

Fondo Europeo de Desarrollo Regional









Barcelona Supercomputing Center Centro Nacional de Supercomputación

Simplifying the lifecycle management of complex application workflows

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12/6/2024

AEiC 2024 workshop, Barcelona

EuroHPC systems

		Status	Country	Peak performance	Architecture
Pre-Exascale	LUMI	Operational	Finland	539.13 petaflops	64-core AMD EPYC™ CPUs + AMD Instinct™ GPU
	Leonardo	Operational	Italy	315.74 petaflops	Intel Ice-Lake ' ohire Rational States of the states of th
	MareNostrum 5	Operational	Spain	295.81 ropean Exast	AMD Epyc 7H12 + Nvidia A100
Petascale	Meluxina	Operational JUPITER	mputer	announce	invIDIA Ampere
	Vega	Operational Superco		recaflops	AMD Epyc 7H12 + Nvidia A100
	Karolina	Operational Superational In Julio	Silduy	12.91 petaflops	AMD + Nvidia A100
	Discoverer	Operational	Bulgaria	5.94 petaflops	AMD EPYC
	Deucalion	Operational	Portugal	5.01 petaflops	A64FX, AMD EPYC, Nvidia Ampere

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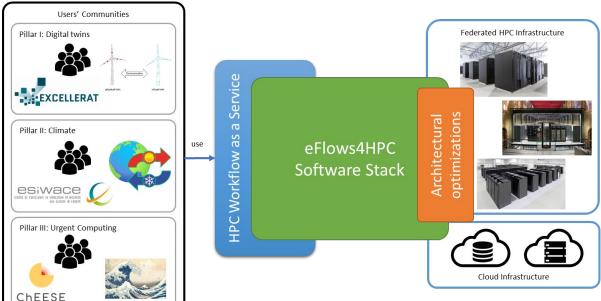
https://eurohpc-ju.europa.eu/about/our-supercomputers_en

