



Fluid

Elastic Dataset for Kubernetes



- **Why Fluid**
- **What's Fluid**
- **Fluid Architecture**
- **Roadmap**

Data-intensive Applications Are Trending on Kubernetes

“By 2023, 70% of AI workloads will use Application container technology like Kubernetes” predicted by Gartner [1]

“Google Replaces YARN with Kubernetes to Schedule Apache Spark” at ApacheCon North America[2]



[1] <https://www.gartner.com/en/conferences/emea/data-analytics-switzerland/featured-topics/topic-ai-machine-learning>

[2] <https://thenewstack.io/big-data-google-replaces-yarn-with-kubernetes-to-schedule-apache-spark/>

Journey of A Data-driven Company



Batch processing only

MapReduce
(distributed computing)

Hadoop

HDFS
(Data locality storage)

tightly coupled storage

2010



many more computes

TensorFlow **Spark** Flink

Disaggregated

hadoop HDFS ceph


diverse data sources

2015



Containerized, elastic env

TensorFlow **Spark** Flink



hadoop HDFS GCP S3 OSS

hybrid environment

2020

Technical Challenges Blocking Faster Adoption



Heterogeneous Data Source



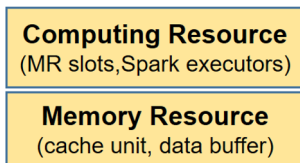
Complexity

Disaggregated Compute & Storage



I/O Bottleneck

Locality-ignoring Scheduling



Static Resource Allocation

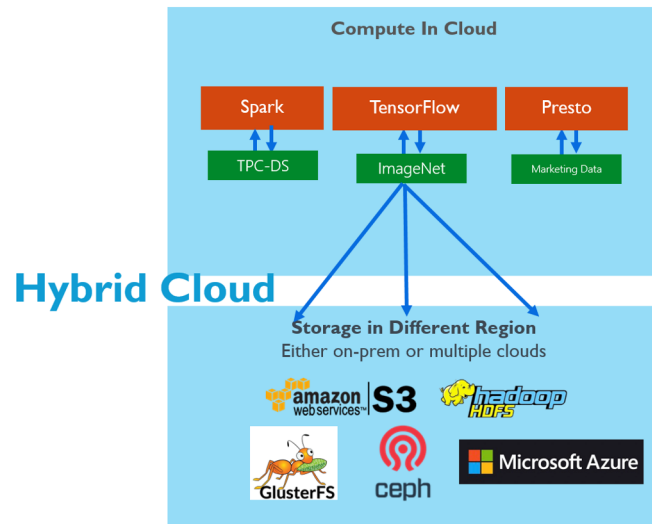


Inefficiency

1st Challenge: Complicated Data Storage



- **Heterogeneous storage:** HDFS, Ceph, S3 ...
- **Various data semantics:** files, objects, ...
- **Low-level data access APIs:** DFS, S3, POSIX, and complicated settings.

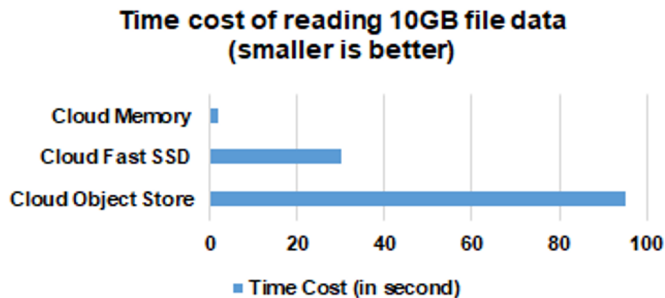


Fluid introduces “DataSet”, a high-level abstraction to applications, hiding details of heterogeneous data sources

2nd Challenge: I/O Bottleneck



- **Disaggregated compute and storage:** more I/O overhead
- **Elastic infrastructure** leads to more **remote data access**
- **Concurrent data applications competing for limited data access bandwidth**



Fluid accelerates data access with distributed cache runtimes

3rd Challenge: Locality-ignoring Scheduling



- **Workload has no aware on scheduler:**

move data or move compute?

