Merlin

Machine Learning Workflow for HPC

Joe Koning
Luc Peterson
The LBPM WF Team
LLNL

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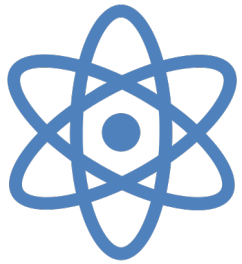
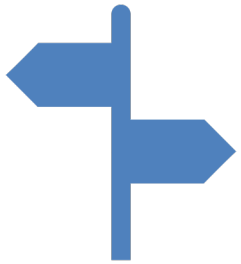
LLNL-PRES-825612

Lawrence Livermore National Laboratory

Outline



Technical Overview



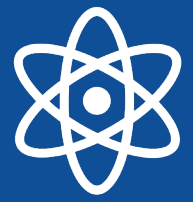
Current Direction

Future Direction

Merlin is the work of many amazing people from across LLNL

- Ben Bay
- Joe Koning
- Jeremy White
- Jessica Semler
- Peter Robinson
- Frank Di Natale
- Bogdan Kustowski
- Vic Castillo
- Yamen Mubarak
- Kevin Athey
- Aiden Keogh
- Brian Spears
- Timo Bremer
- Rushil Anirudh
- Jay Thiagarajan
- Jim Gaffney
- Gemma Anderson
- John Field
- Scott Brandon
- Dave Fox & Livermore Computing (LC)
- Dong Ahn, Steve Herbein & the Flux Team
- Brian Van Essen & the LBANN Team
- Dan Laney, Becky Haluska & the Workflow Team
- Cyrus Harrison & the Conduit team
- Steve Langer & many friendly and patient users

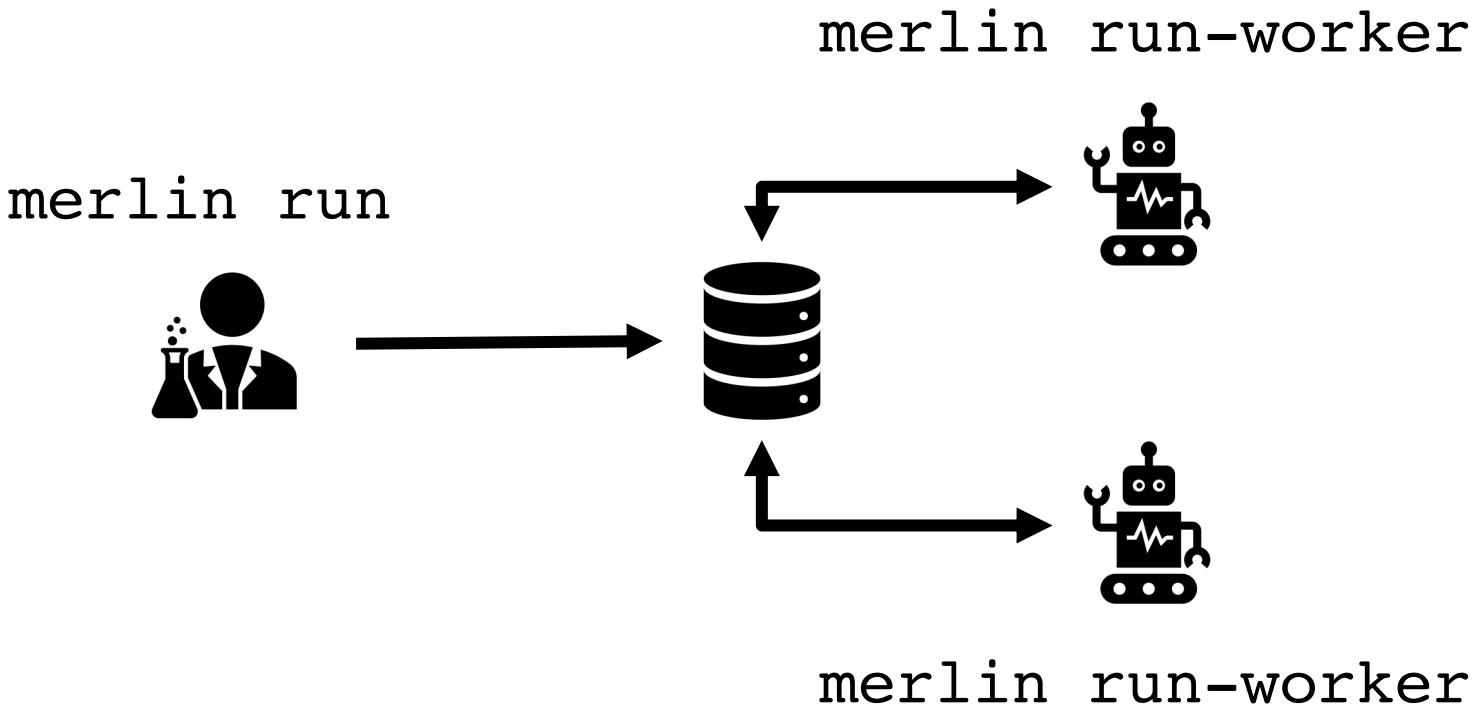
Funding: Learning Based Predicted Models LDRD-SI (Spears PI)



technical overview

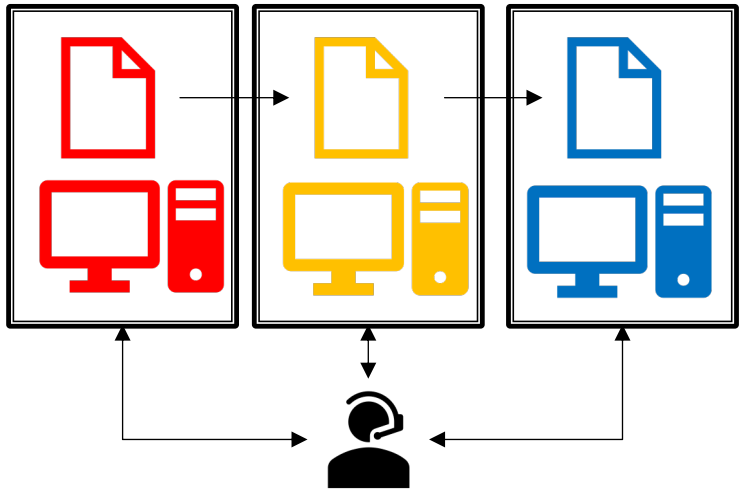

Terminal — ssh -XY lucpeter@quartz — 92×36
(merlin3_7) bash-4.2$

What just happened? A producer-consumer workflow



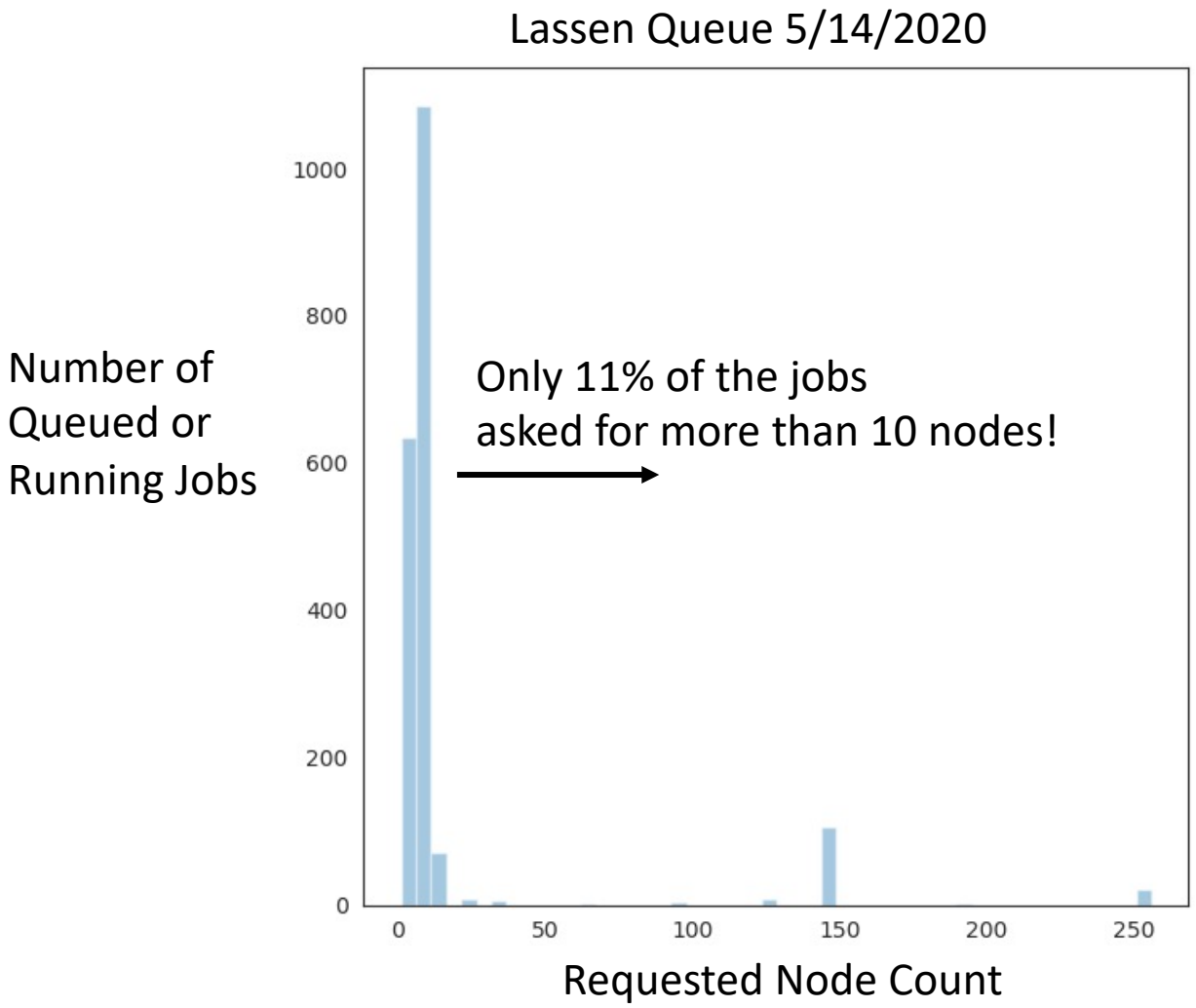
Traditional HPC workflow systems: External or Internal Coordination

External coordination
Independent batch jobs tied to tasks



e.g. maestro

There is a large demand for small jobs



Problem:

HPC schedulers are more efficient with fewer large jobs than with many small jobs

As our systems get better, we can do more with fewer nodes. El Capitan…

Traditional HPC workflow systems: External or Internal Coordination

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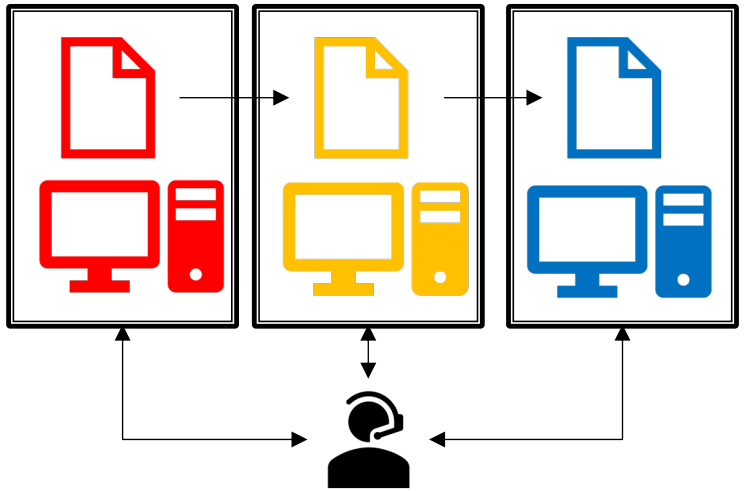
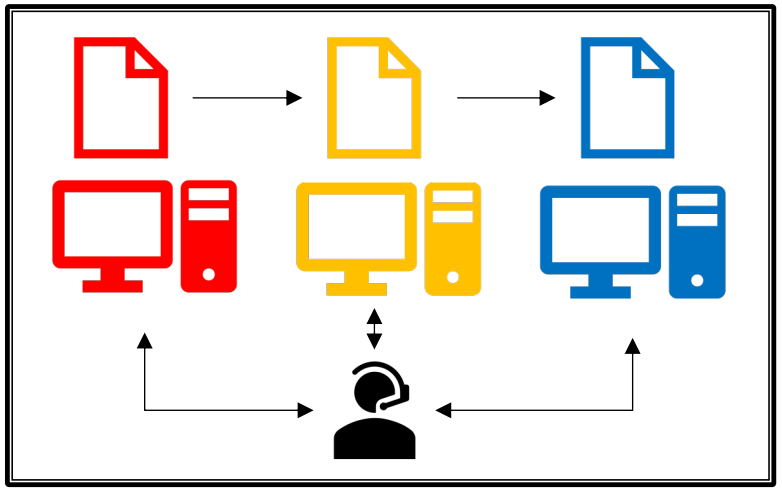


e.g. maestro

Internal coordination
Large batch jobs for multiple tasks

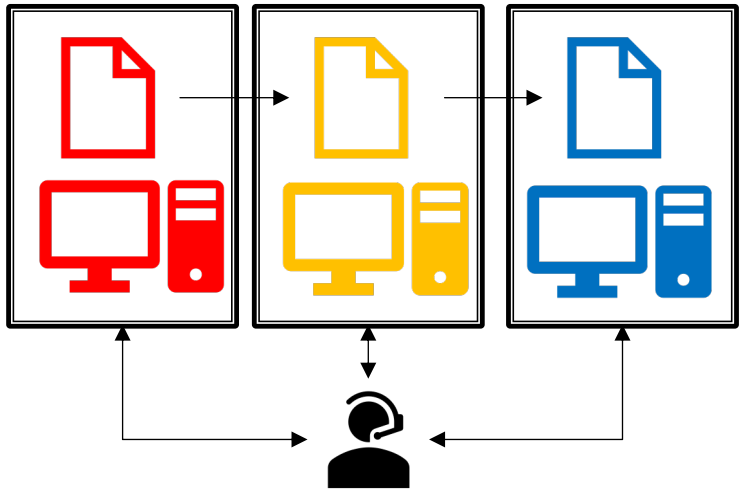


Improves scheduler performance via packing.

