Powering ODH With Ray

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Jupyter & Ray In The Cloud

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Landscape

Ray At 10000 Meters

Jupyter and Open Data Hub in Context

Ray on ODH

Demo! 🎉

Community Collaborations

Ray's Ecosystem Niche



Ray's Compute Model

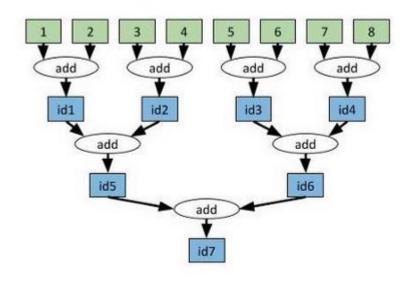
Tasks

Actors

Ray's Compute Model

Tasks Actors @ray.remote def f(x): # body Actors Actors # body Actors # class foo(object): # body

Dependency DAG

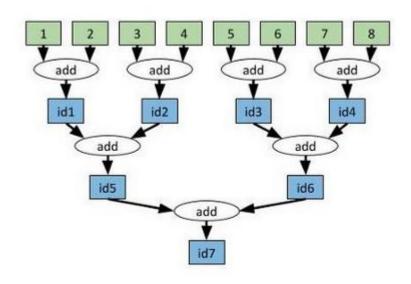


```
id1 = add.remote(1, 2)
id2 = add.remote(3, 4)
id3 = add.remote(5, 6)
id4 = add.remote(7, 8)
id5 = add.remote(id1, id2)
id6 = add.remote(id3, id4)
id7 = add.remote(id5, id6)
result = ray.get(id7)
```

Robert Nishihara

https://towardsdatascience.com/modern-parallel-and-distributed-python-a-quick-tutorial-on-ray-99f8d70369b8

Dependency DAG



```
id1 = add.remote(1, 2)
id2 = add.remote(3, 4)
id3 = add.remote(5, 6)
id4 = add.remote(7, 8)
id5 = add.remote(id1, id2)
id6 = add.remote(id3, id4)
id7 = add_remote(id5, id6)
result = ray.get(id7)
```

Robert Nishihara

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Ray's Data Model

- Plasma Object Store
- Typeless and Schemaless
- "Local First"
 - Pull remote data when needed
 - Read/Write is all local to worker node

Ray's Scheduling Model

- "Local First"
 - Prefers to run tasks on local scheduler
 - Submits to global scheduler if necessary

Native Ray Libraries

- Tune: Scalable Hyperparameter Tuning
- RLlib: Scalable Reinforcement Learning
- RaySGD: Distributed Training Wrappers
- Ray Serve: Scalable and Programmable Serving

Ray Community Integrations

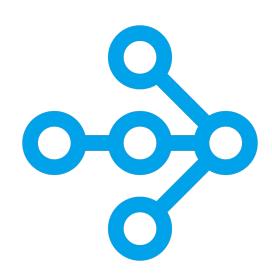
- XGBoost
- Dask
- Horovod
- sklearn
- Spacy
- huggingface

Ray Community Integrations

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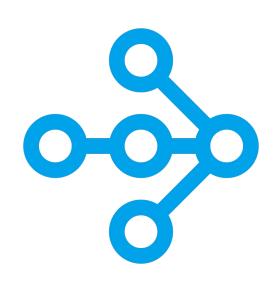
Literate And Interactive Ray...





