Argonne NATIONAL LABORATORY

Academy: Empowering Scientific Workflows with Federated Agents

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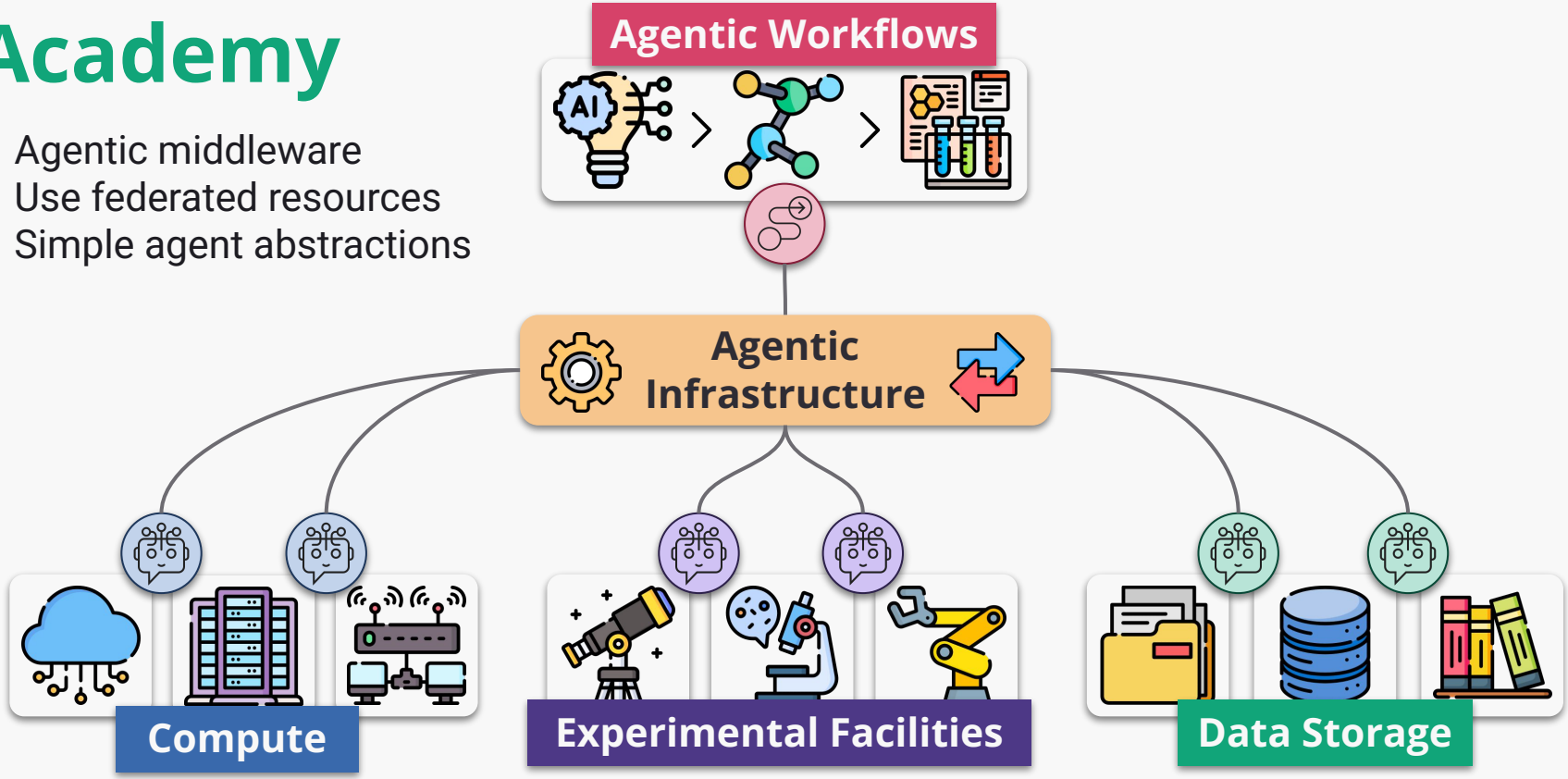
JLESC: AI for Science Breakout

