

Ophidia: High Performance Data Analytics framework for Climate Change at scale

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*IS-ENES3 Virtual workshop on requirements
for a fast and scalable evaluation workflow*



May 19, 2021



Outline

- *Climate data management and analytics challenges*
- *The Ophidia HPDA Framework*
- *Key aspects: execution model, parallelization, HPC integration*
- *Data and metadata operators*
- *Integration with Python*
- *Experimental results*
- *Summary*



Climate analysis challenges & issues

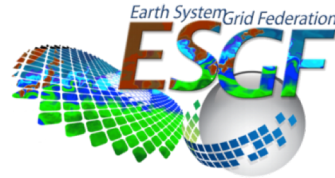
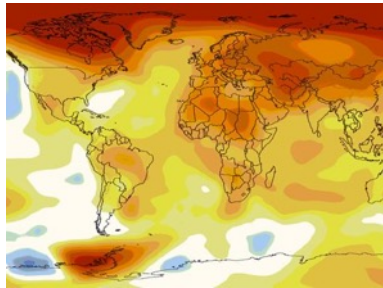
Key challenges and practical issues related to large-scale climate analysis

- Setup of a data analysis experiment requires the **download of (multiple) input data**
 - *Reducing data movement is essential*
- The complexity of the analysis leads to the need for **workflow support**
 - *Data analysis mainly relies on client-side approaches*
- Large data volumes pose **strong requirements** in terms of **computational** and **storage resources**
 - Need for novel HPDA solutions able to scale the analysis
 - Difficulties in exploiting data analysis solutions in HPC environments

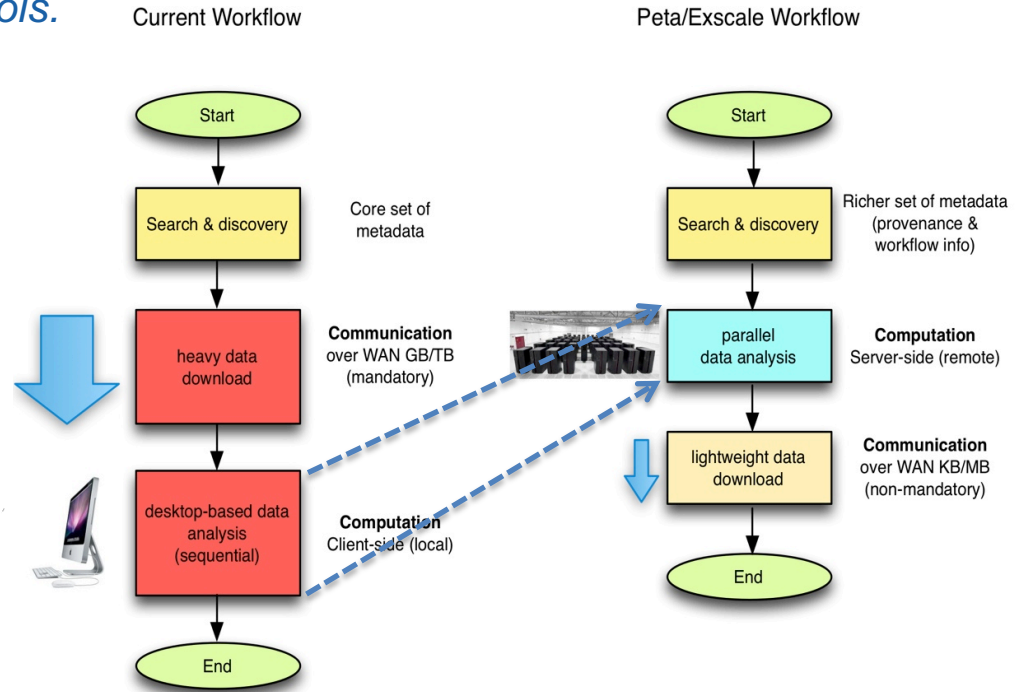
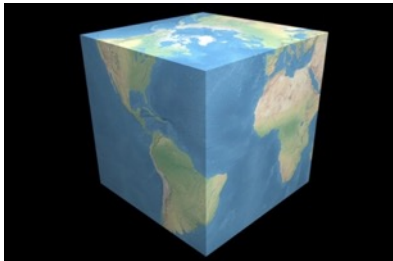


A paradigm shift

Volume, variety, velocity are key challenges for big data in general and for climate change science in particular. Client-side, sequential and disk-based workflows are three limiting factors for the current scientific data analysis tools.



		12.4	11.8	7.8	8.9	
		5.4	2.4	3.1	4.3	
38°	12.4	7.6	13.2	11.3	2.8	6.7
37°	18.4	13.6	14.1	16.3	4.5	3.1
36°	14.4	6.1	9.2	12.4	1.7	5.6
35°	21.3	17.8	23.5	22.1		
	GEN	FEB	MAR	APR		



S. Fiore, A. D'Anca, C. Palazzo, I. Foster, D. N. Williams, G. Aloisio, "Ophidia: toward bigdata analytics for eScience", ICCS2013 Conference, Procedia Elsevier, 2013



Ophidia HPDA framework

Ophidia (<http://ophidia.cmcc.it>) is a CMCC Foundation research project addressing data challenges for eScience

- A **HPDA framework** for multi-dimensional scientific data joining HPC paradigms with scientific data analytics approaches
- **In-memory** and **server-side** data analysis exploiting parallel computing techniques
- Multi-dimensional, array-based, storage model and partitioning schema for scientific data leveraging the **datacube** abstraction
- End-to-end mechanisms to support **interactive analysis, complex experiments** and **large workflows** on scientific data



