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The OpenDeepHealth toolkit

Iacopo Colonnelli (iacopo.colonnelli@unito.it)

2022 DeepHealth Winter School
Wednesday, January 26th



The project has received funding from the European Union's Horizon 2020 research and innovation programme under grant agreement No 825111.



OpenDeepHealth Platform



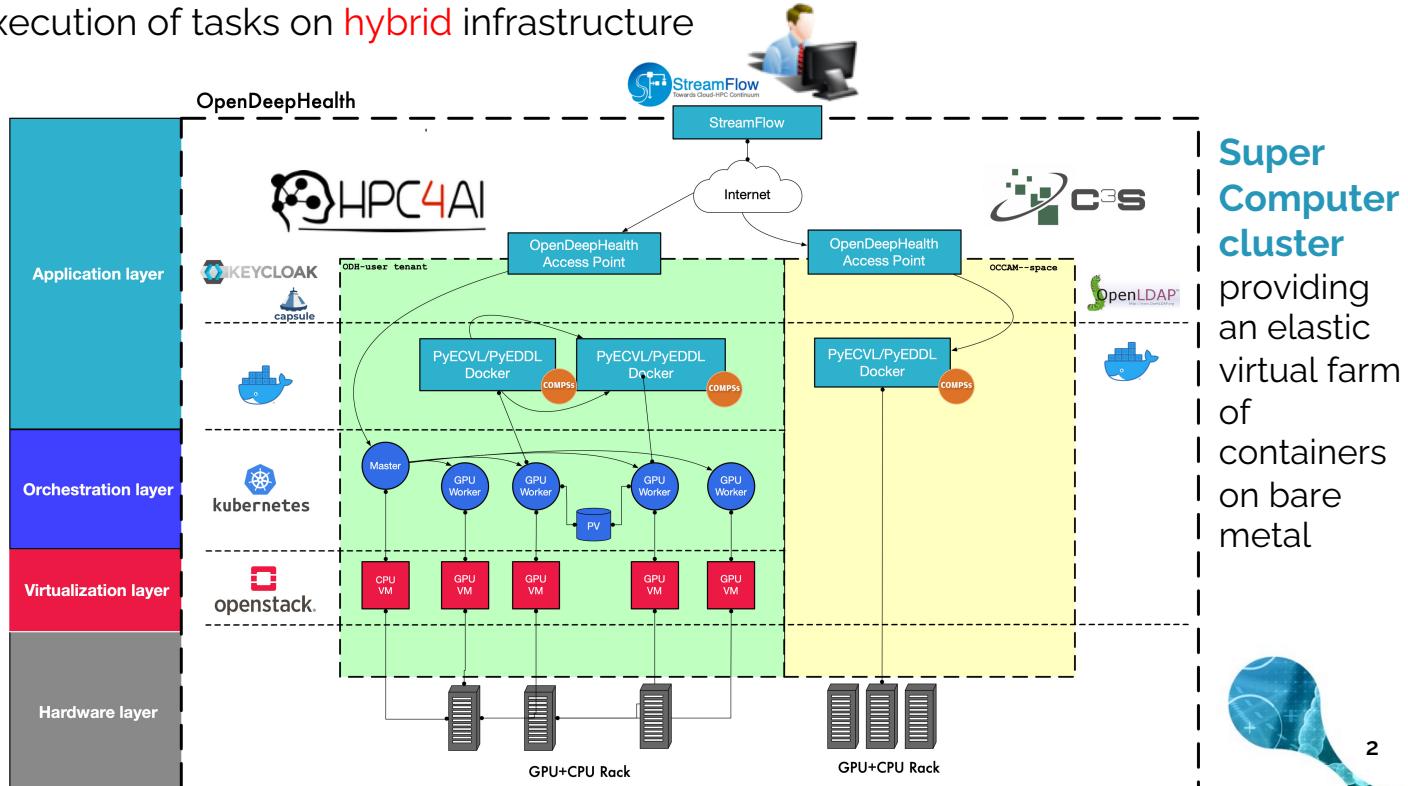
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The **StreamFlow** Workflow Management System (WMS) orchestrates the execution of tasks on **hybrid** infrastructure

Multi-tenant
private Kubernetes
for AI training

Literate workflows to
write complex distributed
pipelines as Jupyter
Notebooks





OpenDeepHealth Platform



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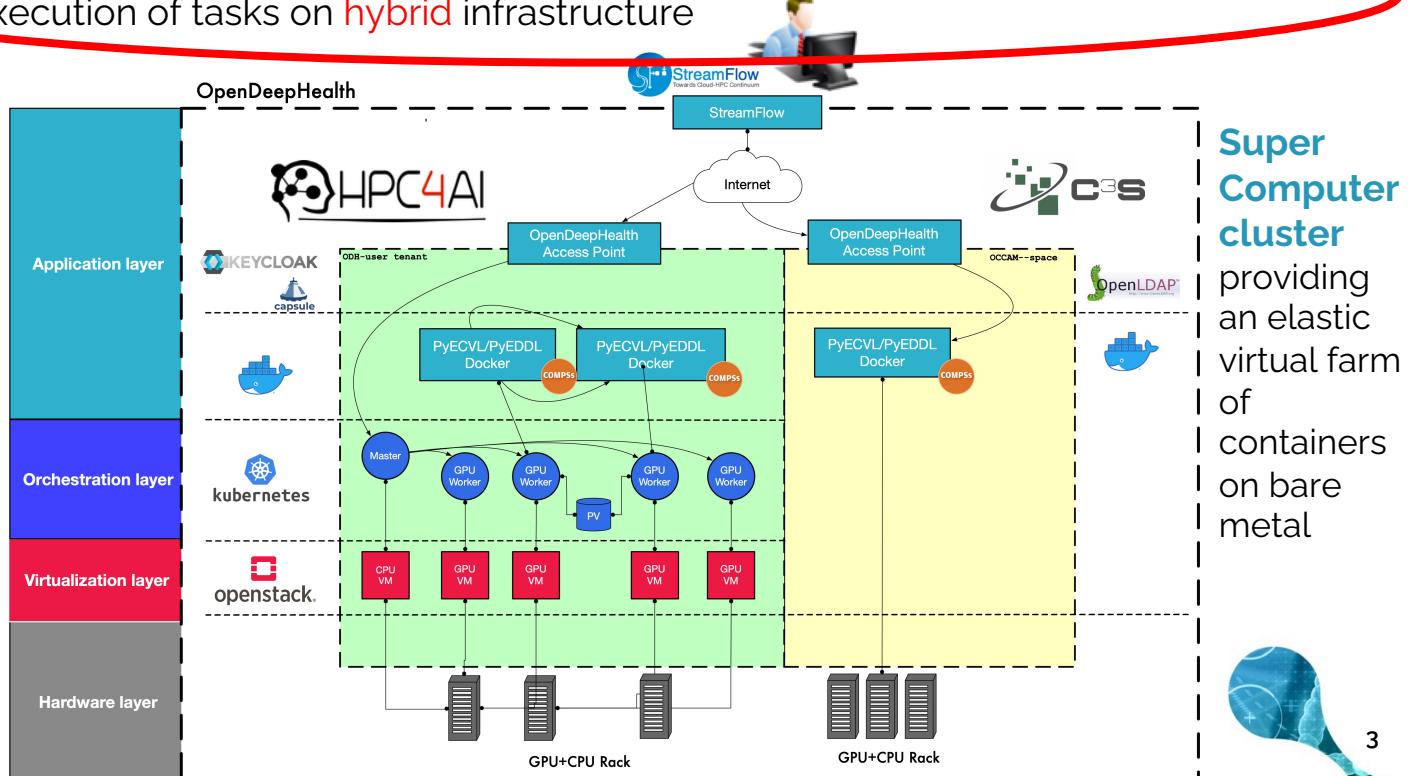


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The StreamFlow WMS

Designed around two main principles:

- Allowing the execution of tasks in **multi-container environments**, in order to support concurrent execution of multiple communicating tasks in a multi-agent ecosystem





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The StreamFlow WMS

Designed around two main principles:

- Allowing the execution of tasks in **multi-container environments**, in order to support concurrent execution of multiple communicating tasks in a multi-agent ecosystem
- Relaxing the requirement of a single shared data space, in order to allow for **hybrid workflow** executions on top of multi-cloud or hybrid cloud/HPC infrastructures.





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Hybrid workflows

A **hybrid workflow** is a workflow whose steps can span **multiple**, **heterogeneous**, and **independent** computing infrastructures.





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- Each location can expose **different methods and protocols** for authentication, communication, resource allocation and job execution;
- Direct communications and data transfers among different locations **may not be allowed**.





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A **hybrid workflow** is a workflow whose steps can span **multiple**, **heterogeneous**, and **independent** computing infrastructures.

PROS:

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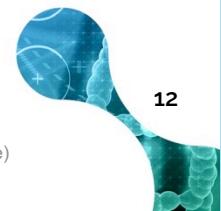


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- A single model describes both the workflow and the execution architecture: **improve portability and reproducibility**;
- Knowing the locations topology can lead to better scheduling algorithms, privileging **data locality** and minimising the **data transfer overhead**;
- Delegating data transfers and computation offloading to the Workflow Management System (WMS) brings **additional features OOTB** (e.g. security, fault-tolerance, cross-stack, urgency-based scheduling).

