



Towards Safe and Fearless Lossy Compression of Weather and Climate Model Data

**Juniper Tyree^[1,5], Sara Faghih-Naini^[2,5], Milan Klöwer^[3], Tim Reichelt^[3,6],
Peter Dueben^[2,5], Karsten Peters-von Gehlen^[4,5], and Heikki J. Järvinen^[1,5]**

[1] INAR, University of Helsinki, [2] ECMWF, [3] University of Oxford, [4] DKRZ, [5] ESIWACE3, [6] Embed2Scale

CONTINENTS Webinar, 28.05.2025



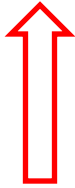
EuroHPC
Joint Undertaking

Funded by the European Union. This work has received funding from the European High Performance Computing Joint Undertaking (JU) under grant agreement No 101093054.



Co-funded by
the European Union





**Feel free to take pictures of
most slides ...
... unless the camera is red**



Towards exascale weather and climate simulations

Supporting the community of weather and climate modelling in Europe



...

Develop tools to tackle the data challenge of high-resolution models

03

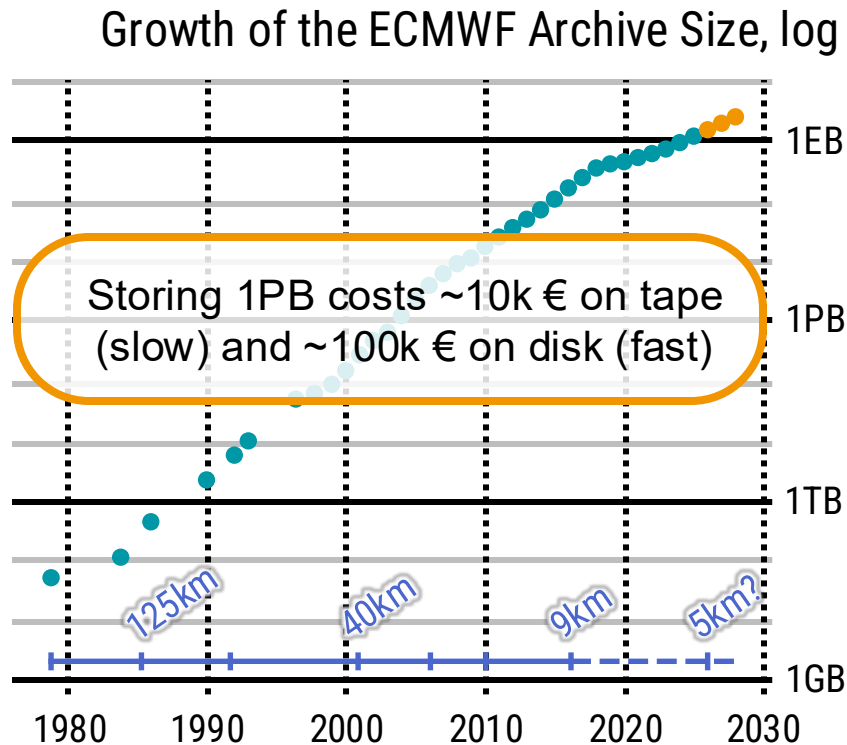
Domain-specific data compression

3.2



**What is Lossy Compression,
Why do we need it,
and Why is it not used?**

Problem: Growth in Weather & Climate Model Data Volumes



European Centre for Medium-Range Weather Forecasts archive size
Dataset provided by Mike Hawkins (ECMWF), Figure adapted from Milan Klöwer (Oxford)

We need
**lossy
compression**
But losing
data accuracy
is *scary*

Problem: Fear of silently losing problem-specific information

Computer scientist: “I can give you great savings by using lossy compression.”

Domain scientist: “Nice! ... Actually, did we remember to keep the ...

- **logarithmic error** norms small for **specific humidity**
- L_2 **error** small for **temperature**
- **budgets** correct when **integrating** over long climate trajectories
- L_∞ **norm** small for **extreme precipitation events**
- **integral in the vertical** correct even though it sums across **several orders of magnitude**
- **delta at high precision** for CO_2
- ...

And how does lossy data compression change the **enthalpy budget** when cold rain is entering a warm ocean and changing the vertical layering of the vertical column in an ocean model???”