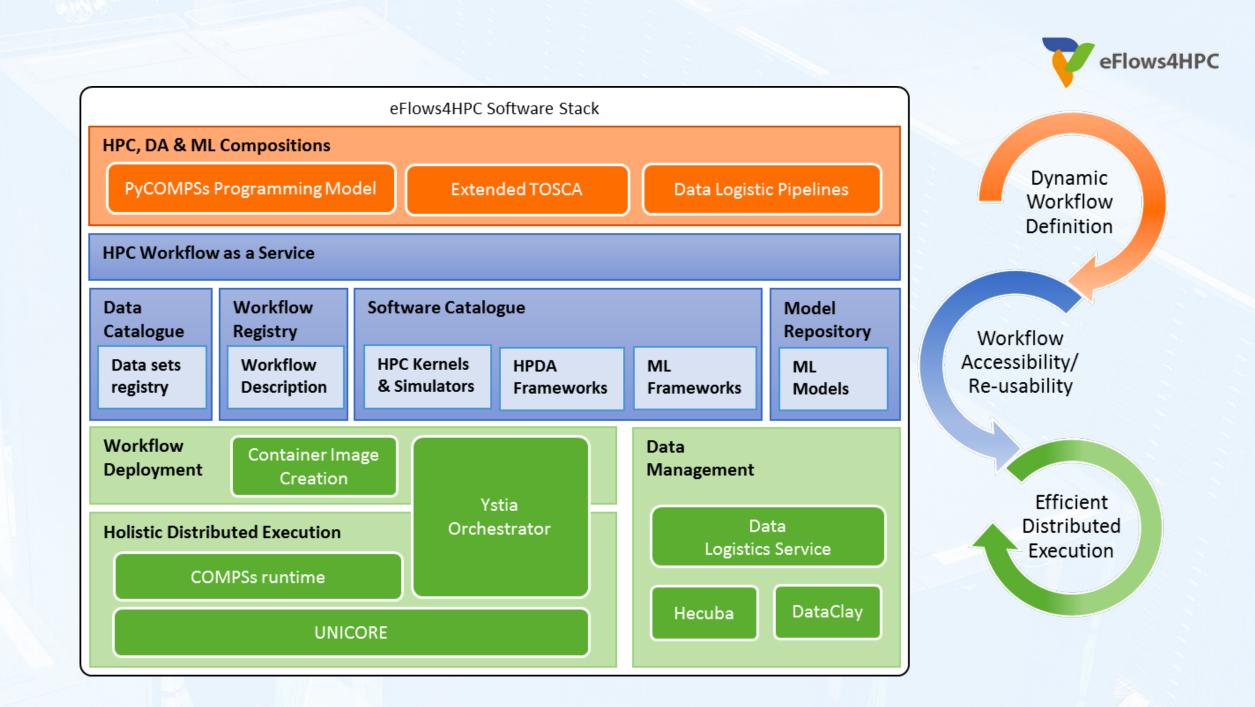
eFlows4HPC in a nutshell



- Software tools stack that makes easier the development and management of complex workflows:
 - Combine different aspects
 - HPC, AI, data analytics
 - Reactive and dynamic workflows
 - Autonomous workflow steering
 - Full lifecycle management
 - Not just execution
 - Data logistics and Deployment
- HPC Workflows as a Service:
 - Mechanisms to make easier the use and reuse of HPC by wider communities
- Architectural Optimizations:
 - Selected HPC AI Kernels Optimized for GPUs, FPGA, EPI
- Validation Pillar's
 - End-user workflows linked to CoEs