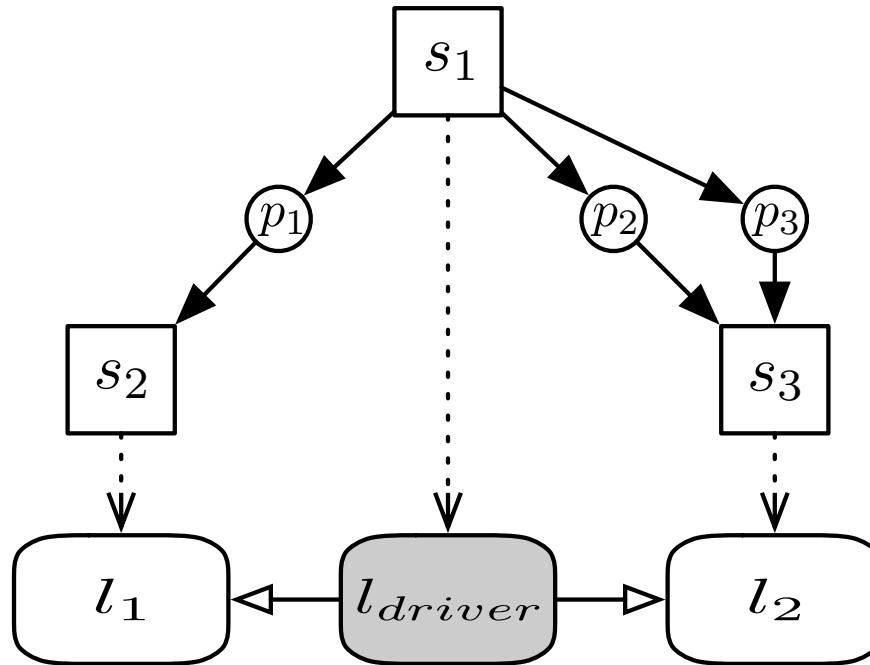




Hybrid workflows



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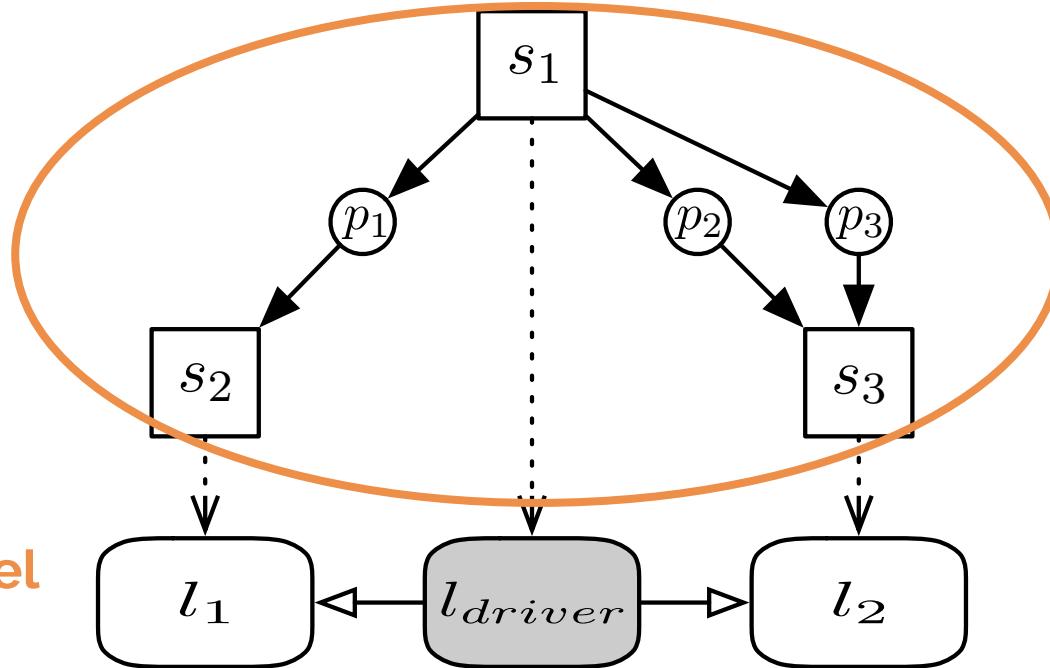




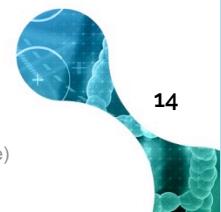
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Hybrid workflows



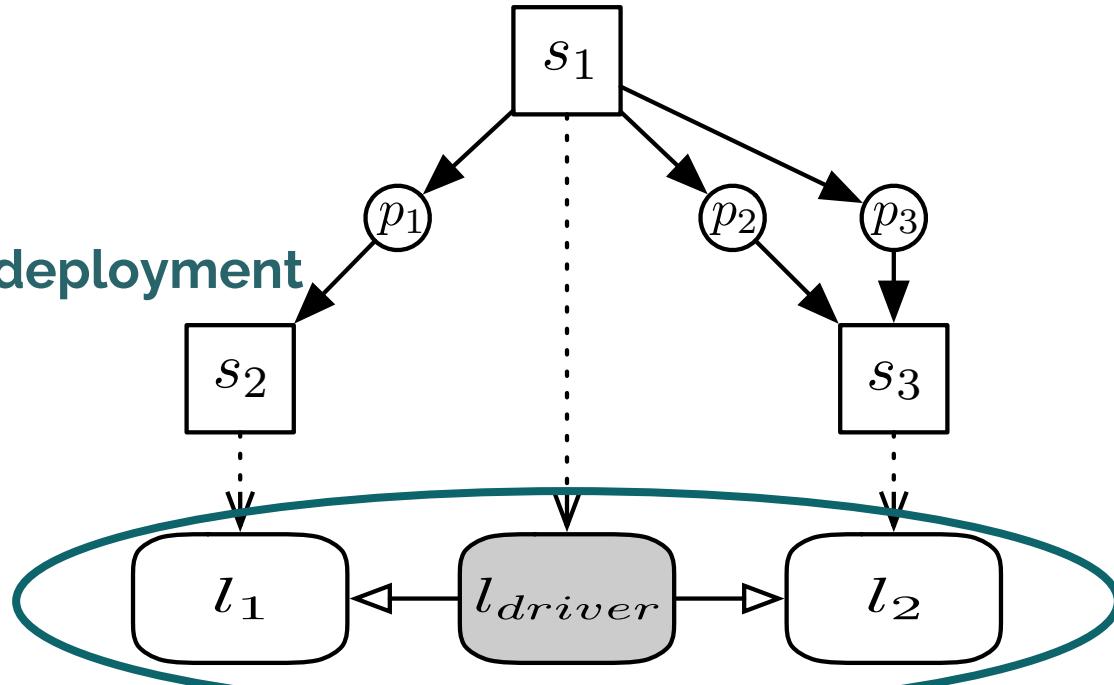
Workflow model





Hybrid workflows

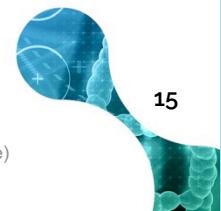
Topology of deployment locations



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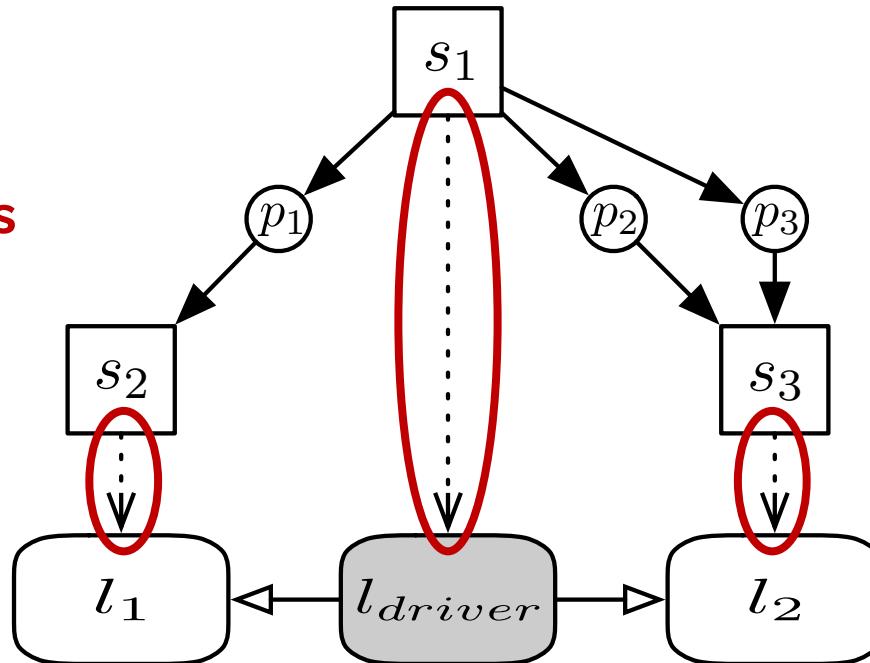