Computer scientist: “Just check the compressed outputs to see if anything broke”



Problem: We need lossy compression but are afraid of using it

Lossy Compression is crucial to handle Data Growth

Unclear safety requirements and error tolerances for lossy compression in weather and climate science

Lack of Trust in Lossy Compression
to uphold the safety requirements
for everyday domain-scientist users



**How do we bridge the gap
between lossy compression
and user needs to build trust?**



Our roadmap towards Safe, Trusted, and Fearless Lossy Compression



The Roadmap to *Fearless* Lossy Compression



Open Science Online Laboratory to Try it out and Convince Yourself

We don't even know what error tolerance we have!

Specify clear safety requirements

What compressor should I use?

Benchmark compressors for safety and performance

But my safety requirements are far stricter than yours!

Create safeguards to use any compressor without fear



Compression should be *Easy* & *Fearless*

**“We don’t want to disrupt the science, but
disrupt the data challenge”**

Heikki Järvinen, ESiWACE3 kick-off, BOG5 session, 01.02.2023



What are our safety requirements? What errors can we tolerate?

Juniper Tyree^[1,4], Milan Klöwer^[2], Sara Faghih-Naini^[3,4], and Peter Dueben^[3,4]

[1] INAR, University of Helsinki, [2] University of Oxford, [3] ECMWF, [4] ESIWACE3

Problem: Unclear which compression errors can be tolerated

Model outputs are stored with excessive precision

Trailing mantissa bits contain false information / noise

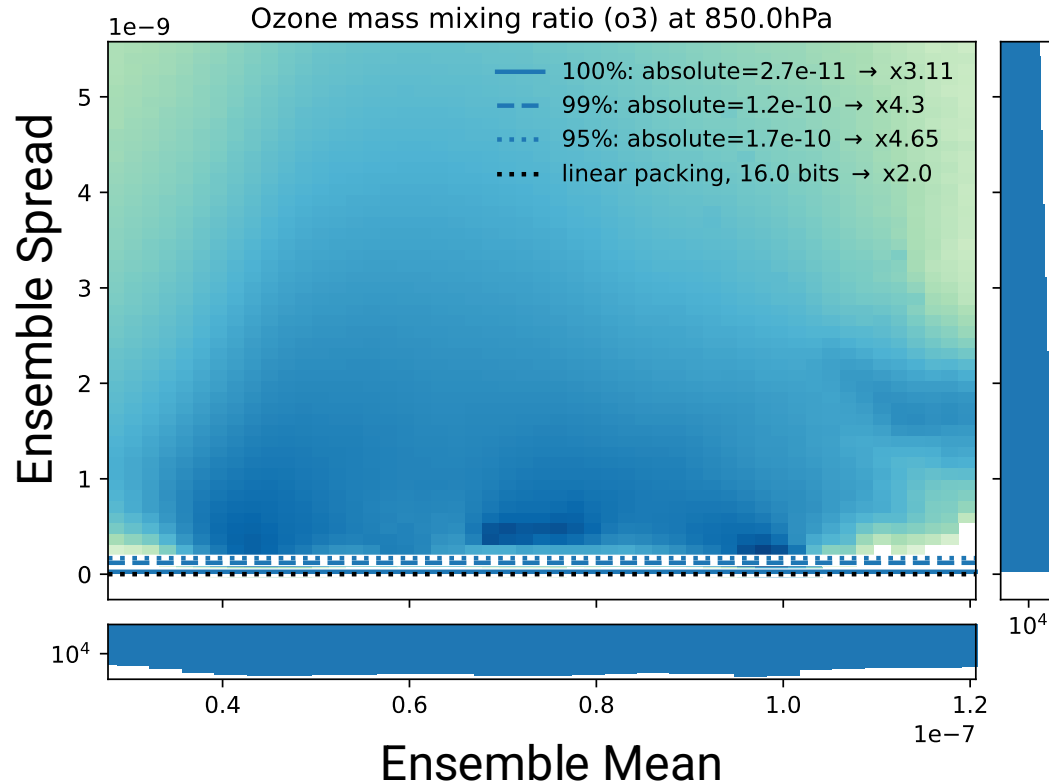
Tolerance varies by variable, resolution, use case

