

DLHub

Data and Learning Hub for Science



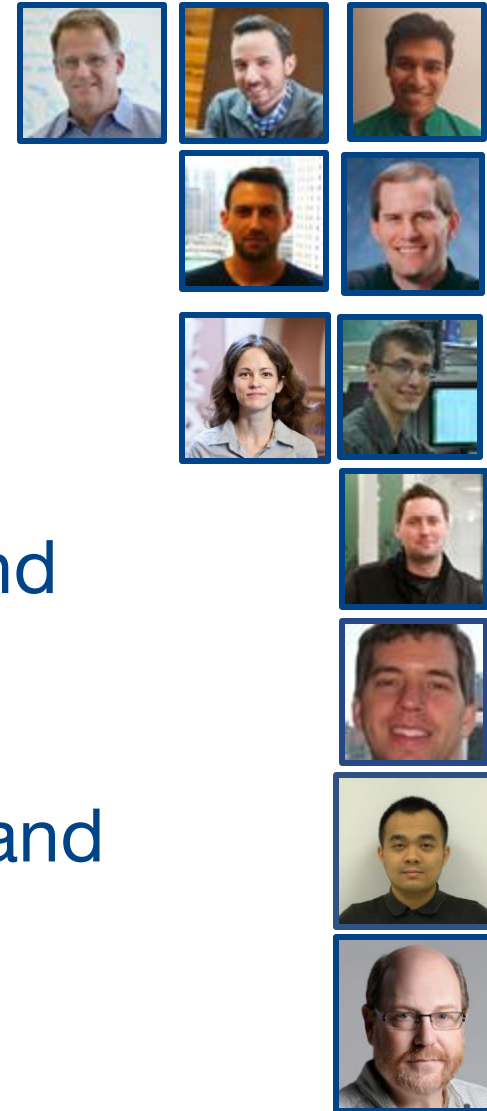
[Ben Blaiszik \(bblaiszik@anl.gov\)](mailto:bblaiszik@anl.gov), Ryan Chard, Logan Ward,
Kyle Chard, Zhuozhao Li, Anna Woodard, Yadu Babuji,
Steve Tuecke, Mike Franklin, Ian Foster

Funding: 2018 Argonne Advanced Computing LDRD

Data and Learning Hub for Science (DLHub)



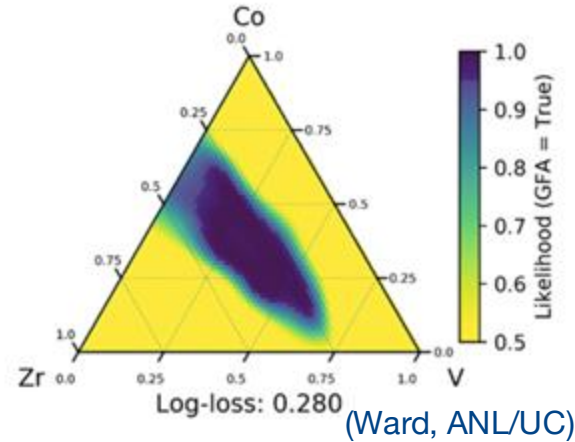
- Collect, publish, categorize models from many disciplines (materials science, physics, chemistry, genomics, etc.)
- Serve models via API to simplify sharing, consumption, and access
- Enable new science through reuse, real-time integration, and synthesis of existing models



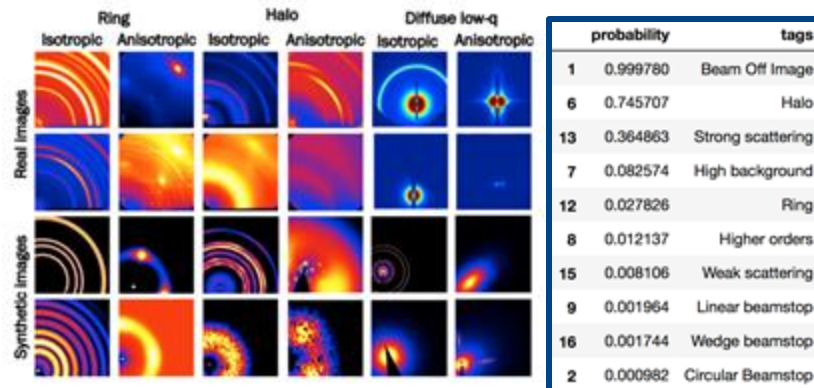
Select DLHub Use Cases

Model-driven Experimentation and Data Tagging

- Metallic glass discovery [active learning]



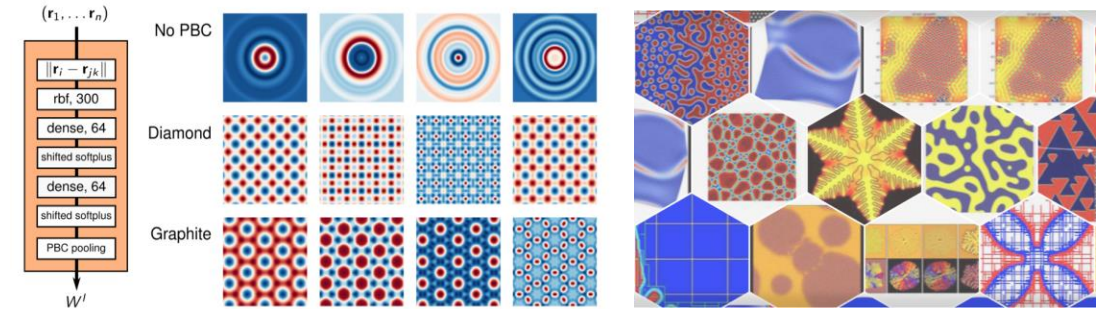
- XRD beamline image tagging



(Yager, BNL)

Community Model Benchmarking

- Crystal structure
- NIST PFHub



(Ward, ANL/UC)

(Wheeler, Warren, Heinonen
NIST/UC/Argonne/NU)

Automated Model Retraining with New Data

- Models linked to dynamic data sources



(Center for Hierarchical Materials
Design NIST/UC/Argonne/NU)

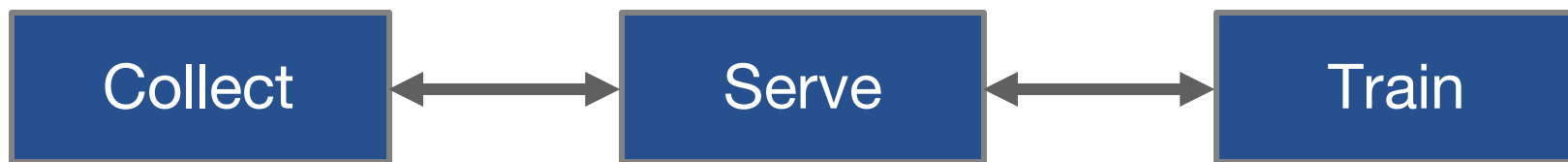


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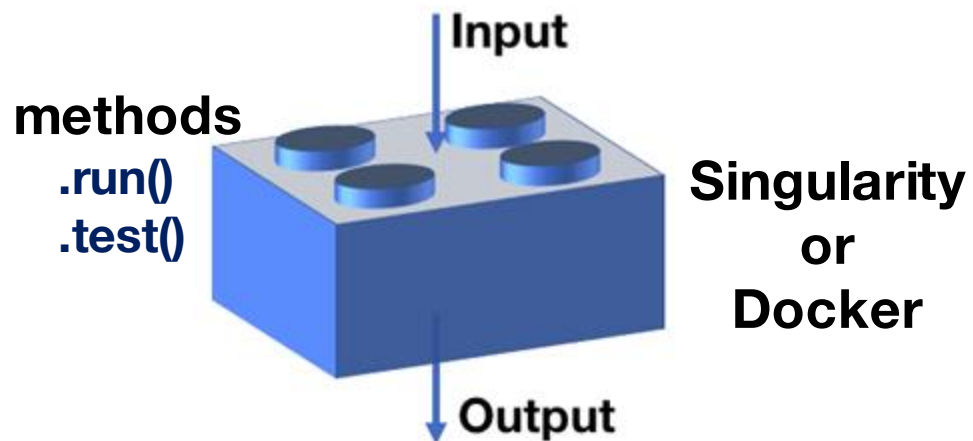