e.g. UQPipeline

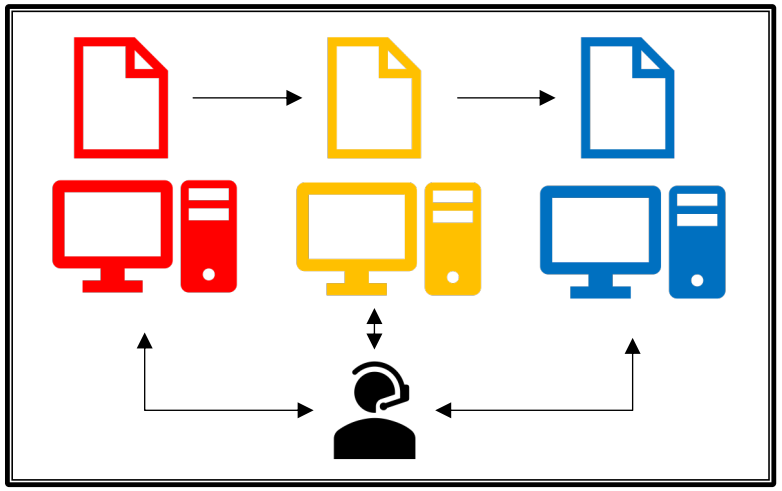
Merlin is a producer-consumer workflow system

External coordination
Independent batch jobs tied to tasks



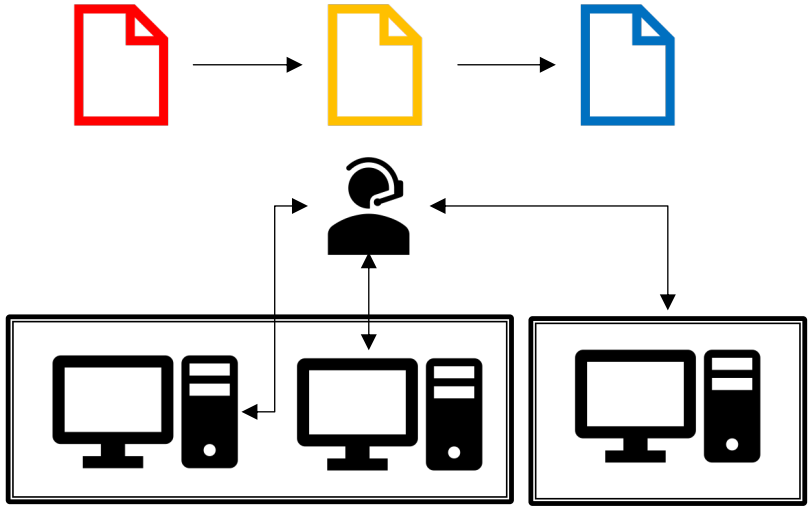
e.g. maestro
https://github.com/LLNL/maestrowf

Internal coordination
Large batch jobs for multiple tasks

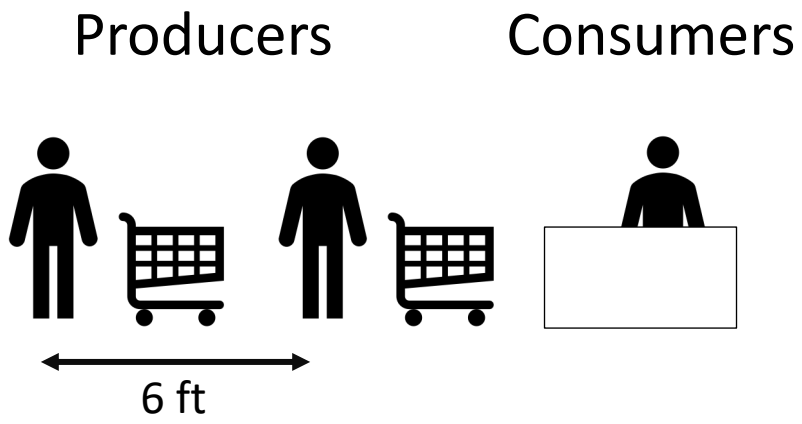


e.g. UQPipeline

Centralized task coordination
Distributed task-agnostic resources



A (useful?) analogy for a producer-consumer workflow model: grocery shopping



"Much of education is cleverly contrived not to produce understanding, but merely the ILLUSION of understanding."
- Donald Simanek, *The Dangers of Analogies*

A (useful?) analogy for a producer-consumer workflow model: grocery shopping

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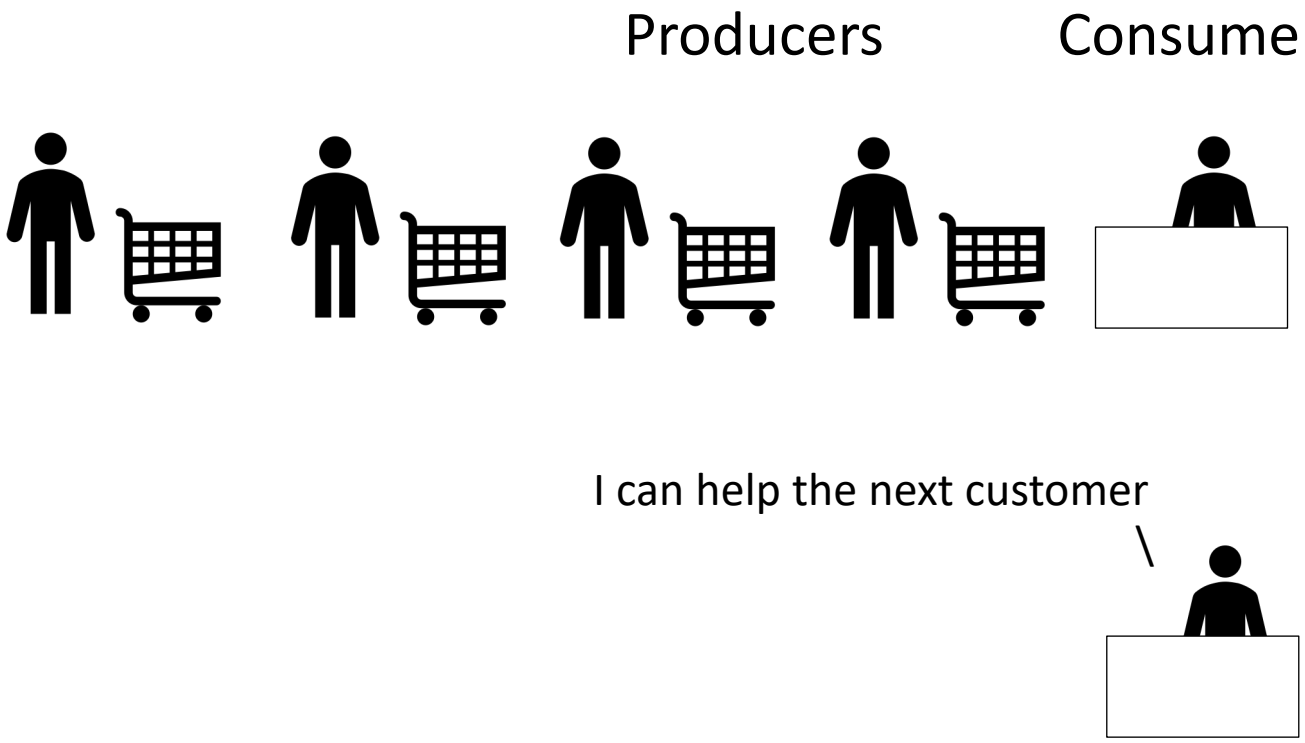

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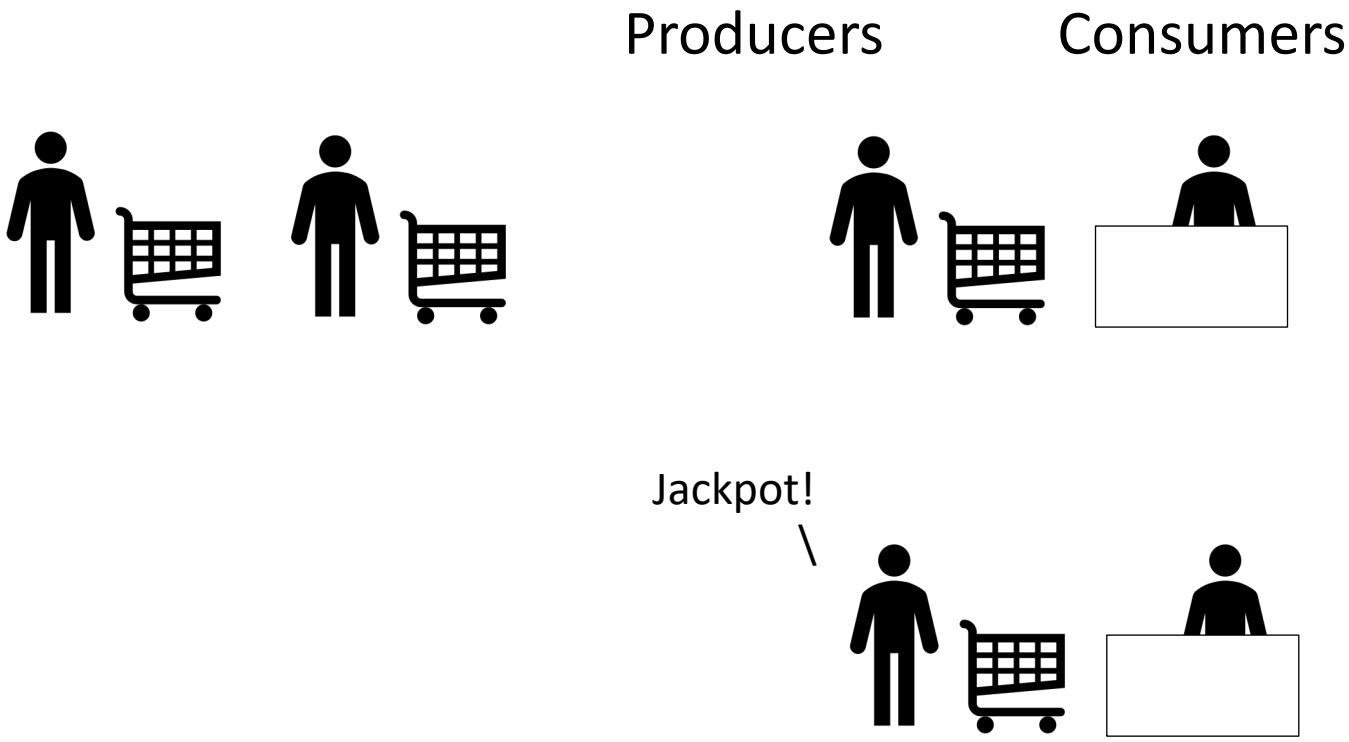

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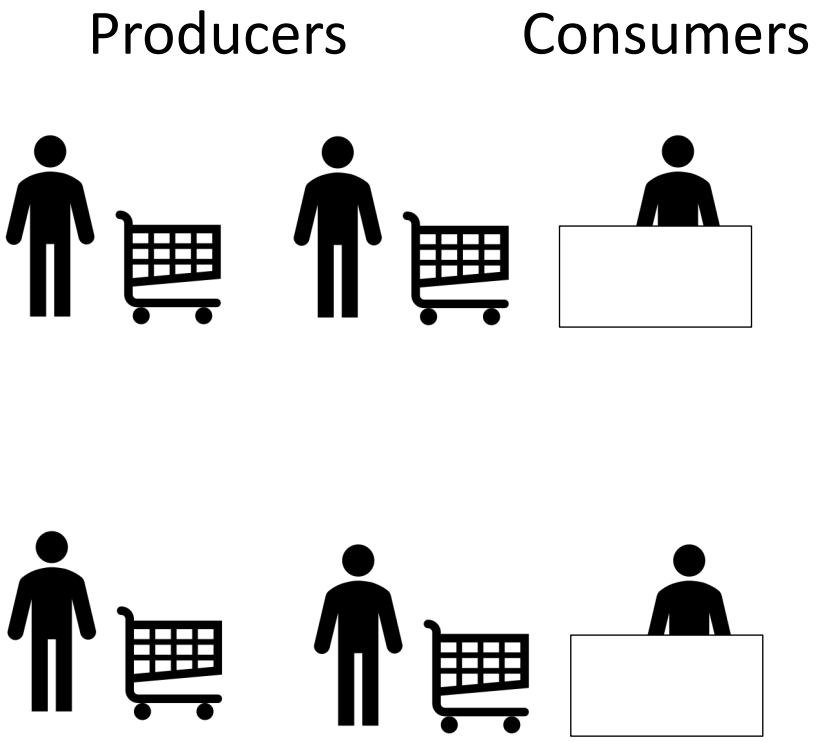
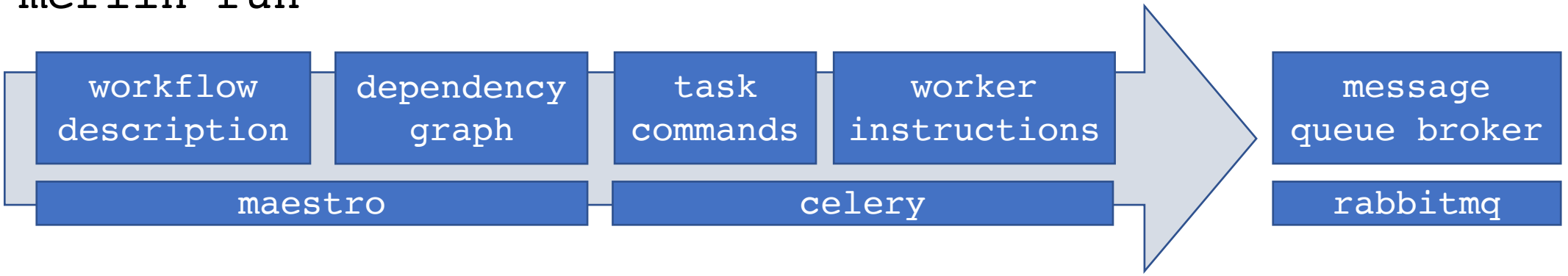
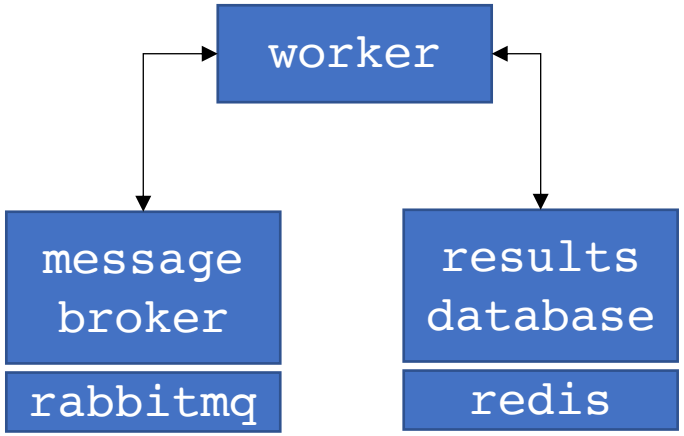