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Hybrid workflows

Mapping relations

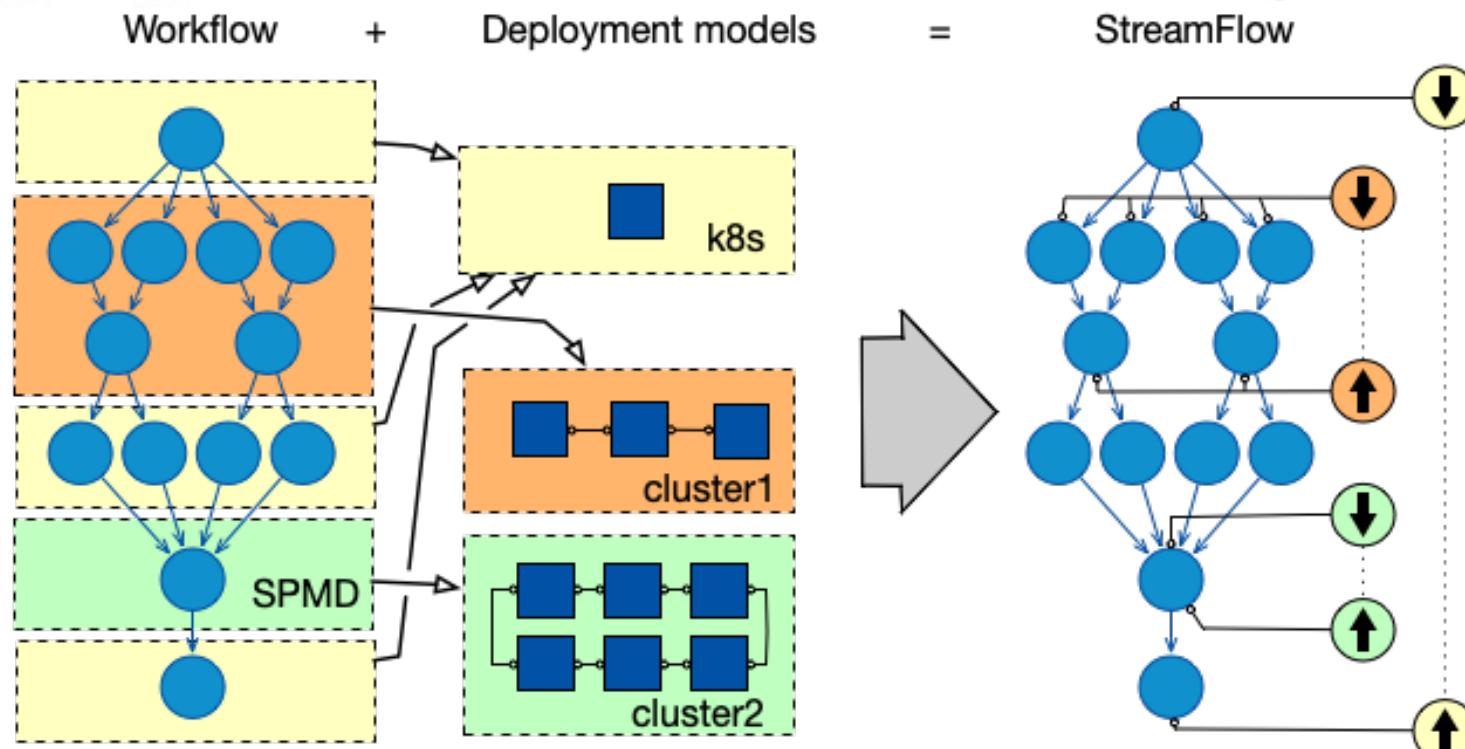




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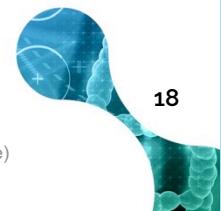
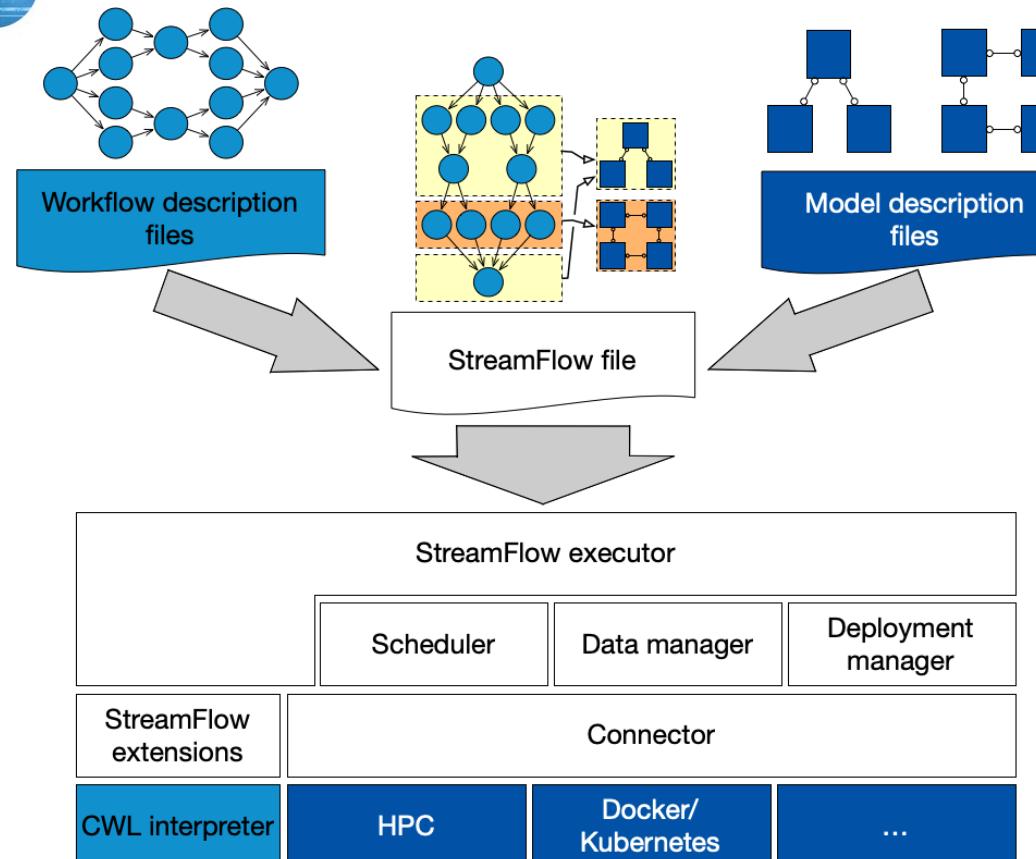




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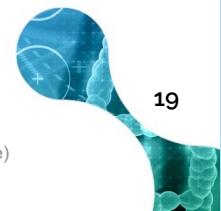
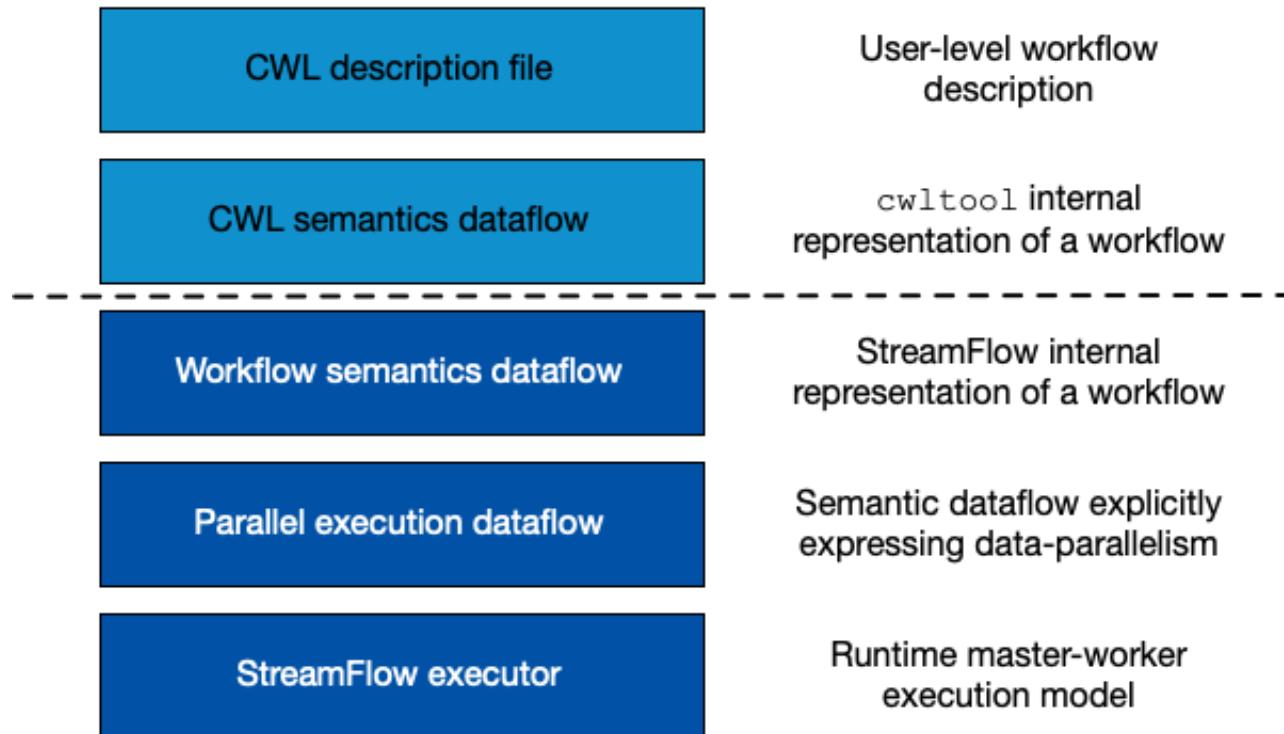


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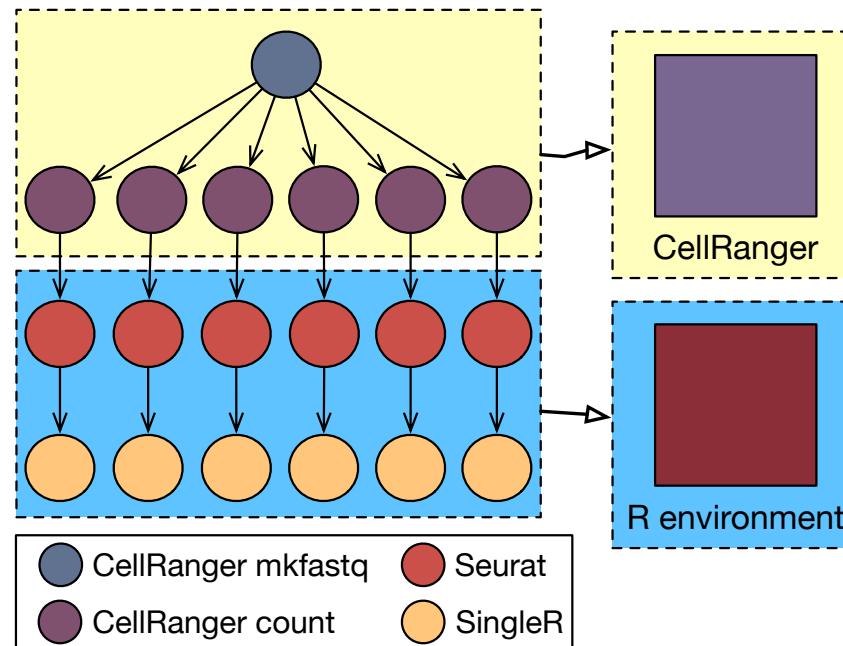




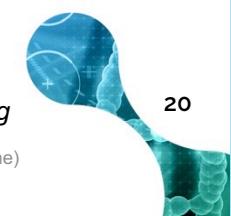
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Single-Cell RNA Sequencing Pipeline



I. Colonnelli, B. Cantalupo, I. Merelli and M. Aldinucci, "StreamFlow: cross-breeding cloud with HPC," in *IEEE Transactions on Emerging Topics in Computing*, doi: [10.1109/TETC.2020.3019202](https://doi.org/10.1109/TETC.2020.3019202).