Academy

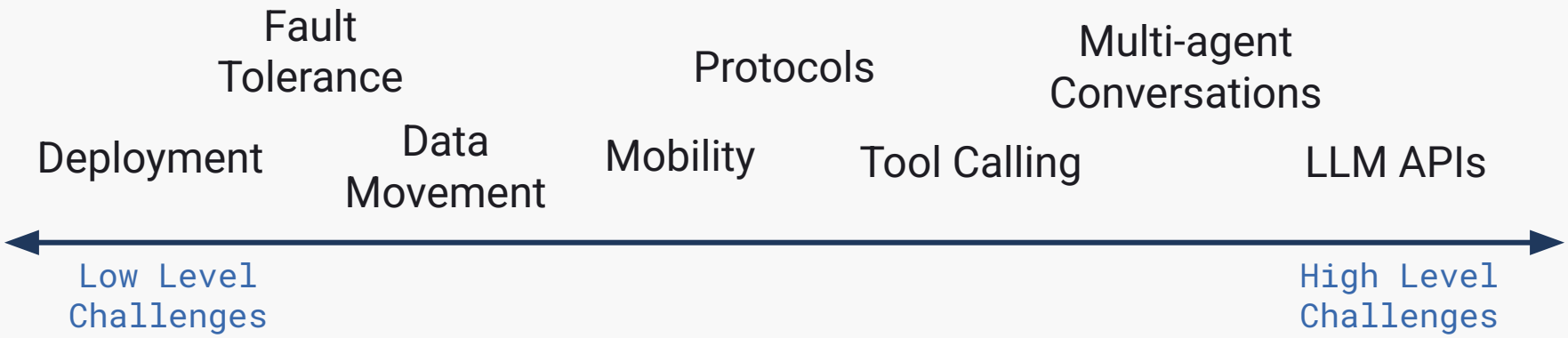
- Agentic middleware
- Use federated resources
- Simple agent abstractions



Agentic Middleware

Software layer that transparently manages the lifecycle, communication, and coordination of autonomous agents across distributed computing environments.

Agentic Middleware: Scope & Challenges



Academy → **Cool Science?** ← **LangChain, AutoGen, Pydantic AI, etc.**

Agentic Middleware: Using Research Infrastructure

Centralized

- Agents co-located (workstation, cloud)
- Research infrastructure available via APIs (REST, SDKs, …)
- Use infrastructure via tool calling

++ Rapidly growing library ecosystem
−− Limited APIs for infrastructure

LangChain, AutoGen, Pydantic AI, etc.

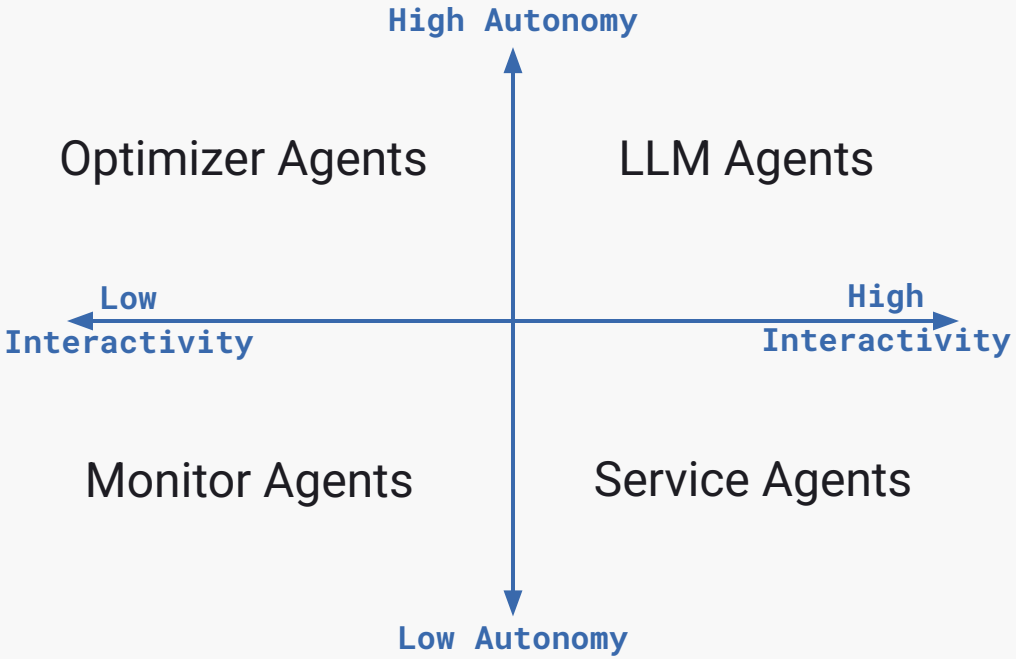
Decentralized

- Agents distributed across infrastructure
- Agents interact asynchronously
- Use infrastructure directly (actuate a robot, submit job, …)