Merlin builds off and leverages 3rd party software

Producer: `merlin run`



Consumer: `merlin run-worker`



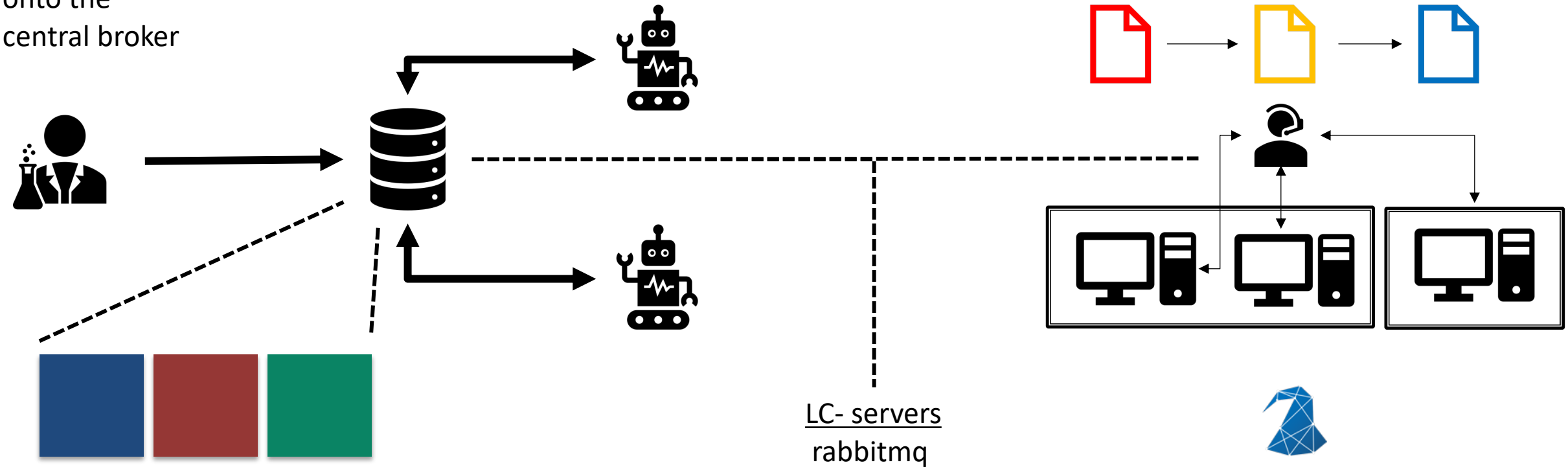
- Airflow
- DoorDash
- Instagram
- Lyft
- Mozilla
- OpenGov
- Nextdoor
- Red Hat
- Read the Docs
- Uber
- Venmo
- …

At LLNL we have centralized servers as brokers (thanks, LC!)

People can push work onto the central broker

Workers can push/pull work

Centralized task coordination
Distributed task-agnostic resources



Broker has multiple queues to sort work

LC- servers
rabbitmq
redis

At LLNL we have centralized servers as brokers (thanks, LC!)

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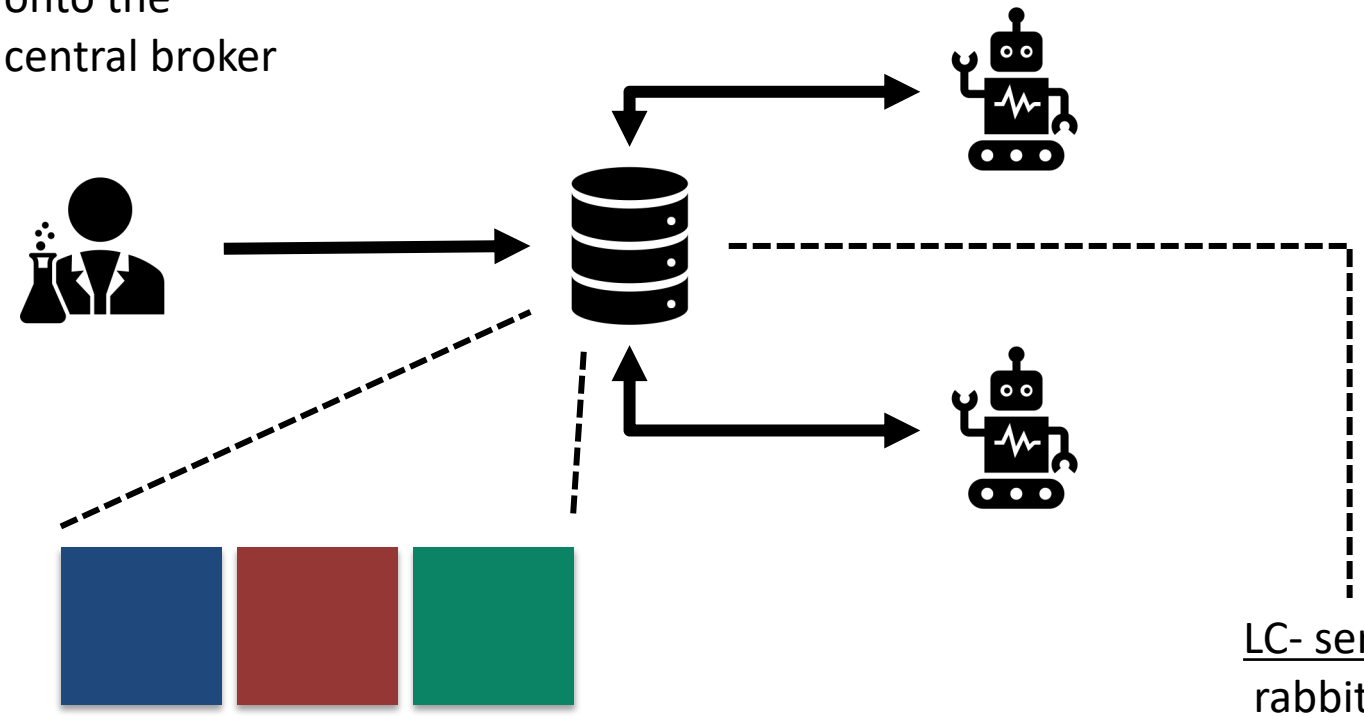


Broker has multiple queues to sort work

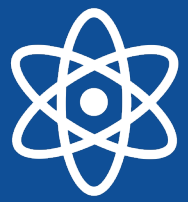
LC- servers
rabbitmq
redis

The central broker lives outside of batch jobs and is visible by all compute nodes on the respective network.

Takeaway: Multiple batch jobs and machines can work on the same work.

We worked hard w/ LC to conform to security requirements, including auto encryption of traffic and results.

Bigger takeaway: this wouldn't have been possible without LC!

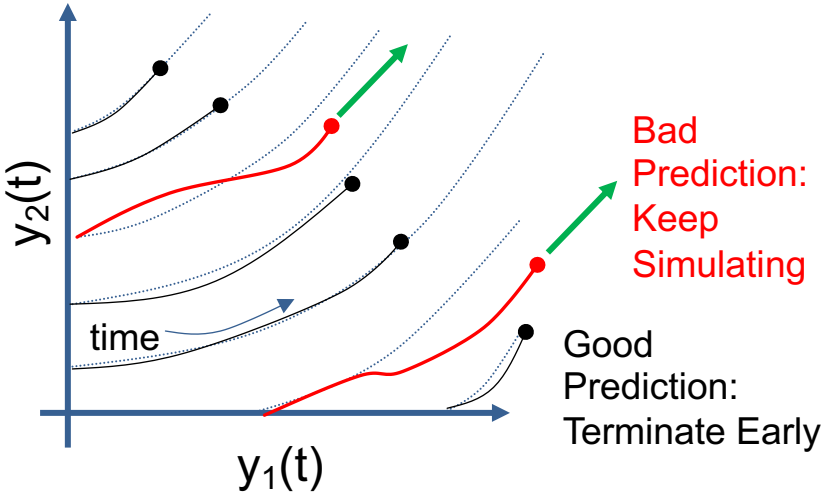


technical overview

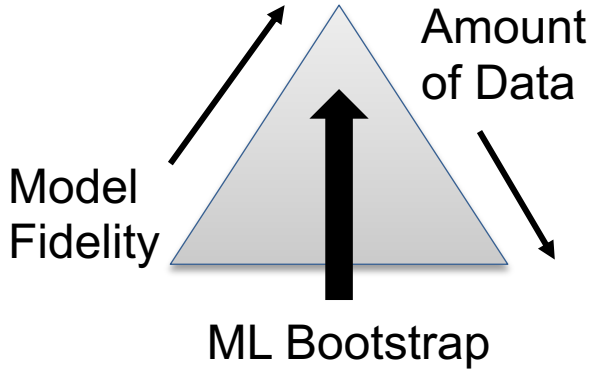
the story

Merlin builds off 5 years of exploring how to augment HPC simulation ensembles with cloud compute technologies

Intelligent Sampling via Forecasting



Multi-Fidelity Modeling via Transfer Learning



LBPM SI LDRD Begins: Is speculative sampling possible?

2015 — 2016 — 2017 — 2018 — 2019 — 2020 — Today

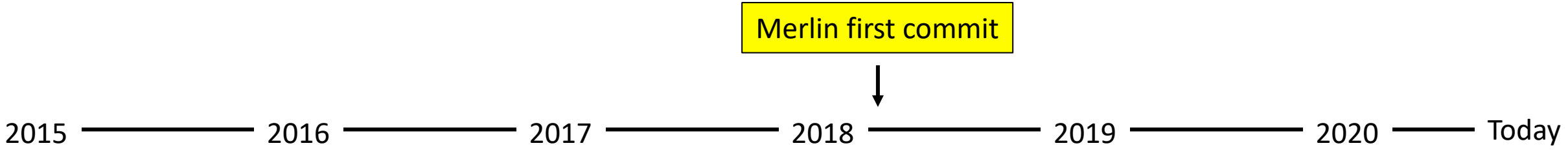
Merlin builds off 5 years of exploring how to augment HPC simulation ensembles with cloud compute technologies

The problem: we would need to be able to coordinate

- Simulations Starting, Stopping and Continuing
- Machine Learning Training
- Machine Learning Inference
- Database Injections
- Database Queries
- Launching New Simulations
- And Operating at Scale b/c deep learning eats data

Team: Could celery work?

We'd need a lot of new infrastructure at LC, but let's try! Why not?