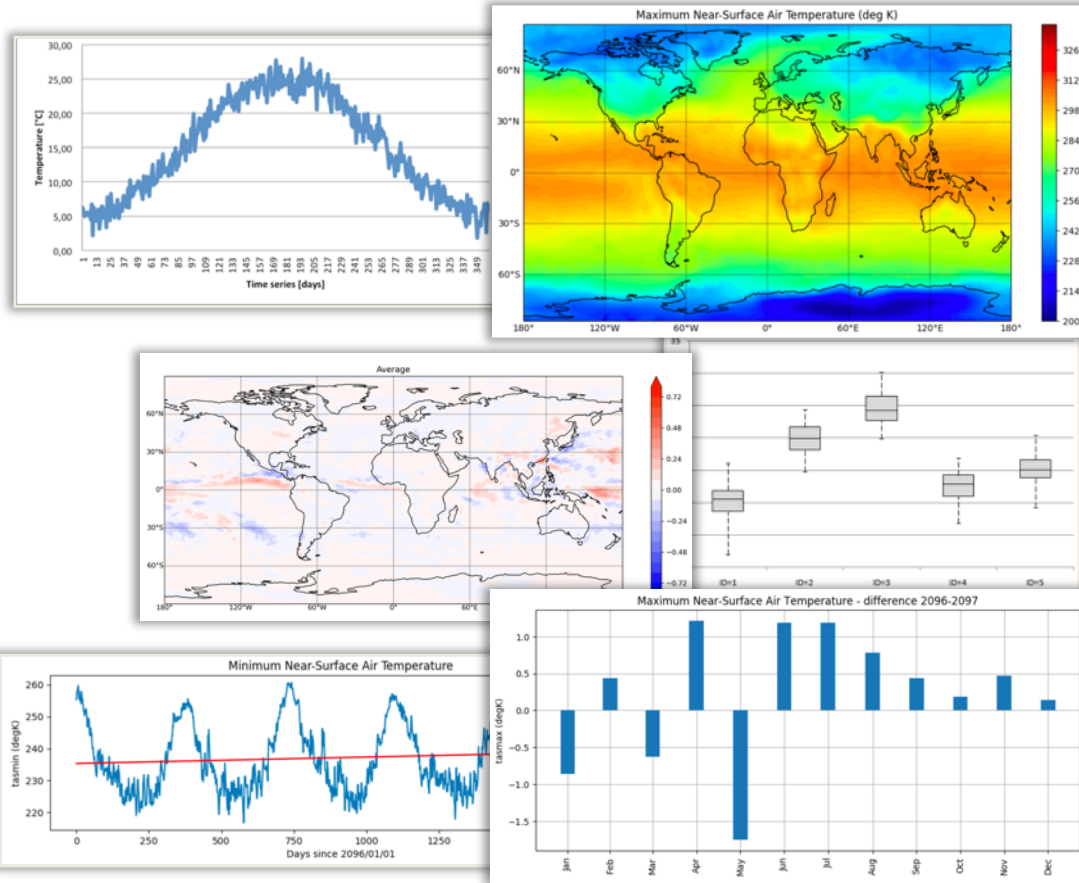
S. Fiore, D. Elia, C. Palazzo, F. Antonio, A. D'Anca, I. Foster, G. Aloisio, "Towards High Performance Data Analytics for Climate Change", ISC High Performance 2019, LNCS Springer, 2019



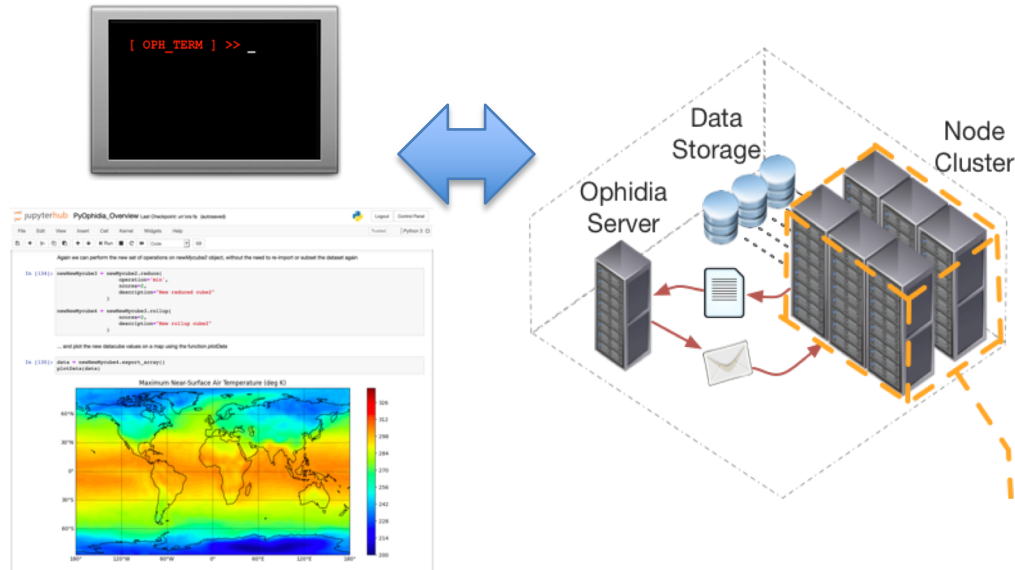
Data analytics requirements and use cases

Requirements and needs focus on:

- *Time series analysis*
- *Data subsetting*
- *Model intercomparison*
- *Multi-model means*
- *Massive data reduction*
- *Data transformation*
- *Parameter sweep experiments*
- *Maps generation*
- *Ensemble analysis*
- *Data analytics workflow support*