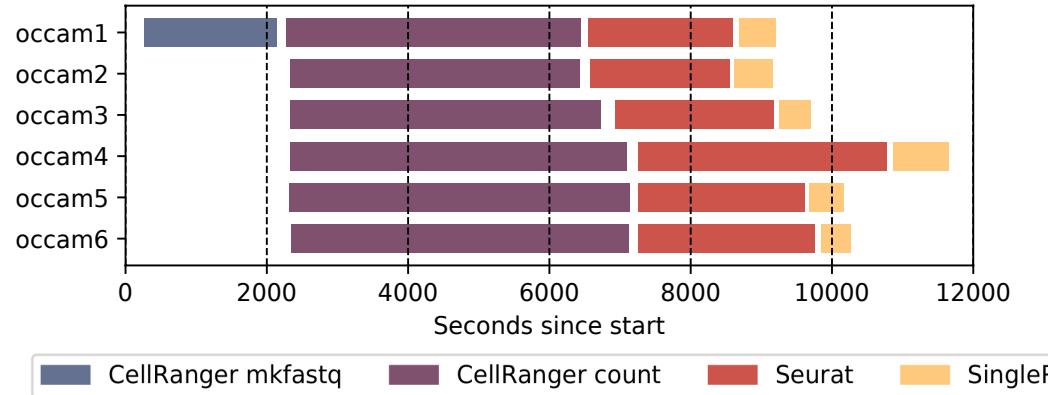




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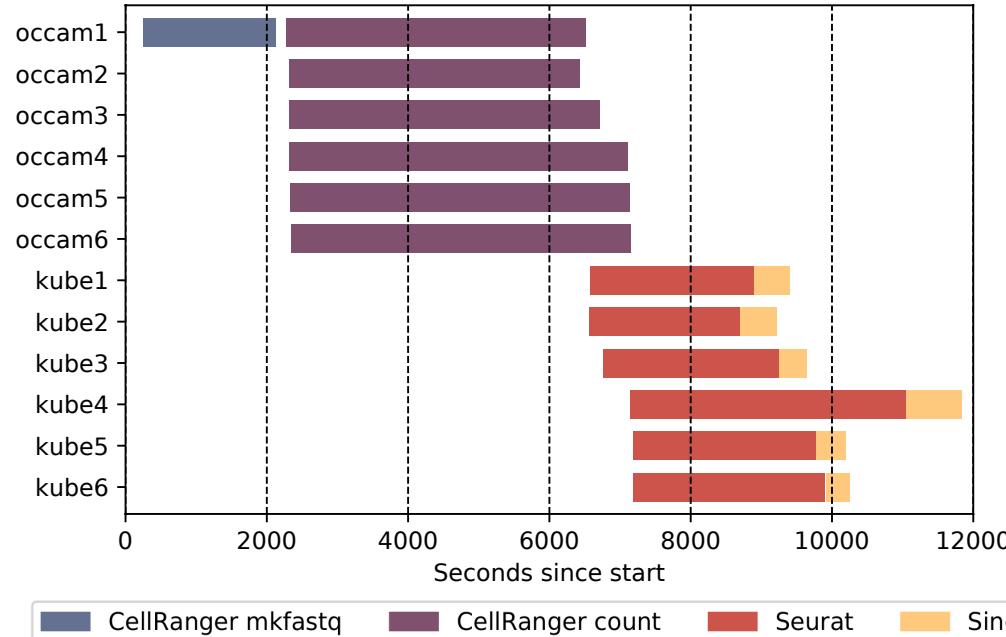
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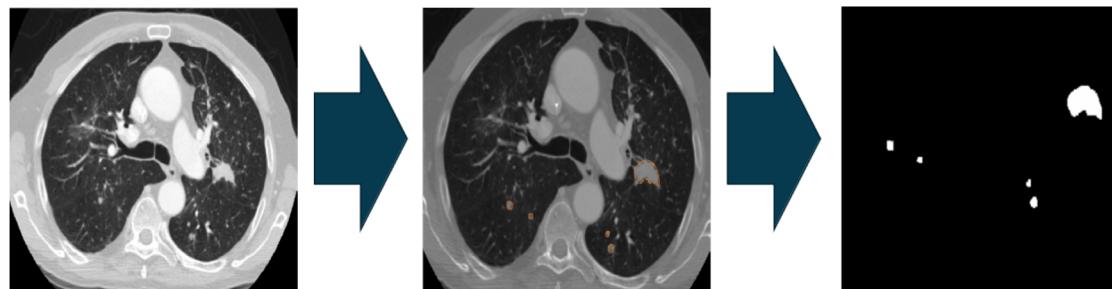


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Lung Nodules Segmentation

- **Lung nodules** are quite common incidental findings in **CT** (Computed Tomography) **scans** and can be defined as **small focal lesions** (ranging from 5 to 30 mm) that can be solitary or multiple.
- The **goal** of DeepHealth UC4 is to train **AI systems to recognize lung nodules** using chest CT scans, providing radiologists an efficient tool for daily activity.



M. Aldinucci et al. "Lung Nodules Segmentation in CT scans by DeepHealth toolkit," in 25th International Conference on Pattern Recognition

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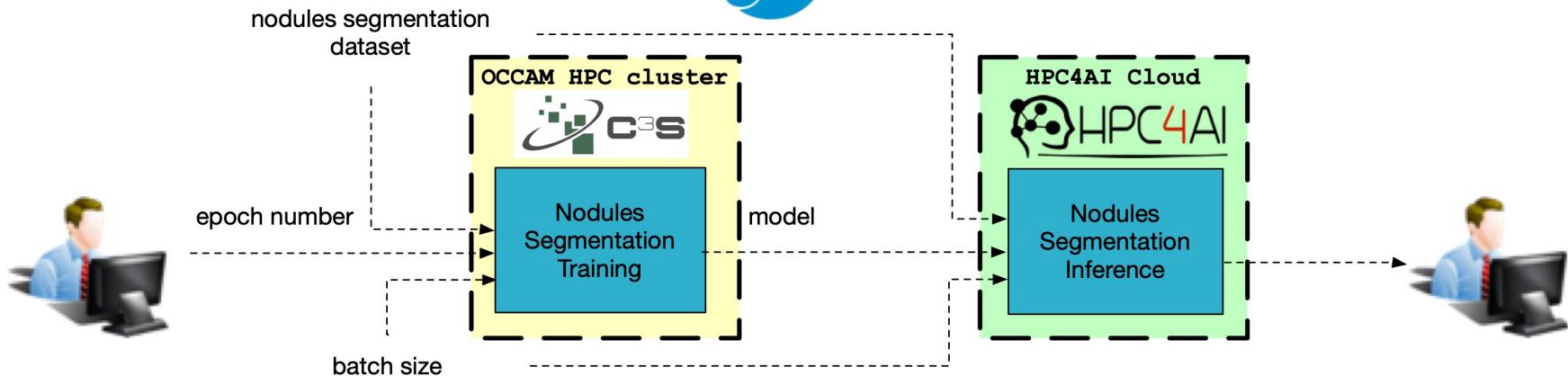




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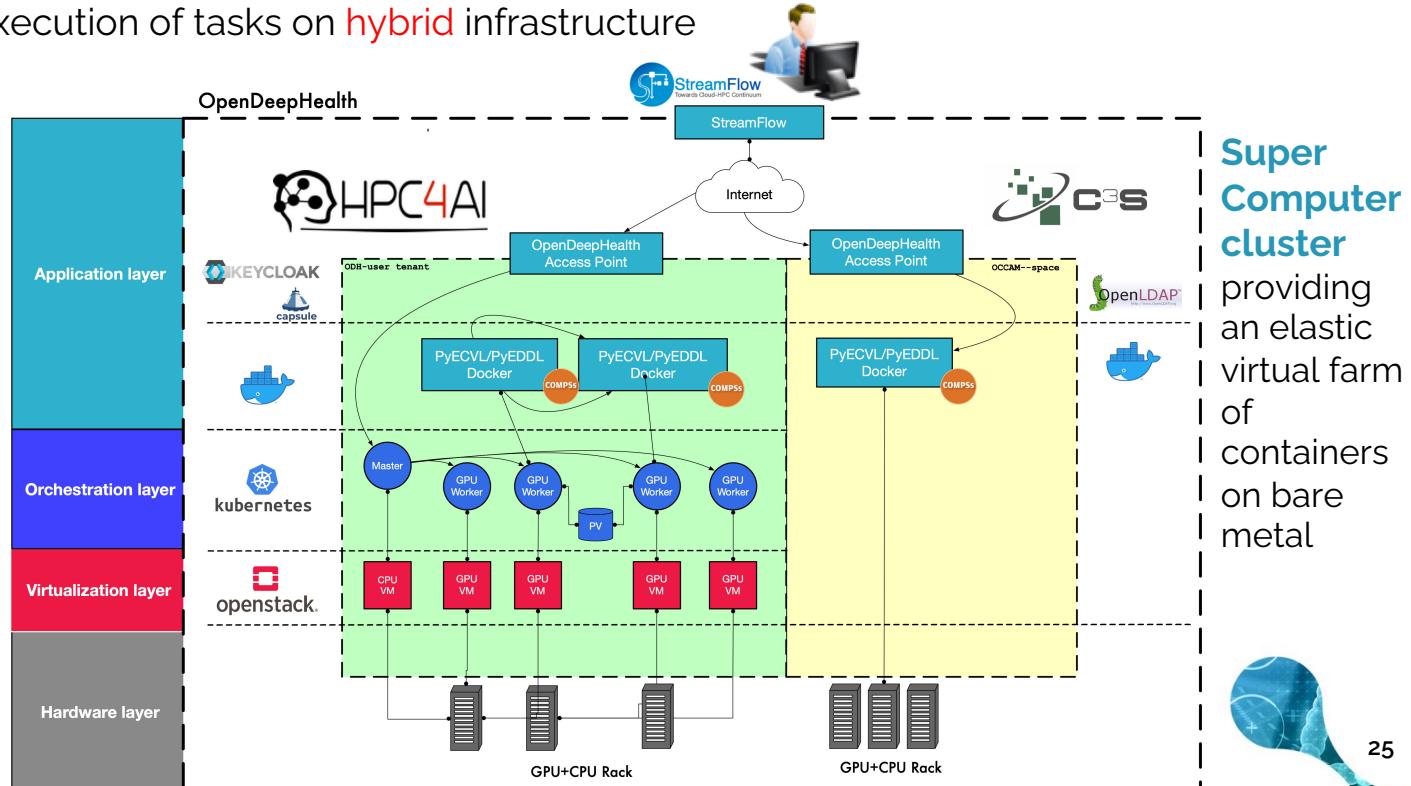
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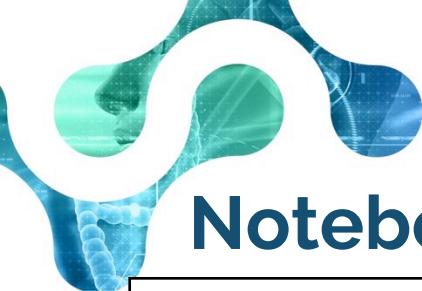
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Notebooks have limits

- Notebooks **are not parallel**: the default execution flow is sequential
- Notebooks **are not distributable**: code is executed in a single backend (kernel)
- Notebooks **are not scalable**: they cannot normally run on HPC environments

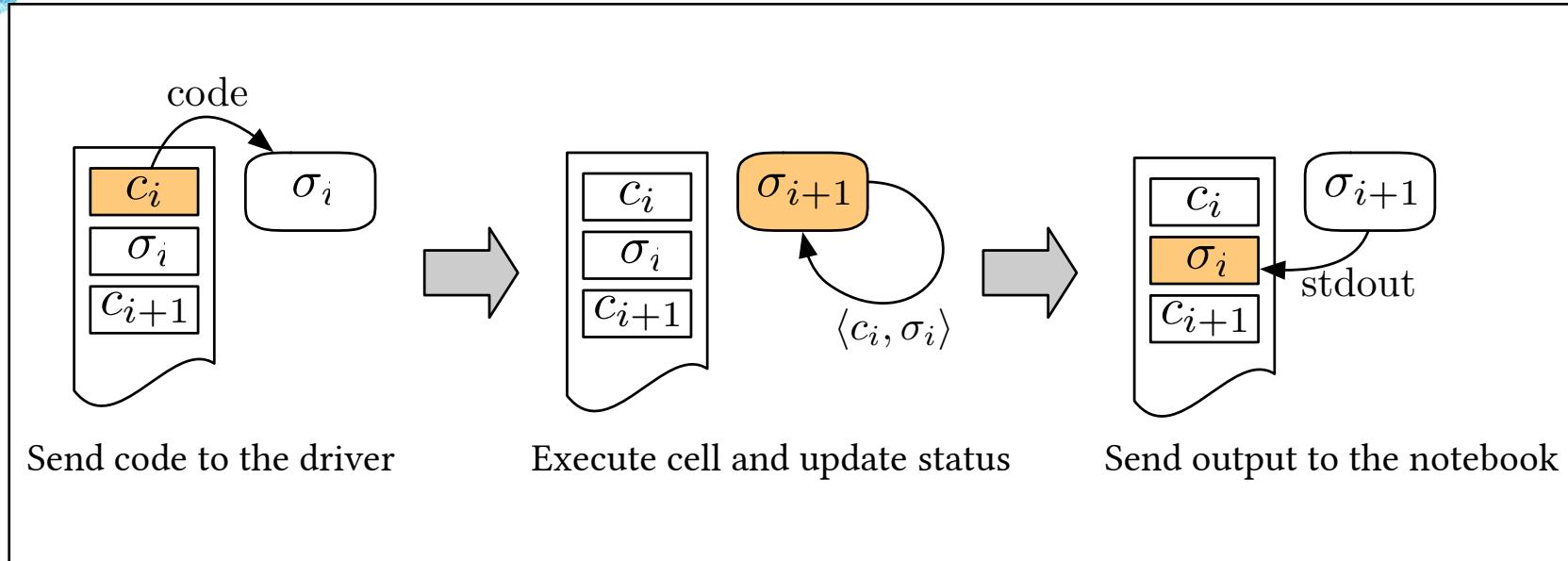




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Solution 1: write scalable code

- Notebooks do not scale, but **user code can scale** (e.g. using parallel libraries as Spark, Tequila, PyCOMPSs, Parsl, Ray, ...)





