

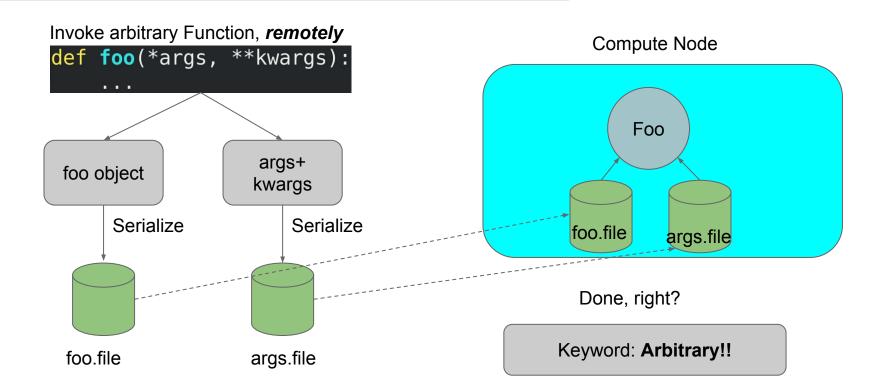
## Parsl+TaskVine: Bridging the Data Gap for High-Performance Scientific Workflows in Python

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### Background





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. . .



# Invoke arbitrary Function, remotely def foo(\*args, \*\*kwargs):

def	<pre>foo(*args, **kwargs): import v</pre>
	import x import y
	from z import a, b, c
	<pre> # input file(s)</pre>
	with open('bar', 'r') as f:
	<pre># output file(s)</pre>
	<pre>with open('tmp', 'w') as f:</pre>



Parsl users see:

- Function object
- args + kwargs

Under-the-hood:

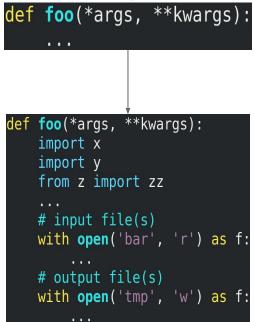
- Package Dependencies
- Input Files
- Output Files
- Intermediate/Temporary
   Files (function chaining, y = f(g(x)))

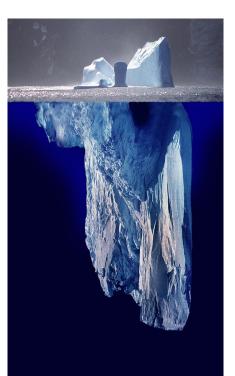
### Background



Tech Stack

#### Invoke arbitrary Function,





Parsl users see: - Function object - args + kwargs

Under-the-hood - Package Dependencies - Input Files - Output Files - Intermediate

/Temporary Files (function chaining, y = f(g(x)))



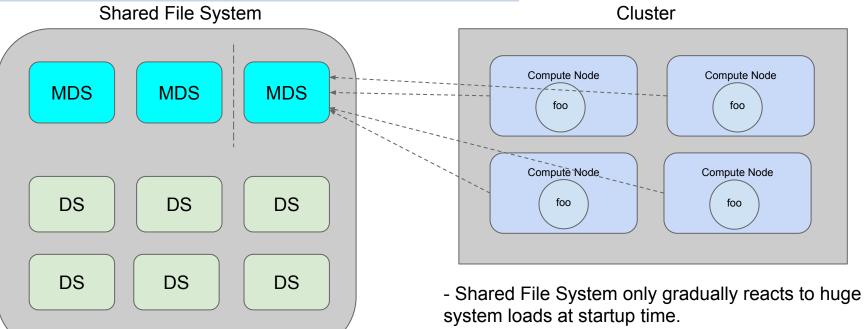
Shared File Systems

GPFS

PANASAS

#### Problem

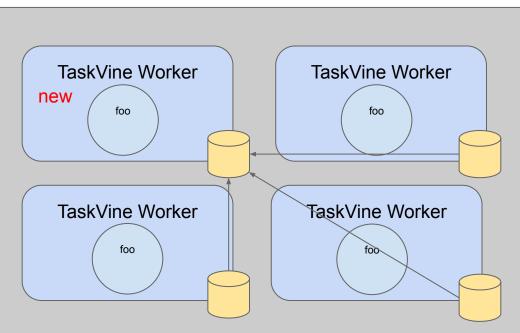




- Caches aren't effective with one-time setups.
- Happens over the course of a workflow execution.