Server-side paradigm and execution modes

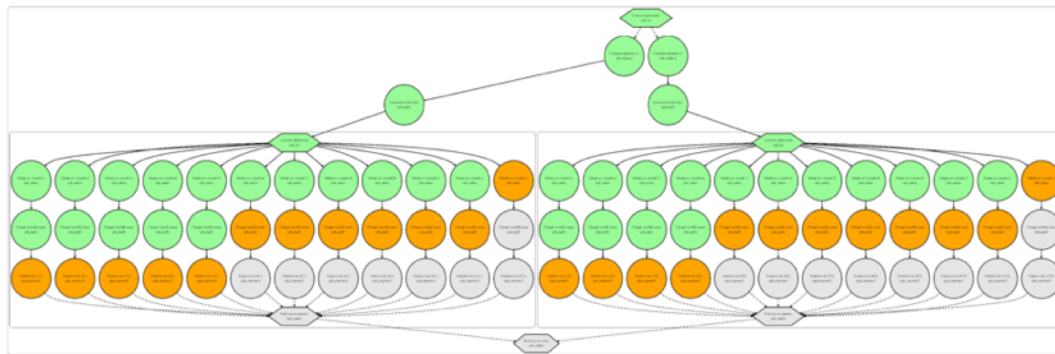


Oph_Term: a terminal-like commands interpreter serving as a client for the Ophidia framework

PyOphidia: a Python interface for datacube management & analytics with Ophidia

Multiple execution modes:

- *Interactive data analysis*
- *Batch processing*
- *Python notebooks and applications*
- *Workflows of operators*



Ophidia Architecture

Multi-interface **interoperable** front-end

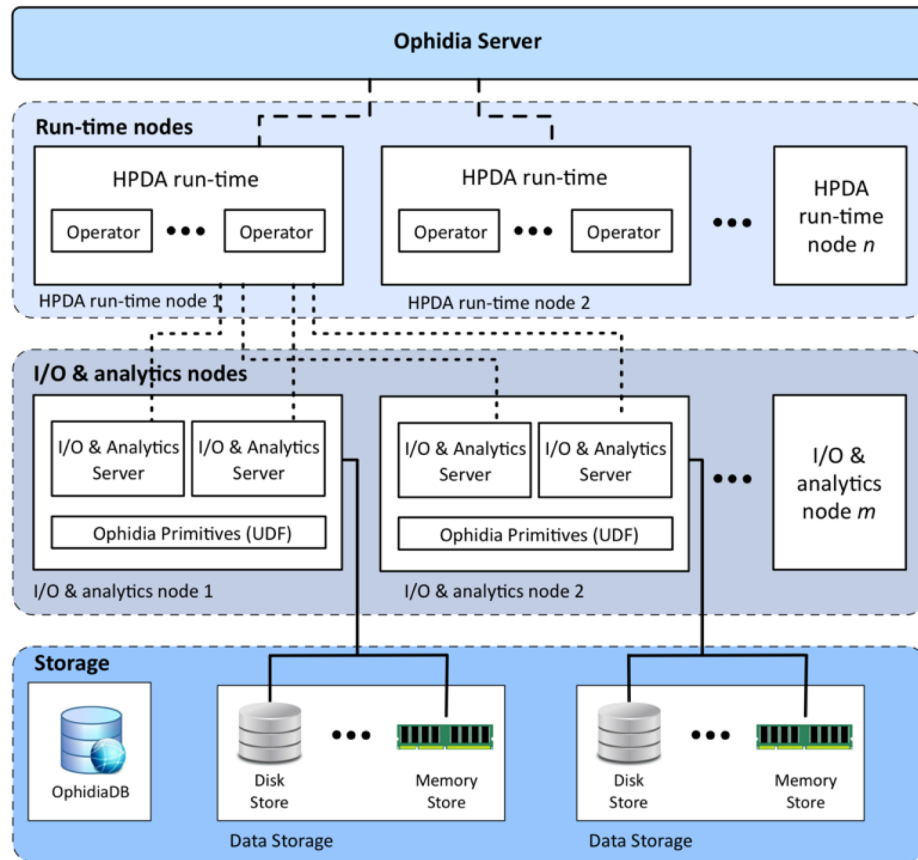
Modular and **extensible** software stack

Multi-level **parallel** execution model:

- Datacube-level (HTC-based)
- Fragment-level (HPC-based: MPI+X)

Support for **in-memory analytics** through array-based primitives (UDF kernels)

Data **partitioned** and **distributed** hierarchically over the storage and managed by the I/O & analytics nodes



D. Elia, S. Fiore and G. Aloisio, "Towards HPC and Big Data Analytics Convergence: Design and Experimental Evaluation of a HPDA Framework for eScience at Scale," in IEEE Access (Early Access), <https://ieeexplore.ieee.org/document/9428012>



