Deploying a Unidata JupyterHub on the NSF Jetstream Cloud, Lessons Learned and Challenges Going Forward

ESIP Summer 2019 Meeting

Julien Chastang Wednesday, 17 July 2019





Outline

Background and Context

Deploying a Geoscience JupyterHub on NSF Jetstream Cloud

Lessons Learned, Challenges Going Forward



Background and Context





Unidata 2024 Proposal: Science as a Service

The Science as a Service concept draws together Unidata's ongoing work to provide geoscience data and software for analysis and visualization with access to workflows designed to take advantage of cloud computing resources.



NSF Jetstream Cloud Collaboration

- What is Jetstream?
 - A National Science and Engineering Cloud funded by an \$11 million NSF grant.
 - Data centers at IU and TACC.
- Attached to fast Internet2 capability.
- Cloud based on OpenStack for creation of VMs, routers, networks, subnets, security groups etc.
- Unidata has been operating on Jetstream for 3 years through a large research grants
- Once you get through granting process, Jetstream is free





Unidata's Exploration of the Jetstream Cloud Thus Far

- Started by containerizing Unidata technology offerings
 - THREDDS
 - IDM
 - McIDAS ADDE
 - RAMADDA
- Deployed containers to create near complete Unidata data center
- Plenty of NCEP at thredds-jetstream.unidata.ucar.edu
- But what about client-side offerings in cloud?
- Next obvious step: "data-proximate" analysis and visualization





Deploying a Geoscience

JupyterHub on NSF Jetstream

Cloud



What is a Jupyter Notebook?

A narrative of:

- Explanatory and expository text
- Software code (Python, R, etc.) and output
- Equations (MathJax, \text{LAT} \text{ET} \text{EX})
- Figures and multimedia

Lorenz System

The Lorenz system is a series of Ordinary Differential equation studied by Edward Lorenz.

```
\dot{x} = \sigma(y - x)
\dot{y} = \rho x - y - xz
\dot{z} = -\beta z + xy
```

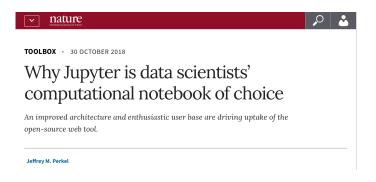
```
[n [10]: def lorenz(x, y, z, s=10, r=28, b=2.667):
             x_dot = s*(y - x)
             y_dot = r*x - y - x*z
             z_dot = x*y - b*z
             return x_dot, y_dot, z_dot
         dt = 0.01; stepCnt = 10000
         xs = np.empty((stepCnt + 1,))
         vs = np.emptv((stepCnt + 1,))
         zs = np.empty((stepCnt + 1,))
         xs[0], vs[0], zs[0] = (0., 1., 1.05)
         for i in range(stepCnt):
             x_dot, y_dot, z_dot = lorenz(xs[i], ys[i], zs[i])
             xs[i + 1] = xs[i] + (x_dot + dt)
             ys[i + 1] = ys[i] + (y_dot * dt)
             zs[i + 1] = zs[i] + (z dot + dt)
          fig = plt.figure()
          ax = fig.gca(projection='3d')
         ax.plot(xs, ys, zs, lw=0.5)
         plt.show()
```







Success of Jupyter in Research and Education

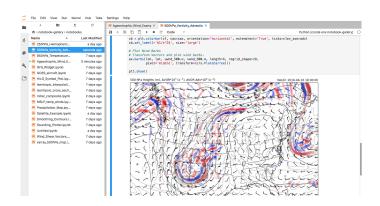


[Jupyter] notebooks are really a killer app for teaching computing in science and engineering - Lorena Barba, Engineering Professor, GWU





JupyterLab: Next Generation UI



- Terminal (git, conda, etc.)
- Text Editor

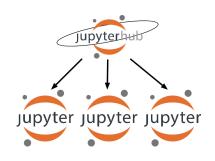




JupyterHub: Multi-user Jupyter Notebook Server

Fernando Pérez: It is infeasible for IT support to assist 800 students install complex software on their laptops.

- Users log in to a JupyterHub server
- Users have their own work space w/ notebooks
- Excellent for workshops or in the classroom
- Administrator can configure ahead of time on behalf of user





Zero to JupyterHub Project

<u>Problem</u>: A single JupyterHub server running on a large VM can only serve a small number of students (< 10).