- **No intelligent data warmup:**

ETL is costly & error-prone

- **No data/app affinity-aware scheduling**

for apps like Spark, Tensorflow

Computing Resource

(MR slots, Spark executors)

Memory Resource

(cache unit, data buffer)

Static Resource Allocation

Fluid builds data-aware scheduling strategies on Kubernetes to coordinate applications with cache runtime automatically

An Elastic Data Abstraction and Acceleration Platform in Cloud Native Environment.

Data Abstraction

for heterogeneous
data sources

Data Acceleration

with autoscaling and
portable cache runtimes

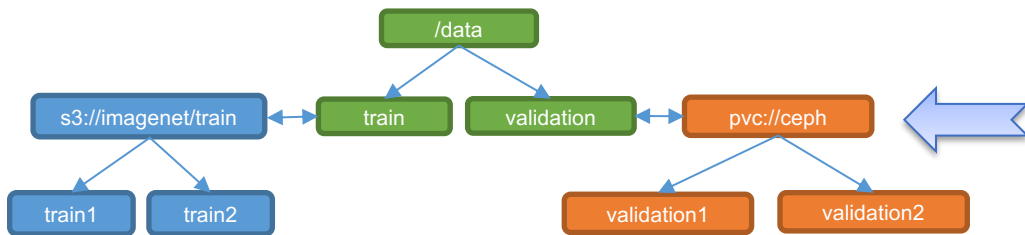
Data-aware Scheduling

to improve data affinity
for apps intelligently

A Quick Start on Fluid: (1) Create a “Dataset”



Edit “**dataset.yaml**” and run
“**kubectl apply -f dataset.yaml**”



```
apiVersion: data.fluid.io/v1alpha1
kind: Dataset
metadata:
  name: imagenet
spec:
  mounts:
    - mountPoint: s3://imagenet/train
      name: train
    - mountPoint: pvc://ceph
      name: validation
  nodeAffinity:
    required:
      - nodeSelectorTerms:
          - matchExpressions:
              - key: GPU
                operator: In
                values:
                  - "true"
```

s3

ceph

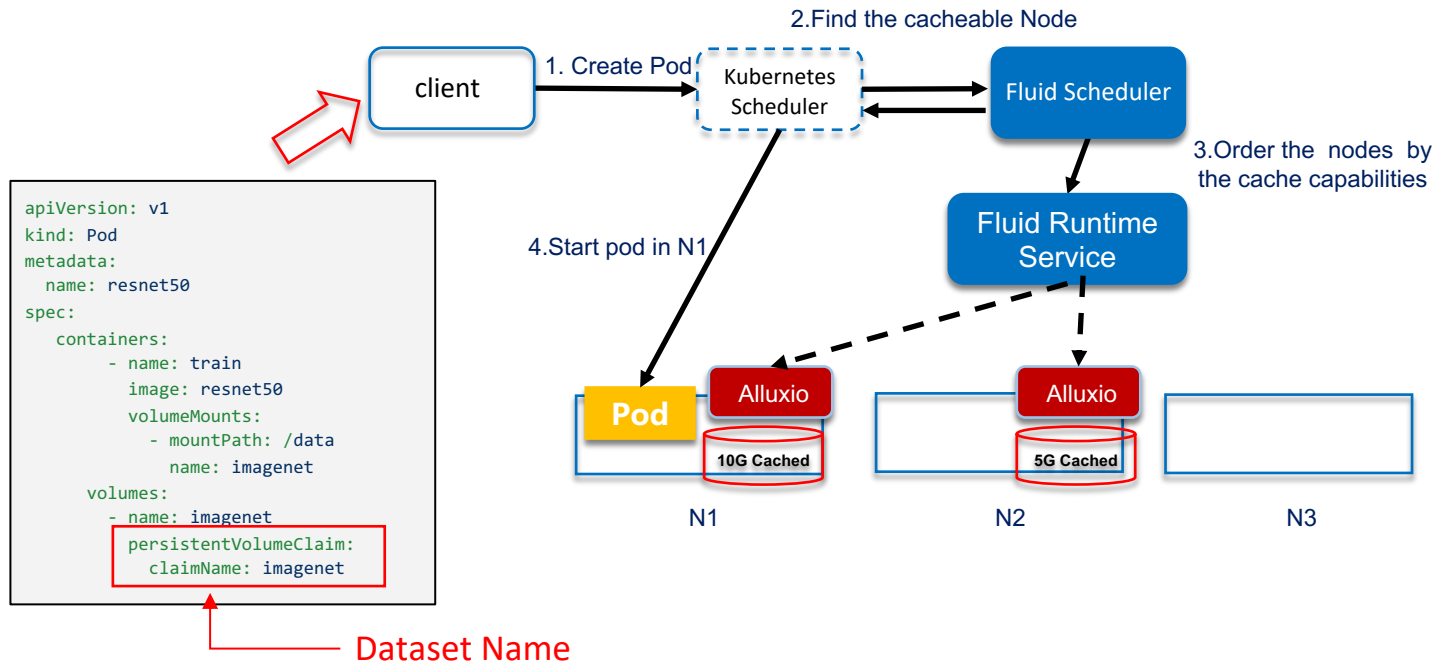
In GPU nodes

```
[root@iZuf6afslw5j6hauyzkw21Z dataset]# kubectl get dataset
NAME          UFS TOTAL SIZE  CACHED  CACHE CAPACITY  CACHED PERCENTAGE  PHASE  AGE
imagenet     183.08GiB      0.00B   100.00GiB      0.0%              Bound  23m
```