On-demand deployment on HPC infrastructures

Target environment: *HPC cluster*

On-demand deployment of I/O & analytics servers

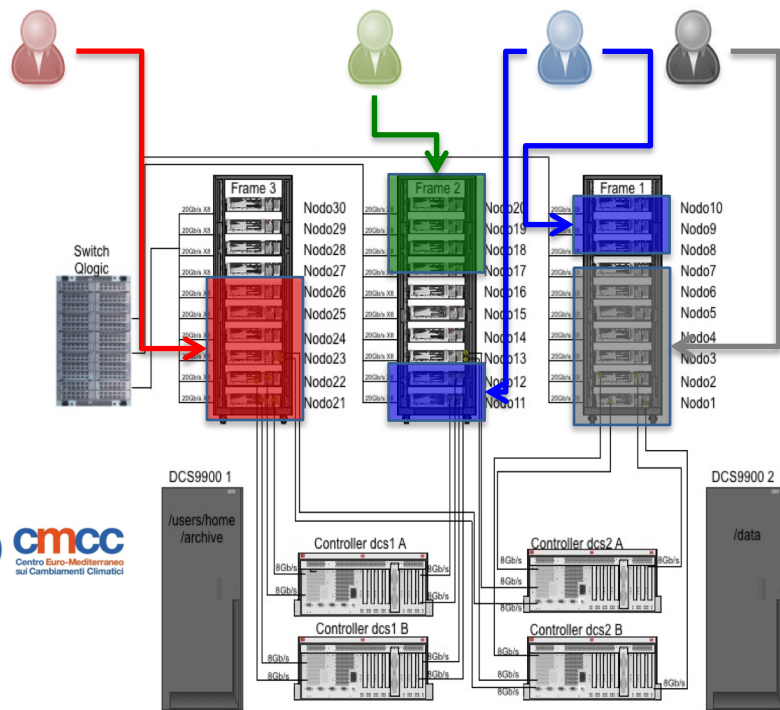
- `oph_cluster`
`action=deploy;nhost=64;cluster_name=new;`
- `oph_cluster` `action=undeploy;cluster_name=new;`

Transparent interaction with scheduling systems

Zeus SuperComputer at CMCC: 1.2 PetaFlops, 348 nodes



Multiple isolated instances can be deployed simultaneously by different teams/users



Ophidia operators

CLASS	PROCESSING TYPE	OPERATOR(S)
I/O	Parallel	OPH_IMPORTNC, OPH_EXPORTNC, OPH_CONCATNC, OPH_RANDUCUBE
Time series processing	Parallel	OPH_APPLY
Datacube reduction	Parallel	OPH_REDUCE, OPH_REDUCE2, OPH_AGGREGATE
Datacube subsetting	Parallel	OPH_SUBSET
Datacube combination	Parallel	OPH_INTERCUBE, OPH_MERGE CUBES
Datacube structure manipulation	Parallel	OPH_SPLIT, OPH_MERGE, OPH_ROLLUP, OPH_DRILLDOWN, OPH_PERMUTE
Datacube/file system management	Sequential	OPH_DELETE, OPH_FOLDER, OPH_FS
Metadata management	Sequential	OPH_METADATA, OPH_CUBEIO, OPH_CUBESHEMA
Datacube exploration	Sequential	OPH_EXPLORECUBE, OPH_EXPLORENC

About 50 operators for data and metadata processing

Ophidia operators documentation: <http://ophidia.cmcc.it/documentation/users/operators/index.html>



“data” operators

```
[12..3289] >> oph_reduce cube=http://127.0.0.1/ophidia/418/12717;operation=avg;ncores=2;nthreads=2;
```

[Request]:

```
operator=oph_reduce;cube=http://127.0.0.1/ophidia/418/12717;operation=avg;ncores=2;nthreads=2;sessionid=http://127.0.0.1/ophidia/sessions/127028404128222463341617004437753289/experiment;exec_mode=sync;cwd=/;cdd=/;host_partition=auto;
```

[JobID]:

```
http://127.0.0.1/ophidia/sessions/127028404128222463341617004437753289/experiment?239#582
```

[Response]:

Output Cube

```
http://127.0.0.1/ophidia/418/12722
```

Execution time: 0.35 seconds

```
[12..3289] >> oph_aggregate operation=avg;ncores=2;nthreads=2;
```

[Request]:

```
operator=oph_aggregate;operation=avg;ncores=2;nthreads=2;sessionid=http://127.0.0.1/ophidia/sessions/127028404128222463341617004437753289/experiment;exec_mode=sync;cube=http://127.0.0.1/ophidia/418/12722;cwd=/;cdd=/;host_partition=auto;
```

[JobID]:

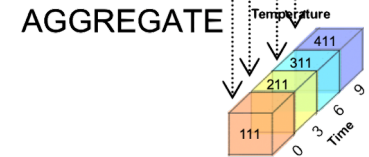
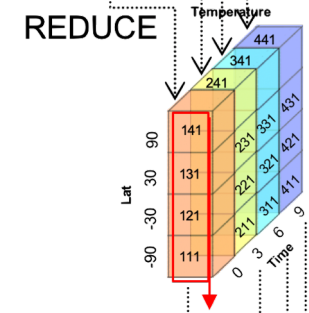
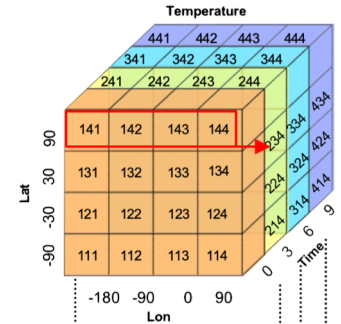
```
http://127.0.0.1/ophidia/sessions/127028404128222463341617004437753289/experiment?240#584
```

[Response]:

Output Cube

```
http://127.0.0.1/ophidia/418/12723
```

Execution time: 0.17 seconds



“data” operators

```
[37..4416] >> oph_explorecube cube=http://127.0.0.1/ophidia/35/67;subset_dims=lat|lon|time;subset_filter=39:42|15:19|1:275;show_time=yes;
[Request]:
operator=oph_explorecube;cube=http://127.0.0.1/ophidia/35/67;subset_dims=lat|lon|time;subset_filter=39:42|15:19|1:275;show_time=yes;sessionId=http://127.0.0.1/ophidia/sessions/374383780832141666641463737283924416/experiment;exec_mode=sync;ncores=1;cwd=/;
```

```
[JobID]:
http://127.0.0.1/ophidia/sessions/374383780832141666641463737283924416/experiment?106#224
```

```
[Response]:
tos
---
```

lat	lon	tos
39.500000	15.000000	1.00000002e+20, 1.00000002e+20, 1.00000002e+20, 1.00000002e+20, 1.00000002e+20, 1.00000002e+20, 1.00000002e+20, 1.00000002e+20, 1.00000002e+20, 1.00000002e+20
39.500000	17.000000	287.3930664062, 286.8287048340, 286.5860595703, 286.9228210449, 288.5254516602, 292.3968200684, 295.8656921387, 297.2062072754, 295.7126464844
39.500000	19.000000	287.6926879883, 287.0508117676, 286.7896118164, 287.0781555176, 288.6802062988, 292.6882629395, 296.4769287109, 297.6632385254, 296.3418273926
40.500000	15.000000	1.00000002e+20, 1.00000002e+20, 1.00000002e+20, 1.00000002e+20, 1.00000002e+20, 1.00000002e+20, 1.00000002e+20, 1.00000002e+20, 1.00000002e+20, 1.00000002e+20
40.500000	17.000000	287.1098632812, 286.5683593750, 286.2949829102, 286.5216674805, 288.0316772461, 291.7698974609, 295.4139709473, 296.8489685059, 295.4132995605
40.500000	19.000000	287.4010009766, 286.7818298340, 286.4914245605, 286.7260742188, 288.3006286621, 292.1842346191, 296.0237731934, 297.2694702148, 295.9751892090
41.500000	15.000000	1.00000002e+20, 1.00000002e+20, 1.00000002e+20, 1.00000002e+20, 1.00000002e+20, 1.00000002e+20, 1.00000002e+20, 1.00000002e+20, 1.00000002e+20, 1.00000002e+20
41.500000	17.000000	286.5835876465, 286.0175781250, 285.7146911621, 285.9142761230, 287.4476623535, 291.1032104492, 294.7090454102, 296.0852355957, 294.7053222656
41.500000	19.000000	286.9717712402, 286.3946838379, 286.0617675781, 286.1446228027, 287.6101989746, 291.2955017090, 295.2700195312, 296.5146179199, 295.3194274902

Summary

Selected 9 rows out of 9



“metadata” operators

[37..4416] >> oph_cubeio

[Request]:

operator=oph_cubeio;sessionId=http://127.0.0.1/ophidia/sessions/374383780832141666641463737283924416/experiment;exec_mode=sync;ncores=1;cube=http://127.0.0.1/ophidia/35/74;cwd=/
;

[JobID]:

http://127.0.0.1/ophidia/sessions/374383780832141666641463737283924416/experiment?82#176

[Response]:

Cube Provenance

INPUT CUBE	OPERATION	OUTPUT CUBE	SOURCE
	ROOT	http://127.0.0.1/ophidia/35/66	/re
	ROOT	http://127.0.0.1/ophidia/35/67	/re
http://127.0.0.1/ophidia/35/66 - http://127.0.0.1/ophidia/35/67	oph_intercube	http://127.0.0.1/ophidia/35/70	
http://127.0.0.1/ophidia/35/70	oph_reduce	http://127.0.0.1/ophidia/35/71	
http://127.0.0.1/ophidia/35/71	oph_merge	http://127.0.0.1/ophidia/35/72	
http://127.0.0.1/ophidia/35/72	oph_aggregate	http://127.0.0.1/ophidia/35/74	

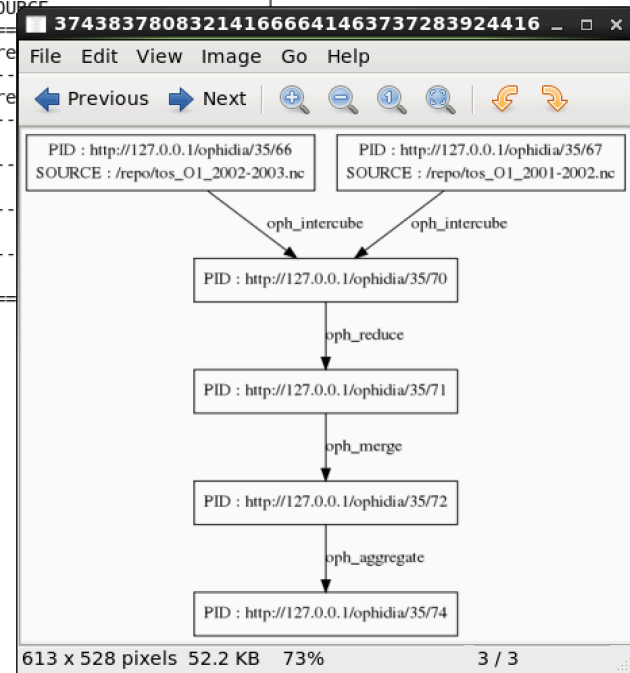
Cube Provenance Graph

Directed Graph DOT string :

```

digraph DG {
  node      [shape=box]
  0         [label="PID : http://127.0.0.1/ophidia/35/74\n"]
  1         [label="PID : http://127.0.0.1/ophidia/35/72\n"]
  2         [label="PID : http://127.0.0.1/ophidia/35/71\n"]
  3         [label="PID : http://127.0.0.1/ophidia/35/70\n"]
  4         [label="PID : http://127.0.0.1/ophidia/35/66\nSOURCE : /repo/tos_01_2002-2003.nc\n"]
  5         [label="PID : http://127.0.0.1/ophidia/35/67\nSOURCE : /repo/tos_01_2001-2002.nc\n"]
  1->0     [label="oph_aggregate"]
  2->1     [label="oph_merge"]
  3->2     [label="oph_merge"]
  4->3     [label="oph_reduce"]
  5->3     [label="oph_intercube"]
}

```



Programmatic support for data science applications

PyOphidia is a Python module to interact with the Ophidia framework.