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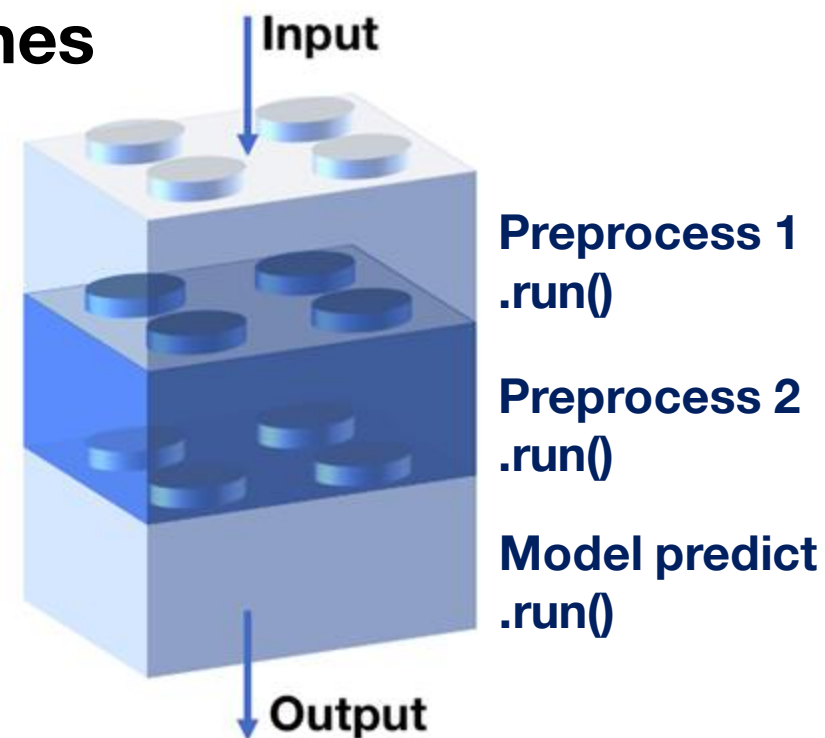
DLHub Servables and Pipelines