++ Data locality, perf., loose coupling
−− Deployment complexity

Academy

Agentic Middleware: Agent Behaviors



Other defining aspects:

- Persistent vs ephemeral
- General vs narrow purpose
- Embodiment

Long-running agentic science apps will incorporate many kinds of agent behaviors.

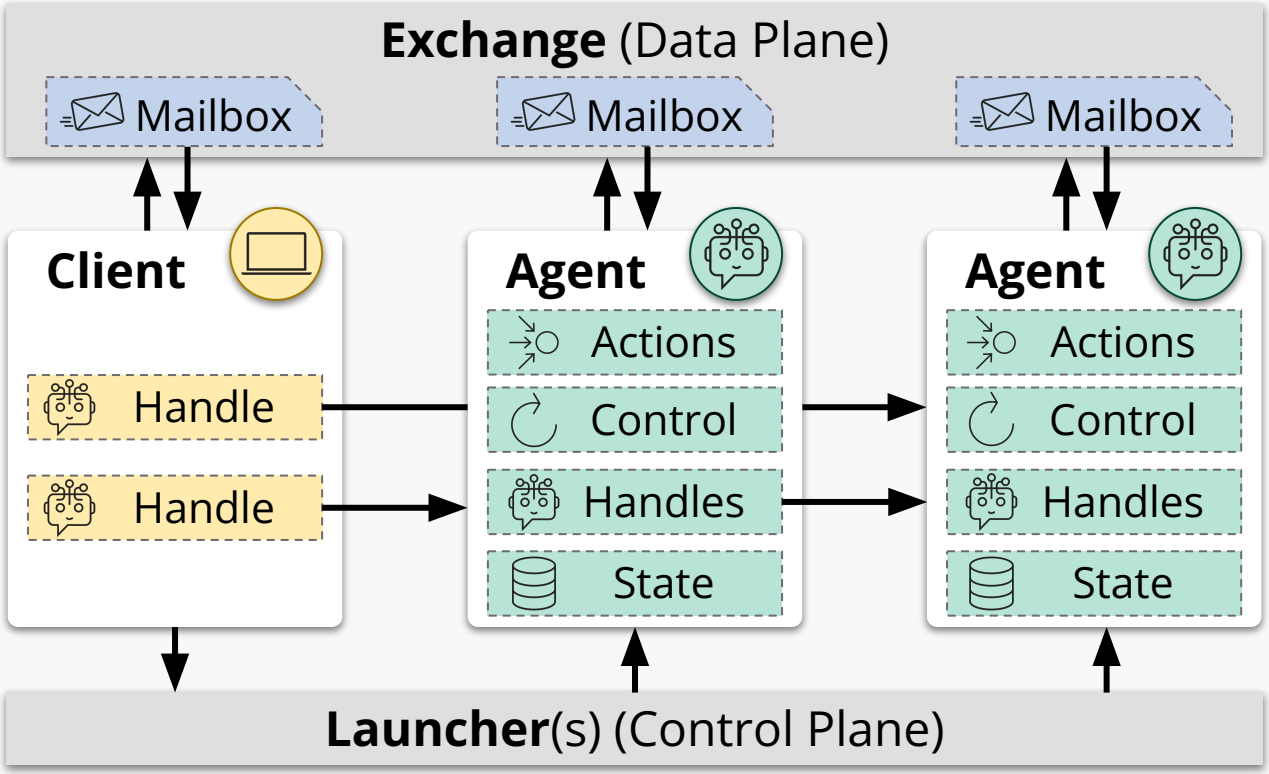
Academy primitives support the creation diverse agent types.