It provides a programmatic access to Ophidia features, allowing:

- *Submission of commands to the Ophidia Server and retrieval of the results*
- *Management of (remote) data objects in the form of datacubes*
- *Easy exploitation from Jupyter Notebooks and integration with other Python modules*

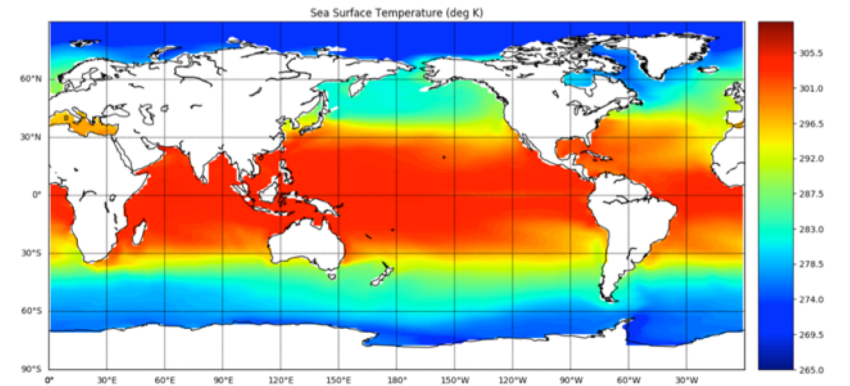
```
from PyOphidia import cube, client
cube.Cube.setclient(read_env=True)

mycube =
cube.Cube.importnc(src_path='/public/data/ecas_training
/file.nc', measure='tos', imp_dim='time',
import_metadata='yes', ncores=5)
mycube2 = mycube.reduce(operation='max',ncores=5)
mycube3 = mycube2.rollup(ncores=5)
data = mycube3.export_array()

mycube3.exportnc2(output_path='/home/test',
export_metadata='yes')
```

<https://pypi.org/project/PyOphidia/>

<https://anaconda.org/conda-forge/pyophidia>



Export result to NetCDF file

```
] : mycube3.exportnc2(output_path='/home/' + cube.Cube.client.username,export_metadata='yes')
```



Python and HPC infrastructure transparency

PyOphidia class hides the HPC environment complexity

```
In [ ]: from PyOphidia import cube, client  
cube.Cube.setclient(read_env=True)
```

Dynamic I/O & Analytics
nodes allocation

```
In [ ]: cube.Cube.cluster(action='deploy', host_partition='test_partition', nhost=4)
```

```
In [ ]: myCube = cube.Cube(src_path='/work/ophidia/tests/tasmax_day_CMCC-CESM_rcp85.nc',  
measure='tasmax', import_metadata='yes', imp_dim='time', description='Max Temps',  
nfrag=16, nhosts=4,  
host_partition='test2',  
ncores=2, nthreads=8  
)
```

Data partitioning
and distribution

Framework
operator
parallelism

```
In [ ]: myCube2 = maxtemp.apply(  
query="oph_predicate('oph_float','oph_int',measure,'x-298.15','>0','1','0')",  
ncores=2, nthreads=8  
)
```

```
In [ ]: myCube3 = myCube2.subset(subset_filter=1, subset_dims='time')
```

Ophidia-notebook data
translation and transfer

```
In [ ]: pythonData = myCube3.export_array(show_time='yes')
```

```
In [ ]: print(pythonData)
```

```
In [ ]: cube.Cube.cluster(action='undeploy', host_partition='test_partition')
```

I/O & Analytics nodes
undeployment

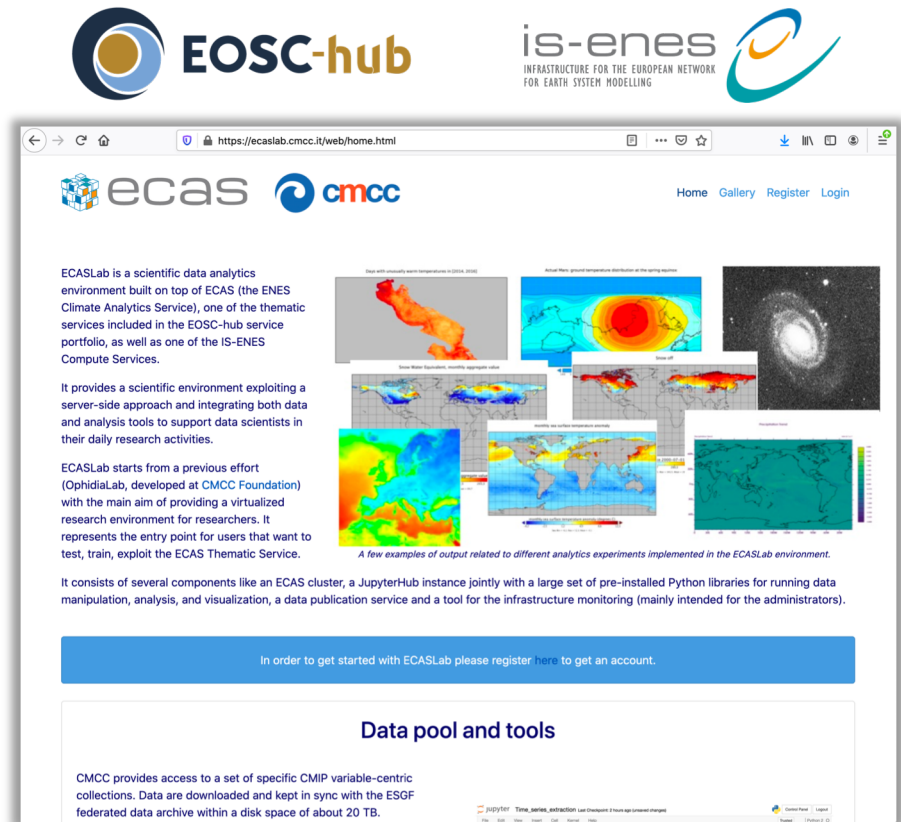


ENES Climate Analytics Service (ECAS)