#### Solution: ParsI+TaskVine





#### Insights

- Costly to move data in and out of cluster.

**CCTools** 

- Unused local disk storage on compute nodes.
- Data from other nodes aren't used.

#### **TaskVine Novelties**

- Explicit Data Placement with Data-to-Task Bindings!
- Data Stays in Cluster until Not Needed.
- No remote I/O! Only local I/O over the course of a workflow execution.
- No shared fs overloading!

#### Solution: Example



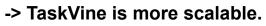
foo(\*args, \*\*kwargs): def Cluster import x import y from z import a, b, c . . . # input file(s) with open('bar', 'r') as f: TaskVine Worker TaskVine Worker # output file(s) new with open('tmp', 'w') as f: foo foo . . . (4) "bar" input file Parsl DFK TaskVine Worker TaskVine Worker (1) args+kwargs (2) x, y, z(3) foo object foo dependencies foo file TaskVine Executor/ Manager Process



### **Evaluation: Data Distribution Method**

Workflow Specification and Setup:

- 2048 function invocations
- Each invocation trains a NN model in 5 minutes.
- Data dependencies include software dependencies (zipped: ~1GB, unzipped: ~3GBs) and dataset (zipped: 17MBs, unzipped: 25MBs).
- Each compute node has 16 cores, 16GBs of memory and disk, and runs 8 tasks concurrently.
- Each worker has access to the local panasas shared file system.
- Varying number of workers -> varying amount of concurrent tasks -> varying pressure on the shared file system.





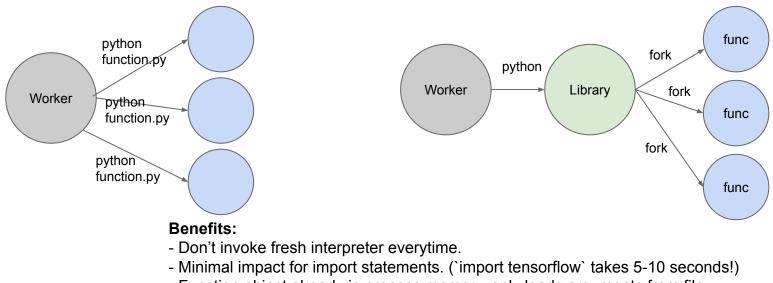
#### Workflow execution time



#### **Optimization: Serverless**

Regular Way: Starting a fresh Python interpreter process for every function call.

Serverless: Fork a current process for every function call.



<sup>-</sup> Function object already in process memory, only loads arguments from file.

#### **Evaluation: Serverless**

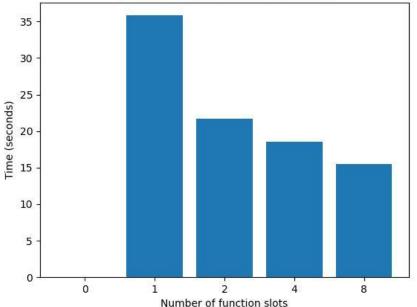


Run of 20,000 simple functions on a local machine with:

- regular way (711s, not shown) and
- serverless (~50x speedup, on right)

-> Great for short-running functions!

20000 simple functions execution time with TaskVine serverless, local machine



TaskVine: Managing In-Cluster Storage for High-Throughput Data Intensive Workflows, WORKS at SC 2023.



### Try it now!

1) Install via conda (recommended):

conda install -c conda-forge ndcctools parsl

2) Change configuration of executor:

3) More examples: https://cctools.readthedocs.io/en/stable/taskvine/#workflow-integration

### **Current Status of TaskVine**



#### This work was supported by NSF Award OAC-1931348

- TaskVine is a component of the Cooperative Computing Tools (cctools) from Notre Dame alongside Makeflow, Poncho, Resource Monitor, etc.
- Research software with an engineering process: issues, tests, manual, examples.
- We are eager to collaborate with new users on applications and challenges!

#### http://cctools.readthedocs.io