https://docs.ray.io/en/master/ray-libraries.html

Hosted In The Cloud

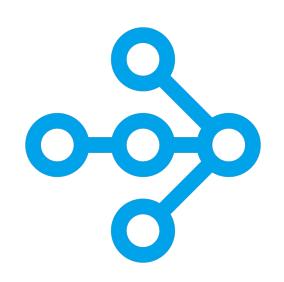






https://docs.ray.io/en/master/ray-libraries.html

Hosted In The Cloud



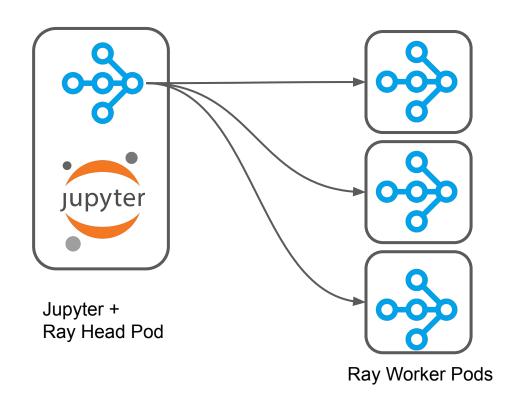




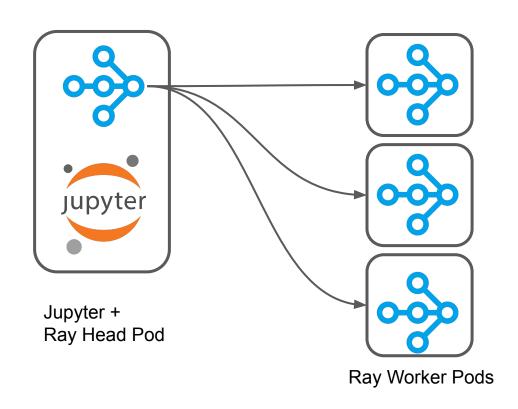


https://docs.ray.io/en/master/ray-libraries.html

Jupyter + ray.init()

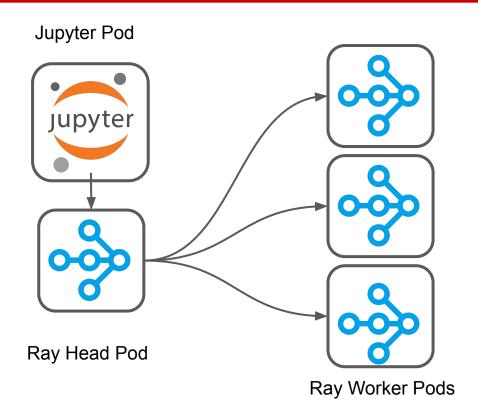


Jupyter + ray.init()





Jupyter + ray.util.connect()

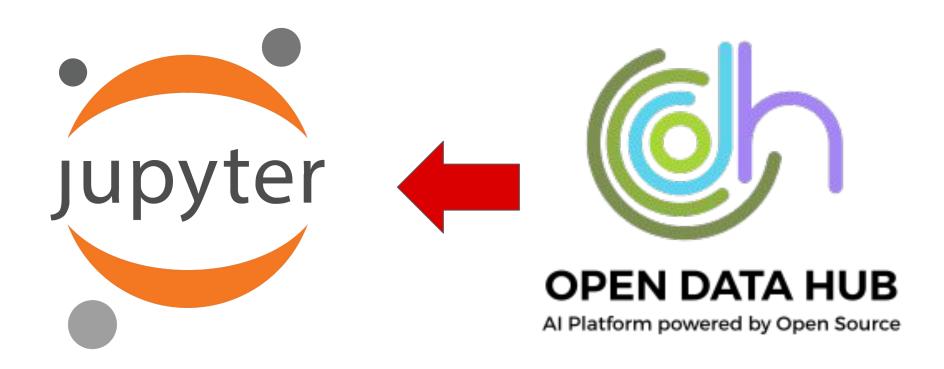




Jupyter ...