- ECAS is one of the *EOSC-Hub Thematic Services* as well as one of *IS-ENES3 compute services*
- ECAS builds on top of the *Ophidia HPDA framework*, integrated with components from *INDIGO-DataCloud*, *EUDAT* and *EGI*
- Integrates computing resources with datasets and data analytics tools
- It provides a user-friendly environment based on *Jupyter Notebooks* and well-known *Python modules*

EOSC Portal Catalogue: <https://marketplace.eosc-portal.eu/services/enes-climate-analytics-service>

Climate Analytics Service: <https://portal.enes.org/data/data-metadata-service/climate-analytics-service>



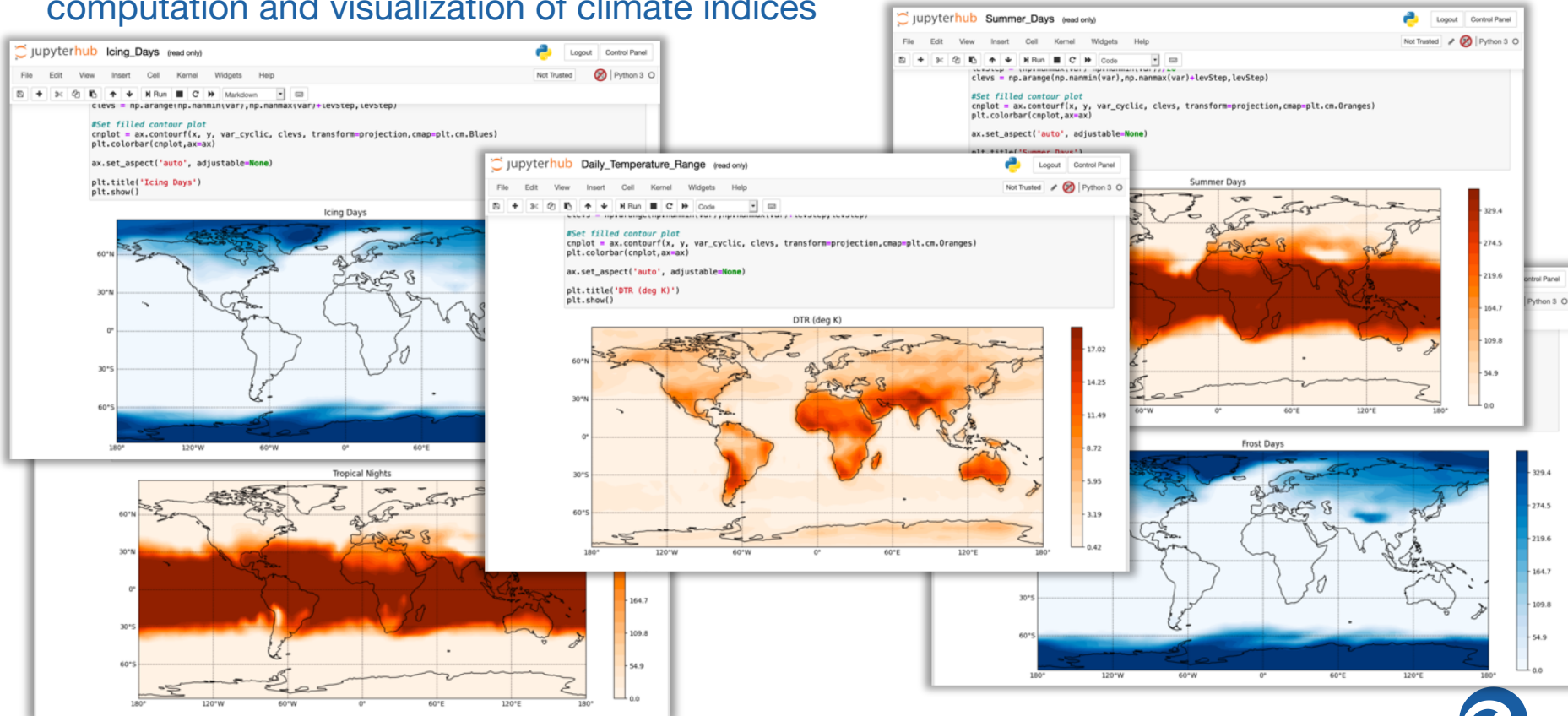
The screenshot displays the ECASLab website interface. At the top, there are logos for EOSC-hub, is-enes (Infrastructure for the European Network for Earth System Modelling), and CMCC. The main content area features a grid of various climate data visualizations, including maps of Europe, global maps, and satellite imagery. Text on the page describes ECASLab as a scientific data analytics environment built on top of ECAS (the ENES Climate Analytics Service), one of the thematic services included in the EOSC-hub service portfolio, as well as one of the IS-ENES Compute Services. It provides a scientific environment exploiting a server-side approach and integrating both data and analysis tools to support data scientists in their daily research activities. ECASLab starts from a previous effort (OphidiaLab, developed at CMCC Foundation) with the main aim of providing a virtualized research environment for researchers. It represents the entry point for users that want to test, train, exploit the ECAS Thematic Service. A blue button prompts users to register here to get an account. Below this, a section titled 'Data pool and tools' states that CMCC provides access to a set of specific CMIP variable-centric collections. Data are downloaded and kept in sync with the ESGF federated data archive within a disk space of about 20 TB. The bottom of the page shows a JupyterLab interface with a terminal window displaying the command 'Time_series_analysis' and a 'Logout' button.