Merlin builds off 5 years of exploring how to augment HPC simulation ensembles with cloud compute technologies

- Tested Infrastructure at Scale
  - Hierarchical Ensemble Tasking
  - Flux for HPC launch
  - Data broker and backend
  - Error recovery
  - Auto cross-machine data-bounce
- 100 Million ICF simulations
- 4.8 Billion Images
- Massive Scientific ML Training Set

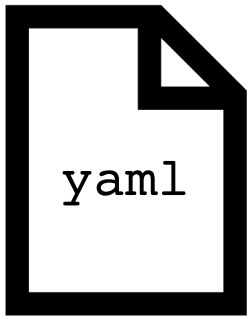
Sierra 100M Run

2015 — 2016 — 2017 — 2018 — 2019 — 2020 — Today

10k sims released to public: https://data-science.llnl.gov/open-data-initiative

Merlin builds off 5 years of exploring how to augment HPC simulation ensembles with cloud compute technologies

- Redesign to use maestro as a library
- Can read and launch any workflow described by maestro
- Opened space of supported workflows
- Expanded upon syntax: control logic, variables, broker-queue



System Redesign: maestro front end

2015 — 2016 — 2017 — 2018 — 2019 — 2020 — Today

Merlin builds off 5 years of exploring how to augment HPC simulation ensembles with cloud compute technologies

`pip install merlin`

`spack install py-merlin`

Open source release: move to GitHub, pypi, spack

2015 — 2016 — 2017 — 2018 — 2019 — 2020 — Today

Merlin builds off 5 years of exploring how to augment HPC simulation ensembles with cloud compute technologies

Merlin

Machine learning for HPC workflows

Watch 5

Navigation

Tutorial
- 0. Before you start
- 1. Introduction
- 2. Installation
- 3. Hello, World!
- 4. Run a Real Simulation
- 5. Advanced Topics
- 6. Contribute to Merlin
- 7. Port Your Own Application

Getting Started
FAQ
Command line

Tutorial

Estimated time:
- 3 hours

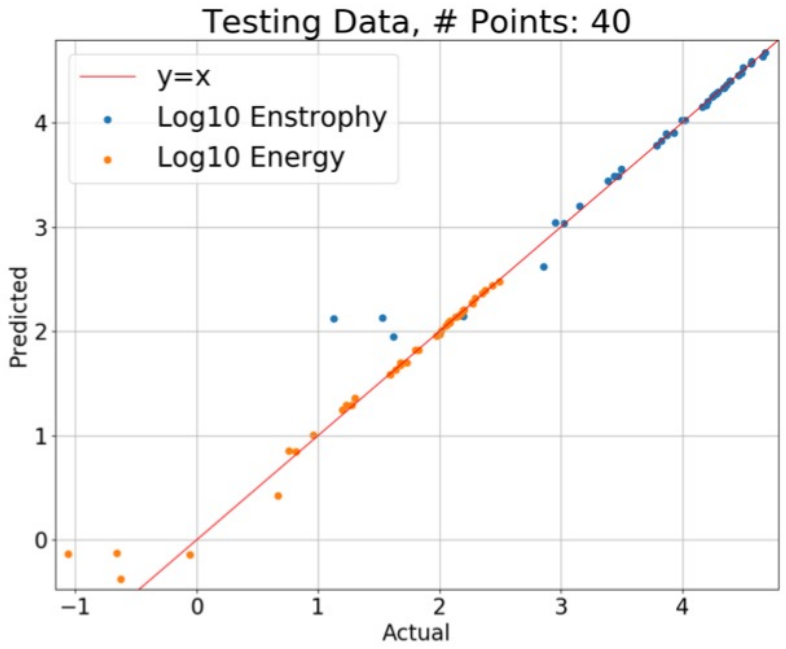
Grab your laptop and coffee, and dive into this 7-module tutorial to become a Merlin expert.

This hands-on tutorial introduces Merlin through some example workflows. In it, you will install Merlin on your local machine, stand up a virtual server and run both a simple workflow and a quasi-real-life physicsy simulation that couples a physics application with visualization and machine learning.

You'll also learn how to use some advanced features and help make Merlin better. Finally we offer some tips and tricks for porting and scaling up your application.

Before you come:

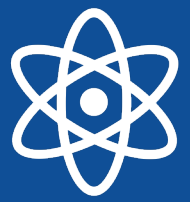
- 0. Before you start



~~WoWoHa~~ Tutorial

2015 — 2016 — 2017 — 2018 — 2019 — 2020 — Today

https://merlin.readthedocs.io/en/latest/tutorial.html



technical overview

tech details

the algorithm

performance

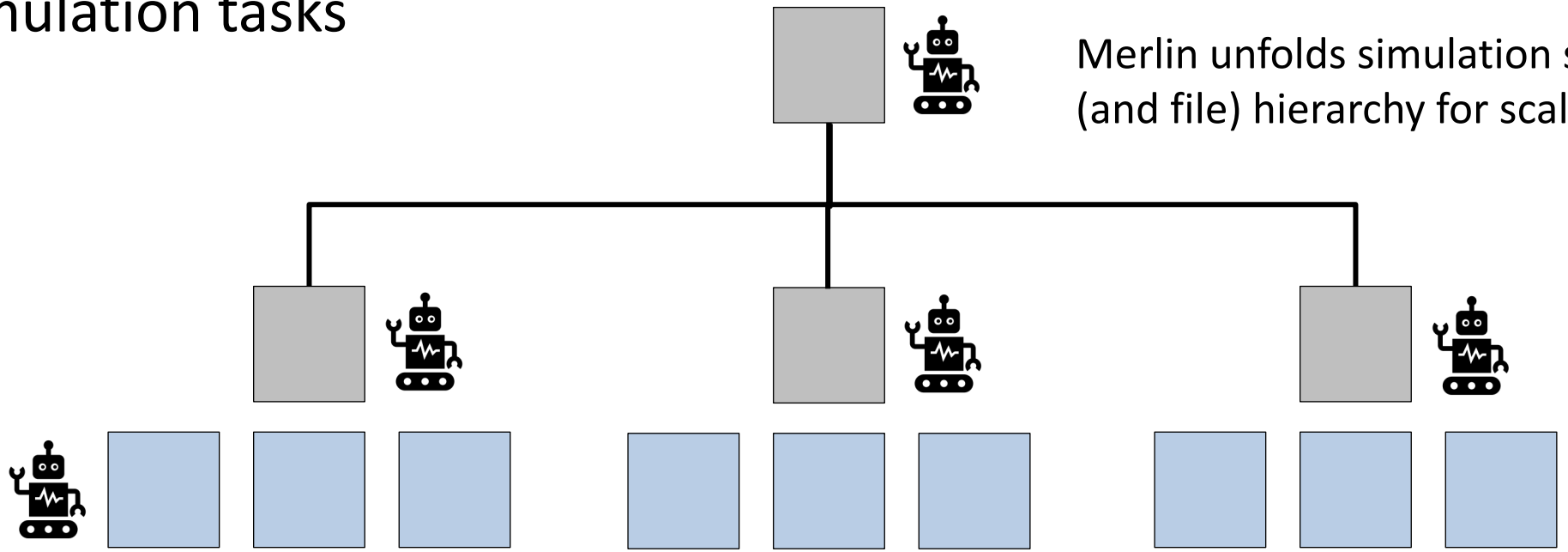
The heart of Merlin is the hierarchical task creation algorithm

task-creation tasks

simulation tasks

In a producer-consumer model, workers can create tasks that create tasks

Merlin unfolds simulation samples into a task (and file) hierarchy for scalability



Workflow creation scales with worker count, significantly reducing time-to-first simulation

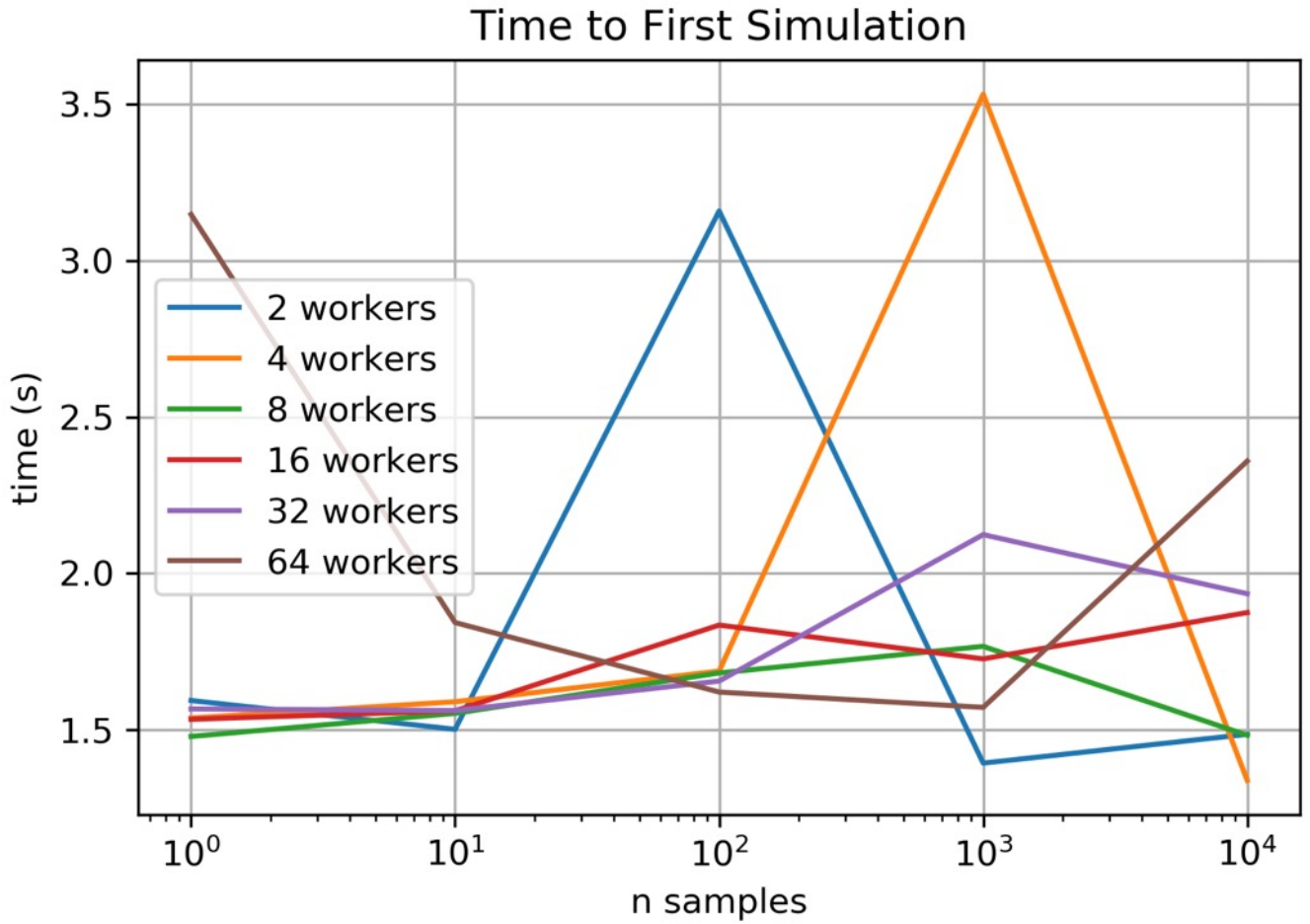
Launching new simulations is very fast, essentially non-blocking

← 300,000 tasks / second

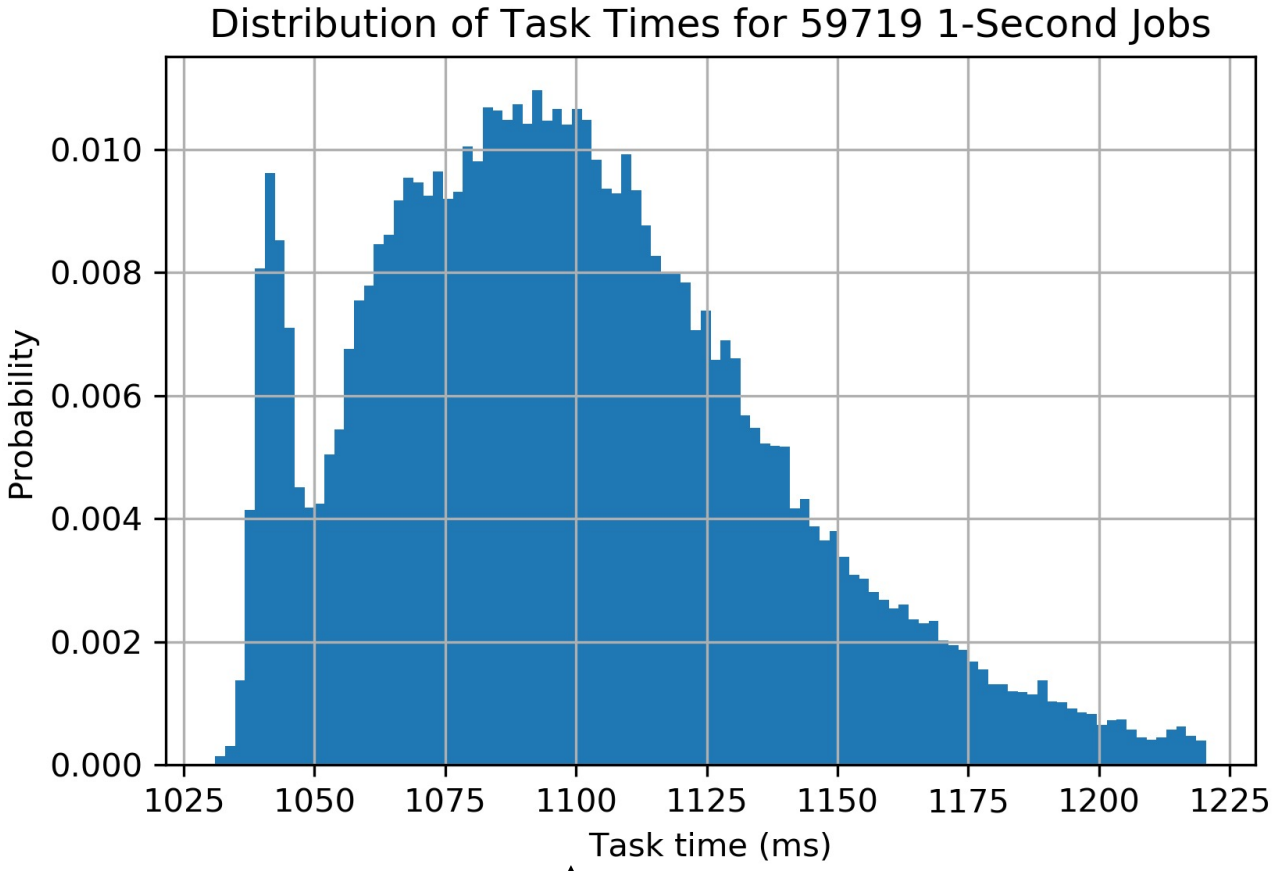
← 10 million samples

Why is it so fast? We only encode the metadata needed to unroll the task hierarchy by the workers

