



# Deploying Julia

Strategies, Architectures and Pitfalls

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## Who am I

- Julia developer since 2012
- Production Julia code since 2016
- Products and services in Julia

## What I'll talk about today

- Packaging – Manifests, Docker, Precompilation
- Sysimages, PackageCompiler
- JuliaC
- Miscellenous Tips and tricks

# What does deployment mean to you?

- Where do you deploy
  - Public or Private Cloud
  - HPC / Supercomputer
  - Desktop / Workstations
- What do you deploy
  - Batch processes (*Model training / Simulations*)
  - Long running service (*Web apps / API services*)
  - Desktop UI
  - OpenSource / Internal / Commercial

# What does deployment mean to me?

- Run code on a machine that is not yours
- Code comes with implicit assumptions about the environment they will run in
  - Operating System
  - Instruction set
  - Dependent libraries and their versions
  - Configuration files

## Send/Publish a file with Julia source code

- Simple 😊
- User needs Julia installed 😬
- User need to manually install packages used 😬
- User may get different versions of packages used 😬
- User may use a different Julia version 😬
- Source code might be proprietary 😬
- Not really feasible except for the very simplest of codes

# Packaging – Project + Manifest

- Project.toml + Manifest.toml
  - Always check in Manifest for Applications
  - User gets the same version of packages 😄
- User needs to install Julia & instantiate 😞
  - The correct version of Julia
    - New Juliaup functionality to pick Julia versions from project
- Same package works across Operating Systems
- Multiple files to track if all you need is a script 😞
  - In 1.13, you can store project+manifest in source files

## Packaging – Registry

- Always use a private registry for private code
  - Unless you are single developer, deploying to single machine
- A registry is only a git repository, so no major infrastructure is required
- LocalRegistry.jl can help with managing a registry
- Makes dependency versioning and version tracking possible

## Considerations for HPC Clusters

- Set your depot to a shared filesystem
  - Shared filesystems are slow
- Make sure to precompile before running a distributed job
  - Don't want thousands of repeated precompiles
  - Consider running with `--compiled-modules=strict`
  - Precompile on login node if possible
- Set CPU target appropriately
  - Especially for heterogenous nodes
- Figure out correct MPI config



# Packaging - Docker

- Public Julia docker images
- Instantiate a manifest during docker build
  - Internet access needed at build time
  - Precompile: `Pkg.precompile()` or use a precompile workload
  - Manage CPU targets unless building on the same machines as deploying
  - `JULIA_CPU_TARGET=generic;sandybridge,-xsaveopt,clone_all;haswell,-rdrnd,base(1)`
- Use memory hints at runtime
  - `julia --heap-size-hint=4G`
- Set thread limits appropriately



## Aside on Precompile workloads

- Julia only compiles the code that runs at top level
- If you have functions that take a long time to compile, but haven't been run at top level (ie at compile time) then you will compile at runtime, and pay a latency cost
- Run those functions at top level or
- Use `PrecompileTools.jl`

## System Images (or sysimage)

- “Serialized snapshot of a julia session”,
- Contains julia objects, compiled code, code yet to be compiled, etc.
- Relatively fast to load
- Julia comes with a sysimage containing the base and (most) standard libraries

# Custom Sysimages solve the compile time latency

```
pkg> add ModellingToolkit
```

```
...
```

```
186 dependencies successfully precompiled in 611 seconds
```

```
julia> @time using ModelingToolkit
```

```
3.443558 seconds (4.96 M allocations: 341.655 MiB, 7.86% gc time)
```

```
julia> @time using ModelingToolkit
```

```
32.879128 seconds (6.31 M allocations: 414.853 MiB, 1.66% gc time,  
1.48% compilation time: 84% of which was recompilation)
```



Enter PackageCompiler.jl



# Custom Sysimages solve the compile time latency

```
julia> using PackageCompiler
```

```
julia> PackageCompiler.create_sysimage(["ModelingToolkit", "GLMakie"];  
sysimage_path="mtk_makie_sys.so")
```

```
C:\> julia +1.12 --sysimage=mtk_makie_sys.so --project=.
```

```
julia> @time using ModelingToolkit  
0.000991 seconds (505 allocations: 27.688 KiB)
```