Jupyter via Open Data Hub



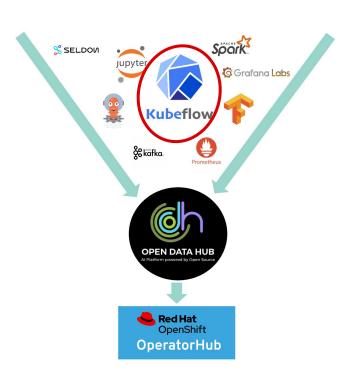
Open Source Downstream

Reference Platform



Open Source Downstream

Reference Platform



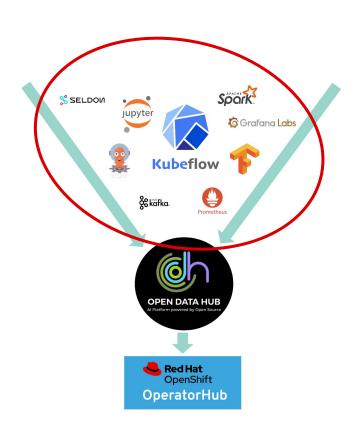
Open Source Downstream

Reference Platform

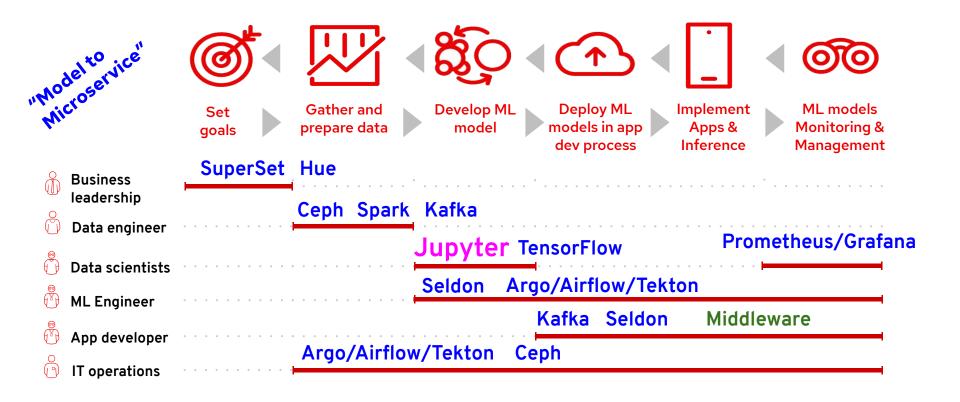


Open Source Downstream

Reference Platform



Data Science with ODH



Dog-Fooding ODH at Red Hat



Application Logs

Applications in the product release pipeline store their **runtime logs** in our system. These groups are also engaged for **anomaly detection**



Cluster Metrics

Operational metrics from OpenShift clusters. AlOps is engaged here.