Time-to-first simulation is a few seconds

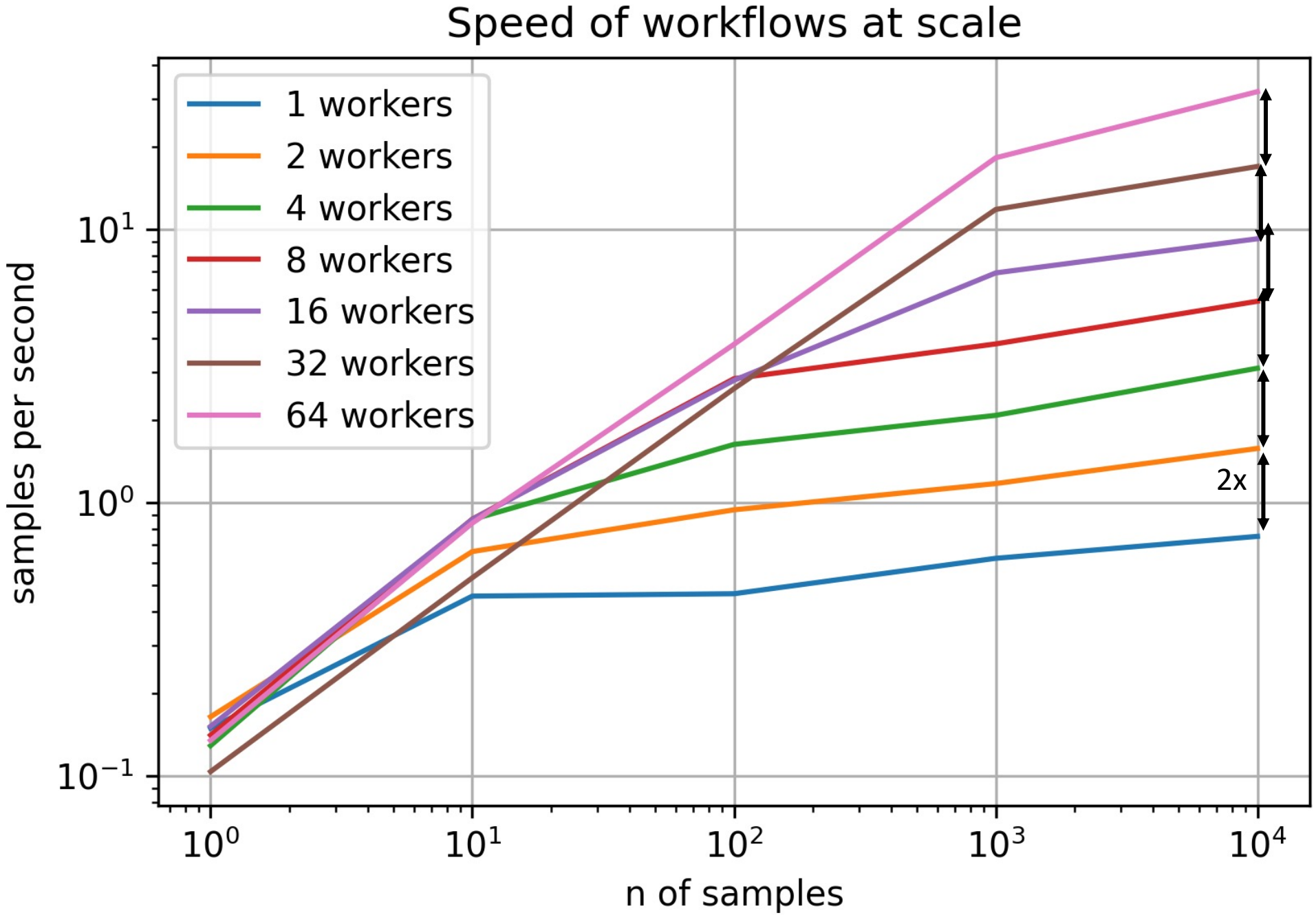


Per-simulation overhead is ~ 100ms



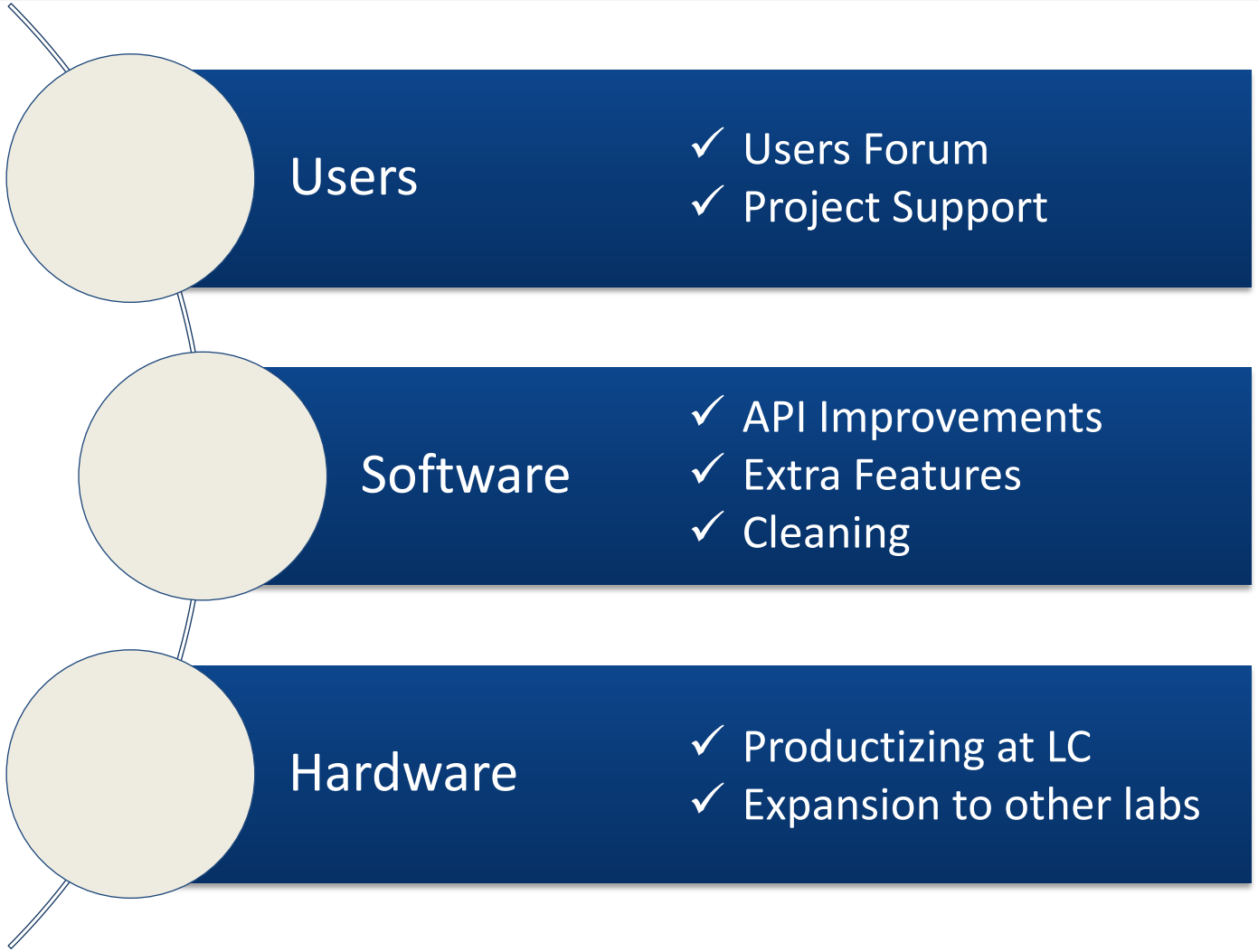
60k “sleep 1” tasks, 1000 ms/task

Overhead is ~100 ms / task

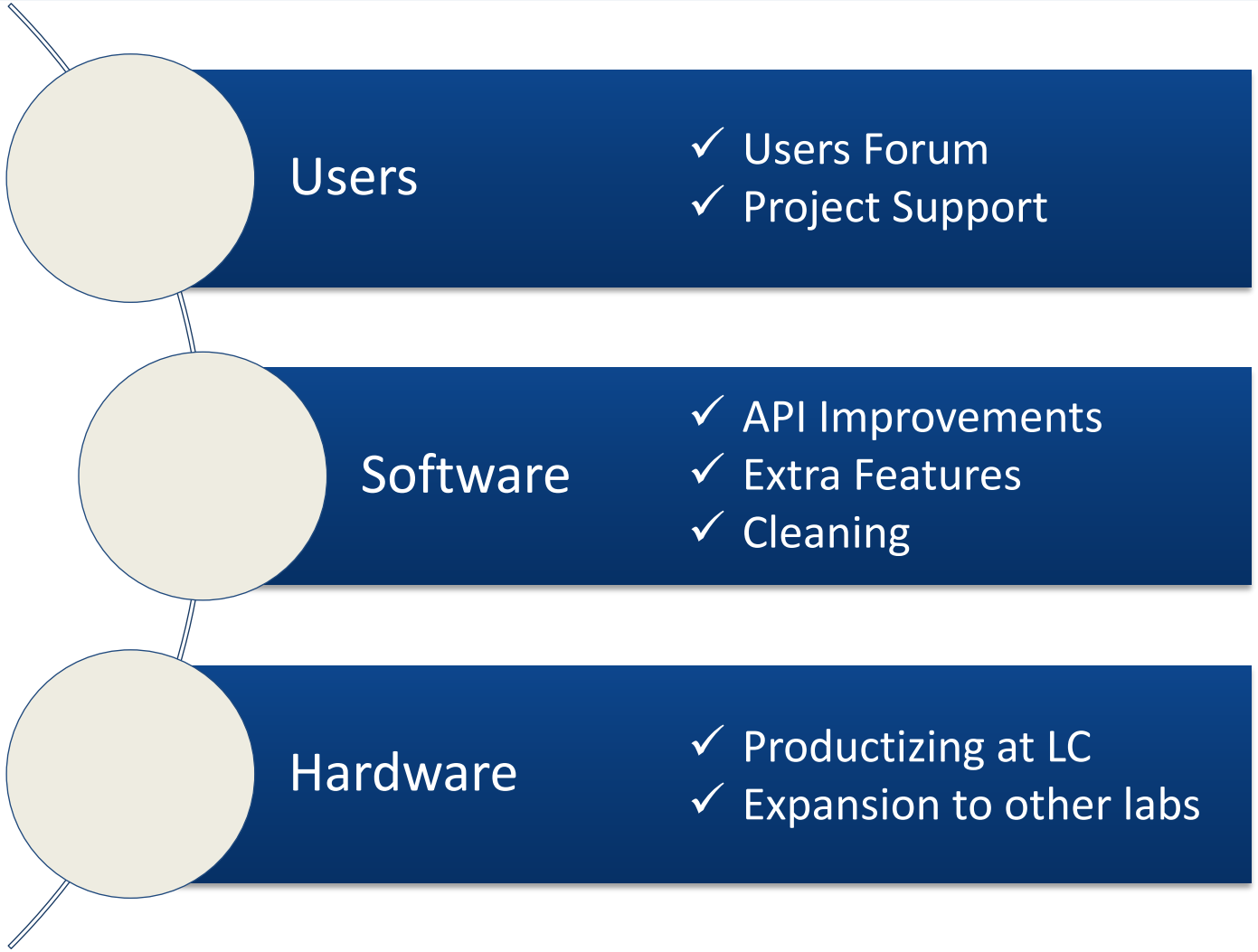

Speed of workflows at scale
1 workers
2 workers
4 workers
8 workers
16 workers
32 workers
64 workers
2x
samples per second
n of samples
10^{-1}
10^{0}
10^{1}
10^{0}
10^{1}
10^{2}
10^{3}
10^{4}

