RADICAL-Pilot and Parsl (RPEX)
Executing Heterogeneous Workflows on HPC Platforms

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Motivation: Why the Integration approach in general?

- Possibility of achieving better capabilities by merging both system capabilities.
- New capabilities serve new use cases or existing use cases in a better way.
- Only develop new capabilities when they are not available in both systems.
- Integrating two independent systems → we are preventing effort duplications → better collaboration.
- Less code.
- RADICAL-Pilot (RP) is a workload runtime system based on the building block approach.
- By design, both Parsl and RP allow the ease of integration with other systems.
RADICAL-Pilot (RP)

- A scalable, modular, and interoperable pilot system that enables the asynchronous execution of heterogeneous workloads on heterogeneous HPC resources.
- RP provides methods for efficiently and effectively scheduling, placing, and launching independent tasks across multiple nodes.
- RP supports the concurrent execution of up to $10^6$ tasks on $10^3$ compute nodes with low overheads.
Parsl and RADICAL-Pilot

What Parsl offers to RADICAL-Pilot:

- Workflows management capabilities.
- Construct and build complex dynamic workflows
- Flexible and lightweight API and programming model.

What RADICAL-Pilot offers to Parsl:

- Single/multi cores/GPUs with MPI, OpenMP, and single/multi-[threaded|process] tasks.
- Heterogeneous resources management (CPUs/GPUs)
- Granular resource specification at the pilot/task level (processes/ranks/threads).
RP Executor (RPEX) Architecture

```python
[ ]: config = Config(
    executors=[RPEX(label='RPEX',
                     resource='tacc frontera',
                     login_method='interactive',
                     walltime=30,
                     managed=True,
                     cores=56,
                     gpus=0)],
    strategy=None,
    usage_tracking=True)
parsl.load(config)

@python_app
def mpi_rpe_x_task(process_type, cpu_processes, pre_exec, comm=None):
    size = comm.Get_size()
    rank = comm.Get_rank()
    name = comm.Get_processor_name()
    print("Hello, World! I am process %d of %d on %s\n" % (rank, size, name))
    result = mpi_rpe_x_task(process_type=rp.HAPI,
                            cpu_processes=3,
                            pre_exec=['conda activate env'], comm=None)
```
Supported Use Cases

Colmena:

- A Python package for machine learning based steering of ensemble computations on HPC platforms.
- Requires RPEX capabilities to launch large MPI workflows (executables and functions) via Parsl programming model.

Ice Wedge Polygons:

- Detect and classify wedge-shaped ice masses create polygonized land surface patterns called Ice Wedge Polygons (IWP) across large Arctic areas.
- Requires RPEX capabilities to launch MPI CPU and GPU tasks across multiple nodes.
Performance results on TACC Frontera

Colmena:

- A single node MPI executables and non-MPI python functions workload.
- Executing 3600 tasks on 256 nodes.
- A resource utilization of ~98% on 256 nodes (~14K CPU cores)
- RPEX OVH of ~500s (RP overheads + Parsl overheads)
- Throughput of ~15 tasks/sec.

Ice Wedge Polygons:

- A multi-node MPI functions workload.
- Executing 32 tasks on 16 nodes.
- A resource utilization of ~99% (~900 cores and 64 GPUs)
- RPEX OVH of 6.5s.
- Throughput of ~1 tasks/sec.
Future work and new use cases

- Develop a bulk mode submission for RPEX to boost the throughput of:
  - Task Submission
  - Task Scheduling
  - Task Execution

- Cylon is a new upcoming use case that requires RPEX capabilities.
  - Cylon is a python and C++ parallel data framework that provides fast and scalable data aggregation while offering the possibility to integrate with other data distribution frameworks for additional capabilities.