Model output archival must use conservative tolerances

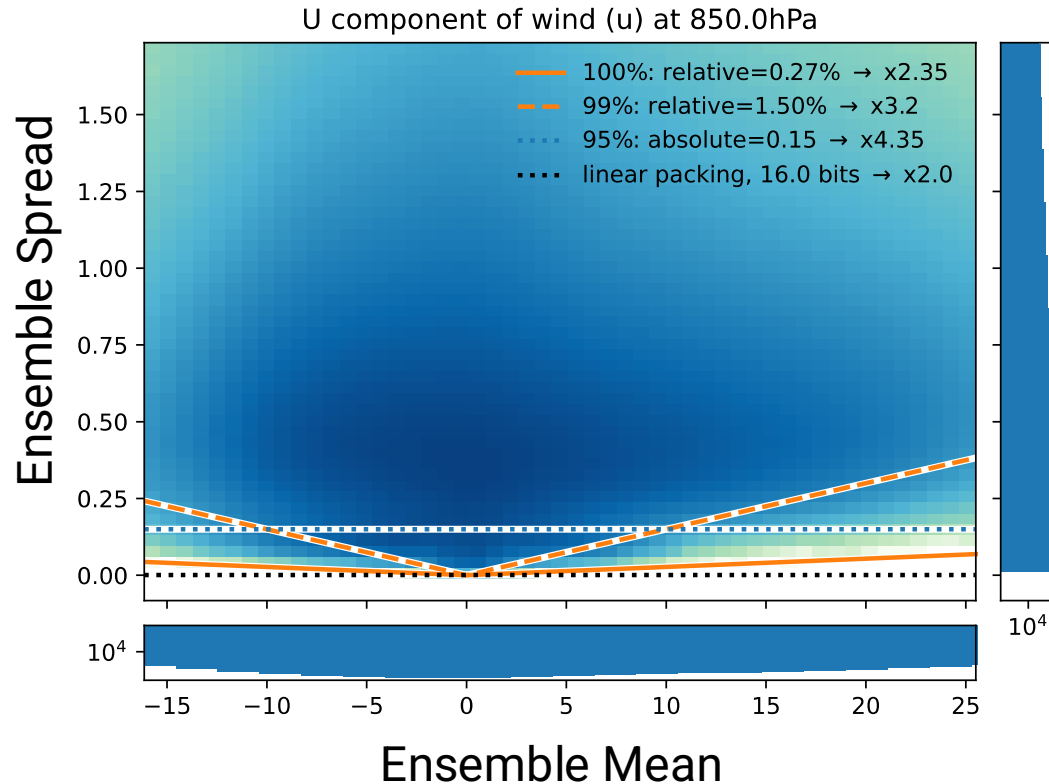
Community needs recommendations and standards

GRIB already uses lossy linear packing, improve on that

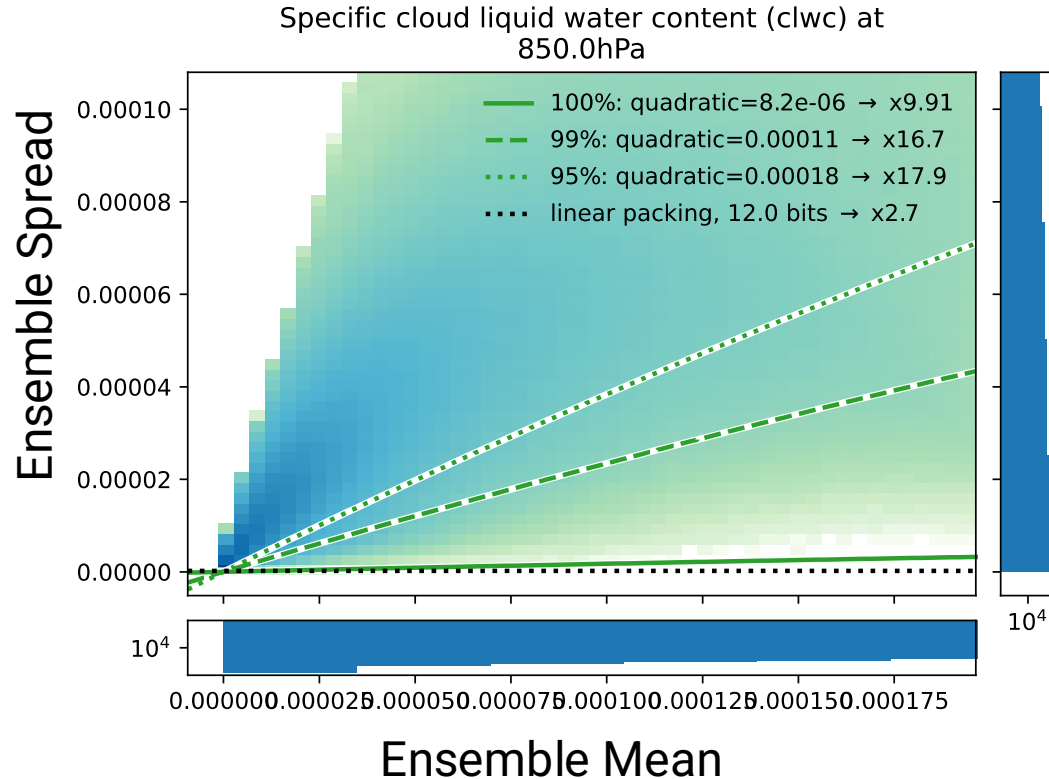
Results: Variables with absolute error bounds, e.g. O3, temp