How does **Academy** support the expression of **diverse agent behaviors** and deployment across **distributed/federated resources**?

Focus 3: Coordinate async agent messaging

Focus 1: Program diverse agents and interactions

Focus 2: Deploy agents on federated resources



https://academy.proxystore.dev/latest/concepts/

Communication & Execution

Exchange

- ➔ Asynchronous communication through mailboxes
- ➔ Every agent/client in system has a unique mailbox
- ➔ Local & distributed implementations
- ➔ Optimized for low-latency
- ➔ Hybrid communication model
- ➔ Prefer direct communication between agents when possible; fall back to indirect communication via object store
- ➔ Pass-by-reference with ProxyStore for large data

Launcher

- ➔ Not required but enables remote execution of agents
- ➔ Returns handle to launched agent
- ➔ Local threads or processes
- ➔ Distributed with Parsl
- ➔ Federated with Globus Compute

Writing Apps in Academy

Agents defined by a behavior

Clients & other agents can request actions

```python
import time, threading
from academy.behavior import Behavior, action, loop

class Example(Behavior):
    def __init__(self) -> None:
        self.count = 0  # State stored as attributes

    @action
    def square(self, value: float) -> float:
        return value**2

    @loop
    def count(self, shutdown: threading.Event) -> None:
        while not shutdown.is_set():
            self.count += 1
            time.sleep(1)
```

Instance of a behavior is state

Control loops for autonomous behavior

https://academy.proxystore.dev/latest/get-started/

Single interface for managing your agents

Launch agent and get handle

Interact with agents via handles

```python
from academy.exchange.hybrid import HybridExchange
...
from academy.manager import Manager

gce = GlobusComputeExecutor('<UUID>')

with Manager(
    exchange=HybridExchange('localhost', 6379),
    launcher=ExecutorLauncher(gce),
) as manager:
    behavior = Example()  # From the prior slide
    handle = manager.launch(behavior)

    future = handle.square(2)
    assert future.result() == 4

    handle.shutdown()  # Or via the manager
    manager.shutdown(handle.agent_id, blocking=True)
```

Launch agents via Globus Compute

Pass handles to other agents

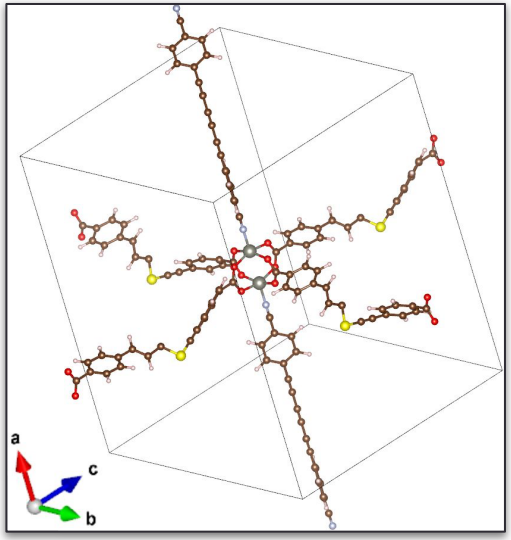
https://academy.proxystore.dev/latest/get-started/

Use Case: MOF Discovery

Metal Organic Frameworks (MOF)

- ➔ Composed of organic molecules (ligands) and inorganic metals (nodes)
- ➔ The sponges of materials science!
- ➔ Porous structures that adsorb and store gases
- ➔ Topologies can be optimized for targeted gas storage → **Carbon Capture**

