



UniFaaS: Programming across Distributed Cyberinfrastructure with Federated Function Serving

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Motivation



Executing scientific workflows across cyberinfrastructure(CI) amortizing queue times, distributed data, specialized accelerator etc.

When executing distributed scientific workflows

• funcX

Pros: Easy to build a distributed computing resource pool

Cons: Independent execution, manual data staging, limitations of input/output size

• Parsl

Pros: Support the DAG workflow, data staging (e.g. FTP, HTTP) Cons: Complicated to execute workflows on distributed CI simultaneously

□ What about funcX as an executor of Parsl?

- Things can be resolved immediately easy to program (in Parsl's way), distributed execution
- Things to be resolved data management, performance (scheduling)

Programming and Architecture

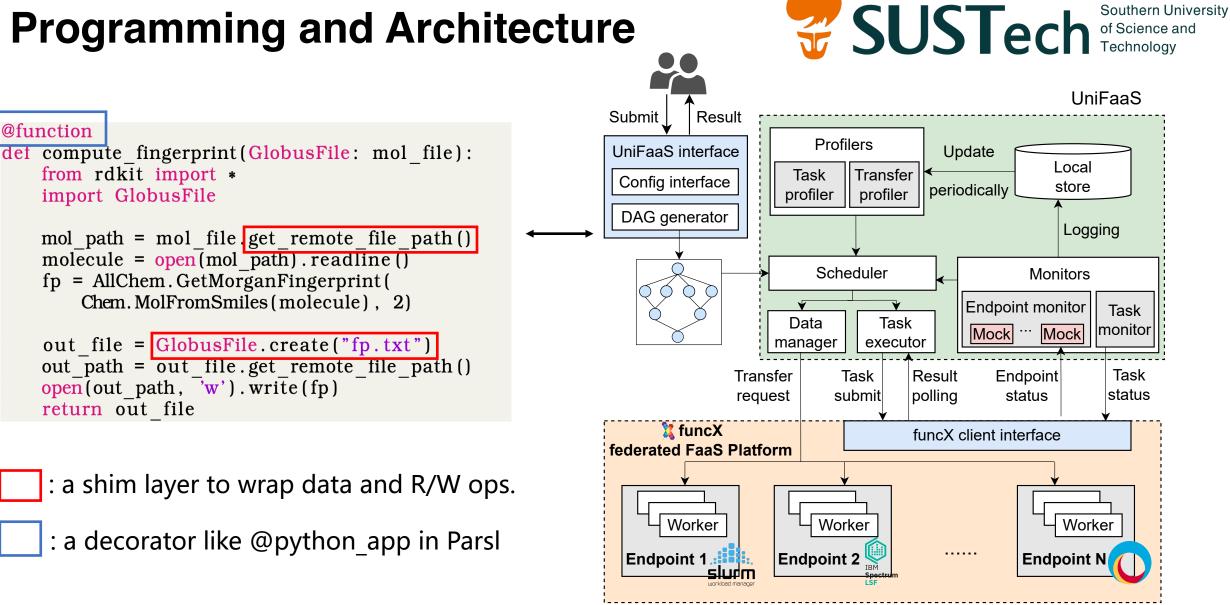
@function

from rdkit import *

open(out path, 'w').write(fp)

import GlobusFile

return out file



UniFaaS architecture

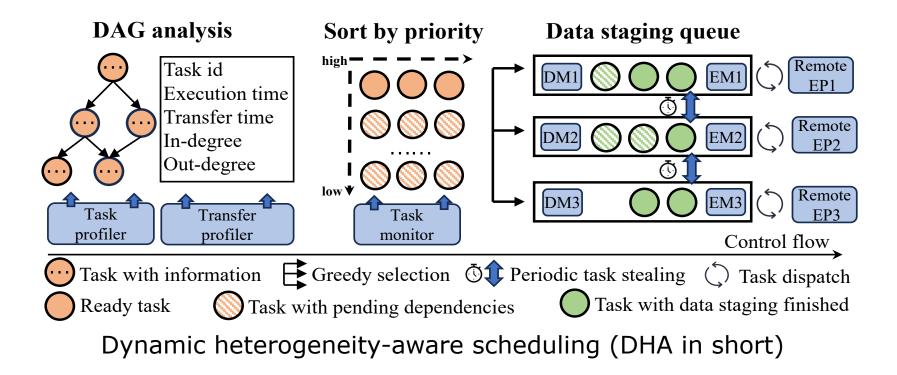
UniFaaS Scheduling

Goal: to minimize workflow's makespan

Challenges: varying data staging time, dynamic resource capacity.

Intuition:

- Data staging problem: start it as early as possible
- Dynamic resource capacity : real-time scheduling