Current Direction

Development Endgame



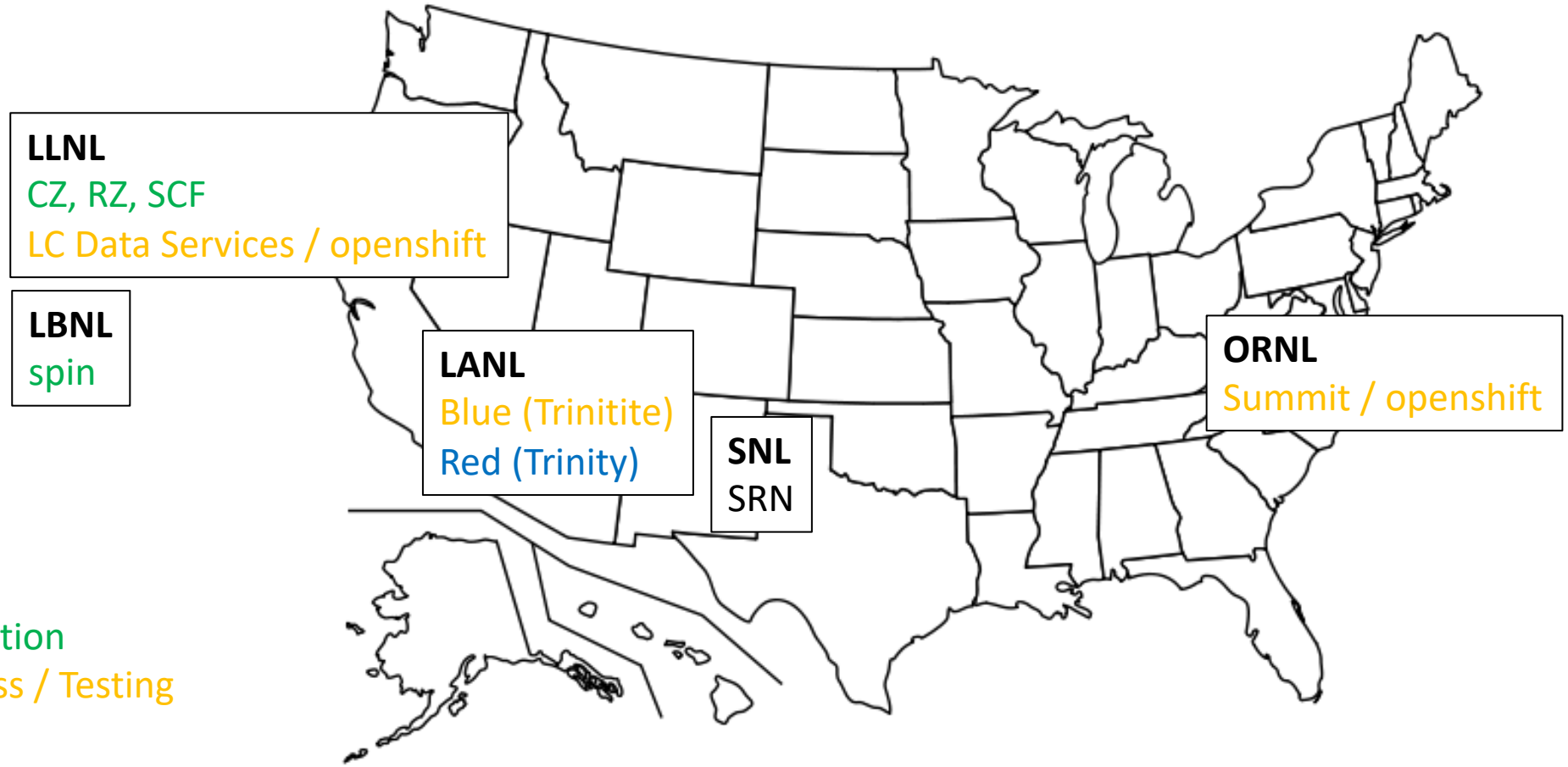
Development Endgame



1. Support for latest version of Maestro
2. Upgrade status command with task results
3. Add auto worker configuration

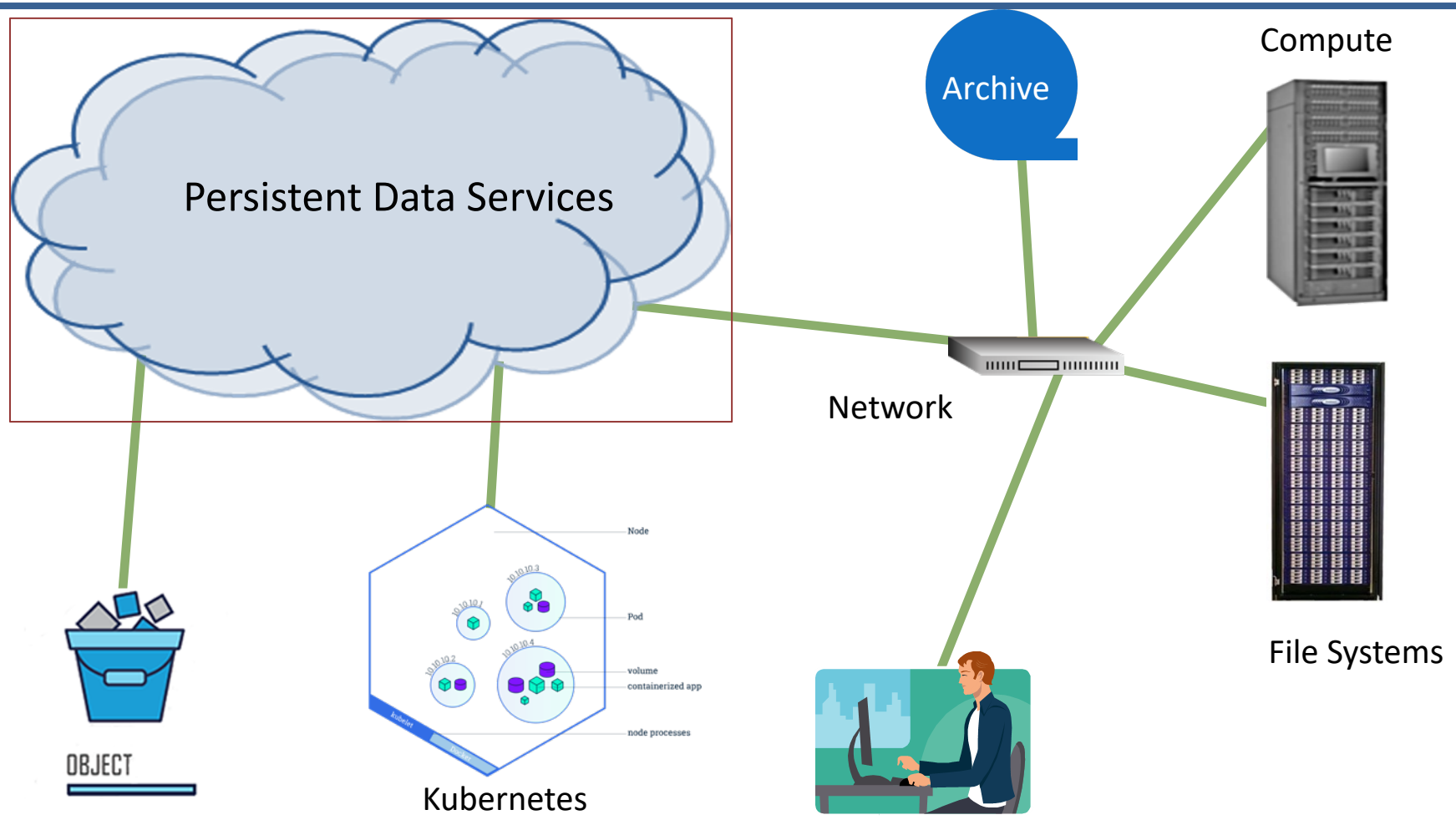
Future Direction

Status of Merlin Infrastructure @ Other DOE Sites



In Production
In Progress / Testing
Planned
Expressed Interest

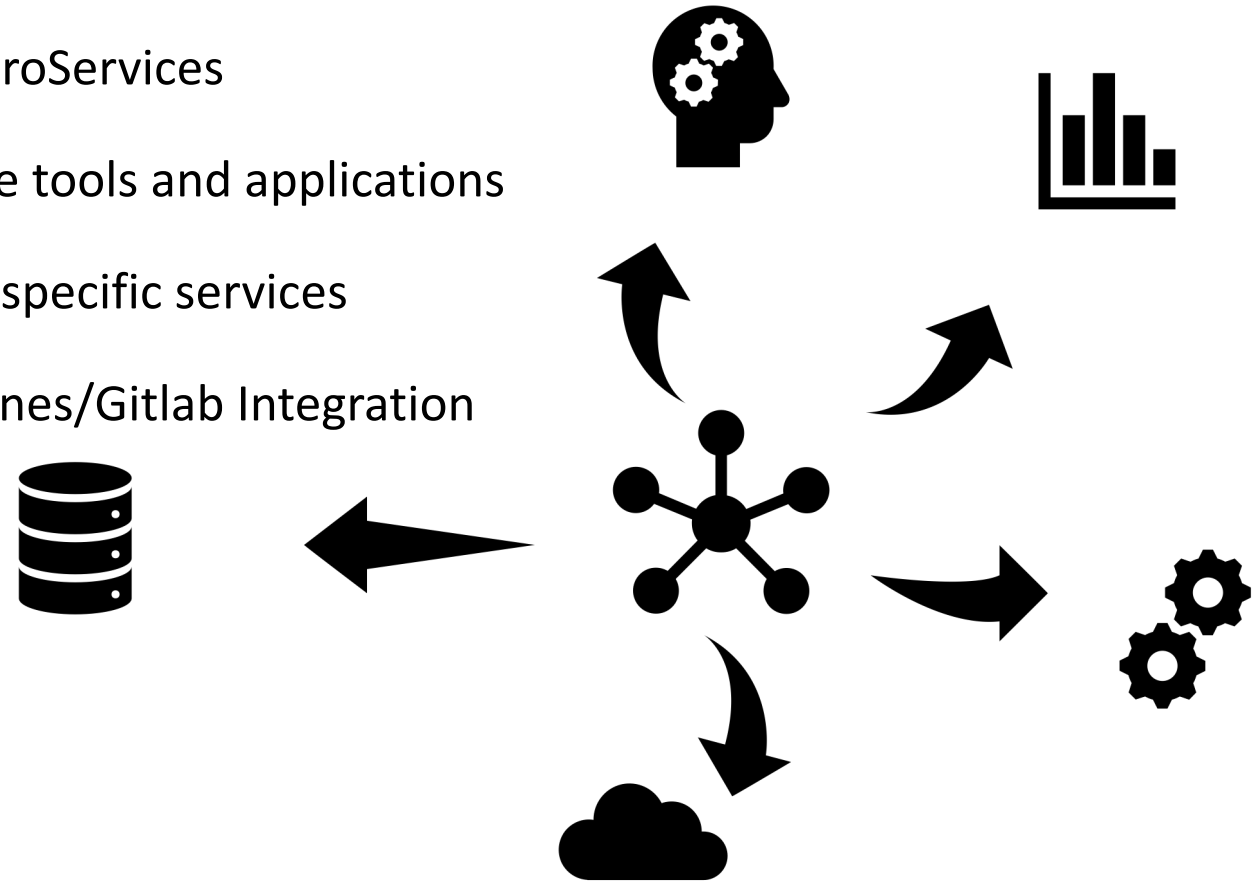
LC plans to productize this: A Persistent Data Services Platform



Coming Soon @ LC

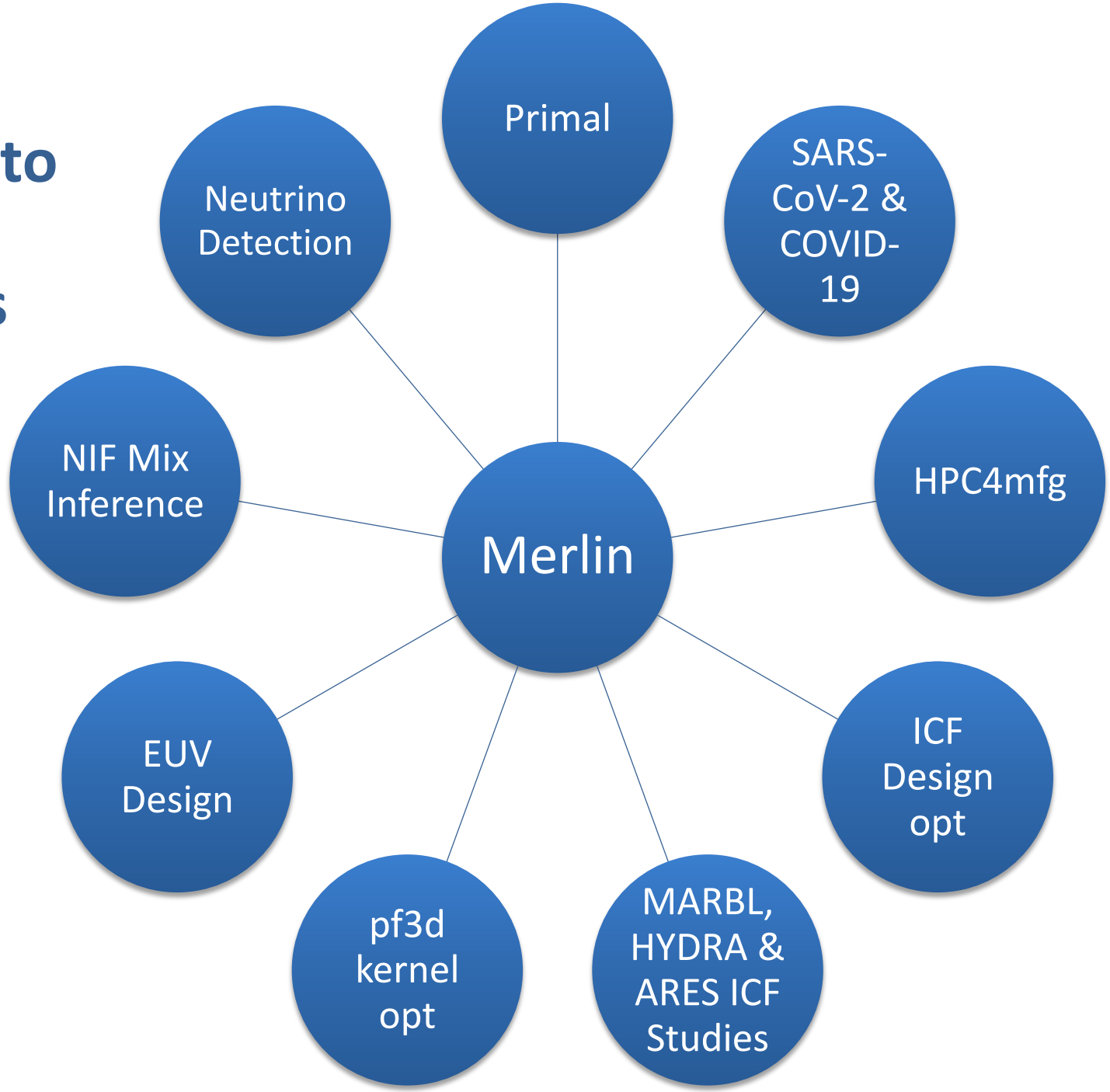
A Persistent Data Services Platform: Advanced Deployments

- Suite of MicroServices
- Data Science tools and applications
- Application specific services
- CI/CD pipelines/Gitlab Integration



Coming Soon @ LC

From LDRD investment to program deliverables



Want to learn more?