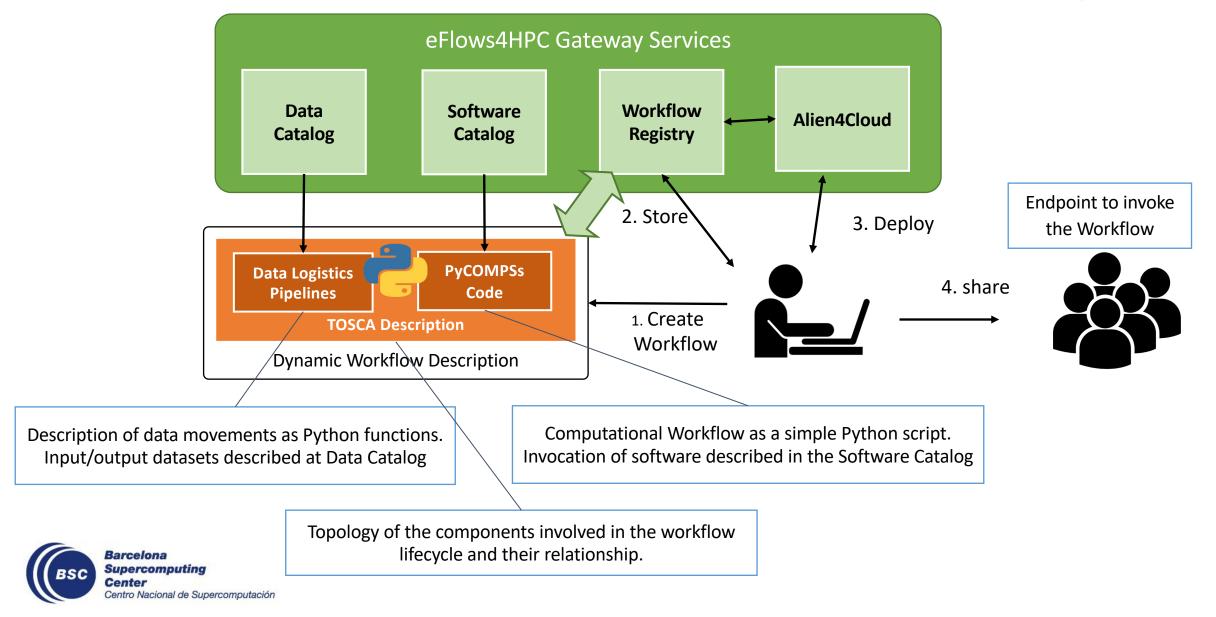






Workflow development overview

eFlows4HPC

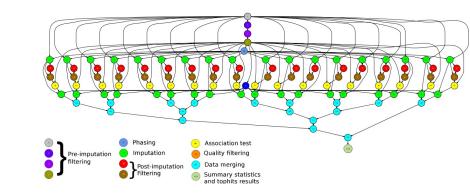


Programming with PyCOMPSs/COMPSs



- Sequential programming, parallel execution
- General purpose programming language + annotations/hints
 - To identify tasks and directionality of data
- Builds a task graph at runtime that express potential concurrency
- Tasks can be complex, parallel, even MPI
- Offers a shared memory illusion to applications in a distributed system
 - The application can address larger data storage space: support for Big Data apps
- Agnostic of computing platform
- Provenance recording
- Syntax extended to better integrate AI and HPDA





```
@task(c=INOUT)
def multiply(a, b, c):
    c += a*b
```

```
initialize_variables()
startMulTime = time.time()
for i in range(MSIZE):
    for j in range(MSIZE):
        for k in range(MSIZE):
            multiply (A[i][k], B[k][j], C[i][j])
compss_barrier()
mulTime = time.time() - startMulTime
```

PyCOMPSs features and runtime

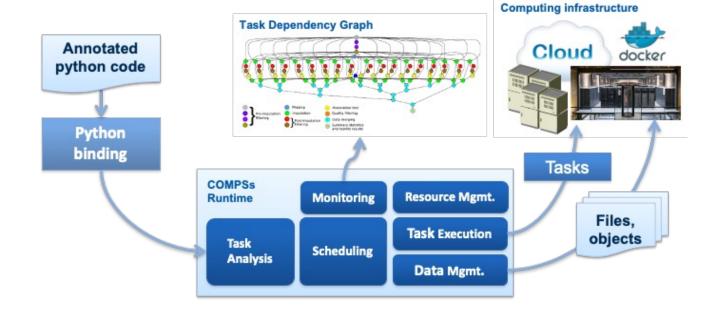
- PyCOMPSs/COMPSs applications executed in distributed mode following the masterworker paradigm
 - Description of computational infrastructure in an XML file
- Sequential execution starts in master node and tasks are offloaded to worker nodes
- All data scheduling decisions and data transfers are performed by the runtime
- All data scheduling decisions and data transfers are performed by the runtime
- Support for elasticity

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Heterogeneous Tasks

- A task can be more than a sequential function
 - A task in PyCOMPSs can be sequential, multicore or multi-node
 - External binary invocation: wrapper function generated automatically (@binary)
 - Supports for alternative programming models: MPI (@mpi)
- Can be combined with other decorators
 - @constraint: To indicate amount of memory, number of processors or GPUs per binary or MPI process
 - @container: When software is distributed as a container

```
@container(engine=`SINGULARITY', image="/path/to/app.sif")
@binary(binary="app.bin" args="-in {{f in}} -out {{f out}})
                                                                    @binary(binary="app.bin" args="-in {{f in}} -out {{f out}})
@task(f in=FILE IN, f out=FILE OUT)
                                                                    @task(f in=FILE IN, f out=FILE OUT)
def app task(f in, f out):
                                                                    def app task(f in, f out):
    pass
                                                                        pass
@constraint(processors=[{'processorType':'CPU', 'computingUnits':'1'},
                         {'processorType':'GPU', 'computingUnits':'1'}])
@task(returns=1)
def func(a, b, c):
                                                                    @constraint (computingUnits= "8")
    . . .
    return result
                                                                     @mpi (runner="mpirun", processes= "16", ...)
                                                                     @task (returns=int, stdOutFile=FILE OUT STDOUT, ...)
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                                                                    def nems(stdOutFile, stdErrFile):
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                                                                         pass
```