## Custom sysimages with packagecompiler

- Reduce compile time latency 😄
- Needs a C compiler on the build machine
- Need to use the same version of Julia to load it 😞
- Artifacts need to be downloaded separately 😞
- Need sources to resolve new packages in environment
- Build for each OS (and no ability to cross-compile)
- Needs a loooong time to create, and results are biiiiiig 😞

```
.rwxr-xr-x  1.5G avik   29 Dec 19:07  my_custom_sys.so
```

# Apps with PackageCompiler

- Package Julia+sysimage+artifacts+entrypoint

```
julia> PackageCompiler.create_app( "App", "AppCompiled")
```

```
$ tree AppCompiled/
AppCompiled/
├── bin
│   ├── HelloApp
│   └── julia
├── lib
│   └── julia
├── libLLVM.so.18.1jl
├── libatomic.so.1.2.0
├── libblastrampoline.so.5
├── libdSFMT.so
└── libgcc_s.so.1
```



# Relocatability

- Application should not depend on particular filesystem path
- Paths are used to find different kinds of resources
  - Binaries / Libraries / Data artifacts
- Your package and ALL its dependencies must be relocatable
- Use jll packages and artifacts for native dependencies
- Don't store file paths or pointers in module globals – use `__init__` functions instead.
  - Actually, use `OncePerProcess` instead (1.12+)

# JuliaC

- Create small, standalone binaries (~1.5Mib)
- All code needs to be statically inferable
- 1.12+ only
- Key innovation is ***trim*** support in the compiler
- Needs infrastructure around the compiler, including PackageCompiler

# JuliaC

```
pkg> app add JuliaC
```

```
$ juliac HelloApp --output-exe hello -trim
```

```
$ ls -al hello
```

```
-rwxr-xr-x 1 avik avik 1691304 Jan  5 10:54 hello
```

```
$ ./hello
```

```
Hello, world
```

# JuliaC

- JuliaC binaries depend on `libjulia`

```
$ ldd ./hello
```

```
    libjulia.so.1.12 => /home/avik/.julia/juliaup/julia-  
1.12.3+0.x64.linux.gnu/lib/libjulia.so.1.12 (0x00007cac188a1000)
```

```
    libjulia-internal.so.1.12 => /home/avik/.julia/juliaup/julia-  
1.12.3+0.x64.linux.gnu/lib/julia/libjulia-internal.so.1.12  
(0x00007cac18200000)
```

```
    libunwind.so.8 => /home/avik/.julia/juliaup/julia-  
1.12.3+0.x64.linux.gnu/lib/julia/libunwind.so.8 (0x00007cac17a00000)
```

# Packaging with JuliaC

```
$ juliac HelloApp --output-exe hello --trim --bundle hellodir
```

```
$ tree hellodir
```

```
├─ bin
│   └─ hello
└─ lib
    ├── julia
    │   └─ libatomic.so
    .....
    │   └─ libz.so.1.3.1
    └─ libjulia.so.1.12.3
```

```
4 directories, 57 files
```

## Considerations for long running services

- Latency of first request
  - Good precompile workload, or
  - Warmup requests
- Concurrency – threads or tasks or processes?
- Recommend not putting Julia web services directly on the public internet
  - Proxy via apache/nginx/caddy etc
  - Pay attention to package vulnerabilities

## Considerations for cloud native usage

- Interfaces to cloud platforms - AWS.jl / Azure.jl
- Memory is the biggest cost on public cloud services
- Shared disks are slow, fast IO costs more
- Consider running Julia processes on Kubernetes
  - Kuber.jl provides complete set of entities and operations



Tell us your stories





# References

- <https://github.com/GunnarFarneback/LocalRegistry.jl>
- <https://juliahpc.github.io/>
- Package compiler webinar: <https://youtu.be/J6h6Tj8lluE>
- <https://github.com/JuliaLang/JuliaC.jl/>
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