Servables

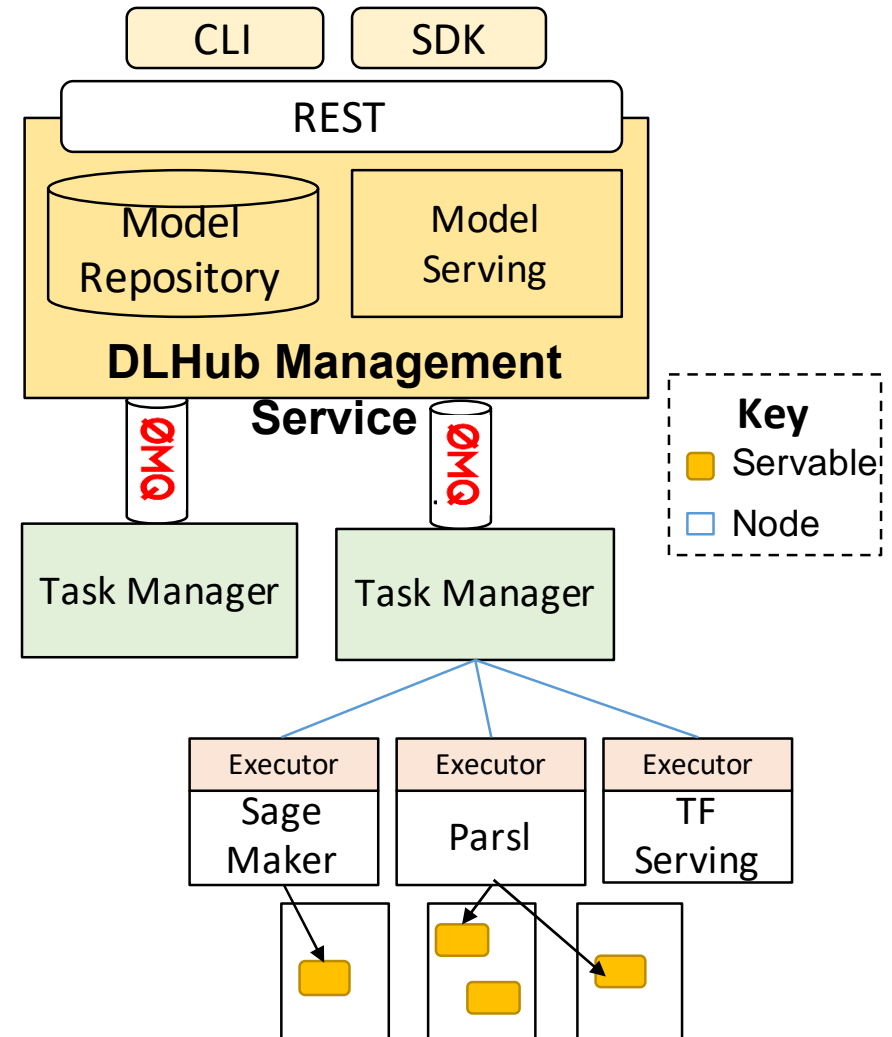


Pipelines



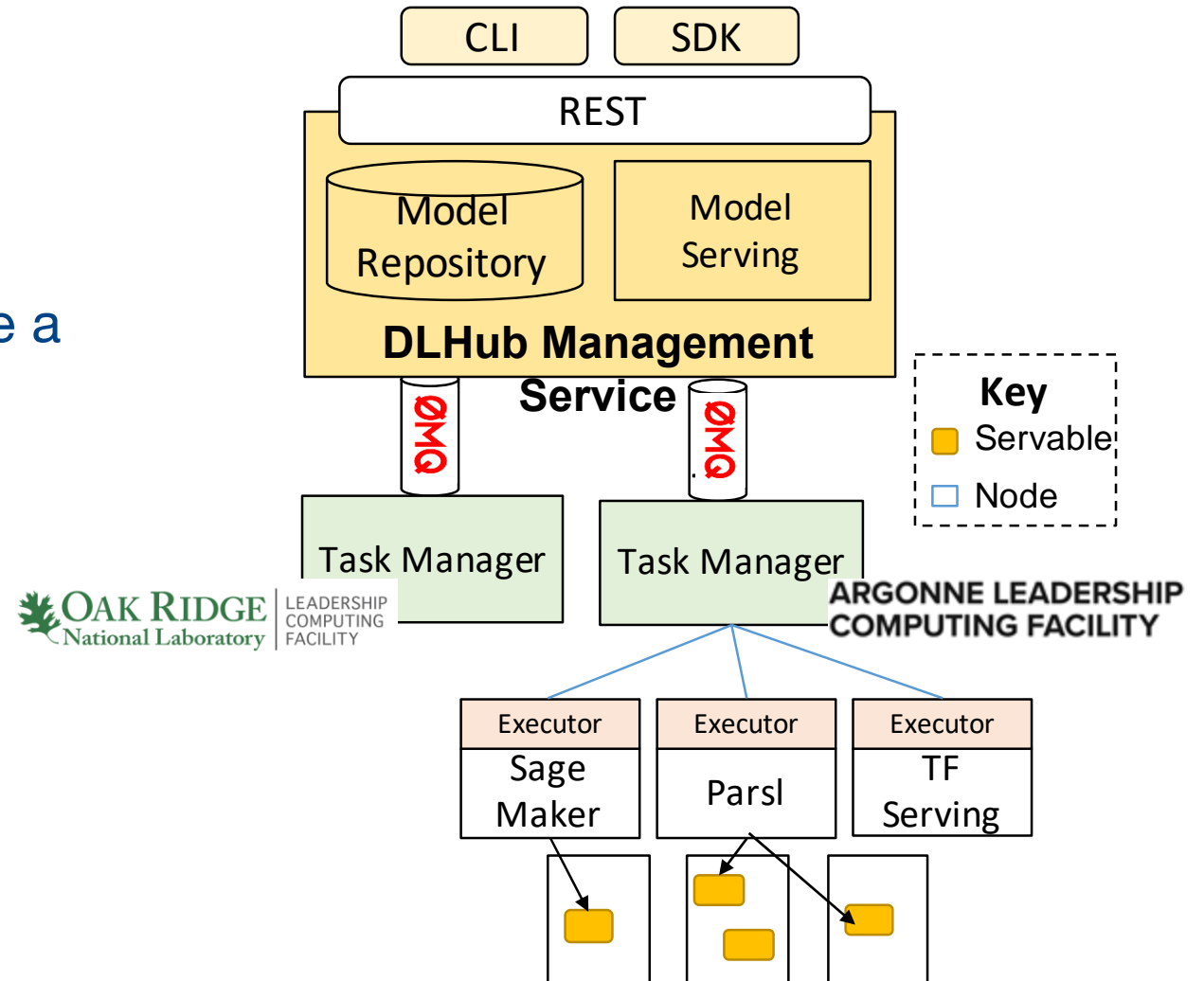
DLHub Architecture

- REST API with Python SDK (available) / CLI (delivery in Nov. 2018)
 - Support model markup, data staging, registration, and invocation
- Model Repository
 - Container registry
 - Advanced search functions
 - Identifier minting capabilities



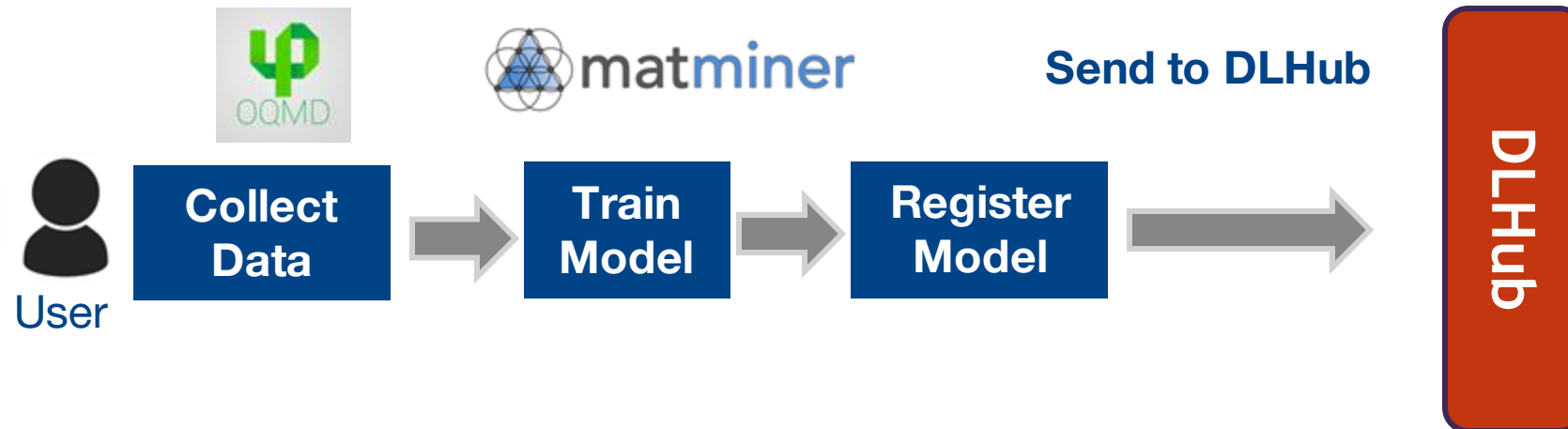
DLHub Architecture

- Task Managers (TM) to support execution on various compute resources
- Executors chosen by TM to invoke a given servable'
- Caching at TM
- Data staging with Globus
- Batch submissions
- Scalability through deployment of model replicas

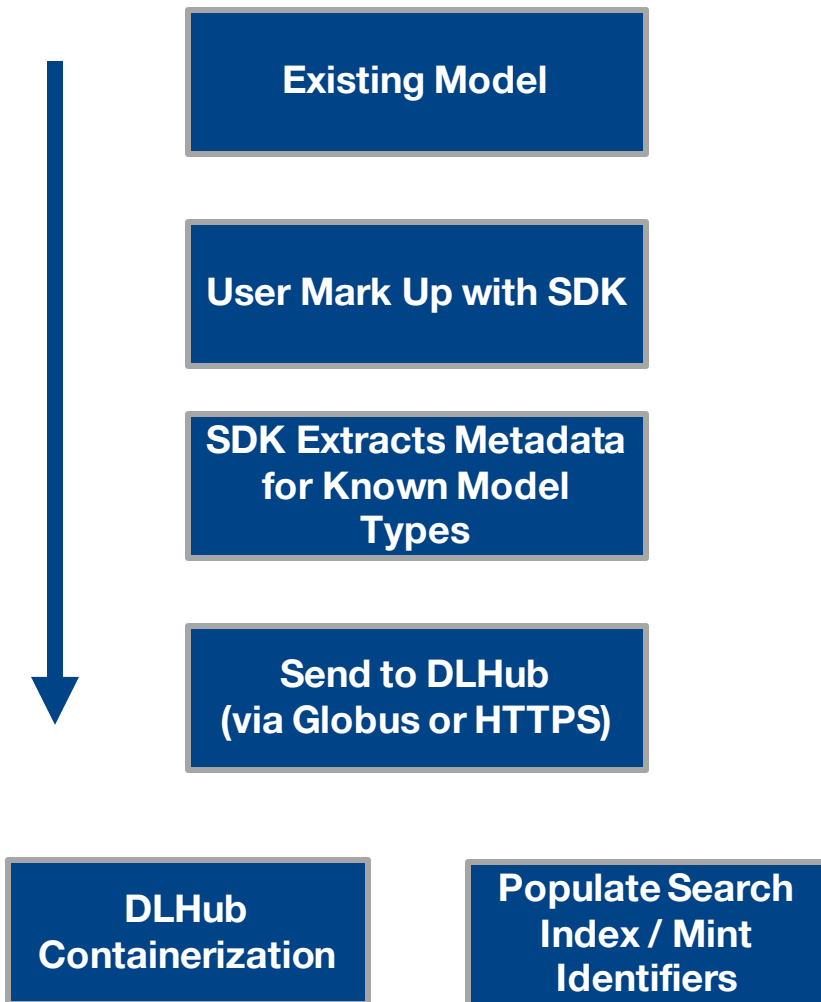


DLHub Model Registration and Publication

- **Register** model metadata, weights, and files to improve discoverability and reusability
- **Containerize** model to enhance interoperability
- **Identify** model with a permanent identifier (e.g., DOI, minid)
- **Version** model and data pre/post processing steps
- **Deploy** model with simplified interfaces for users
- **Control** access to model metadata and usage
- (future) **Automate** retraining and testing when new data are available



Marking up a Model - Python SDK



```
from dlhub_toolbox.models.servables.keras import KerasModel
import pickle as pkl
import json

# Describe the keras model
model = KerasModel('model.hd5', list(map(str, range(10))))

# Describe the model
model.set_title("MNIST Digit Classifier")
model.set_name("mnist_tiny_example")
model.set_domains(["general", "digit recognition"])

# Add link to paper describing the dataset
model.add_related_identifier("10.1109/CVPR.2007.383157", "DOI",
                             | | | | | | | "IsDescribedBy")

model.set_authors(["Lecunn, Yann", "Cortes, Corinna"])

# Describe the outputs in more detail
model.output['description'] = 'Probabilities of being 0-9'
model.input['description'] = 'Image of a digit'
```