Failure management

• Interface than enables the programmer to give hints about failure management

```
@task(file_path=FILE_INOUT, on_failure='CANCEL_SUCCESSORS',
time_out=`$task_timeout')
def task(file_path):
    ...
    if cond :
        raise Exception()
```

- Options: RETRY, CANCEL_SUCCESSORS, FAIL, IGNORE
- Implications on file management:
 - i.e, on IGNORE, output files are generated empty
- Possibility of ignoring part of the execution of the workflow, for example if a task fails in an unstable device
- Opens the possibility of dynamic workflow behaviour depending on the actual outcome of the tasks



Timeouts and exceptions

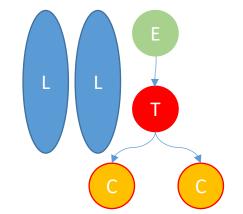
• Timeouts can be defined for a task

• Tasks can raise exceptions

```
@task(file_path=FILE_INOUT)
def comp_task(file_path):
    ...
    raise COMPSsException("Exception
raised")
```

• Combined with groups of tasks enables to cancel the group of tasks on the occurrence of an exception





```
def test_cancellation(file_name):
    try:
    with TaskGroup('failedGroup'):
        long_task(file_name)
        long_task(file_name)
        executed_task(file_name)
        comp_task(file_name)
        cancelledTask(FILE_NAME)
        cancelledTask(FILE_NAME)
    except COMPSsException:
        print("COMPSsException caught")
        write two(file_name)
```

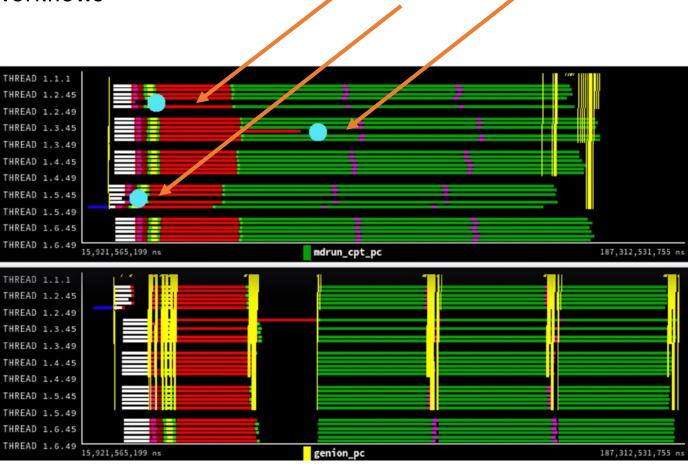
Validation Example

- Protein Mutants workflow from BioBB workflows
- Three types of failures
 - incorrect data,
 - incorrect SW configuration,
 - longer execution time
- On_failure = CANCEL_SUCCESSORS

COMPSs managing failures

Application managing failures





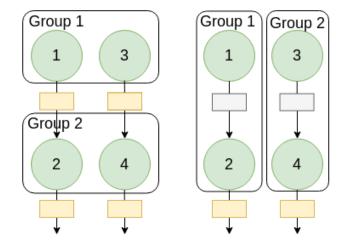
Successor tasks are

cancelled

Checkpointing

- Allows the workflow re-execution avoiding the re-execution of finished tasks
- Asynchronous but with some overhead
 - Save tasks results in a persistent storage
 - Trade-off between performance and time to recover
 - Establishing the right checkpoint granularity is important
- 3 mechanisms for automatic checkpointing
 - Time: periodically, COMPSs saves the last version produced for every value
 - Finished tasks : after the completion of X tasks, COMPSs saves the last version produced for every value
 - Instantiation task groups: Defines groups of tasks, COMPSs saves those data versions that are final results for the group compss_snapshot()
- Indicated by the developer with API
- No checkpoint inside the task: Drawback for very large tasks.
 - Possible integration with internal checkpointing