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- Users are required to **change their code** (more complexity, lot of boilerplate code)
- Deployment of the parallel environment (e.g. a remote Spark cluster) is commonly **out of the scope of the notebook**, i.e. not documented





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Solution 2: write workflows of notebooks

- A single **notebook** represents an atomic workflow step





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- Code (and therefore application logic) must be **fragmented** in multiple notebooks according to the desired grain of parallelism
- Single notebooks are **not self-contained**: they only make sense in combination with the entire workflow



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Solution 3: write workflows as notebooks

- A single **notebook cell** represents an atomic workflow step





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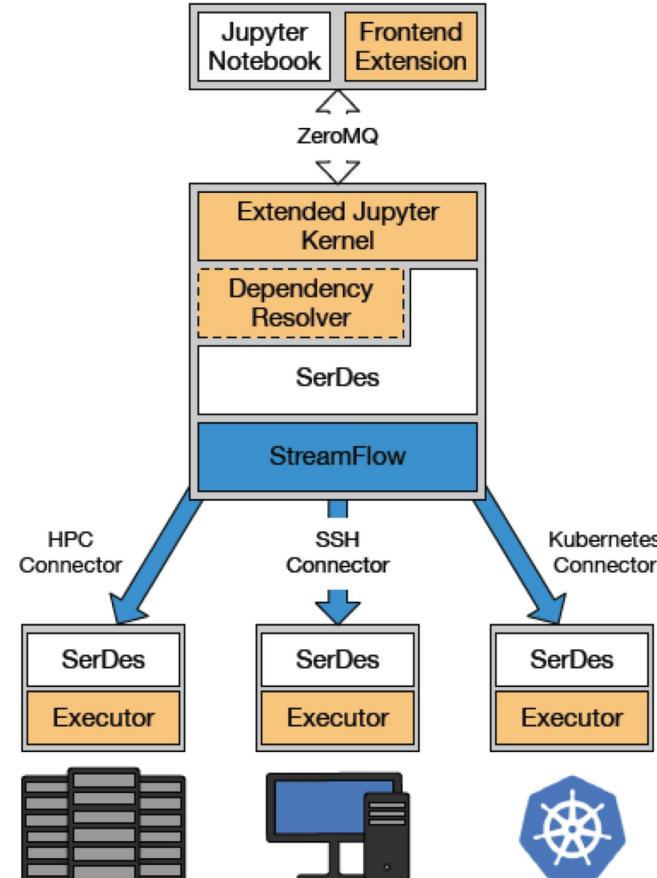
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- Each notebook is **self-contained**: both host and coordination logics are packed in the same format
- Coordination logic is **transparent**: user code can still run sequentially as usual in an environment without workflow metadata interpreters



Jupyter-workflow

The **Jupyter-workflow** kernel extends the IPython software stack to support hybrid literate workflows in the Jupyter stack. It consists of three main components:

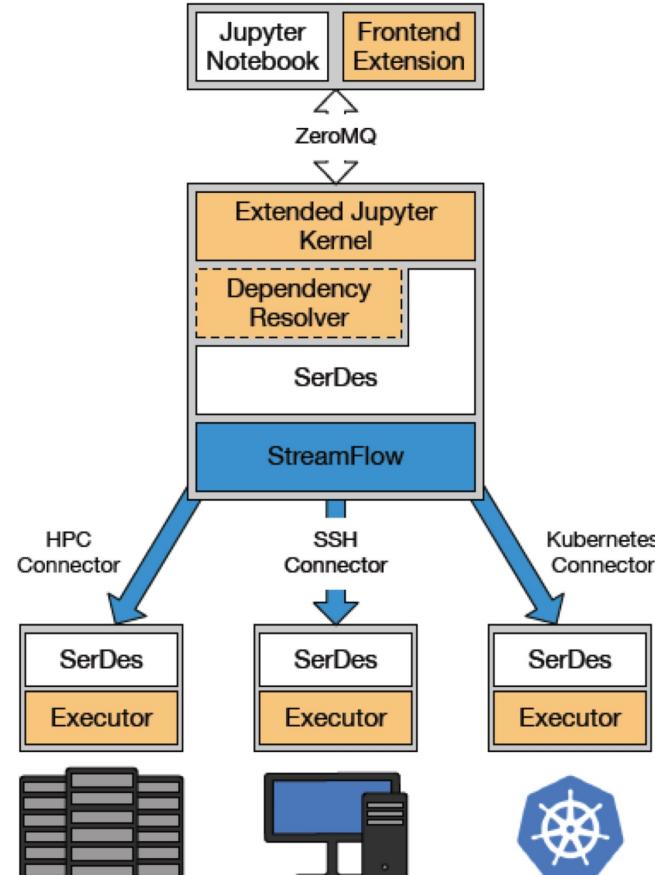




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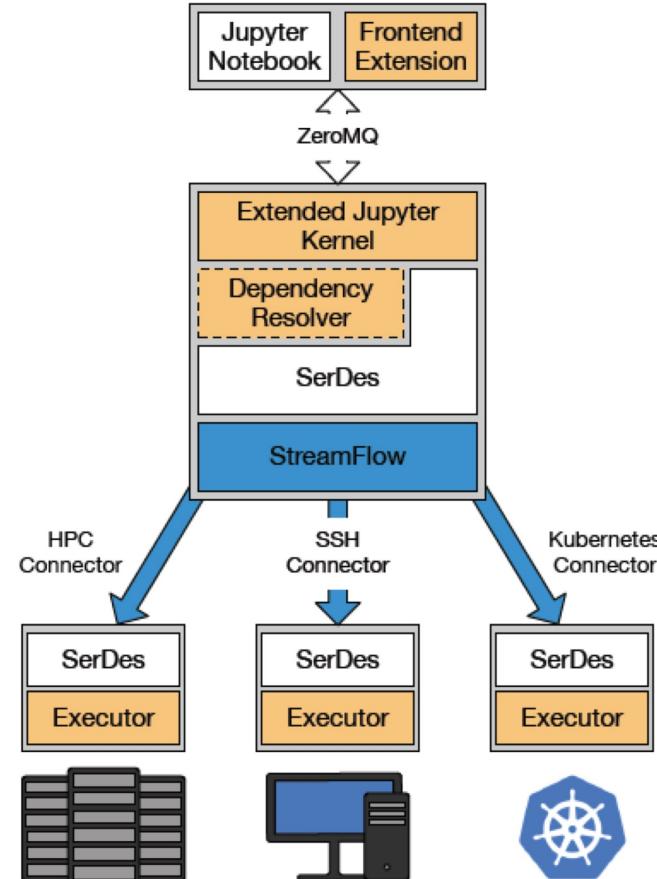




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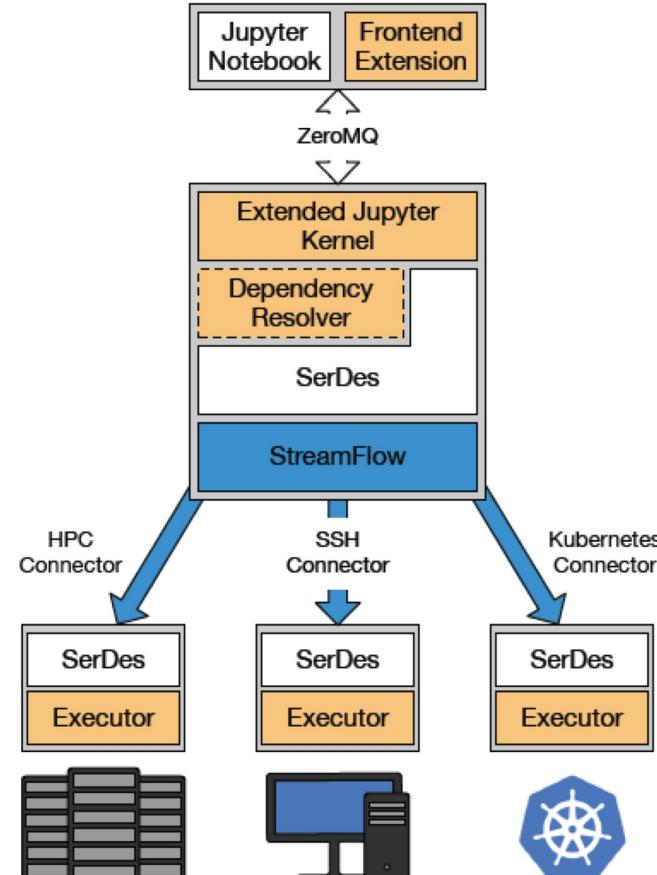


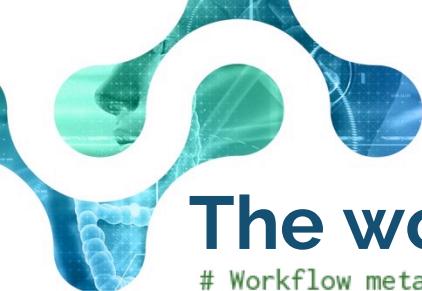


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- A **coordination metadata format** to model global cells configurations and location topologies;
- A **dependency resolver component** to help users identify the input dependencies of each cell;
- A **Jupyter stack extension** to handle coordination metadata, execute cells remotely and manage data transfers (through StreamFlow).