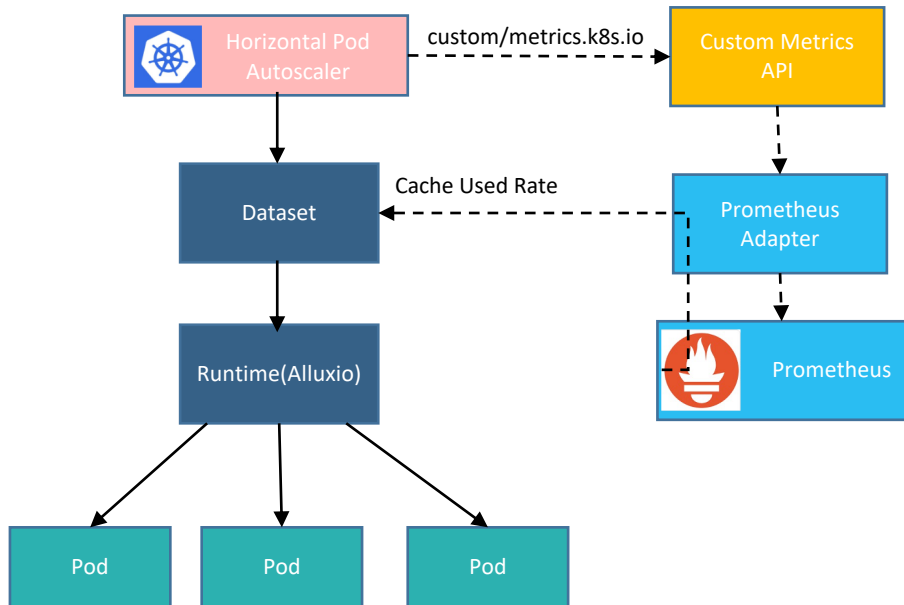
A Quick Start on Fluid: (2) Run Apps with Data Affinity



Create “**pod.yaml**” and run “**kubectl create -f pod.yaml**”

