepts a ParamVector.

11 Unit Testing (1.1)

All codes should be in modules because code in Main runs slower, pollutes Main, and it harder to revise. This also applies to test code.

However, placing the <code>@test</code> or <code>@testset</code> portions into the test module causes them not to run sometimes (why?). It also implies that using the test module runs all tests, which is generally unwanted. I therefore place the <code>@test</code> code into a separate file (not inside a module).

Errors in the code to be tested (but not caught by @test) cause the entire test run to crash. Preventing this requires all tests to be enclosed in a @testset. A sequence of @testset does not do the trick. An error in one prevents all others from being run. Nested @testsets produce nested error reports (nice).

@test statements can be placed inside functions. To preserve result reporting, the function should contain a @testset and return its result.

12 Workflow (1.1)

Revise is key. It is now possible to simply use using on any module once. Revise then automatically keeps track of changes. Using includet creates problems for me.