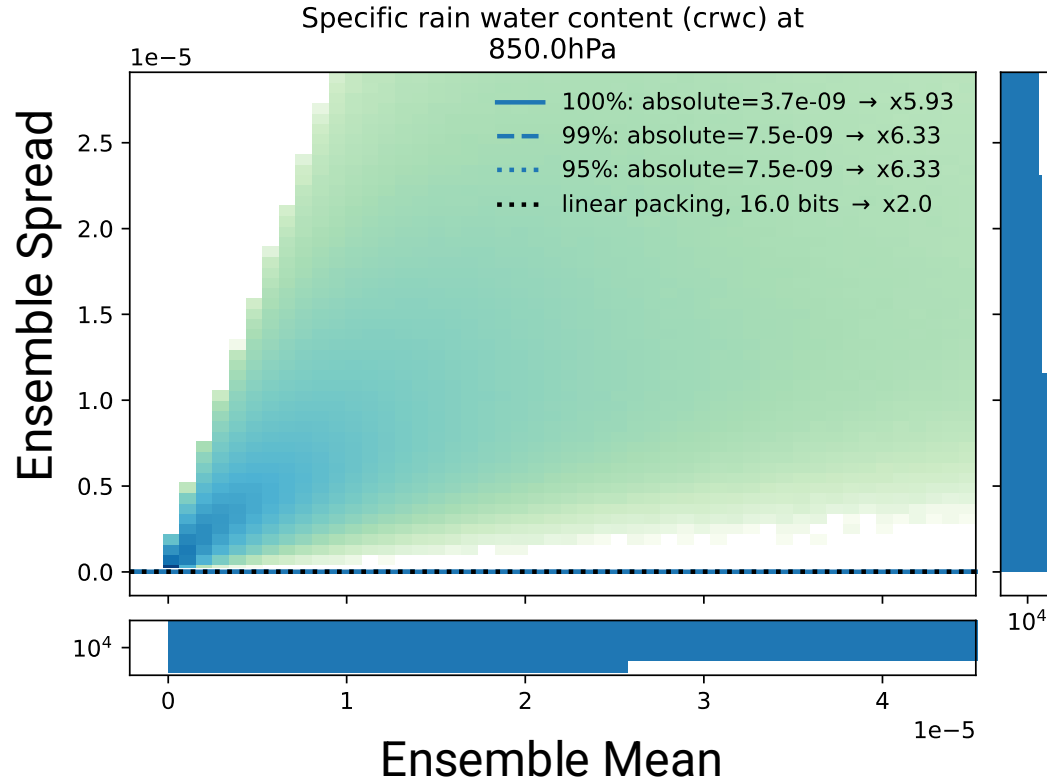
Results: Variables with relative error bounds, e.g. u/v wind