Python SDK - Automated Metadata Generation

Citation Metadata

```
"datacite": {
  "creators": [{
    "givenName": "Yann",
    "familyName": "Lecun",
    "affiliations": []
  },
  {
    "givenName": "Corinna",
    "familyName": "Cortes",
    "affiliations": []
  }
],
  "titles": [{
    "title": "MNIST Digit Classifier"
  }],
  "publisher": "DLHub",
  "publicationYear": "2018",
  "relatedIdentifiers": [{
    "relatedIdentifier": "10.1109/CVPR.2007.383157",
    "relatedIdentifierType": "DOI",
    "relationType": "IsDescribedBy"
  }],
  "identifier": {
    "identifier": "10.YET/UNASSIGNED",
    "identifierType": "DOI"
  },
  "resourceType": {
    "resourceTypeGeneral": "InteractiveResource"
  }
},
```

DLHub Metadata

```
"dlhub": {
  "version": "0.1",
  "domains": [
    "general",
    "digit recognition"
  ],
  "visible_to": [
    "public"
  ],
  "id": null,
  "name": "mnist_tiny_example",
  "files": {
    "other": [],
    "model": "model.hd5"
  }
},
```

Access Control

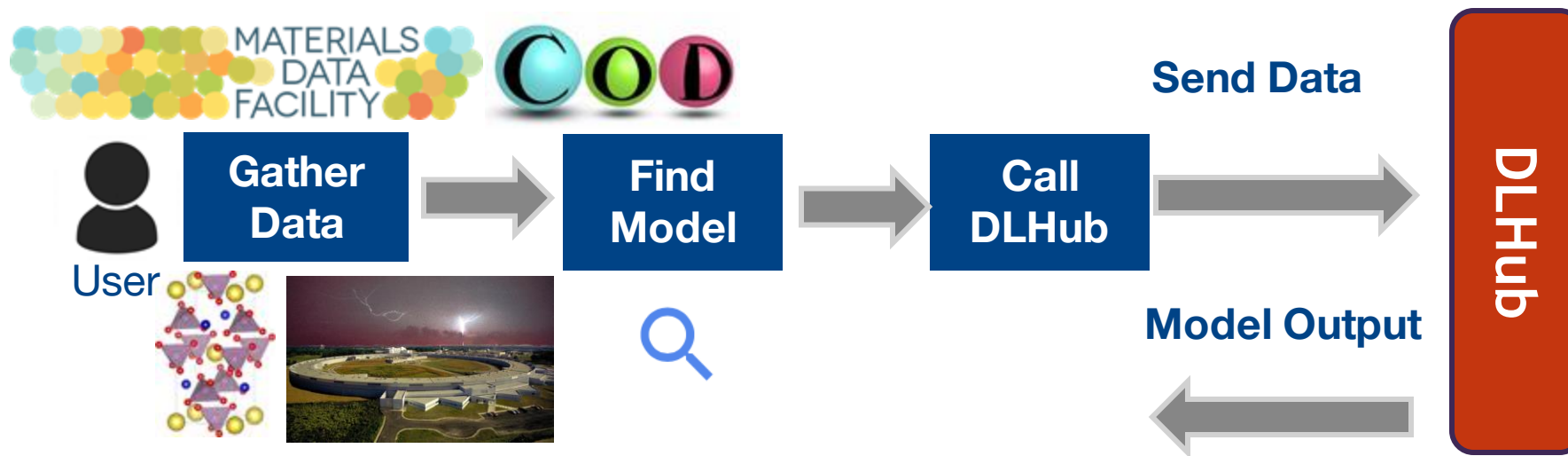
- **Public**
- **Globus users**
- **Globus groups**

Servable Metadata

```
"servable": {
  "methods": {
    "run": {
      "input": {
        "type": "ndarray",
        "description": "Image of a digit",
        "shape": [null,28,28,1]
      },
      "output": {
        "type": "ndarray",
        "description": "Probabilities of being 0-9",
        "shape": [null,10]
      },
      "parameters": {},
      "method_details": {
        "method_name": "predict",
        "classes": ["0","1","2","3","4",
          "5","6","7","8","9"]
      }
    }
  },
  "shim": "keras.KerasServable",
  "language": "python",
  "dependencies": {
    "python": {
      "keras": "2.2.4",
      "h5py": "2.8.0"
    }
  },
  "type": "Keras Model",
  "model_type": "Deep NN"
}
```

DLHub Model Discovery and Usage

- Find curated and tested models
- Use models through simple interfaces

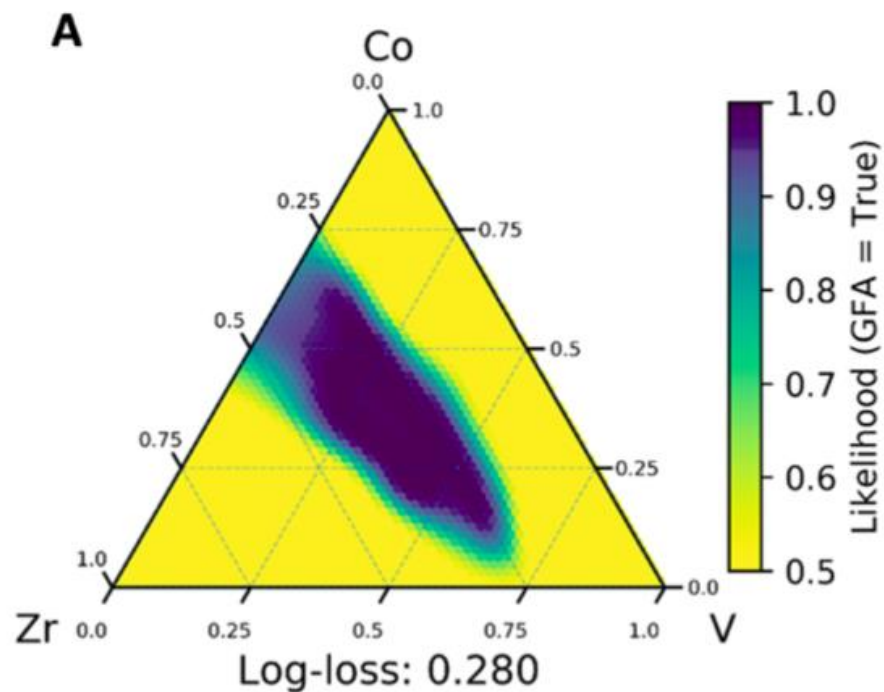


Predicting Glass-forming Ability

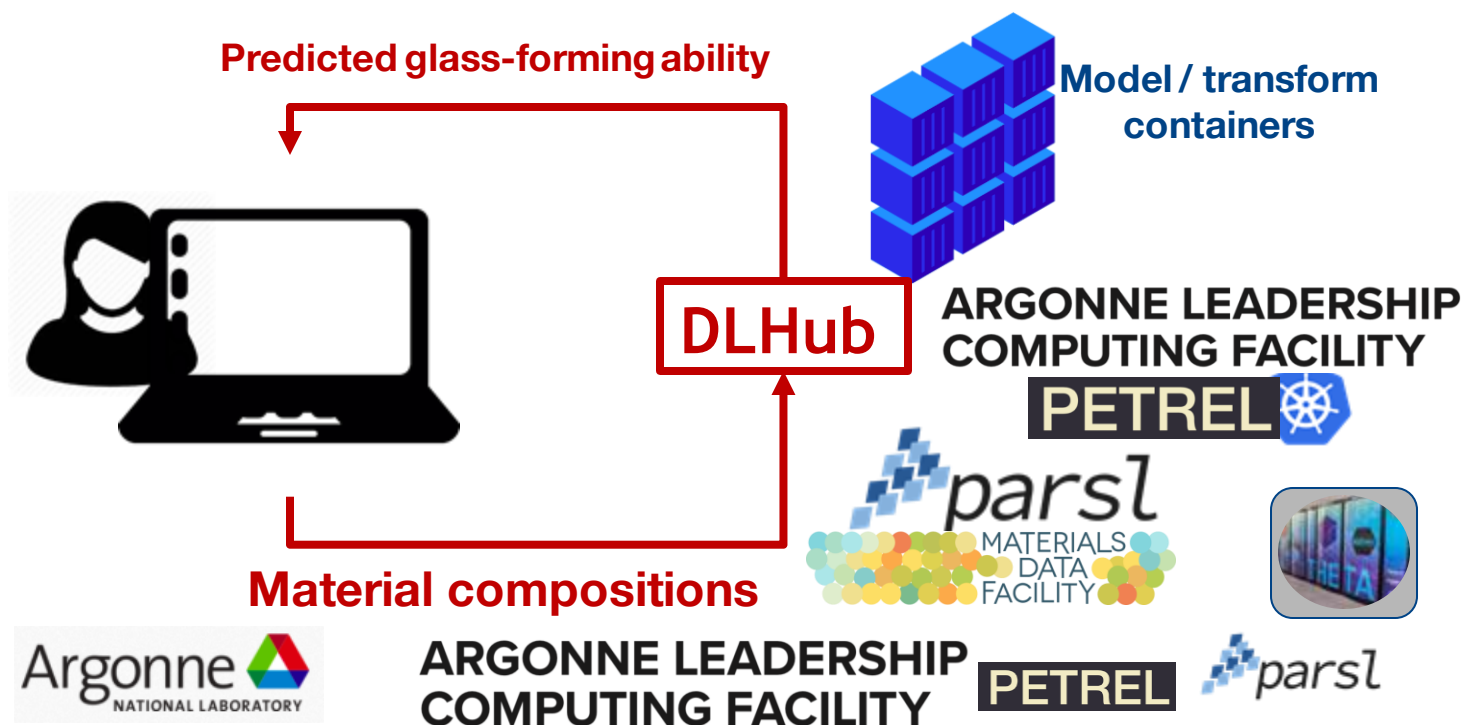
Accelerated discovery of metallic glasses through iteration of machine learning and high-throughput experiments