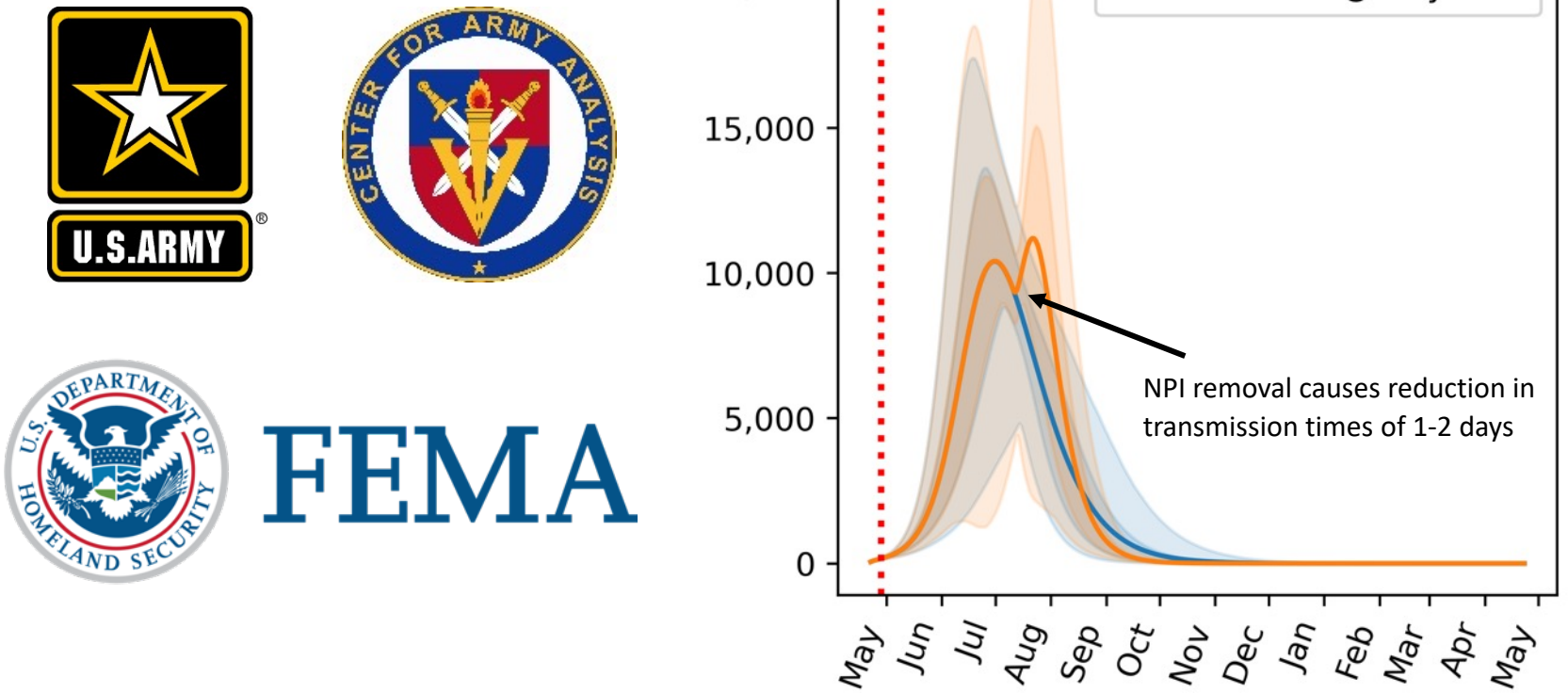
- Full Documentation: https://merlin.readthedocs.io/
- ArXiv article: **Enabling Machine Learning-Ready HPC Ensembles with Merlin**
- Tutorial: https://merlin.readthedocs.io/en/latest/tutorial.html
- GitHub: https://github.com/llnl/merlin
- `pip install merlin`
- `spack install py-merlin`

Thank you!

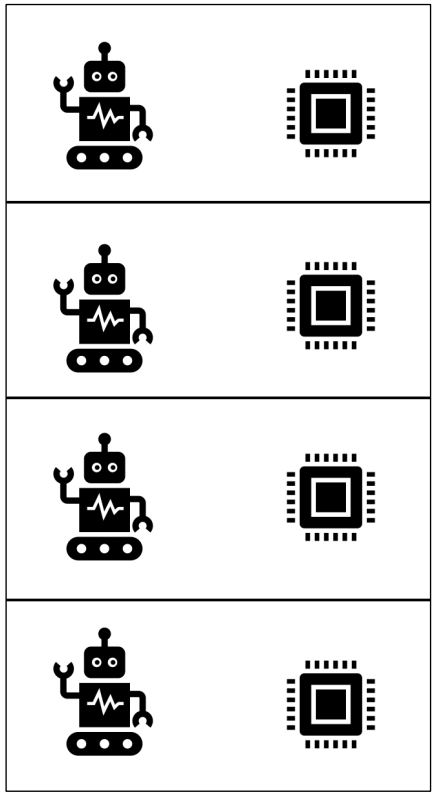
github.com/llnl/merlin

Lawrence Livermore National Laboratory

We are using Merlin for large-scale COVID19 Scenario Modeling



4 workers per Lassen Node
1/GPU, each running tensorflow

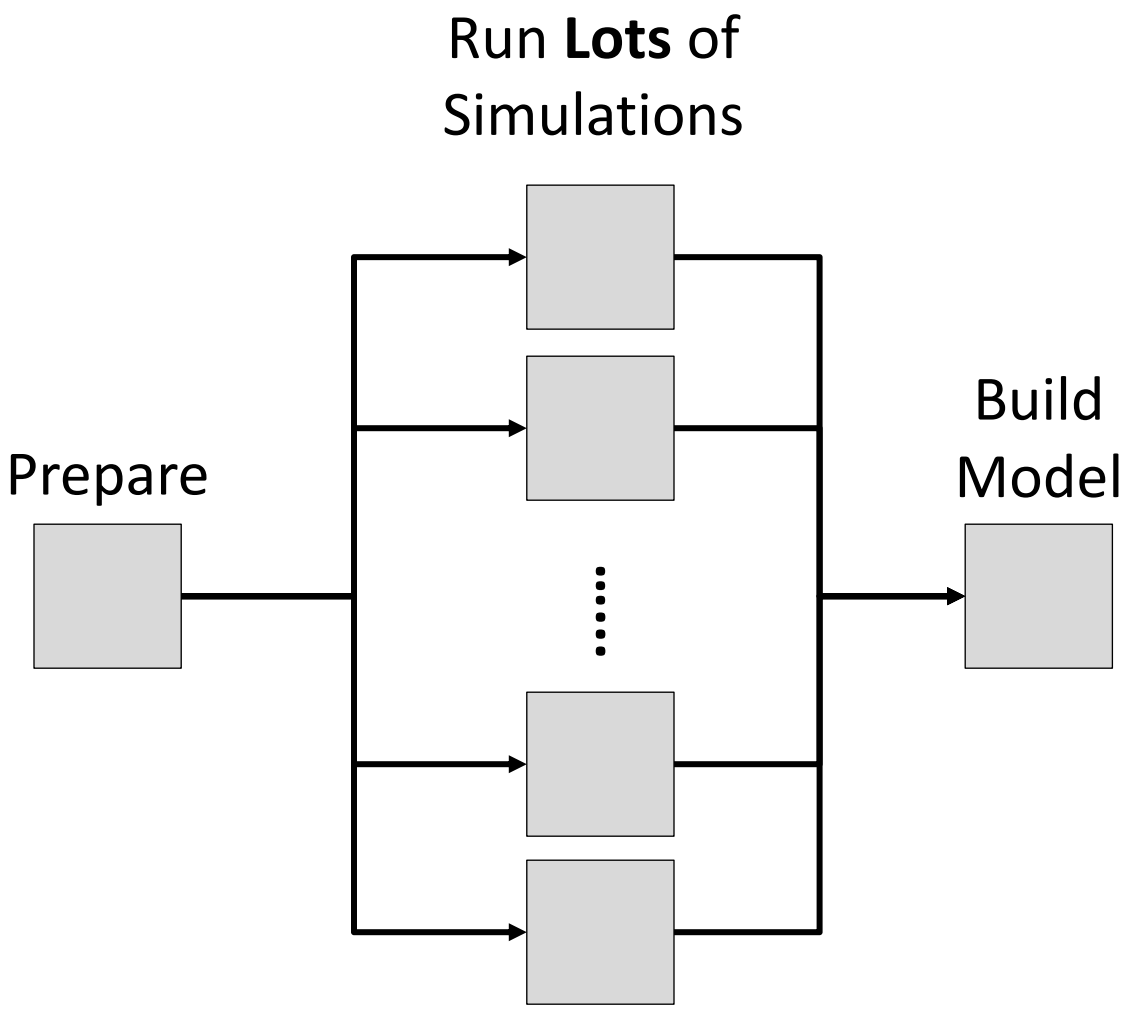


With flux we can model one scenario w/ UQ for the entire country in ~5 minutes on a few nodes: near real-time feedback

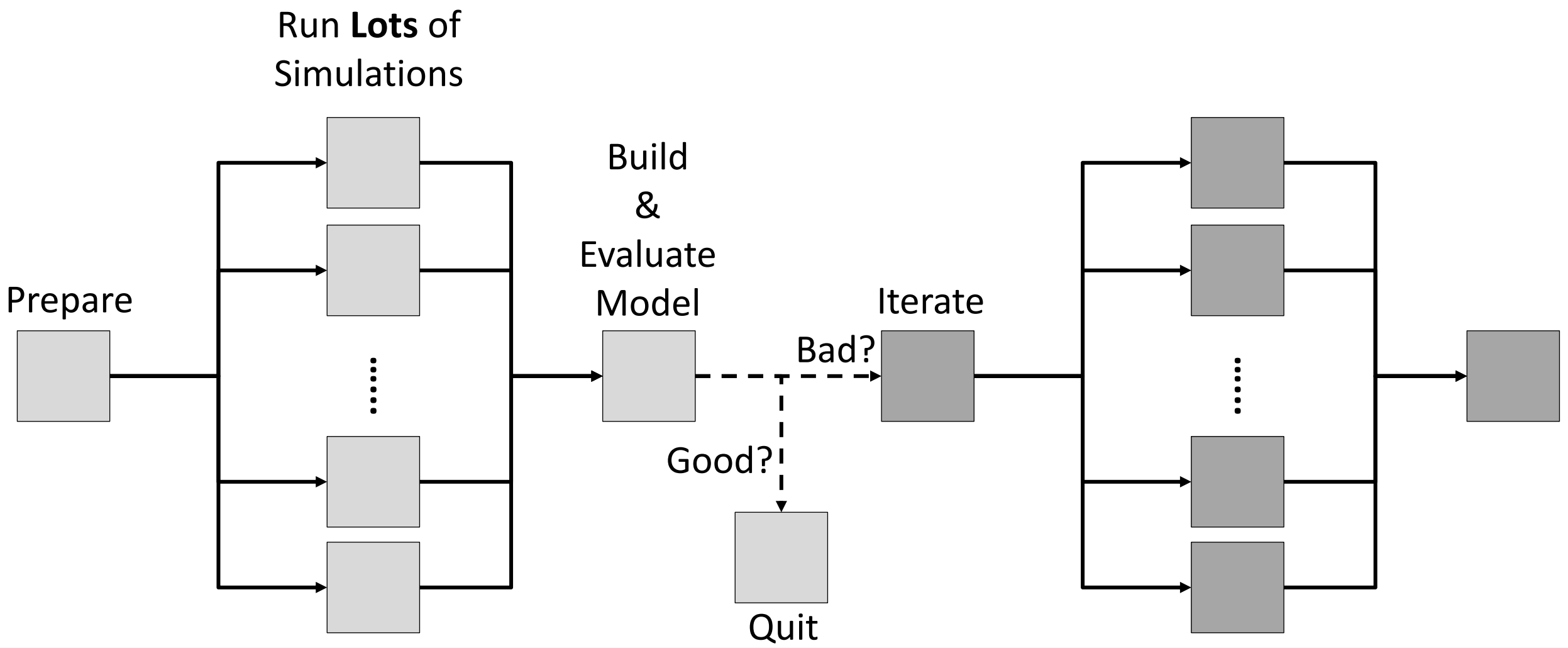
What can you do with Merlin?

What use cases did we target?