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The workflow metadata format

```
# Workflow metadata
{
    "step": [
        "in": { # List the members of In( $c_i$ )
            "type": "name" | "env" | "file" | "control",
            "name": "variable name",
            "scatter": {
                "size": "token num"
            },
            "serializer": {
                "predump": "code executed before serialising",
                "postload": "code executed after serialising"
            },
            "value": "value to assign to the name",
            "valueFrom": "can take value from a different variable"
        ],
        "autoin": True | False, # Resolve In( $c_i$ ) automatically
        "out": [ # List the members of Out( $c_i$ )
            ...
        ]
    },
}
```

**Input and output
dependencies**





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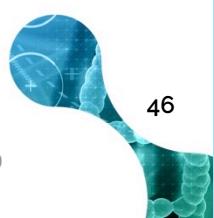


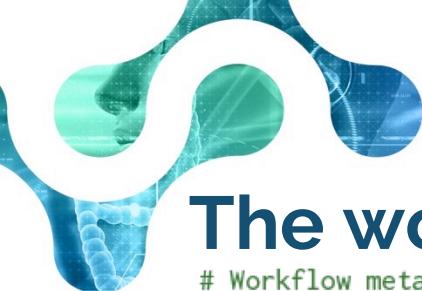
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        "out": [ # List the members of  $Out(c_i)$ 
            ...
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```

Explicit data-parallel
constructs





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Custom object serialization





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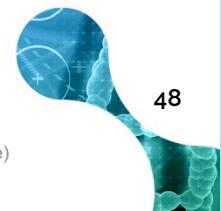


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```

Automatic
dependency
retrieval





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The workflow metadata format

Description of remote environments

```
"target": {# Part of the StreamFlow format
    "model": {# Description of the execution environment
        ...
    },
    "service": "target service inside the model",
    "resources": "number of workers to reserve",
    "workdir": "working directory on the remote resource"
},
"version": "v1.0"
}
```

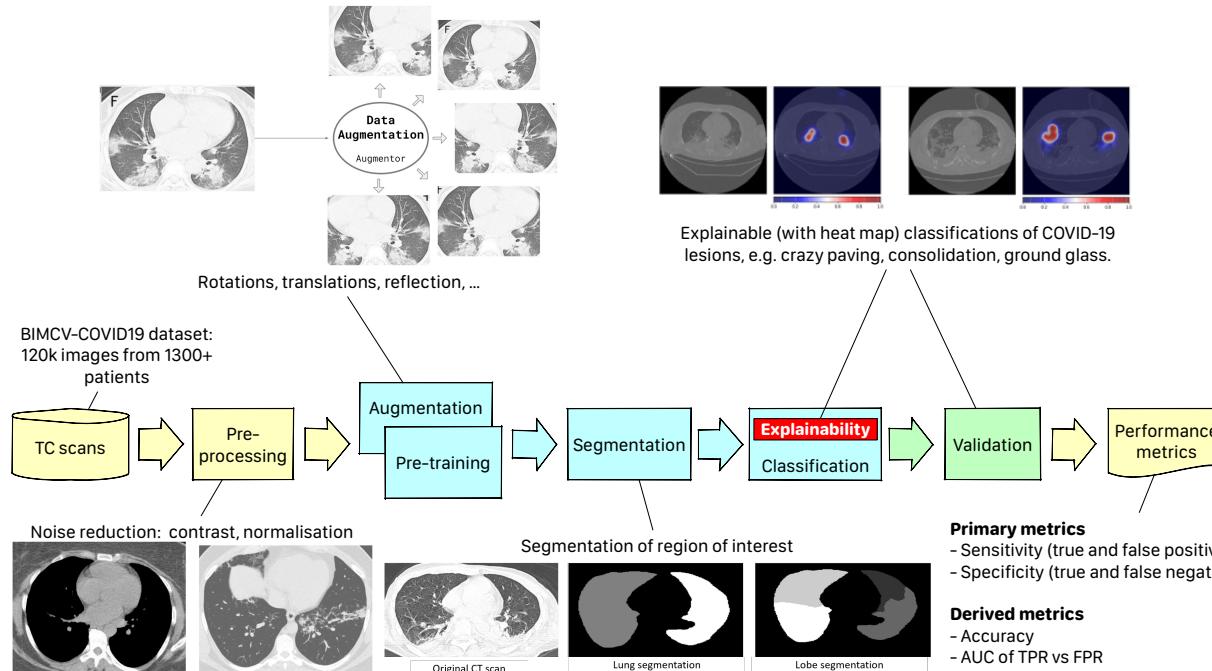




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CLAIRE-COVID19 Universal Pipeline



M. Aldinucci, “[Invited Talk] HPC application cloudification: the streamflow toolkit,” in *PARMA-DITAM 2021*



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CLAIRE-COVID19 Universal Pipeline

PreDNN (2 varianti)

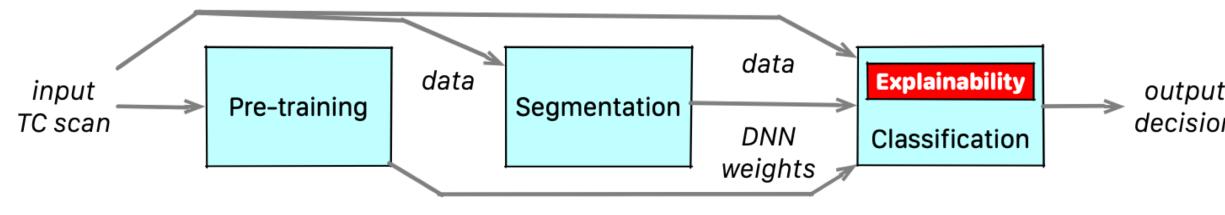
1. Pre-training
2. Self-training

SegDNN (5 varianti)

1. None
2. Fully-convolutional (ResNet)
3. DeepLabV3 (ResNet)
4. U-net
5. Tiramisu

ClassDNN (11 varianti)

1. Inception (3 varianti)
2. ResNet (3 varianti)
3. DenseNet (3 varianti)
4. AlexNet
5. Inception-ResNets



Iper-parametri (9 varianti)

- A. Learning rate (3 valori)
- B. Weight decay (3 valori)
- C. Learning rate decay step (3 valori)

M. Aldinucci, “[Invited Talk] HPC application cloudification: the streamflow toolkit,” in *PARMA-DITAM 2021*

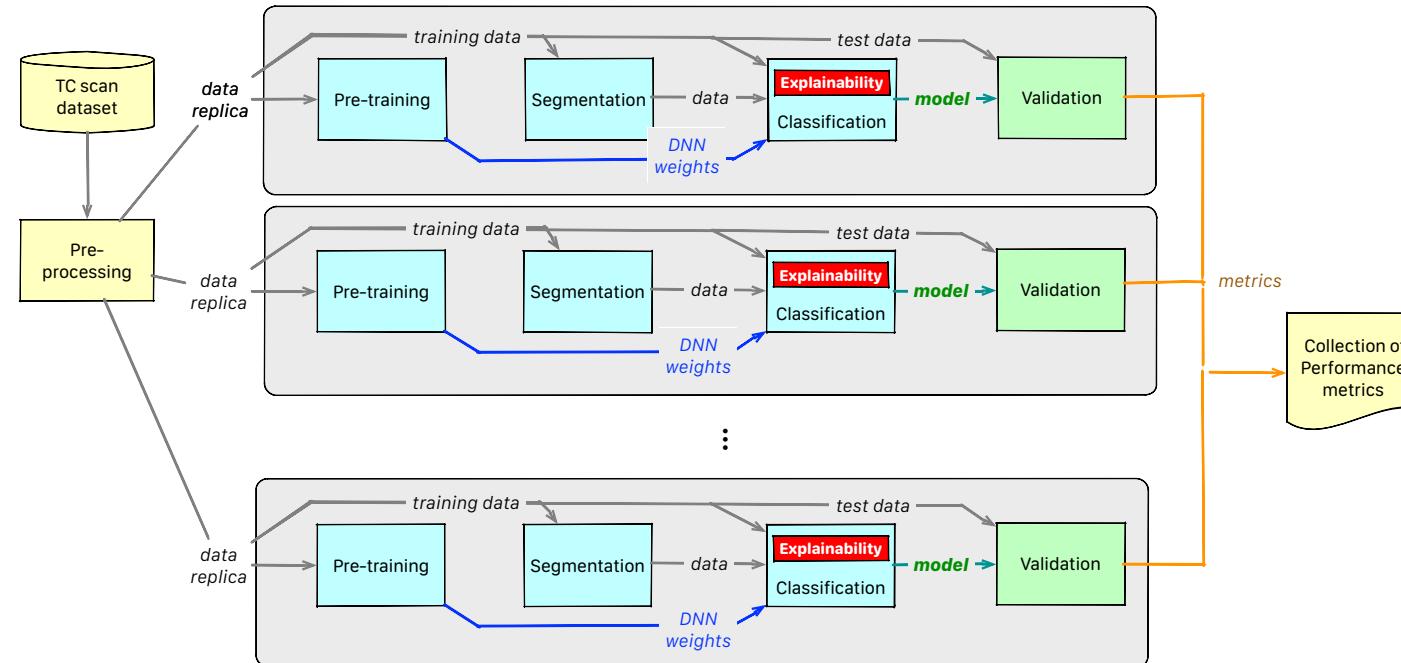
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CLAIRE-COVID19 Universal Pipeline



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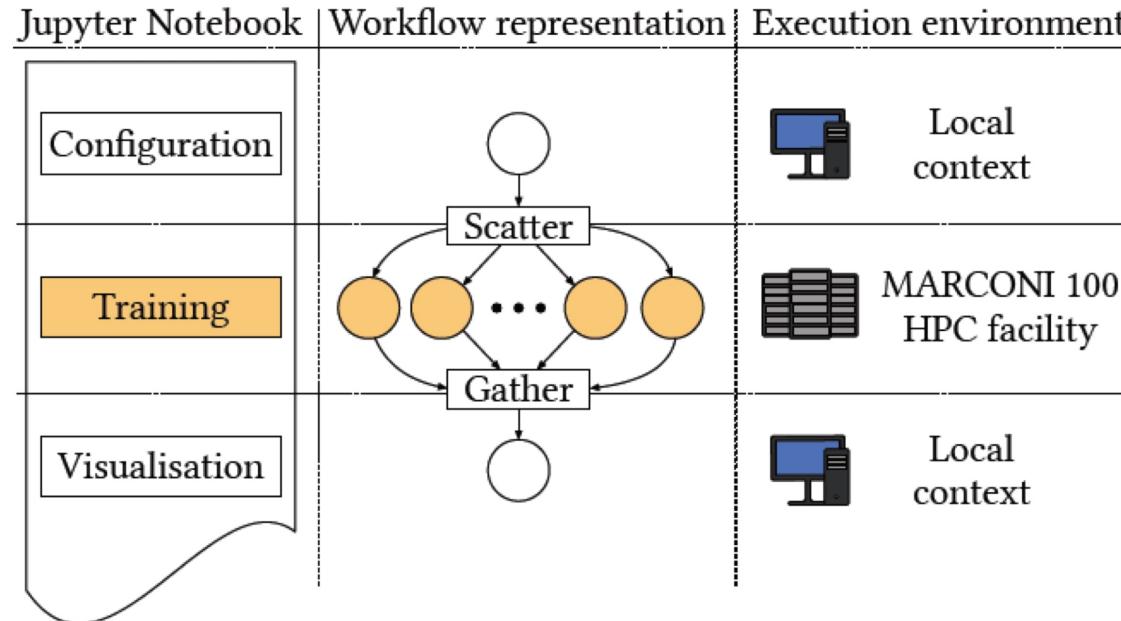
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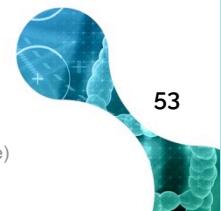
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I. Colonnelli et al., "Distributed workflows with Jupyter," in *Future Generation Computer Systems*





