

# Performance study of Earth System Models

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### **Motivation**

Global **ESMs are complex** systems made of heterogeneous components, that require a substantial amount of effort to work correctly and use the resources wisely. The goal of running these models using grids on the order of 1 km will be impossible to achieve without using adequate optimization techniques.

**Earth System Models** (ESMs) consume billions of computing hours every year. Our motivation is to optimize these models to **save both time and energy**, and **allow further increases in resolution and complexity** to improve their capabilities of simulating small-scale features and reduce the impact of the parametrizations.



Scalability Study: How efficient is your model?

Daily mean of: sea surface temperature (NEMO, left) and 2m air temperature (IFS, right) from a ORCA12-T1279 EC-Earth coupled simulation. The influence of the sea surface temperature (cold eddy detaching from the Gulf Stream) on lowest layer of the atmosphere can be clearly observed.

Methodology to evaluate Earth System Models

Complex models require complex studies. Several techniques are needed to optimize an ESM:

#### Mathematical and Computational study

Understand the algorithms used and the overhead or drawbacks introduced by the parallel implementation.

#### **Scalability Study**

The evaluation of an ESM requires deep knowledge about the complex workflows and experiments done [1].

#### **Profiling Study**

General and oriented profiling analysis for ESMs [2].

#### Introducing optimizations

Computational improvements, optimizing or introducing new approaches [3].

#### **Reproducibility study**

Keeping in mind the chaotic nature of climate models, evaluate the impact in accuracy and reproducibility of the model after applying the proposed optimizations [4]. Climate predictions have complex workflows. New metrics are needed to evaluate \_\_\_\_\_\_ the computational efficiency.

## 

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 The simulation does not only involve the execution of a model during a sequence of time steps (represented by the sim job).

• The experiment adds complexity in the horizontal for ensembles (members with perturbations in the initial conditions).

 The experiment adds complexity in the vertical, running long simulations divided into chunks and including pre- and post-processing.

|  | Metric                      |            | used to evaluate  |
|--|-----------------------------|------------|---|
|  | Simulation Year Per D       | Day (SYPD) | how efficient is your sim job per each year of the simulation                           |
|  | Core-hours Per Year (CHPY)  |            | how efficient is your sim job with respect to the number of parallel resources used     |
|  | Queue Time                  |            | how much time are all the jobs waiting in the queue                                     |
|  | Actual SYPD                 |            | how affect queue time to the complete experiment, from the first to the last sim job    |
|  | Energy Cost Per Year (JCPY) |            | how much energy is needed per each year of simulation                                   |
|  | Memory Bloat                |            | how much memory is needed for the execution of the sim job                              |
|  | Data Output Cost            |            | how much time is used for the output post-processing, only done for specific time steps |
|  | Coupling Cost               |            | how much time is wasted in waiting time, produced by irregular executions               |

![](_page_0_Figure_28.jpeg)

The execution of a coupled model is complex. Different components run in parallel, exchanging some information and adding some extra overhead (Communication and Interpolation Time). Load imbalances produce extra waiting time as well.
The execution of each component could involve some irregular extra overhead, such as output processing or calculations occasionally done, such as radiation.

Profiling Study: What can you see in a trace?

Performance tools [5] are essential

![](_page_0_Picture_32.jpeg)

![](_page_0_Picture_33.jpeg)

to study the behavior of ESMs: **-Extrae**: is a package used to instrument the code. It generates trace-files with hardware counters, MPI messages and other information.

- **Paraver**: is a browser used to analyze both visually and analytically trace-files.

-Dimemas, Clustering, Folding...: Other tools to evaluate ideal conditions, gather performance data...

-The view of a trace consists of **threads or processes** on the Y axis and the **timeline** on the X axis.

-The base trace (figure on the right) shows **MPI functions**, where each type of call is identified by a color. Blue color represents a computation area.

-Other traces can contain **PAPI** events to collect information regarding the microprocessor performance. Colors represent in this case a gradient between maximum and minimum values.

References

#### About us

The **Performance Team** is part of the **Computational Earth Sciences** Group (CES), at the **BSC Earth Sciences Department**.

The mission of **CES** is to ensure the efficient use of the computational resources by the Earth scientists, to develop Earth Sciences related HPC user-friendly software and to provide computational services to the department and the Earth Sciences community.

![](_page_0_Picture_44.jpeg)

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![](_page_0_Picture_46.jpeg)

**Marenostrum IV Supercomputer** 

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BSC

![](_page_0_Picture_54.jpeg)

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This poster can be downloaded at https://earth.bsc.es/wiki/doku.php?id=library:external:posters Corresponding authors: <u>mario.acosta@bsc.es</u>

![](_page_0_Picture_57.jpeg)

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