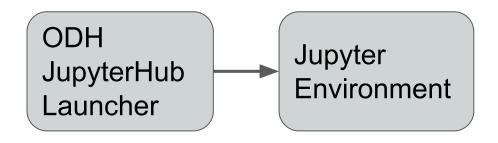


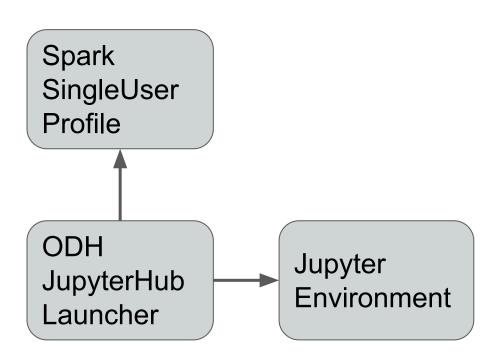
Customer Support Data

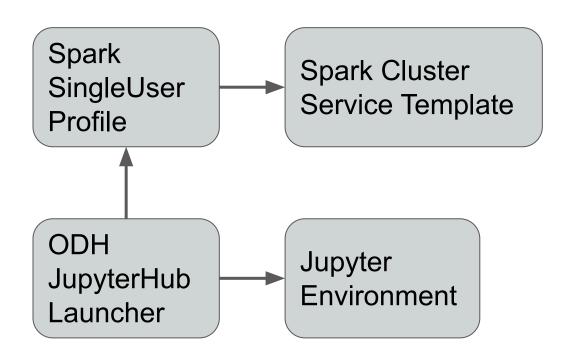
Storage of customer data like

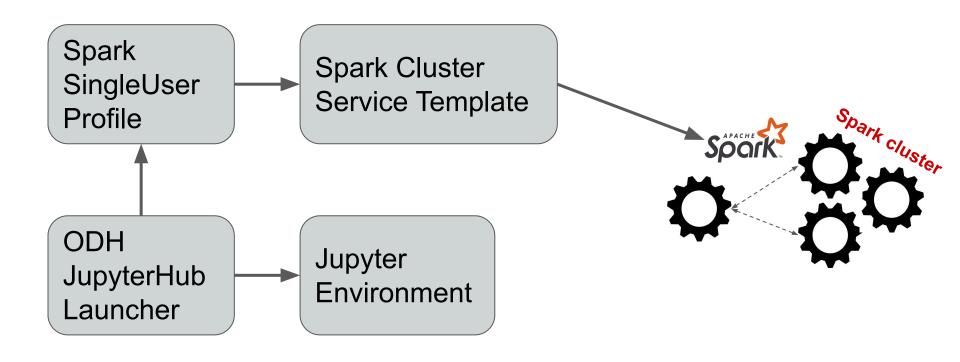
SOSReports, customer feedback, etc.