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CLAIRE-COVID19 Universal Pipeline

Pre-trained DenseNet-121 Fine-tuning
with Adam on BIMCV-COVID19

Hyperparameter	Values
Batch size	64
Learning rate (LR)	$[10^{-3}, 10^{-4}, 10^{-5}]$
LR decay step	[10, 15]
LR decay gamma	10^{-1}
Weight decay	$[5 \cdot 10^{-4}, 5 \cdot 10^{-5}]$

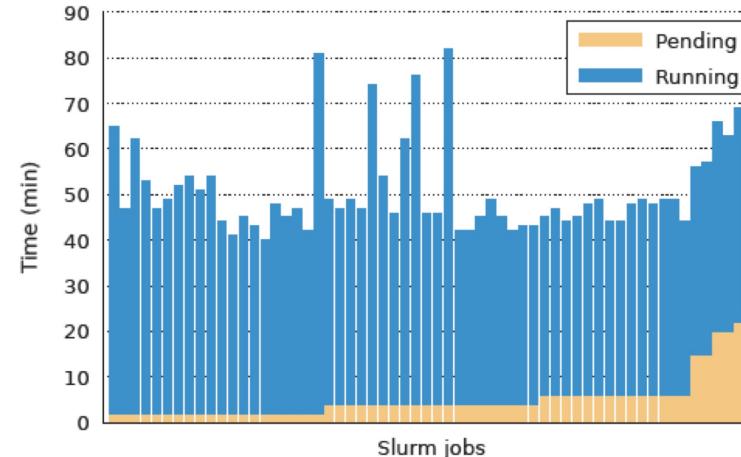
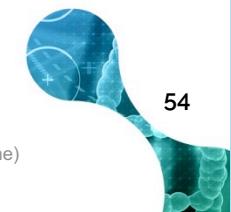


Figure 4: Execution times for the 60 Densenet-121 training experiments running in parallel on 60 GPU-equipped nodes of the CINECA MARCONI 100 facility.

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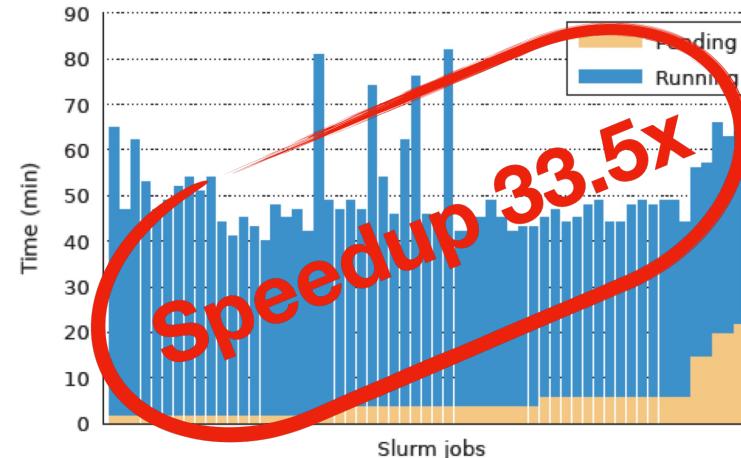


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OpenDeepHealth Platform



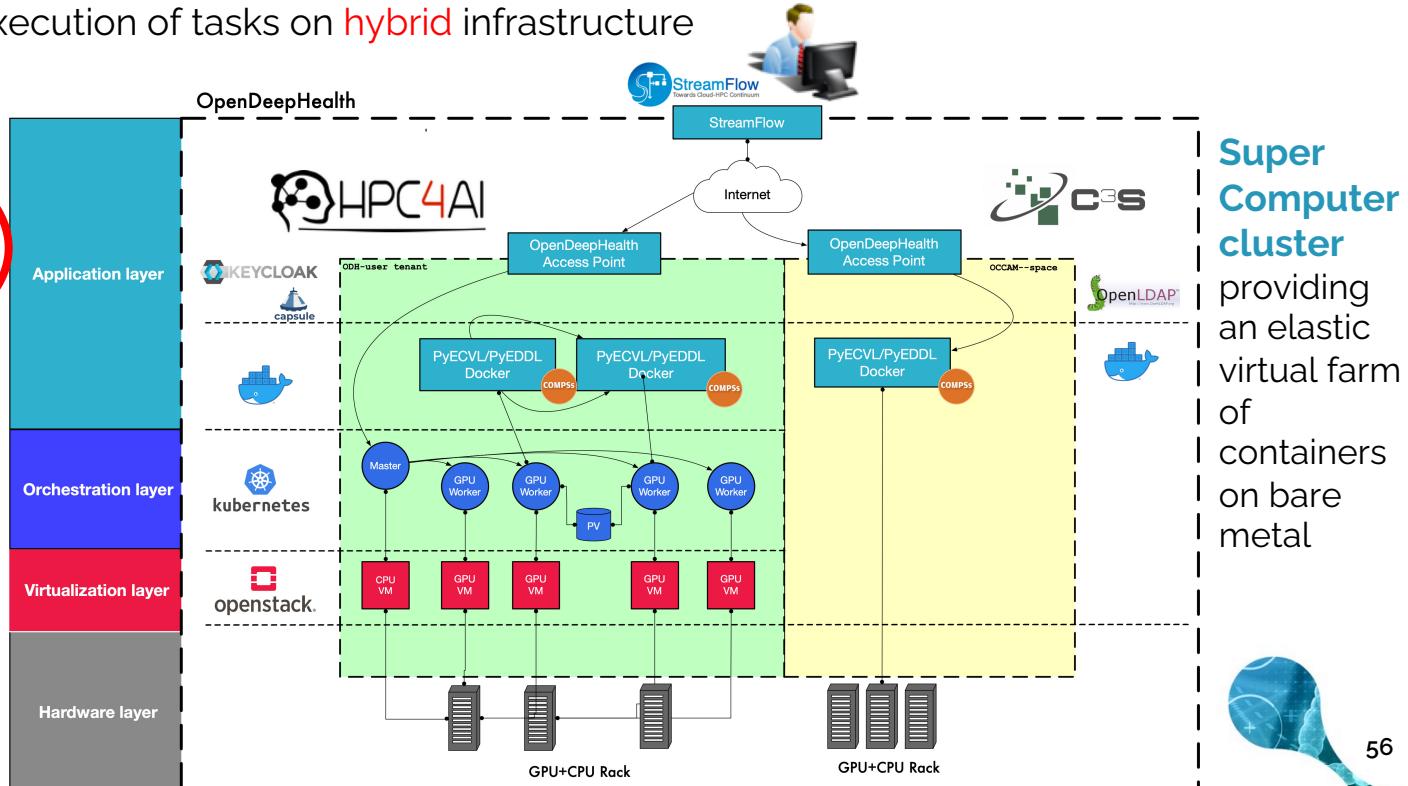
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The **StreamFlow** Workflow Management System (WMS) orchestrates the execution of tasks on **hybrid** infrastructure

Multi-tenant
private Kubernetes
for AI training

Literate workflows to
write complex distributed
pipelines as Jupyter
Notebooks





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Capsule: the Kubernetes multi-tenancy operator

- Capsule leverages the **Kubernetes resources** to handle multi-tenancy at the cluster level;





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 - **Security** (NetworkPolicies, PodSecurityPolicies, allowed Service Kinds, ...)



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 - **Security** (NetworkPolicies, PodSecurityPolicies, allowed Service Kinds, ...)
 - **Resources** (ResourceQuotas, LimitRanges, NodeSelectors, StorageClasses, ...)
 - **Services** (available hostnames, CustomResource quotas, integration with running applications through ServiceAccounts, ...)

```
apiVersion: capsule.clastix.io/v1beta1
kind: Tenant
metadata:
  annotations:
    capsule.clastix.io/enable-node-listing: 'true'
spec:
  additionalRoleBindings:
    - clusterRoleName: odh-policy
      subjects:
        - apiGroup: rbac.authorization.k8s.io
          kind: Group
          name: system:authenticated
  ingressOptions:
    allowedHostnames:
      allowedRegex: .opendeephealth.di.unito.it$ 
    hostnameCollisionScope: Disabled
  namespaceOptions:
    quota: 3
  networkPolicies:
    items:
      - egress:
          - to:
              - ipBlock:
                  cidr: 0.0.0.0/0
                  except:
                    - 252.0.0.0/8
                    - 172.17.7.0/24
                    - 172.20.7.0/24
    ingress:
      - from:
          - namespaceSelector:
              matchLabels:
                capsule.clastix.io/tenant: unito
          - namespaceSelector:
              matchLabels:
                name: projectcontour
          podSelector:
              matchLabels:
                app: envoy
          - podSelector: {}
```

```
policyTypes:
  - Ingress
  - Egress
nodeSelector:
  pool: unito
owners:
  - kind: Group
    name: unito-tenant-owners
serviceOptions:
  allowedServices:
    externalName: true
    loadBalancer: false
    nodePort: false
  externalIPs:
    allowed: []
```

Practically, a Tenant is **a single YAML file to manage all the aspects of multi-tenancy**



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Multi-tenant architecture

- Capsule (<https://capsule.clastix.io/>) for **cluster-level segregation**





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Multi-tenant architecture

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- JupyterHub (<https://jupyter.org/hub>) for **single-user notebooks**