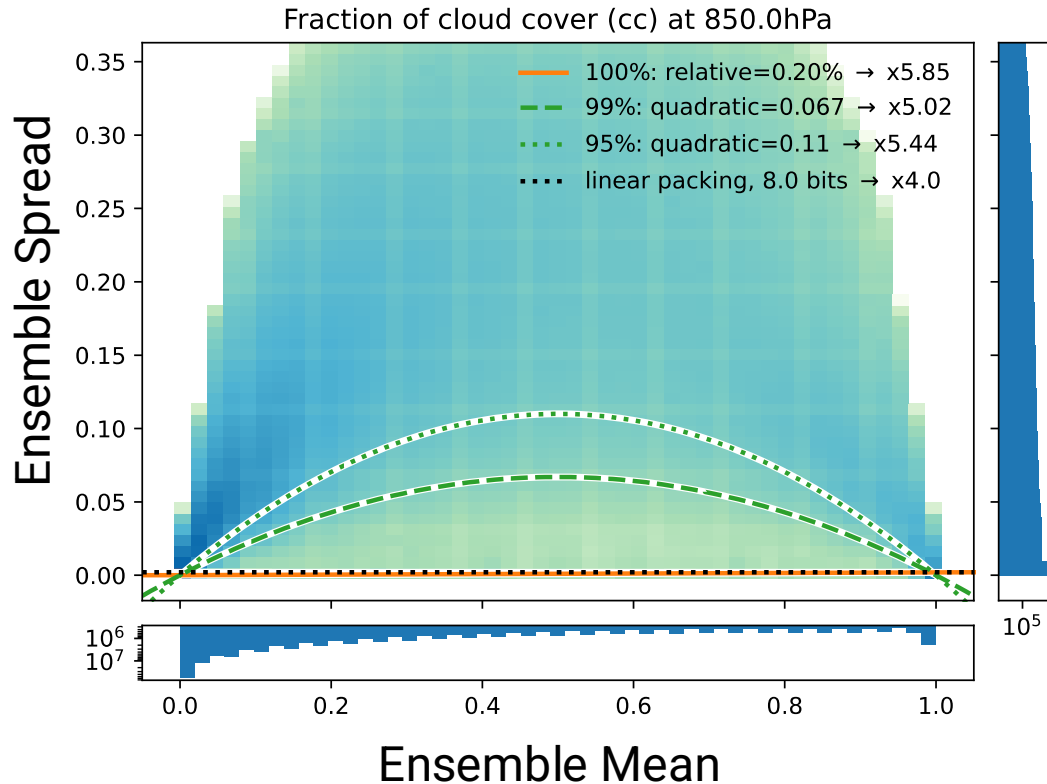
Results: Variables with relative error bounds, e.g. cloud water



Results: Variables with unclear error bounds, e.g. rainwater



Results: Variables with quadratic error bounds, e.g. cloud cover





Outlook: Next steps, publication, and future work

-Q2

Comparing expert and automatic bounds

Experience vs data analysis

Q2-Q3

Initial publication and table of error bounds

Three suggested levels of error bounds per variable

...

Outreach, Feedback, Refinement, Standards

Further feedback and analyses, replace linear packing



**How do existing
compressors perform?
Are they safe to use?**



ClimateBenchPress: A Benchmark for Compression of Climate Data

**Tim Reichelt^[1,7], Juniper Tyree^[2,8], Milan Klöwer^[1], Peter Dueben^[3,8], Bryan Lawrence^[4],
Dorit Hammerling^[5], Allison Baker^[6], Sara Faghieh-Naini^[3,8], and Philip Stier^[1,7]**

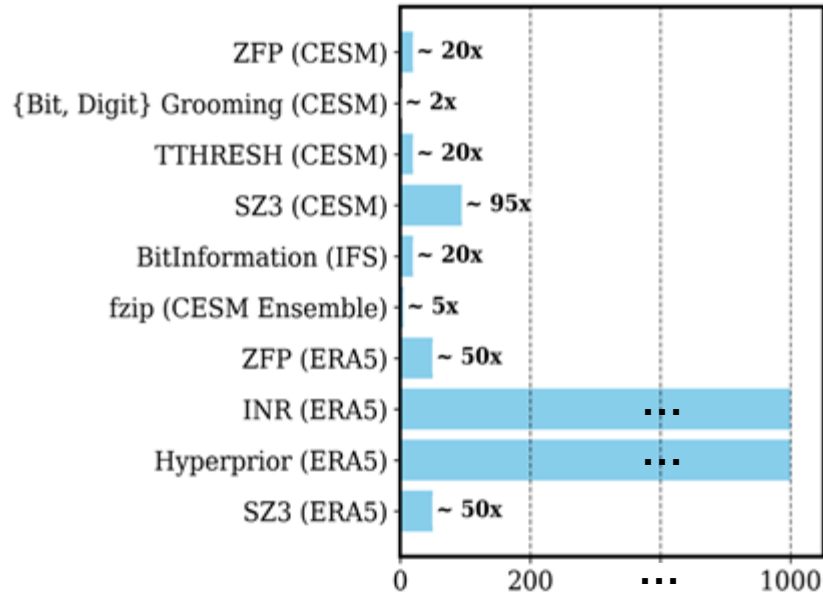
[1] University of Oxford, [2] INAR, University of Helsinki, [3] ECMWF, [4] NCAS, University of Reading,