Fang Ren^{1,*}, Logan Ward^{2,3,*}, Travis Williams⁴, Kevin J. Laws⁵, Christopher Wolverton², Jason Hattrick-Simpers⁶ and Apurva Mehta^{1,†}

10.1126/sciadv.aag1566



- Where are the model and trained weights?
- How do I run the model on my data?
- How can I retrain the model on new data?
- How can I build on this work?

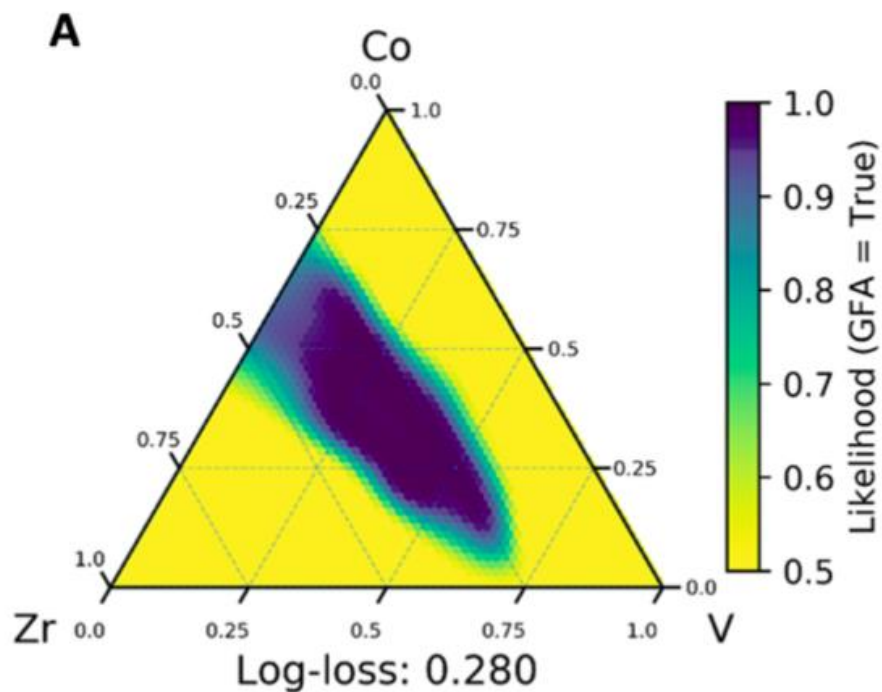


Predicting Glass-forming Ability

Accelerated discovery of metallic glasses through iteration of machine learning and high-throughput experiments