- Task not checkpointed
- Task checpointed
- Task output not copied
- Task output copied

Checkpointing Overhead

- Benchmarks:
 - K-means clustering
 - PMXCV19, bio workflow that evaluates changes in the binding affinity between SARS-Cov-2 spike protein and Human ACE2 receptor.
 - Principle Component Analysis (PCA)
- Policies:
 - Instantiated Tasks Groups ITG (Grouping every 10 instantiated tasks)
 - Finished Tasks FT (Every 10 finished tasks)
 - Periodic Time PT (15 seconds interval)

	NC	ITG	FT	PT
K-Means	$220.12~\mathrm{s}$	222.56 s (1%)	229.34 s (4.5%)	228.44 s (4%)
PMXCV19	33 m	$33.9 \mathrm{~m} (2.7\%)$	34.1 m (3.3%)	33.6 m (1.8%)
PCA	$883.13~\mathrm{s}$	1075.13 s (21.7%)	1026.21 s (16.1%)	1284.07 s (45.4%)



Checkpointing Overhead

- Importance on the policy and frequency choices depending on the application
- Policies:
 - Instantiated Tasks Groups ITG (Grouping every 10, 50, 100 instantiated tasks)
 - Periodic Time PT (15, 30, 60 seconds interval)
 - Finished Tasks FT (Every 10, 40, 100 finished tasks)

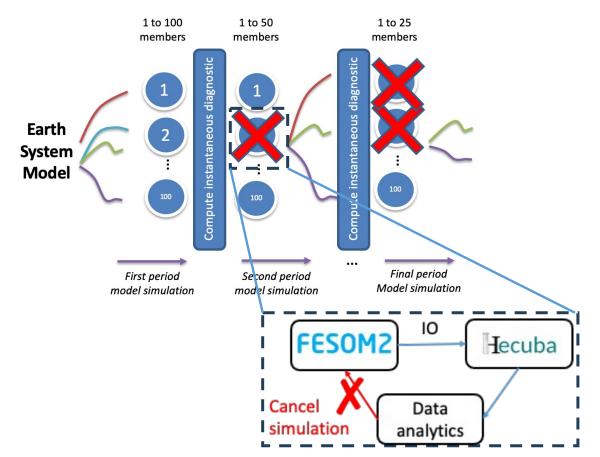
	Fine-grain	Medium-grain	Coarse-grain
Kmeans (ITG)	222.56 s (1.1%)	244.25 s (11%)	264.36 s (20%)
PMXCV19 (PT)	33.6 m (1.8%)	33 m (0%)	33.1 m (0.3%)
PCA (FT)	1026.21 s (16.1%)	1016.20 s (15%)	1103.94 (24.9%)



DA-driven ensemble member pruning

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Earth-System Model (ESM) workflow



Task-groups and **exceptions** used to dynamically prune ensemble members based on data analytics

Hecuba support for a lambda architecture, allowing both batch processing and stream processing