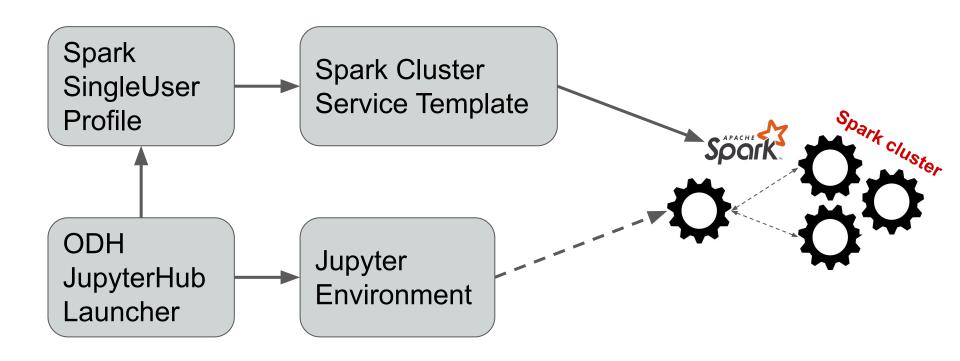
ODH
JupyterHub
Launcher

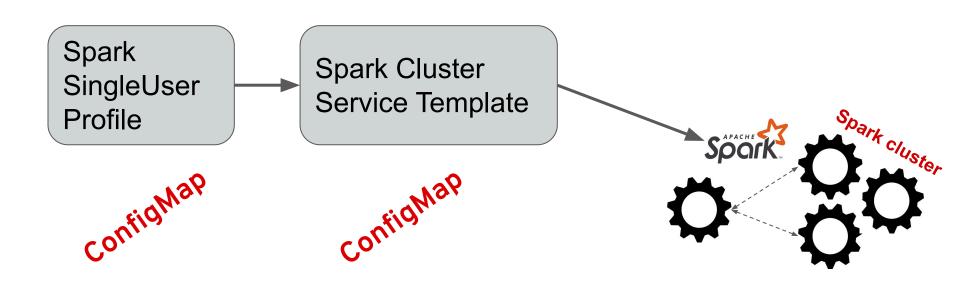




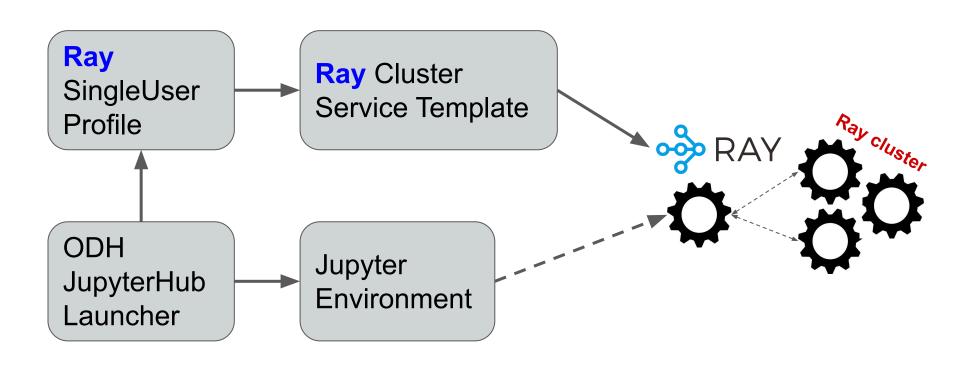








Ray on ODH?



Ray Single User Profile

```
data:
  jupyterhub-singleuser-profiles.yaml: |
    profiles:

    name: Ray ML Notebook

        images:
          'ray-ml-notebook:experimental'
        services:
          ray-cluster:
            resources:

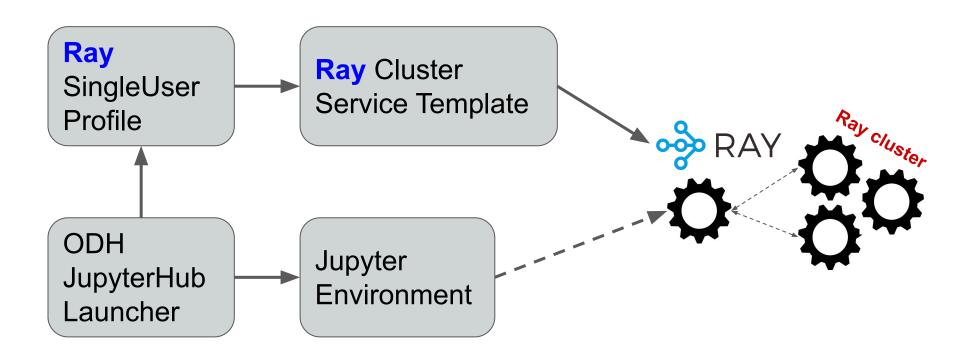
    name: ray-cluster-template

                 path: rayClusterResourceTemplate
            configuration:
               ray_image: 'ray-ml-node:experimental'
              memory_request: '1024Mi'
              cpu_request: '1'
            return:
               RAY_CLUSTER: 'metadata.labels[odh-ray-cluster-service]'
```

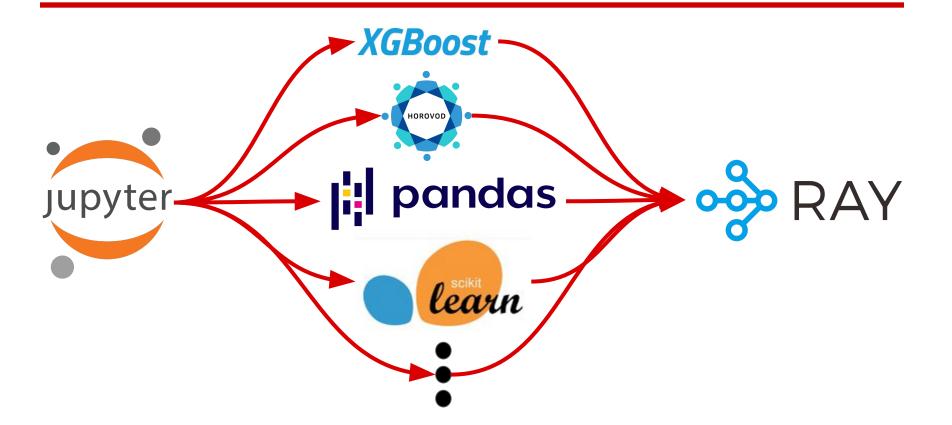
Ray Cluster Service Template

```
data:
  rayClusterResourceTemplate: |
    kind: RayCluster
    apiVersion: cluster.ray.io/vl
    metadata:
      name: 'ray-cluster-{{ user }}'
      labels:
        odh-ray-cluster-service: 'ray-cluster-{{ user }}-ray-head'
    spec:
      maxWorkers: 5
      upscalingSpeed: 1.0
      idleTimeoutMinutes: 5
      # much much more ...
```