Solution: Zero to JupyterHub project aims to install JupyterHub across several orchestrated VMs to accommodate many more users

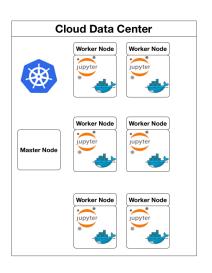
- Virtual Machines
- Software Containers (i.e., Docker)
- Data center software orchestration (i.e., Kubernetes)

Zero to JupyterHub allows for many more users.



Zero to JupyterHub







Zero to JupyterHub on Jetstream/OpenStack

 z2j ported to Jetstream by <u>Andrea Zonca</u> SDSC, w/ help from Jeremy Fischer at IU







K8s Deployed on Jetstream with KubeSpray Project

- Deploy Kubernetes clusters with:
 - Terraform: creation on VMs, routers, networks, subnets, security groups
 - Ansible: kubernetes cluster software installation
- Added layer of scripts to streamline deployment
 - setup-kube.sh
 - setup-kube2.sh
- Initially, must decide on size/number of VMs via terraform.tf
- Can be scaled "manually" thereafter (no autoscaling)



Zero to JupyterHub Customization and Configuration

- YAML configuration file
- HTTPS available with LetsEncrypt or custom certificates
- Authentication via oauth (github, globus, etc.)
- Custom Unidata Docker Container:
 - Gallery: PyAOS examples
 - Workshop: PyAOS training
 - Online Python Training
- Environments to run these projects already installed
- JupyterLab





Persistent Storage Allocation for Each User

- Each user gets a 10 GB disk allocation
- This disk space remains available to them for ? amount of time



Unidata Community JupyterHub

- jupyterhub.unidata.ucar.edu
- 5 "m1.medium" size VMs totaling 30 CPUs 80 RAM
- 60 users (not concurrent) most of which try it one time, though some return customers

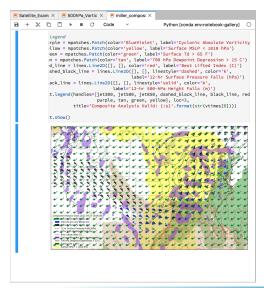


Other Unidata JupyterHubs

- Notre Dame of Maryland University (no kubernetes)
- Southern Arkansas University (no kubernetes)
- JupyterHub for UCAR SOARS summer internship program

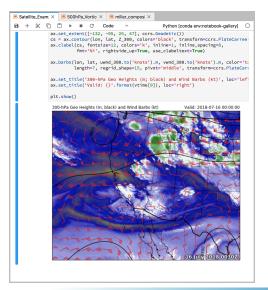


Example Notebook: Miller Composite



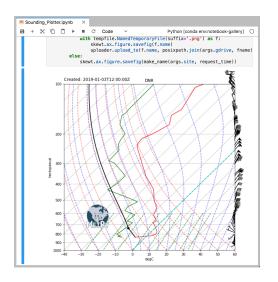


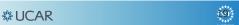
Example Notebook: Satellite + GFS Model





Example Notebook: Upper Air SkewT





Lessons Learned, Challenges Going Forward



Problems with JupyterHub Deployments and Maintenance

- Finicky VMs during cluster creation (takes several tries)
- Timeout errors at every level (deployment and running)
- Complexity associated with software running on clusters
- JupyterHub spawn errors
- Disk allocation errors
- Network problems at TACC
- General lack of reliability throughout entire tech stack





Trough of Sorrow

the startup curve



Source: Paul Graham via andrewchen.co







Additional Caveats

- What to do with user data over long term? What guarantees?
- This project is not Pangeo
 - Goals are more modest and Unidata focused
 - Would be happy to deploy Pangeo on Unidata Jetstream allocation



Lessons Learned

- Scriptifying your deployments to make your life easier
- Interview your users before so cluster can be accurately sized
- Tech stack is new and fragile and will take time before it is stable
- Don't advertise too early, scale gradually by introducing to incrementally wider audiences to address problems
- Need to be persistent to overcome tech challenges: ask for help on github issues and gitter
- Science professionals have a high threshold for problems as long as they can arrive at a desired objective





Future Plans

- Get technology problems under control, don't build on a Swiss cheese foundation
- Experiment more with IU Jetstream data center
- Autoscaling on OpenStack with Zonca collaboration
- Address github issues
- Once things stabilize, promote to wider community



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Resource and Questions

https://github.com/Unidata/xsede-jetstream