UniFaaS Scheduling

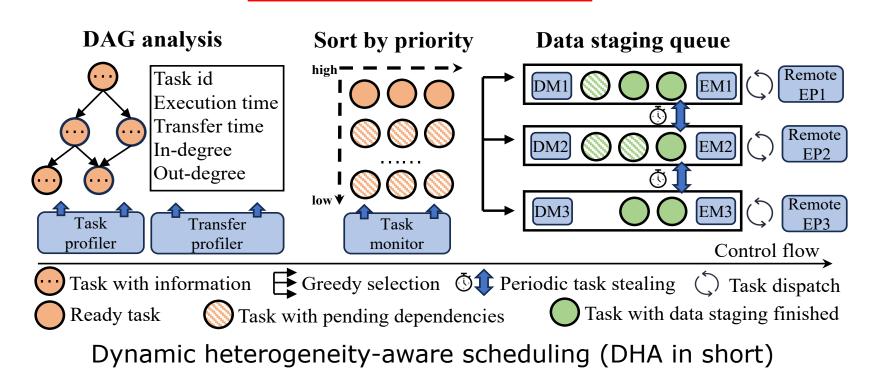


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UniFaaS Scheduling



No prior knowledge

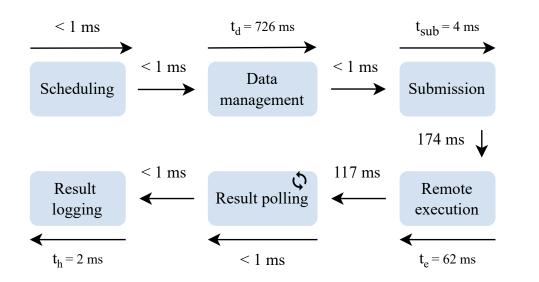
- Locality-aware scheduling for dynamic resource capacity schedule based on real-time status (real-time)
- Capacity-aware scheduling for static resource capacity schedule when the DAG enters our system (offline)

SUMMARY OF THE SCHEDULING ALGORITHMS.

| | Capacity | Locality | DHA |
|----------------------------|----------|--------------|--------------|
| Scheduling type | Offline | Real-time | Hybrid |
| Dynamic DAG supported | X | \checkmark | 1 |
| Dynamic resource supported | X | 1 | 1 |
| Knowledge required | × | X | \checkmark |



Latency



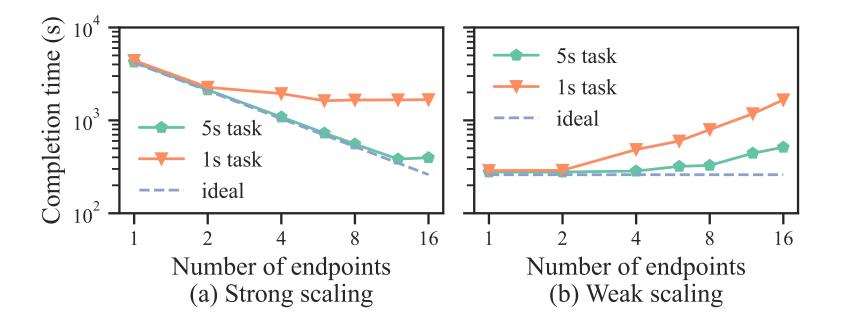
| OVERHEAD OF DIFFERENT ALGORITHMS. | | | |
|-----------------------------------|-----------------------|--|--|
| Scheduling algorithm | Overhead (s) | | |
| Capacity | 1.72×10^{-4} | | |
| Locality | $3.00 	imes 10^{-3}$ | | |
| DHA | 3.46×10^{-3} | | |

One "hello world" task with a 1 MB input totally costs 1087 ms.

All algorithms have a modest overhead.



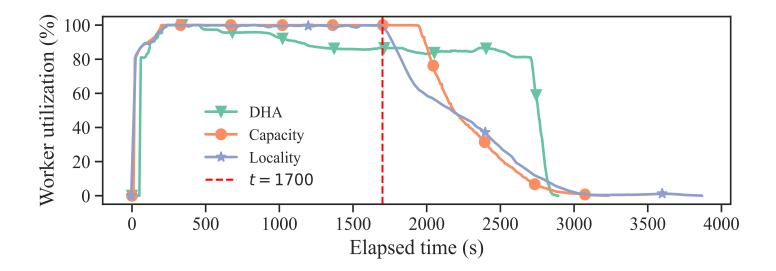
Scalability



Scalability of 5-second tasks is close to the ideal for up to 12 endpoints longer-duration tasks, better scaling



- 1. DHA has the best performance and highest worker utilization.
- 2. Improved performance by 22.99%, while utilizing only an additional 19.48% of resources.



| Experiment | Makespan (s) | Transfer size (GB) |
|----------------------|--------------|--------------------|
| Capacity | 3,240 | 4.86 |
| Locality | 3,882 | 53.46 |
| DHA | 2,898 | 44.94 |
| Baseline: HPC-I only | 3,763 | 0 |

Execute the drug screening workflow under static resource capacity.







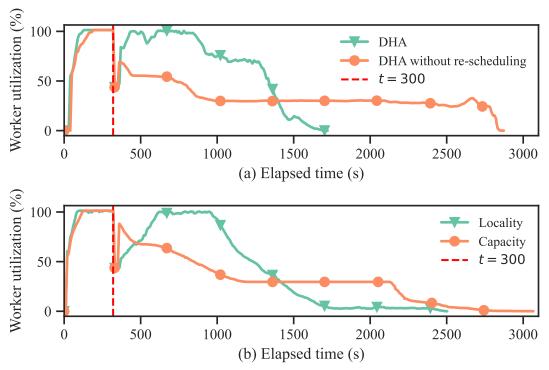


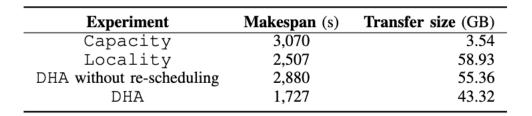
Questions?

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Case study: dynamic capacity

- 1. DHA has the best performance.
- 2. Locality is better than DHA without re-scheduling.





Execute the drug screening workflow under dynamic resource capacity.