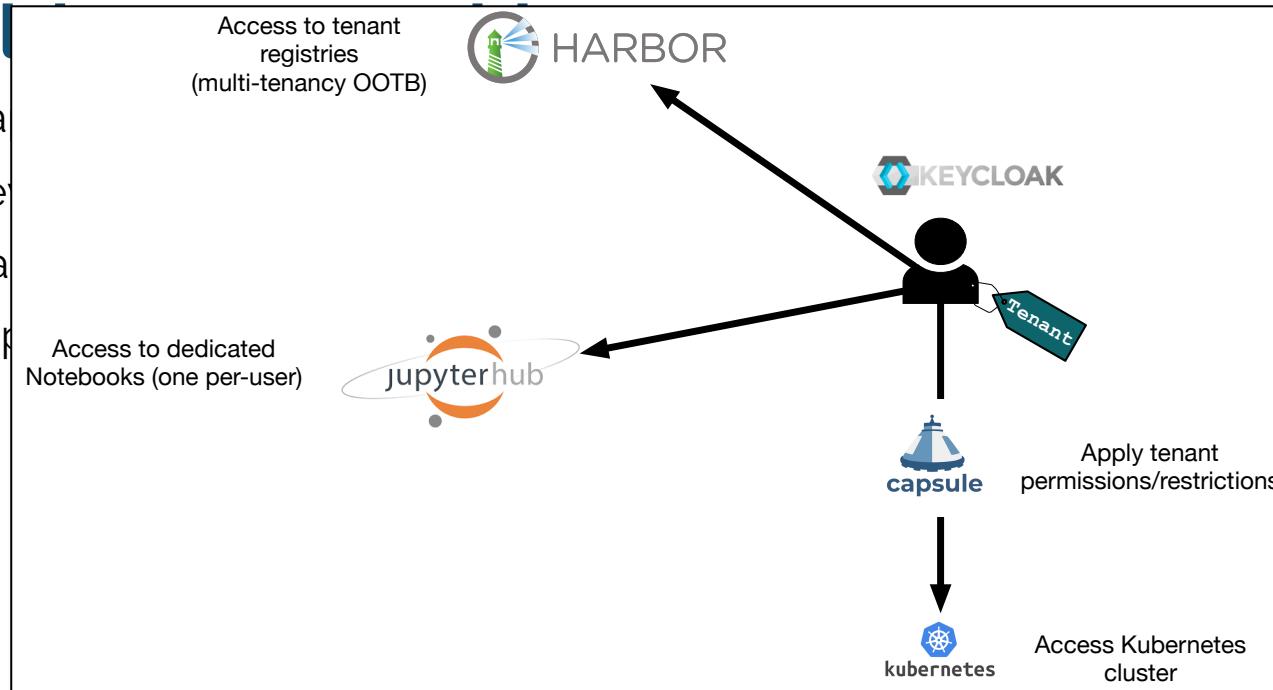


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DEEPHEALTH

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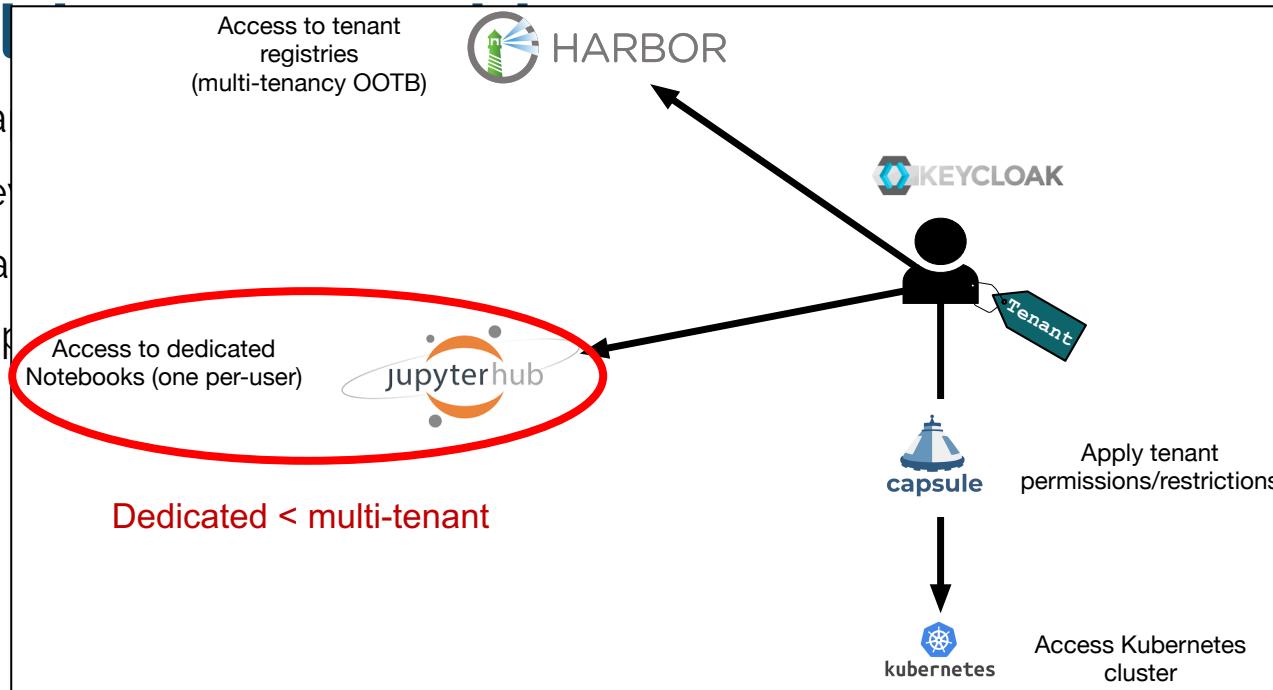


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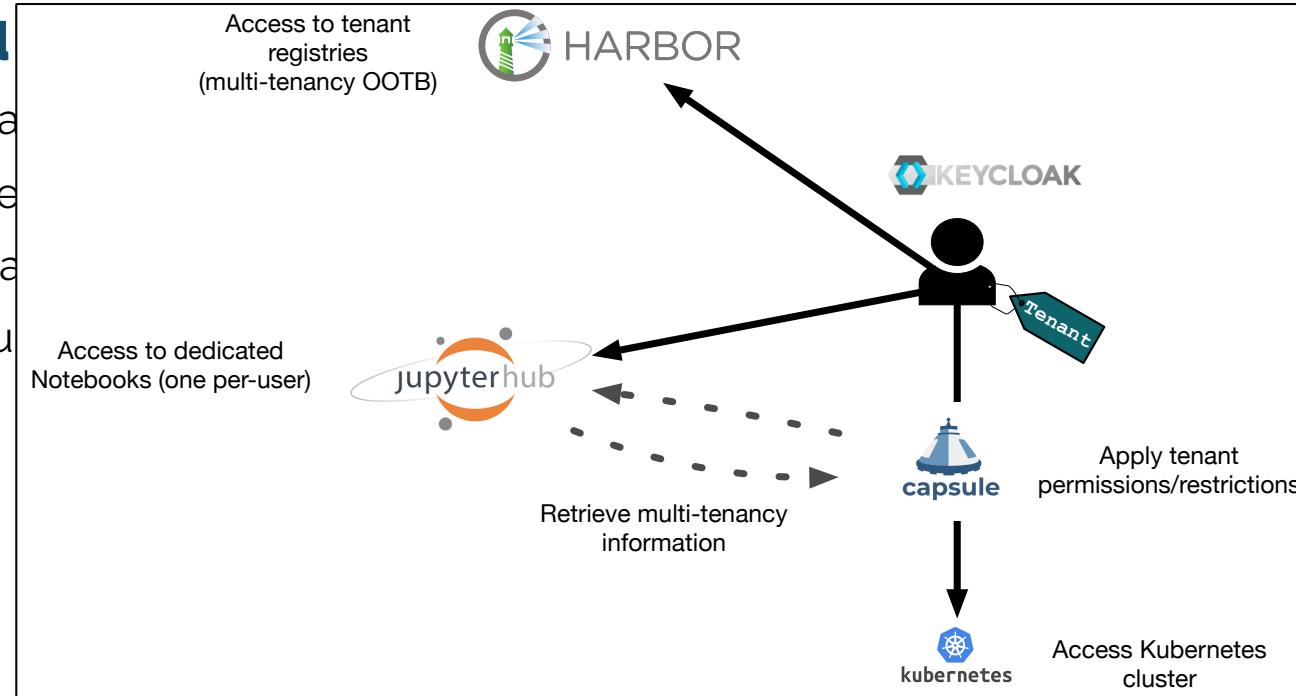
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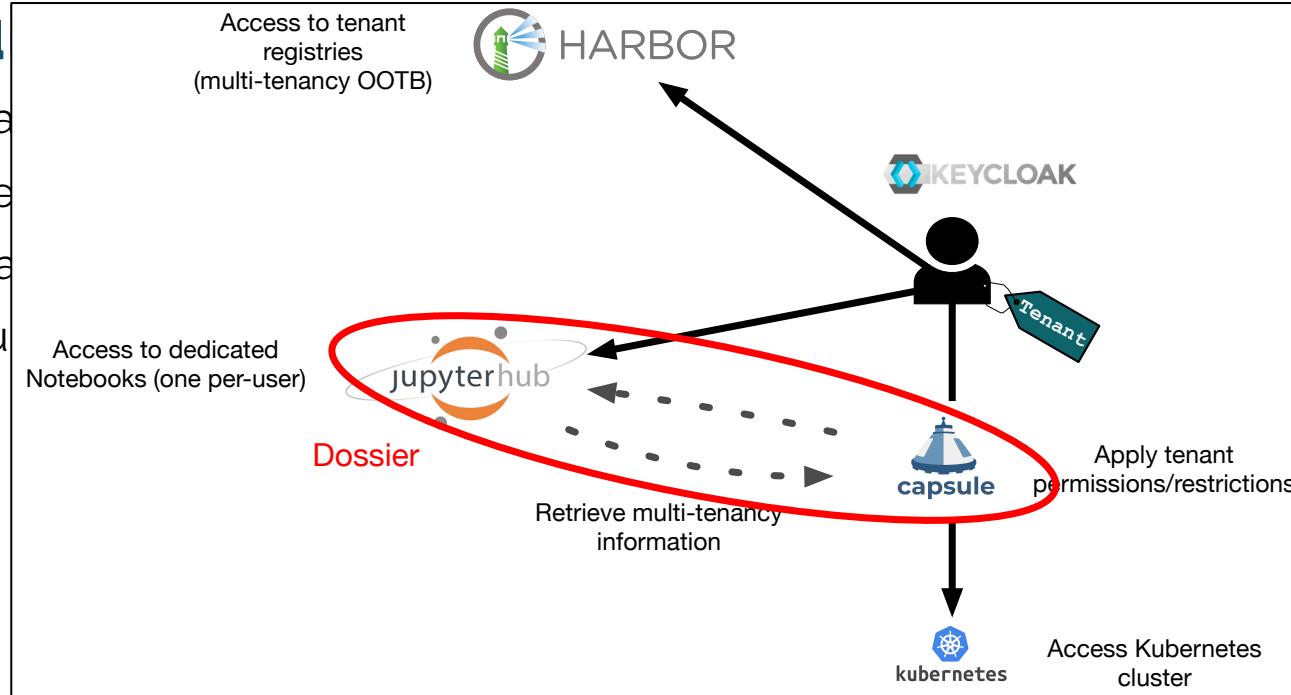
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- Keycloak (<https://www.keycloak.org/>) for **identity management**
- Harbor (<https://goharbor.io/>) for **multi-tenant container registry**
- Dossier (<https://github.com/alpha-unito/dossier>) for **multi-tenant distributed Jupyter Notebooks**





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Dossier: Multi-tenant Distributed Jupyter as a Service

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- Dossier leverages **Capsule** to segment resources and manage permissions at the Tenant level;
- Dossier users can offload computation to other infrastructures, either provided by OpenDeepHealth itself (the **Super Computer cluster**) or independently managed by users;



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- Dossier leverages **Capsule** to segment resources and manage permissions at the Tenant level;
- Dossier users can offload computation to other infrastructures, either provided by OpenDeepHealth itself (the **Super Computer cluster**) or independently managed by users;
- System administrators can easily manage accounting and security at a **coarse-grain** (namespace groups) and in a **centralised way**.



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Thank you!

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iacopo.colonnelli@unito.it

<https://streamflow.di.unito.it/>

<https://jupyter-workflow.di.unito.it/>

<https://github.com/alpha-unito/dossier>



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