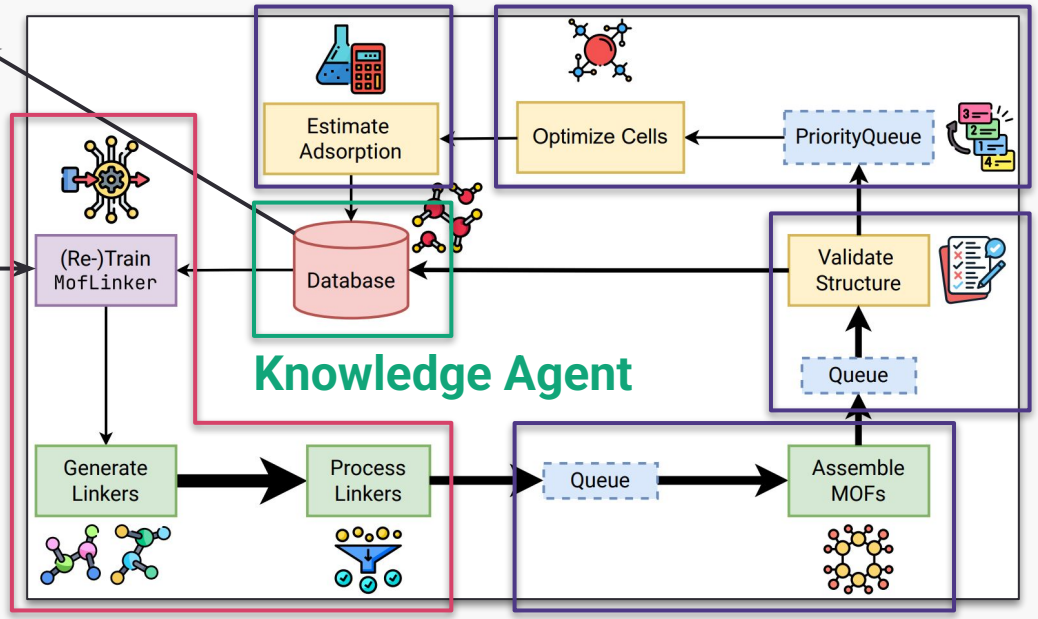
How to efficiently discover MOFs with desirable properties for target applications?