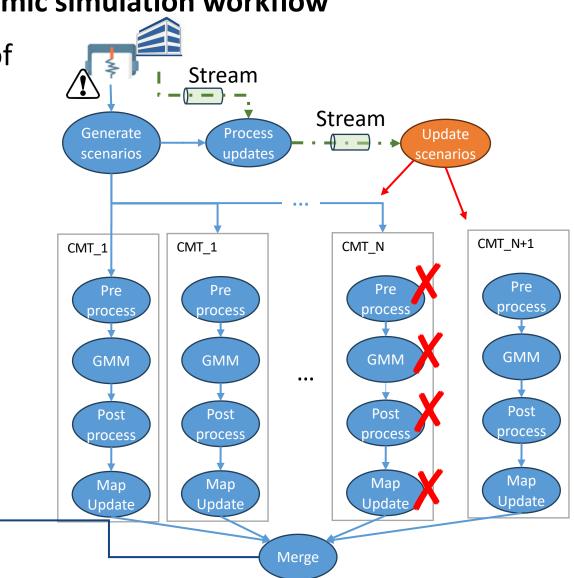


Event-driven cancellation/creation

UCIS4EQ: HPC-based urgent seismic simulation workflow

- Evaluation of scenarios after the occurrence of a seismic event
- Combines multiple web services and HPC simulation (Salvus)
- Workflow Dynamicity:
 - Usage of data streaming for communication of events
 - On event occurrence API supports:
 - Dynamic cancellation of task groups
 - Dynamic creation of new set of tasks