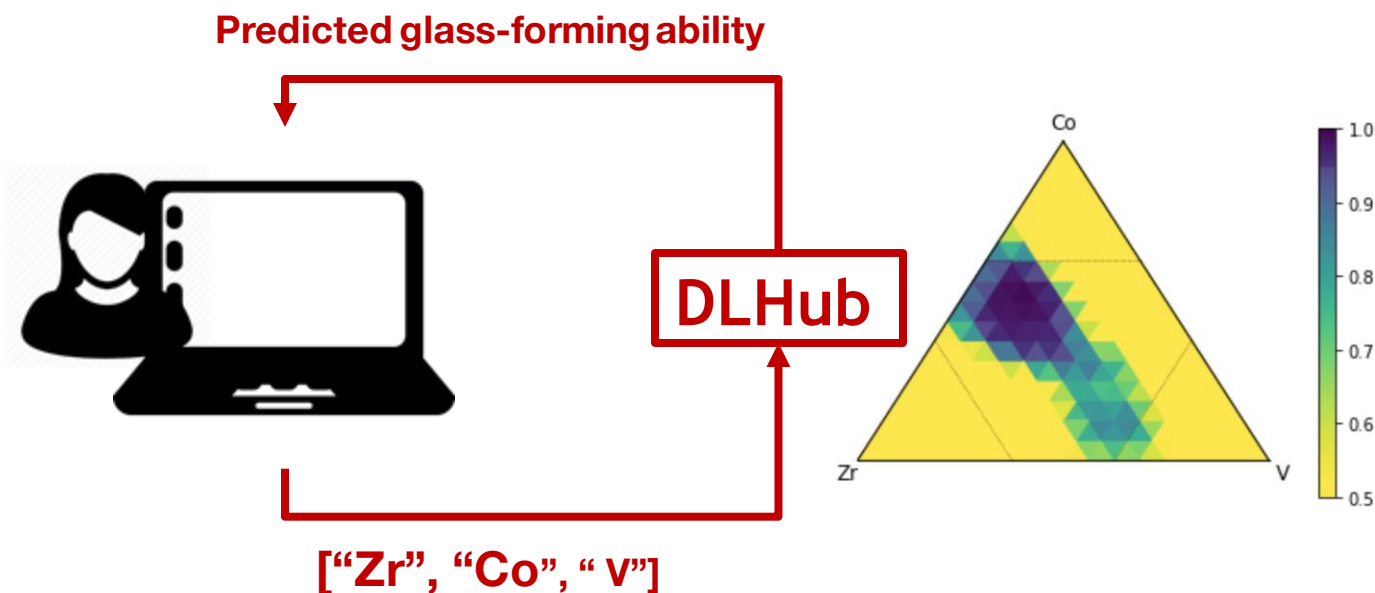
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```
servable_name = "metallic_glass"  
servable_id = dl.get_id_by_name(servable_name)  
elems = ["V", "Co", "Zr"]
```

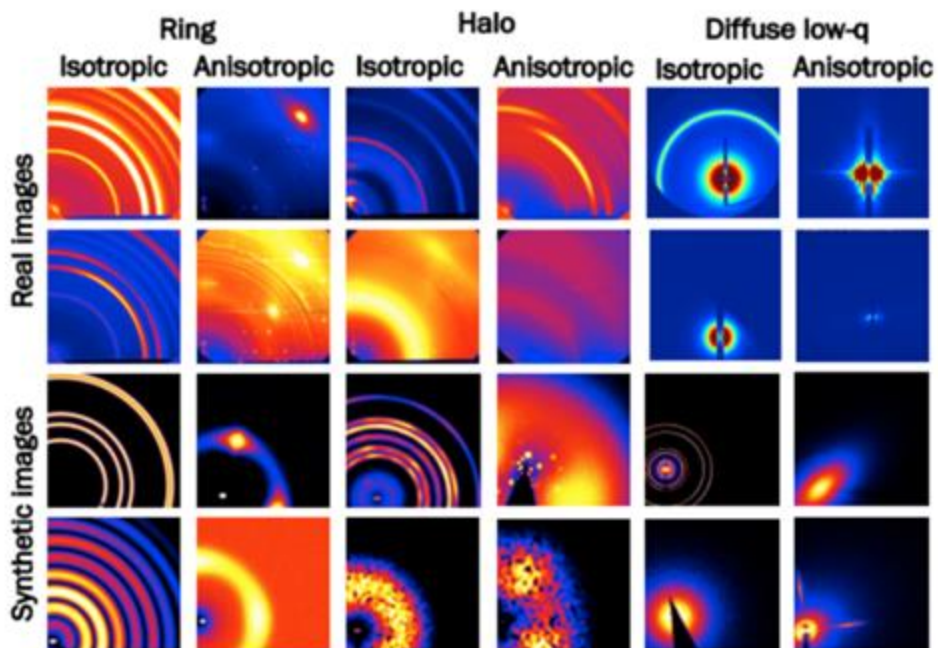
```
res = dl.run(servable_id, {"data":elems})
```



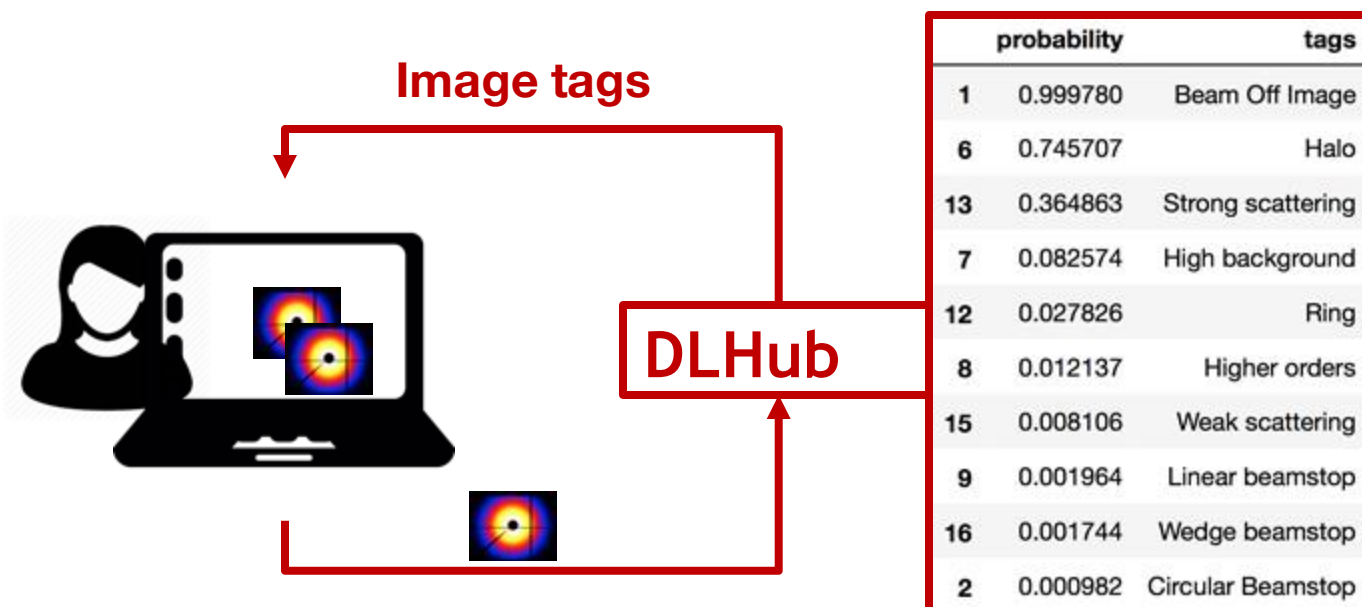
Analyzing Beamline Images

Robust and Scalable Deep Learning for X-ray Synchrotron Image Analysis

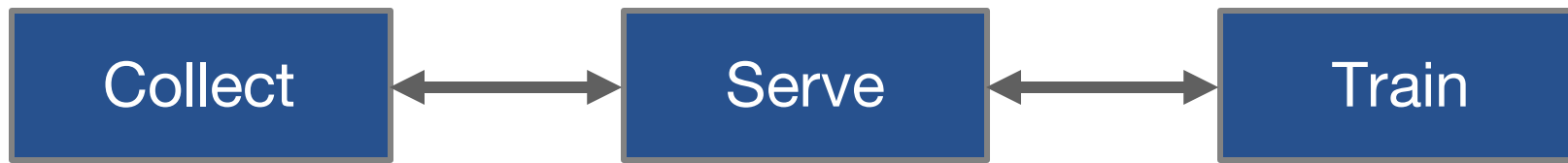
Nicole Meister^{1*}, Ziqiao Guan^{2*}, Jinzhen Wang³, Ronald Lashley⁴,
Jiliang Liu⁵, Julien Lhermitte⁵, Kevin Yager⁵, Hong Qin², Bo Sun⁶, Dantong Yu³



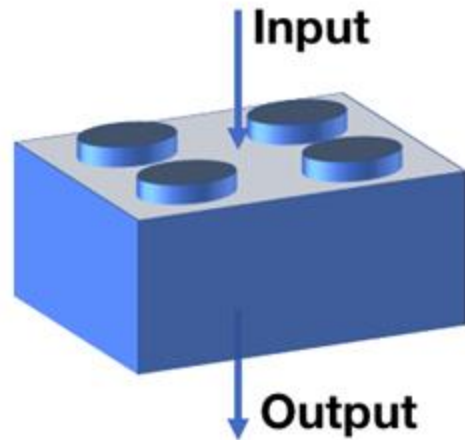
- Stage data into containers via Globus HTTPS
- Pass valid token and data location



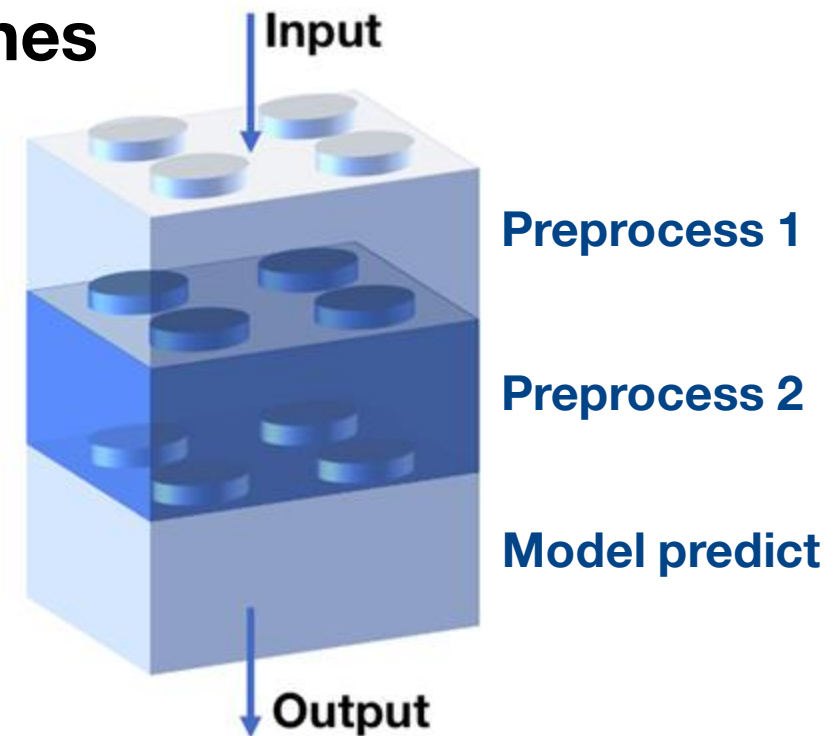
Data and Learning Hub (DLHub): Pipelines



Servables

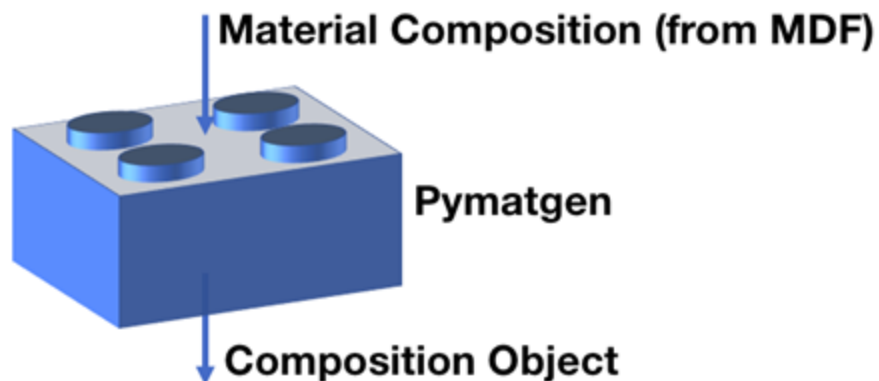


Pipelines

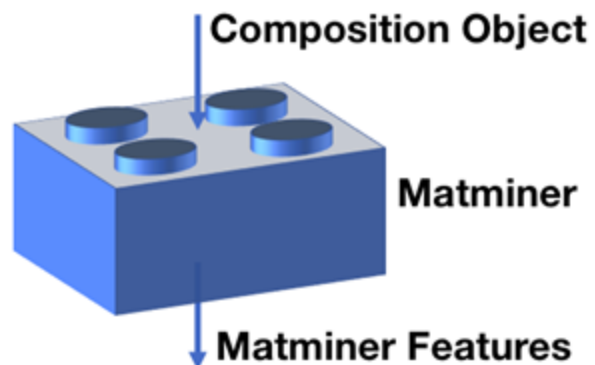


Pipelines: Predicting Formation Enthalpy