Intractable search space of ligand, node, & geometry combinations

MOFA: Online learning + GenAI + Simulation

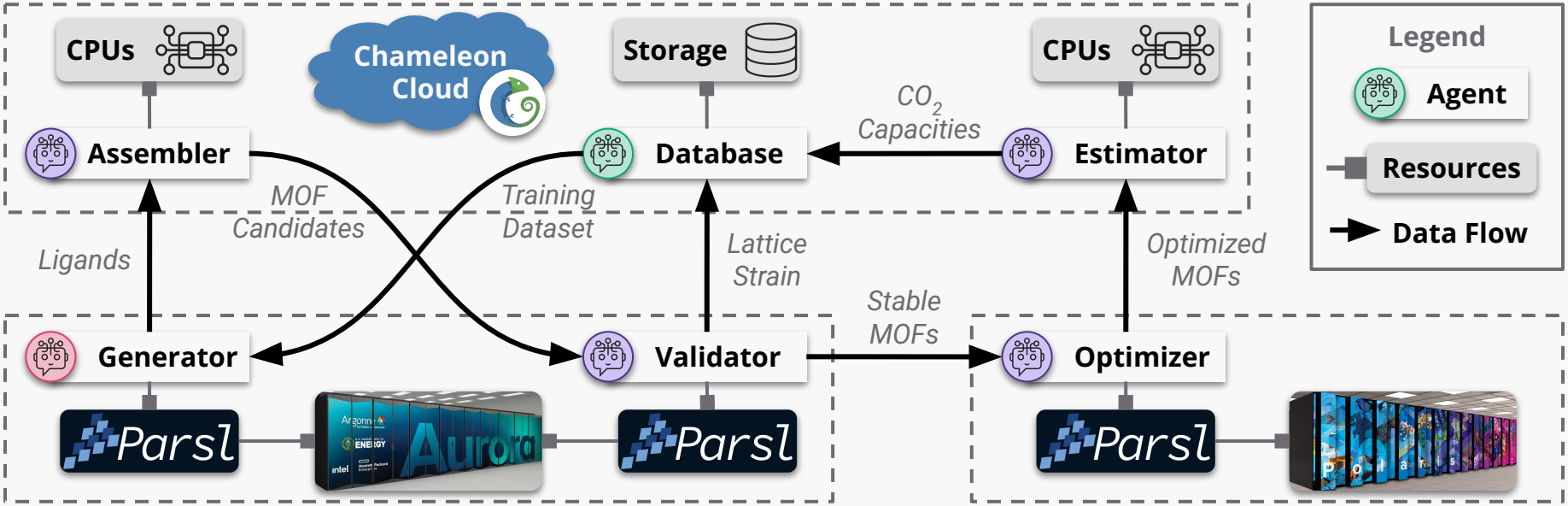
Embodied Agents*



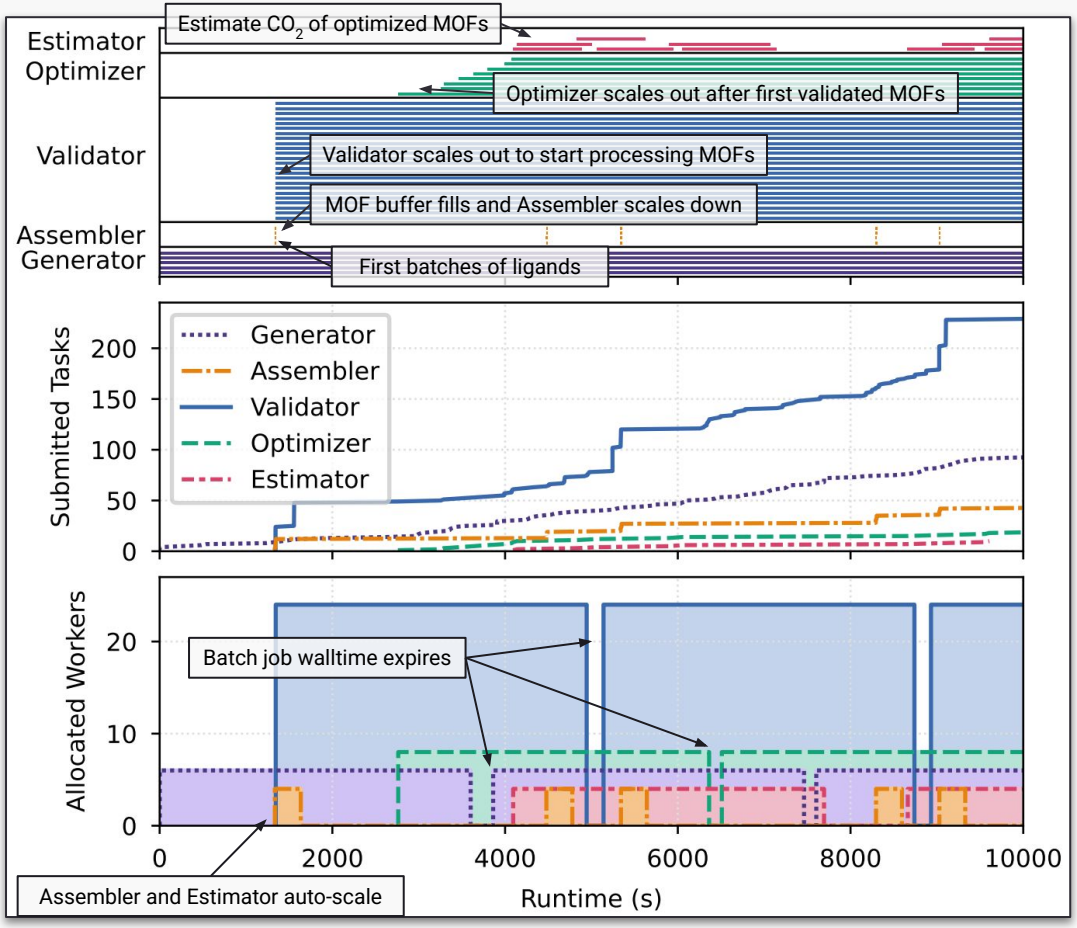
Computational Agents

Yan et al., "**MOFA: Discovering Materials for Carbon Capture with a GenAI- and Simulation-Based Workflow**" (Under Review)

MOFA through Autonomous Agents



Agents executed remotely via Globus Compute



MOFA Agents Trace

Why is this agentic model better?

- ➔ **Placement:** Move agents to resources
- ➔ **Separation of concerns:** Resource acquisition and scaling based on local workload
- ➔ **Loose coupling:** Swap agents or integrate new agents (e.g., SDL)
- ➔ **Shared agents:** Multiple workflows can share agents (microservice-like)

Questions?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Learn more/stay up to date:

- arxiv.org/abs/2505.05428
- github.com/proxystore/academy
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⭐ Academy on GitHub!