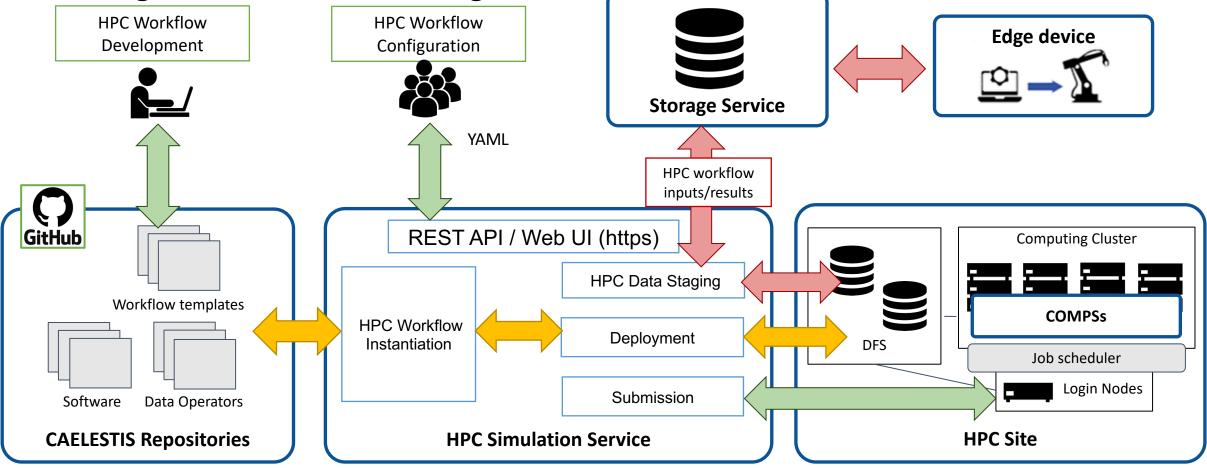




eFlows4HPC

CAELESTIS Simulation Ecosystem Architecture caelestis

Towards a digital twin for aircraft design

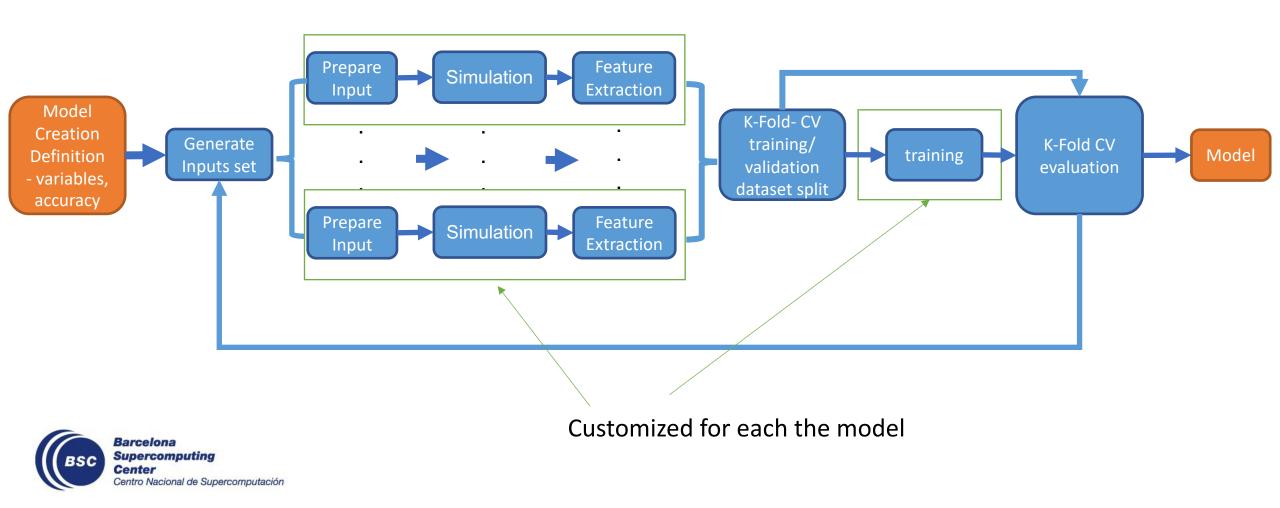


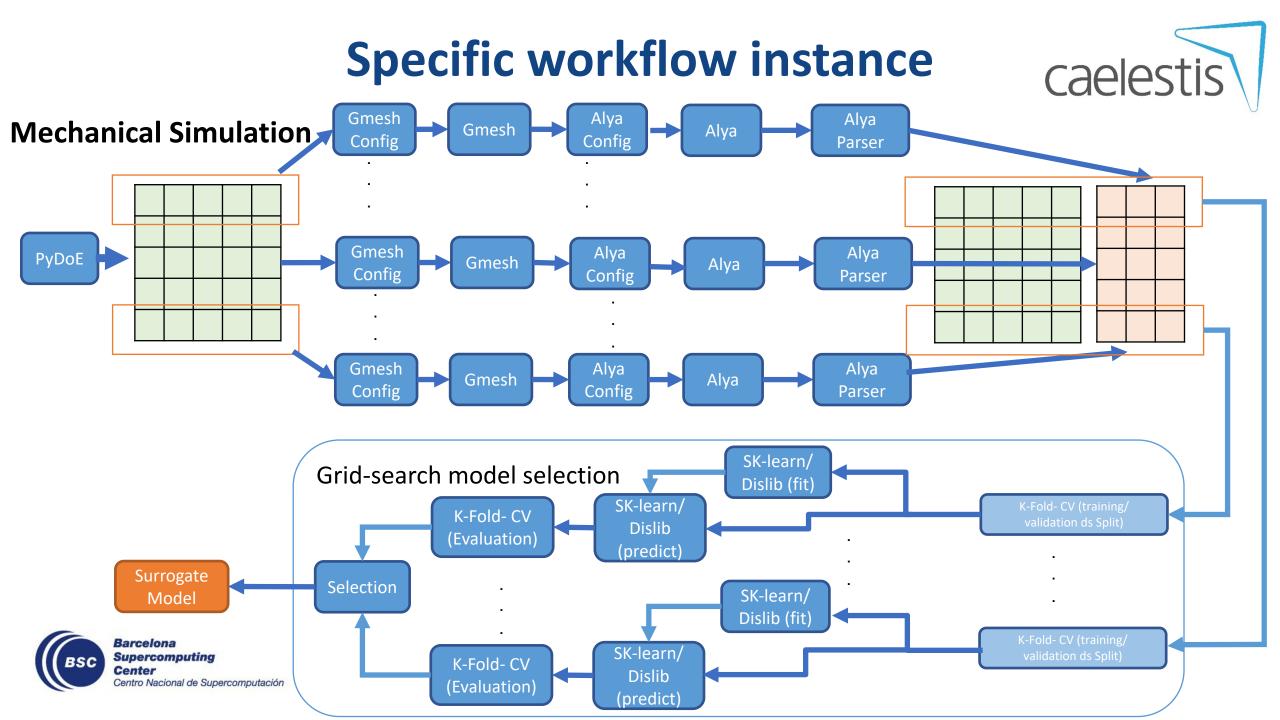


Workflow templates



Surrogate Model Creation Workflow

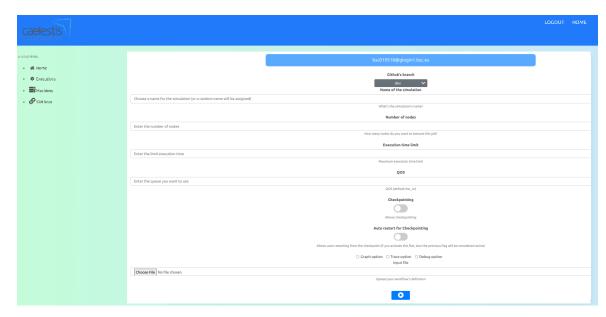




Checkpoint management



- CAELESTIS service offers two management options of checkpointed jobs:
 - Automatic handling: If job terminates due to time constraint, the service reinitiates the workflow submission, which is rerun from the checkpoint. This process can continue until the job concludes successfully.
 - Manual handling: users have the option to handle checkpoints manually. If a job ends due to a time constraint, it will appear in the execution dashboard under the "timeout jobs" list. Users can then initiate a new job from the checkpoint.





Final thoughts

- Coarse-grained task-based programming models provide suitable environments for developing HPC + AI workflows