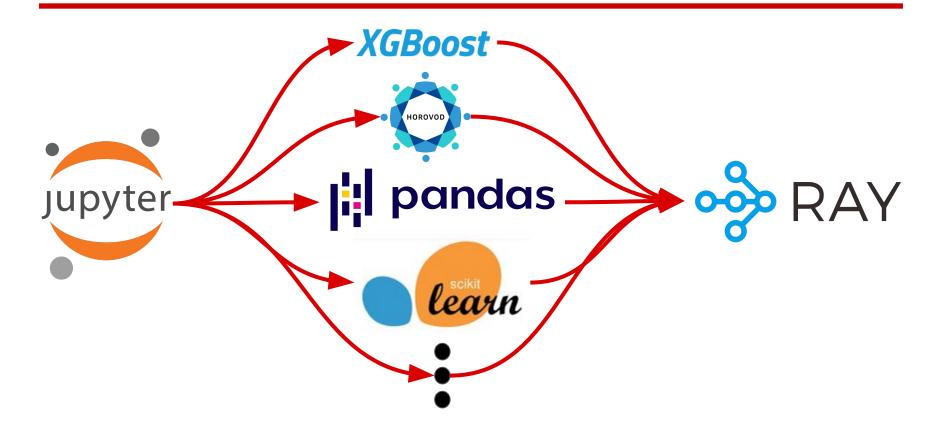
Demo: Ray on ODH!



Unified Scaling for Data Science



Unified Simplified Scaling



Ray on ODH at the Mass-Open Cloud

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Governor Patrick Announces Funding to Launch Massachusetts Open Cloud Project

(b) Mon, 04/28/2014 - 12:07pm

by Mass Open Cloud Project

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Led by Boston University, the MOC is a collaborative effort among BU, Harvard, UMass Amherst, MIT, and Northeastern University, as well as the Massachusetts Green High-Performance Computing Center (MGHPCC) and Oak Ridge National Laboratory (ORNL).

It is supported by a broad alliance of industry partners, including Red Hat.

Ray on MOC

- Maximum 5 workers + 1 head
- 1 CPU, 3 GB memory
- Pre-installed: tensorflow==2.4.1

```
horovod[ray]==0.21.3
keras==2.4.3
scikit-learn==0.24.1
dask==2021.2.0
pandas==1.1.5
```

Operate First

Developing Software In The Open



Operating
Software and
Services In
the Open



https://www.operate-first.cloud/

Operate First PRs for Ray

The components being installed are:

- the ray notebook imagestream (operate-first/apps#302)
- the ray operator CRD (operate-first/apps#321)
- the ray operator pod itself (operate-first/apps#336).
- a singleuser profile for ray ML notebook (operate-first/apps#361)

Other PRs that were part of this integration:

- operate-first/apps#346
- operate-first/apps#350

https://github.com/operate-first/support/issues/102

Roadmap

- Community Ray Operator in Catalog
- Maintain Ray Images via Project Thoth
- Community Use Cases With Jupyter
- Formal Integration With KF and ODH
- KF Pipeline Nodes Backed by Ray

Call To Action

- Play with Ray on Jupyter up on MOC
- File issues and PRs with op-1st
- Report Back! eje@redhat.com

https://www.operate-first.cloud/users/moc-ray-demo/README.md

https://odh.operate-first.cloud/