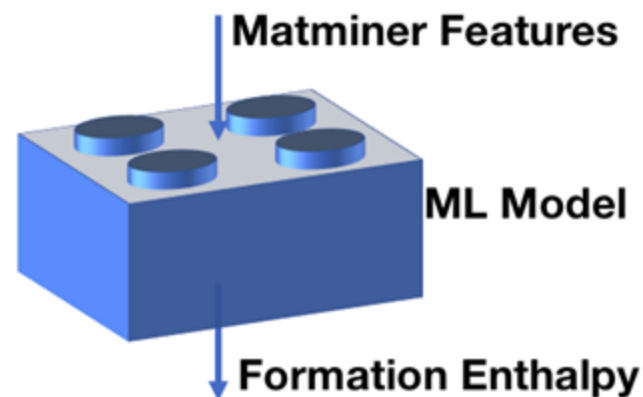
Step 1



Step 2

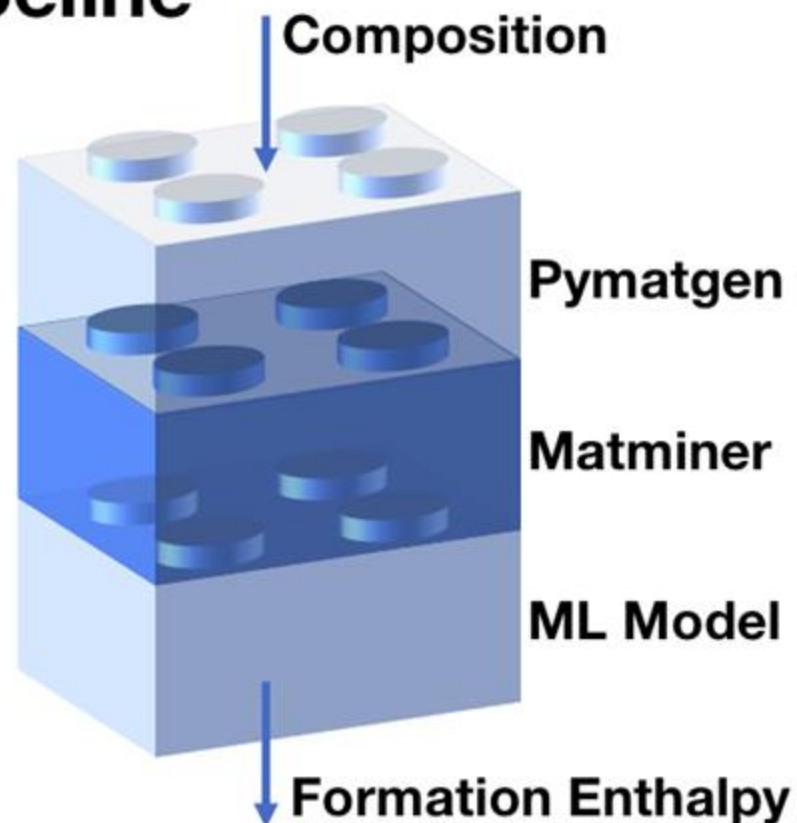


Step 3



Predicting Formation Enthalpy

Pipeline

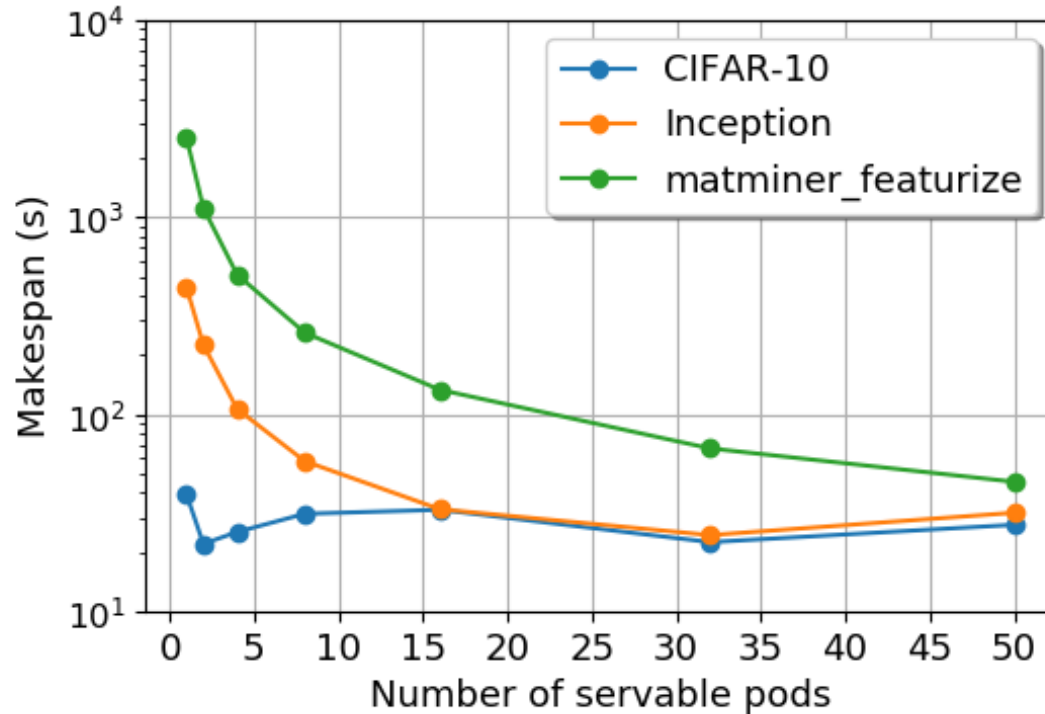


DLHub

Performance

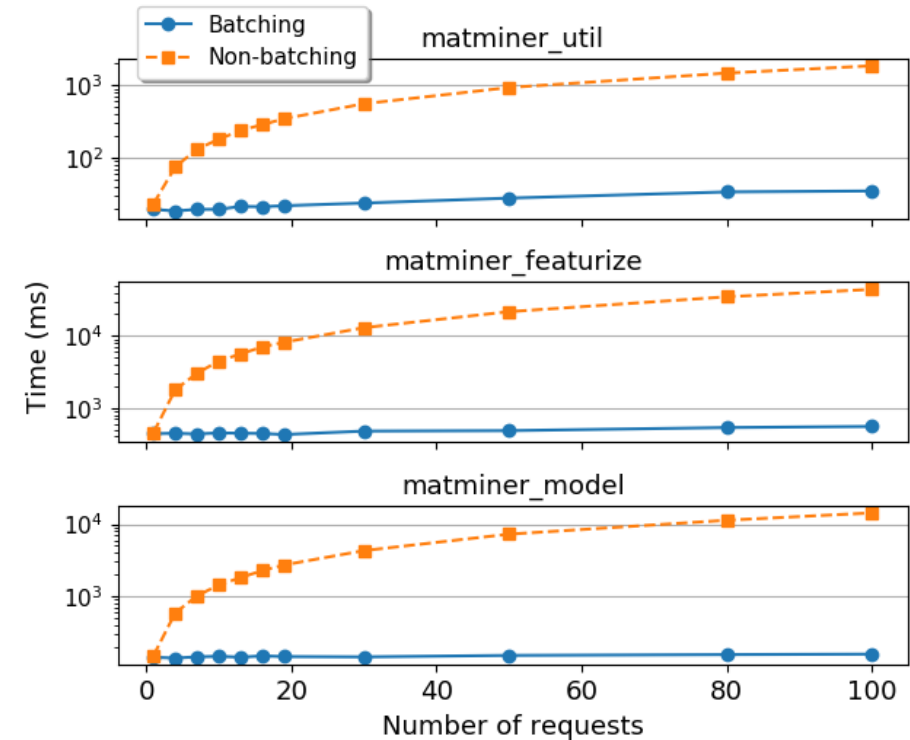
DLHub Performance

Scale Testing



The time required for the Inception, CIFAR10, and Matminer-featurize models to process 5000 inferences with varying numbers of replicas.

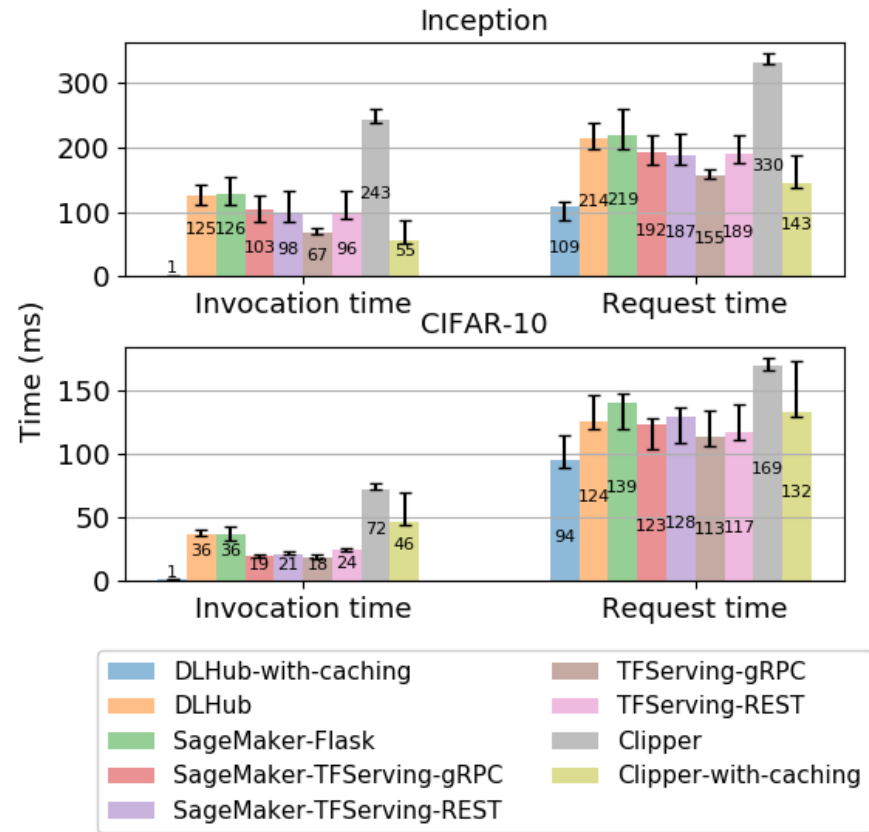
Batching



Servable invocation time, with and without batching.

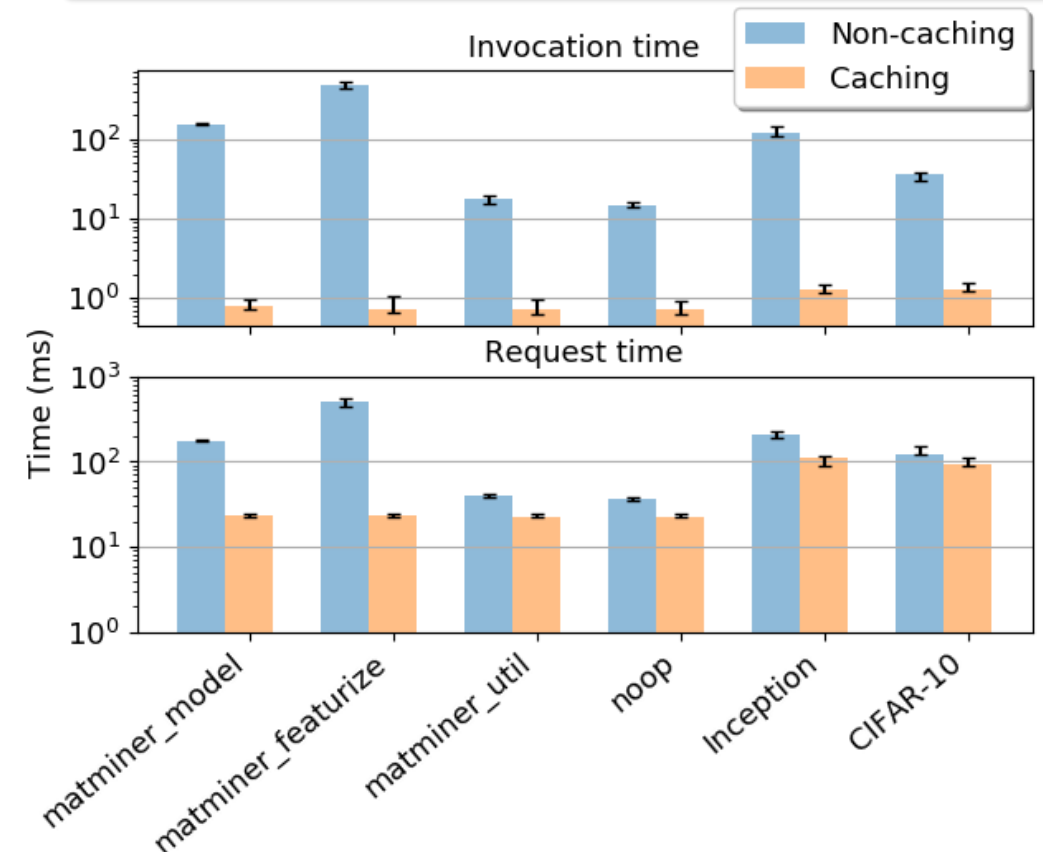
DLHub Performance

Serving General Models



Performance of different serving systems on the Inception and CIFAR-10 problems.

Caching



Performance impact of caching in DLHub. Bars and error bars show median and 5th/95th percentiles

DLHub Summary

Model deposit and discovery

- Developed a model schema to promote discovery
- Implemented advanced search and filtering
- Built ingest flow: models are dynamically staged, packaged, dockerized, published, and indexed

Model serving

- Deployed capabilities for users to run inference with SDK and CLI
- Automated testing of containers
- Implemented caching and batching

Support for multiple execution sites

- PetrelKube: Parsl, TF serving, Sagemaker
- Other: AWS, OSG

Authentication

- Protected model metadata and inference with GlobusAuth
- Secured data staging

Monitoring and statistics

- Request, invocation, data staging

Future work

- Dynamic scaling by load
- Build Web UI to create pipelines and invoke models
- Cache at the servable level within pipelines
- Couple DLHub to data sources (MDF, etc.)
- Integrate with ML frontend tools (DeepForge), optimization tools (DeepHyper), and more
- Create interface for training and retraining of models

Thanks to our sponsors!

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Data Management and Sharing Pilot

Parsl



Globus

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