ECAS @ CMCC <https://ecaslab.cmcc.it>



Computation of climate indices

PyOphidia can be combined with other Python libraries (e.g., cartopy/matplotlib) for the computation and visualization of climate indices



Ophidia in ESiWACE2 project

ESiWACE2 – Excellence in Simulation of Weather and Climate in Europe, Phase 2

- (1) Enabling leading European weather and climate models to leverage the available performance of pre-exascale systems with regard to both compute and data capacity in 2021.*
- (2) Preparing the weather and climate community to be able to make use of exascale systems when they become available.*

Ophidia is one of the applications considered in the frame of **ESiWACE2** (WP4, WP5):

- Integration with **Earth System Data Middleware (ESDM)** (<https://github.com/ESiWACE/esdm>) for I/O over heterogeneous storage systems
- One of the applications (HPDA) targeted by the **ESDM-PAV**
 - Extensions for **in-flight analytics** are being developed
 - Some HPDA **scientific use cases** defined (preliminary implementation)



Ophidia



esiwace
CENTRE OF EXCELLENCE IN SIMULATION OF WEATHER
AND CLIMATE IN EUROPE

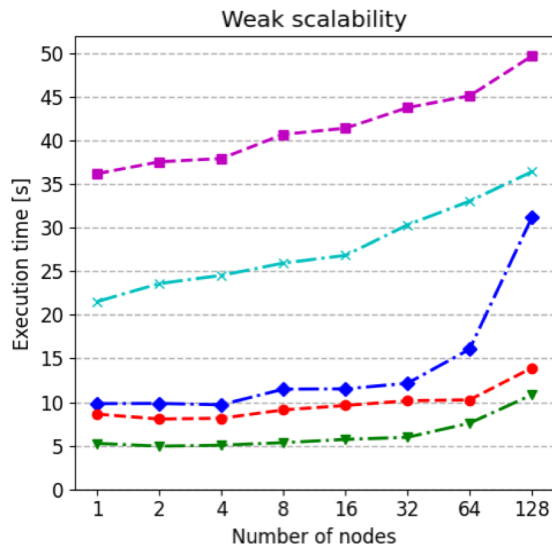


Ophidia HPDA framework benchmark

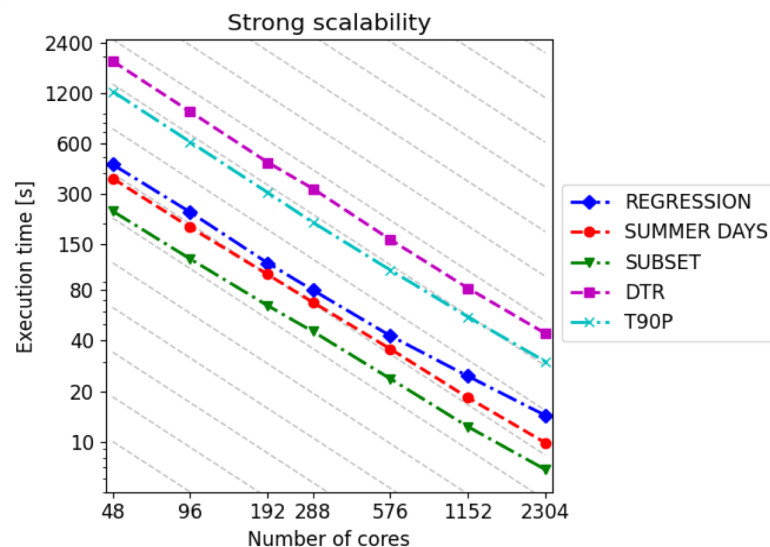
Goal: benchmarking, tuning and optimisation over a large-scale HPC machine of the Ophidia HPDA framework

Evaluate the scalability of Ophidia analytics kernels on a few thousands of cores:

- various **strong** and **weak scalability** tests performed
- **good scalability** in the majority of cases until **3k cores**



Data size per node: 67GiB



Data size fixed: 3.2TiB

We acknowledge PRACE for awarding access to MareNostrum 4 at Barcelona Supercomputing Center (BSC), Spain and the support provided by BSC (PRACE resources for CoE, in the context of ESiWACE).

D. Elia, S. Fiore and G. Aloisio, "Towards HPC and Big Data Analytics Convergence: Design and Experimental Evaluation of a HPDA Framework for eScience at Scale," in IEEE Access (Early Access), <https://ieeexplore.ieee.org/document/9428012>



Summary and future activities

Recap

- Ophidia HPDA framework address climate data challenges at scale
 - Tailored for HPC infrastructures
 - Provides (user-friendly) features for building complex analysis
 - Experimental results show scalability up to few thousand cores

Future activities

- Further benchmarks of the HPDA framework with large-scale applications/workflows
- Address portability with HPC-enabled containerization
- Full integration of Ophidia with the ESDM/ESDM-PAV (ESiWACE2 activities)



Thank you for the attention!

***ESiWACE2** has received funding from the European Union's Horizon 2020 research and innovation programme under grant agreement No. 823988*

***IS-ENES3** has received funding from the European Union's Horizon 2020 research and innovation programme under grant agreement No. 824084*



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