 - eFlows4HPC
 - CAELESTIS
 - DT-GEO
- These programming environments should provide means to guarantee resiliance in the workflows' executions
- Fault tolerance and exceptions mechanisms can be used to provide maleability and dynamicity to the workflow applications
- Checkpointing provides resiliance, but also mechanisms to avoid job execution constraints



Further Information

- Project page: <u>http://www.bsc.es/compss</u>
 - Documentation
 - Virtual Appliance for testing & sample applications
 - Tutorials
- Source Code
 - https://github.com/bsc-wdc/compss
 - Docker Image
 - https://hub.docker.com/r/compss/compss
 - Applications

https://github.com/bsc-wdc/apps https://github.com/bsc-wdc/dislib

https://dislib.readthedocs.io/en/latest/

• Dislib

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HP2C-DT

CyclOps











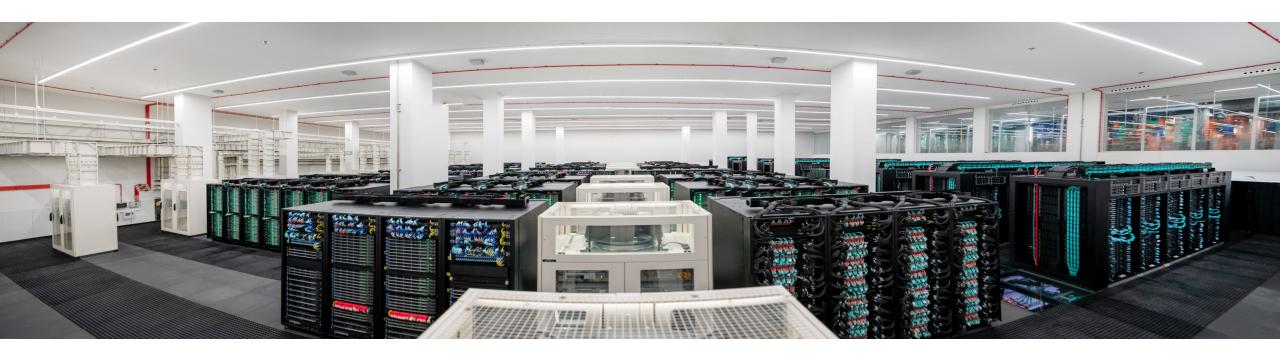




MareNostrum 5



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Thanks!



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