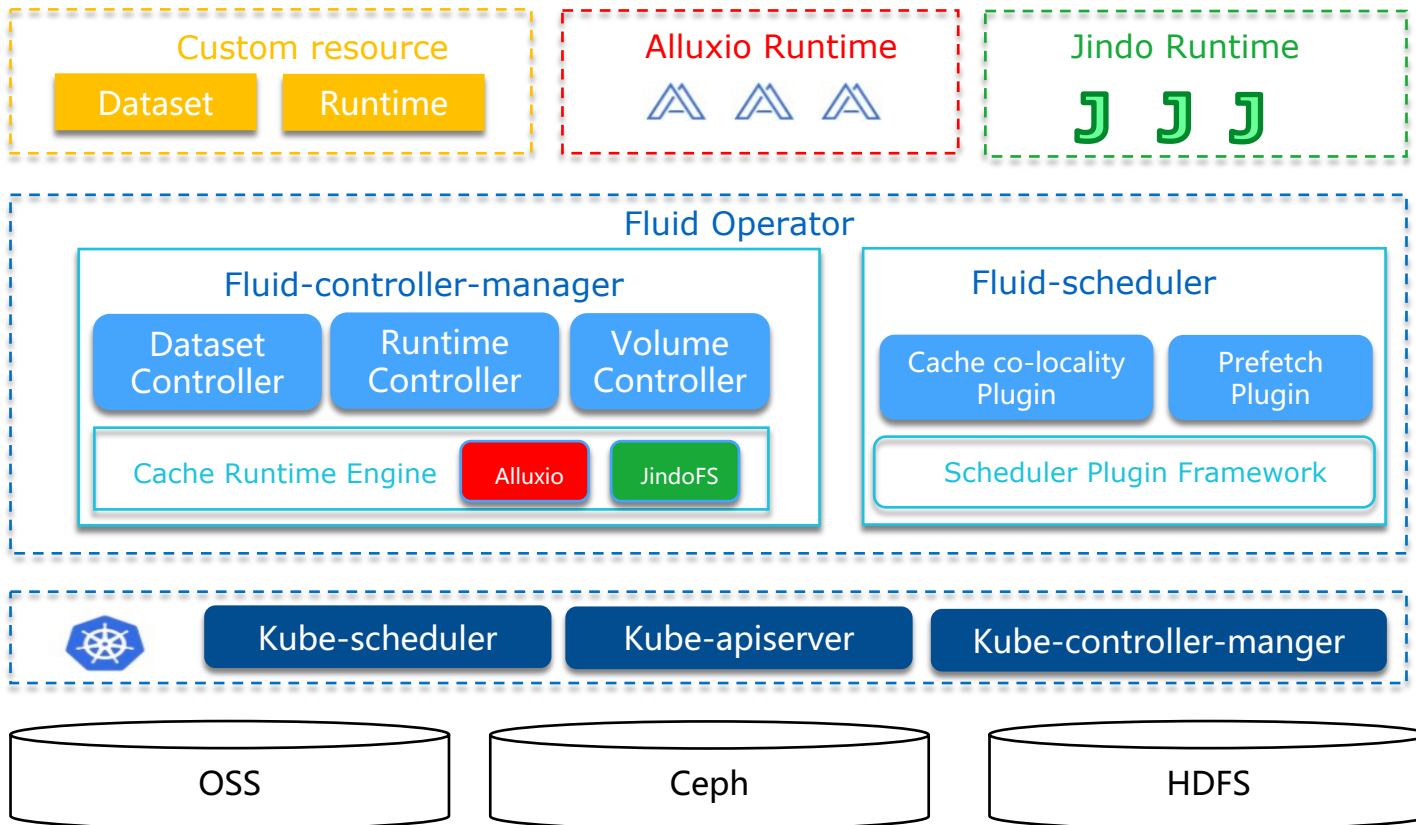


Quick Start on Fluid: Auto-scale Dataset

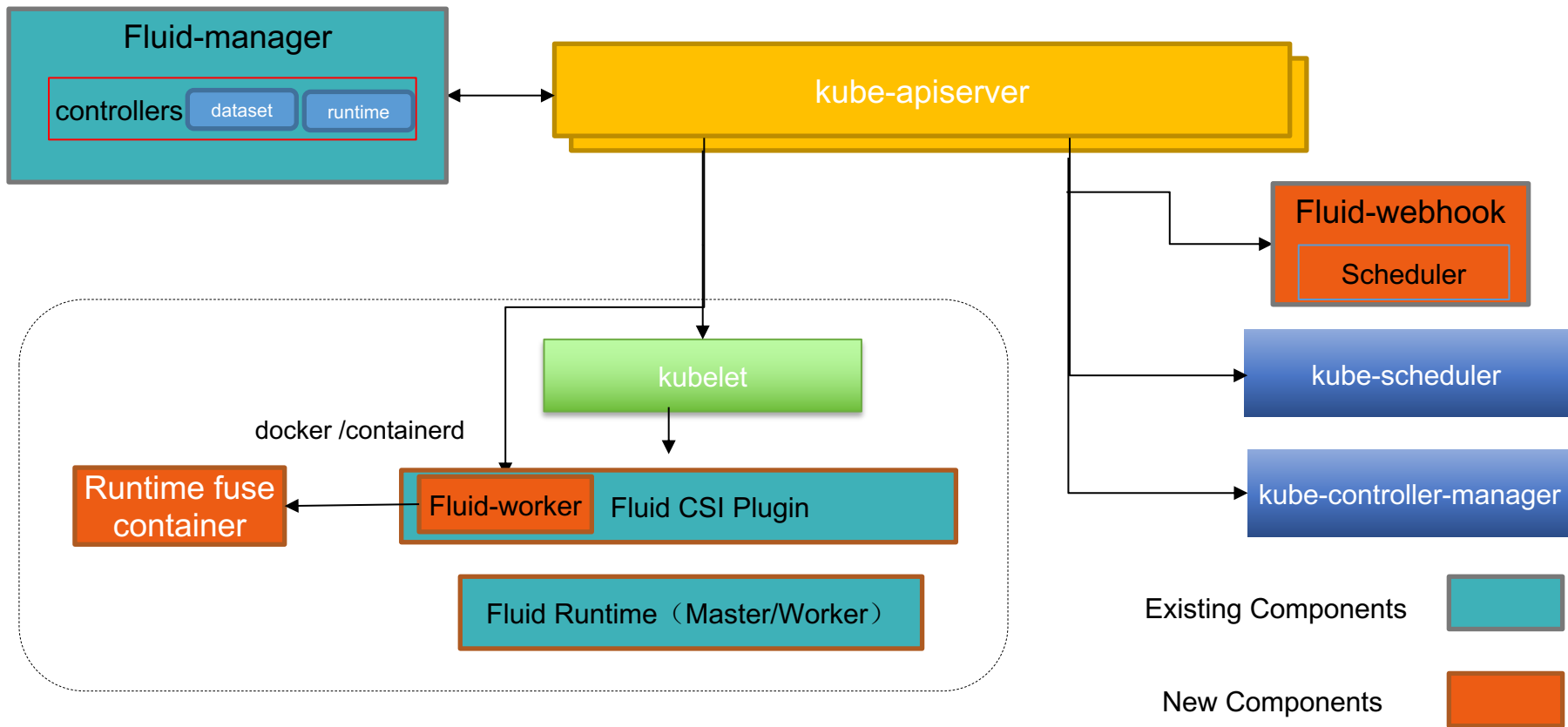


```
apiVersion: autoscaling/v2beta2
kind: HorizontalPodAutoscaler
metadata:
  name: imagenet-hpa
spec:
  scaleTargetRef:
    apiVersion: data.fluid.io/v1alpha1
    kind: AlluxioRuntime
    name: imagenet
  minReplicas: 2
  maxReplicas: 10
  metrics:
    - type: Object
      object:
        metric:
          name: capacity_used_rate
        describedObject:
          apiVersion: data.fluid.io/v1alpha1
          kind: Dataset
          name: imagenet
      target:
        type: Value
        value: "75"
```

Fluid Architecture



Fluid Architecture(New)



- **Deep learning model training on K8s with high computing bandwidth**
 - The same training dataset is used repeatedly
 - Accessing training data is the bottleneck
- **Unified data processing pipeline with analytic engines and AI systems**
 - Saving data to external storage leads to huge I/O cost
 - Applying cross-task optimization (pipelining) on tasks is challenging
- **Elastic distributed deep learning (e.g. computing resources demand varies)**
 - Data cache capacity requirement varies during app run time

- **Manage and operate datasets in large scale K8s Cluster to improve the performance of accessing of data**
 - **Long time** Dataset for **regular usage**
 - **Multiple** datasets co-exist

- **Feedbacks**
 - **Scheduling for large-scale cluster**
 - **Self-maintained**
 - **Big data**
 - **Presto/Spark**

Fluid 0.5

- **Accelerate DL on Kubernetes in a simple way**
- **Scale out/in Dataset**
- **A small set of datasets**

Use case

As a data infrastructure engineer, I want to be able to setup dataset cache for AI immediately that can be easily and efficiently accessed.

Fluid 1.0

- **Enable big data on Kubernetes**
- **Comprehensive schedule capability for data and workload**
- **Dataset lifecycle management in large scale of Kubernetes**

Use case

As a data infrastructure engineer, I want to run both AI and Big Data in the same K8s with data acceleration. And I don't want to take care of the data management.