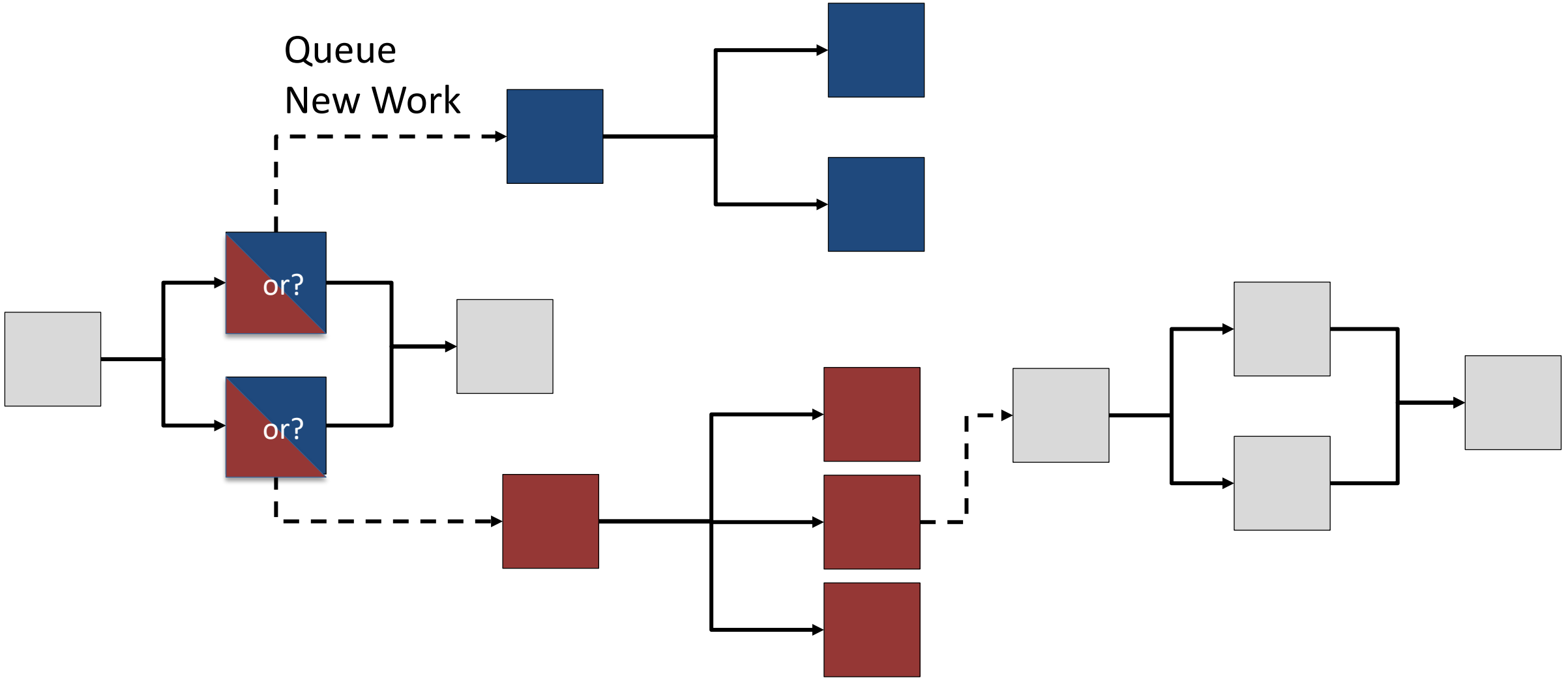
Surrogate Model Building



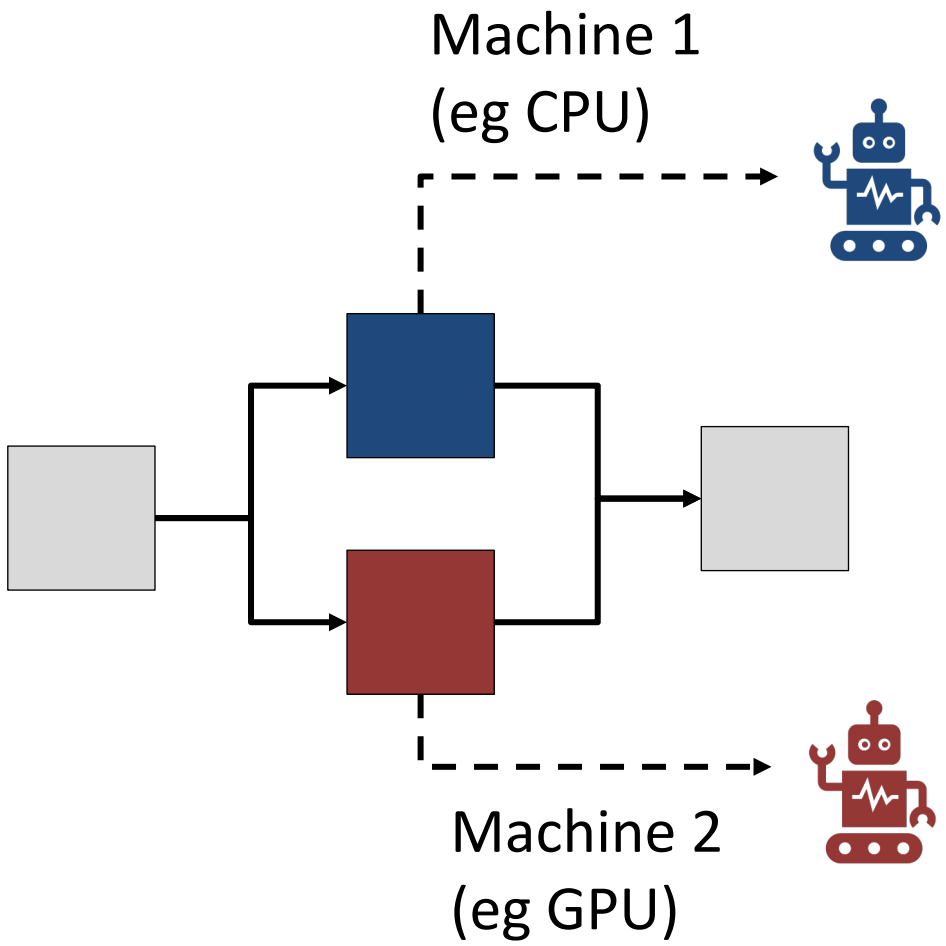
Active Learning



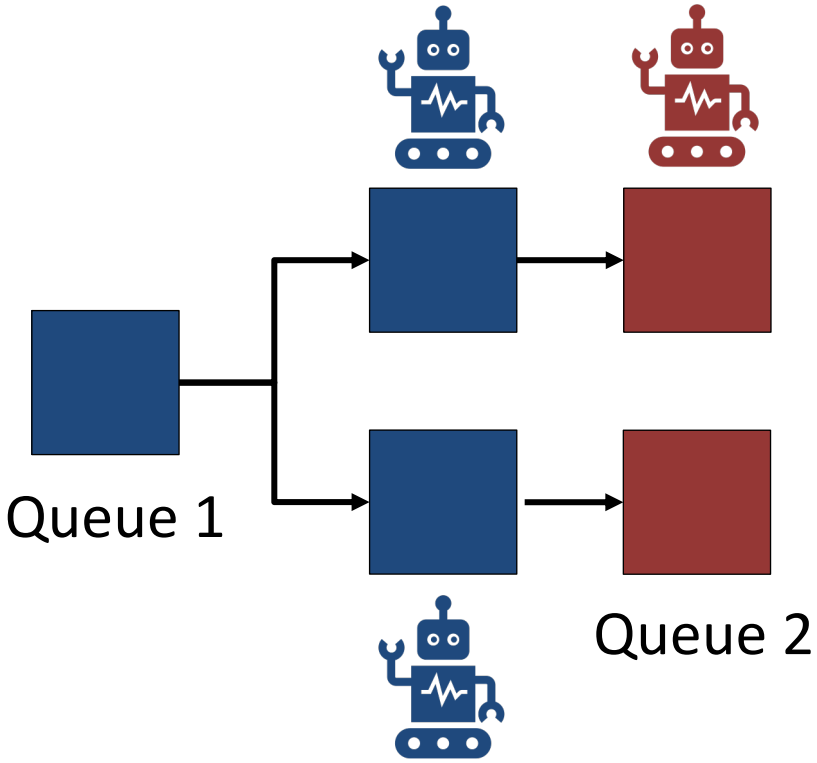
Dynamic Workflow Branching and Launching



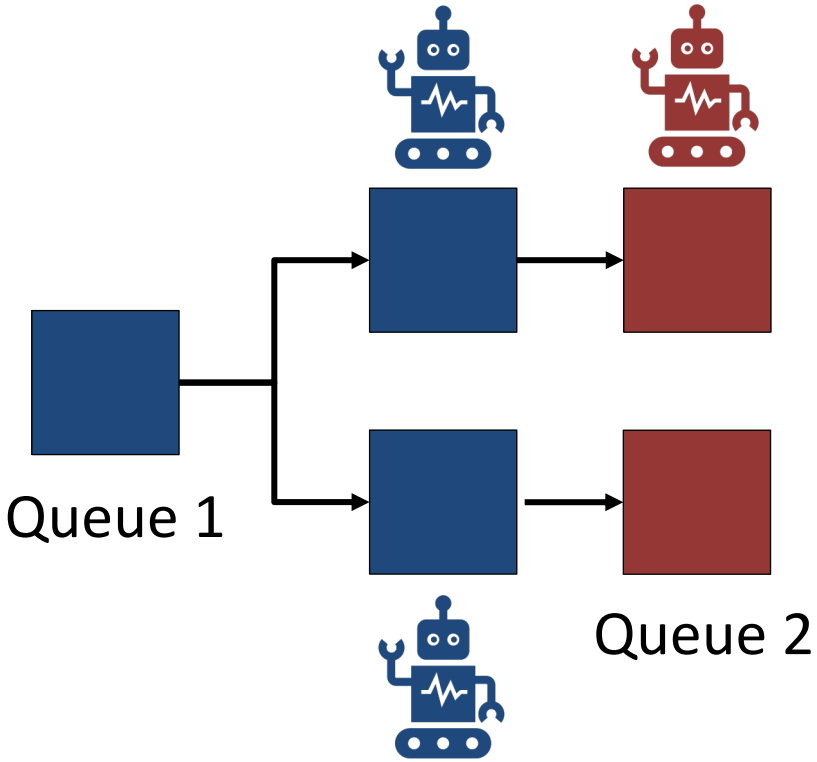
Heterogenous Workflows

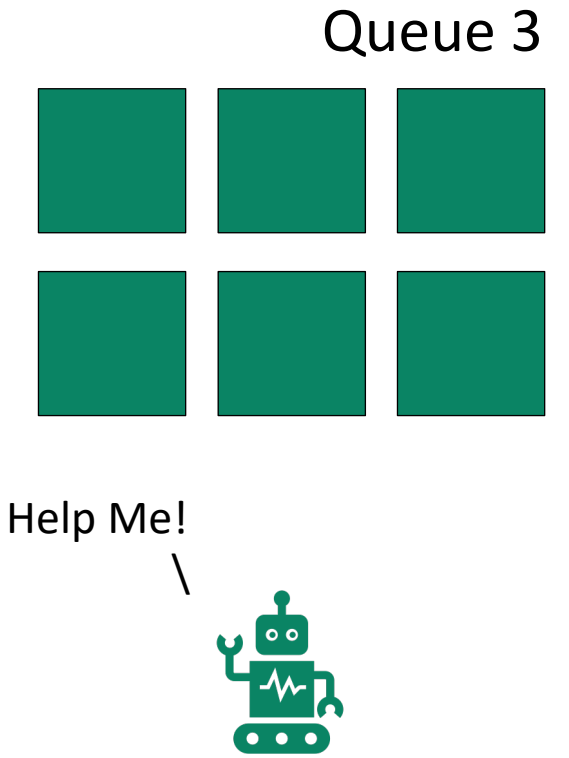


Task Prioritization

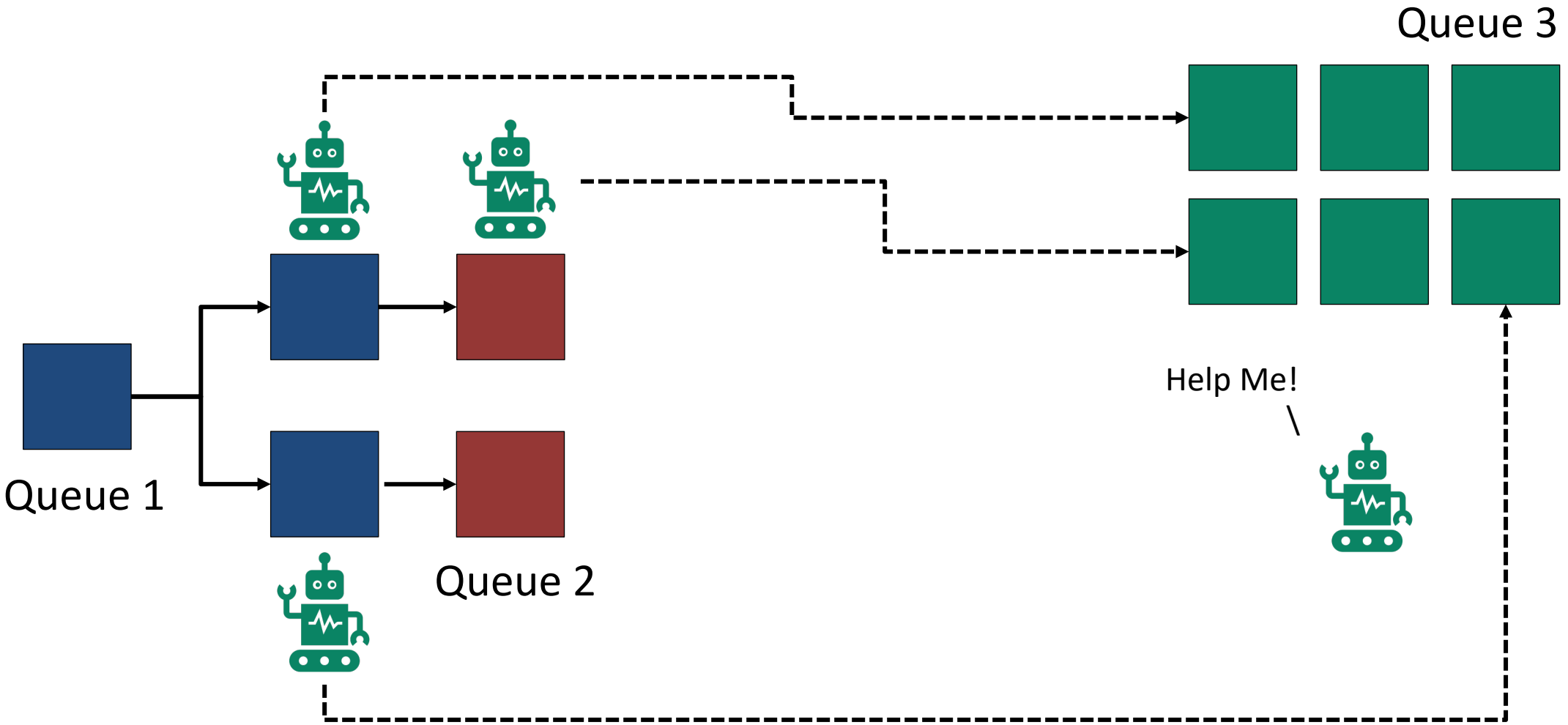


Task (Re)Prioritization





Task (Re)Prioritization



Task (Re)Prioritization