[5] Colorado School of Mines, [6] NCAR, [7], Embed2Scale, [8] ESIWACE3

Problem: Why do we need a benchmark?

Compression ratios reported in the literature ...

are **difficult to compare** because they use different



- Datasets and **resolutions**
- **Variables** and dimensions
- Compression **error bounds**
- Measurement techniques

Aims: What do we want from a good benchmark?

Fair and Comparable Results

Same data, same error bounds (3 levels), same tests

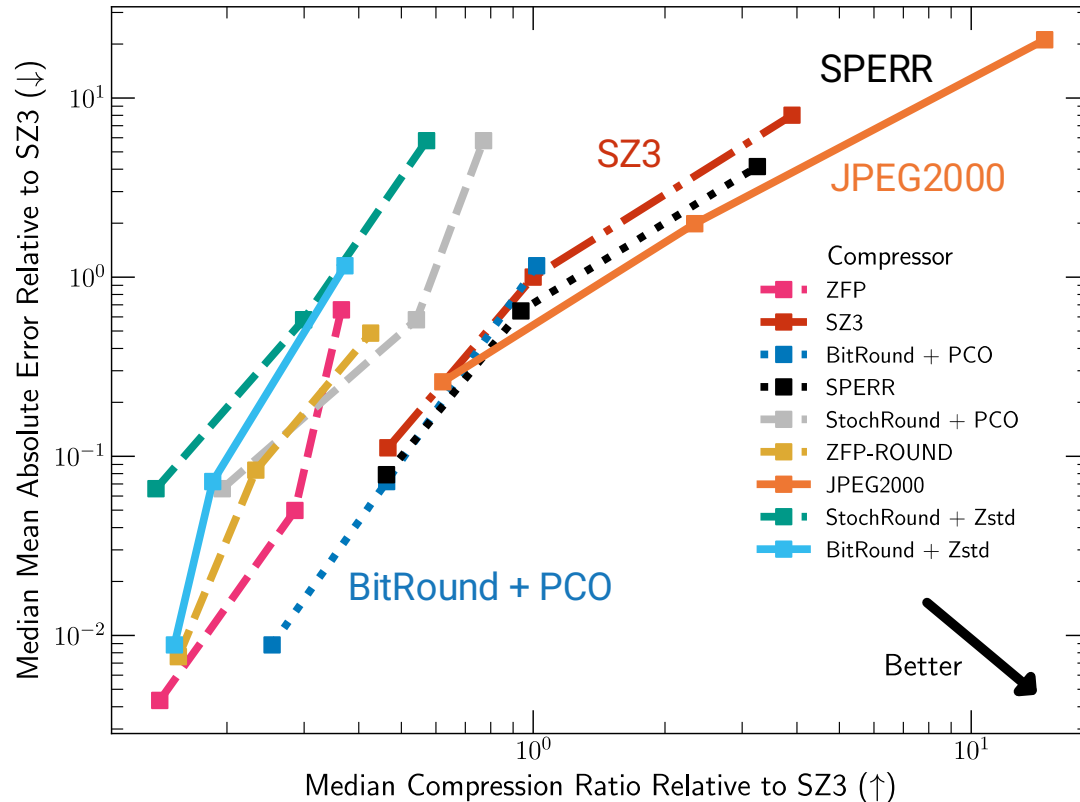
Interpretable Results from Representative Sources

2D & 3D from CMIP6, ERA5, NextGEMS, CAMS, Satellite

Ease of use and expansion

Modular and extensible, several dataset sizes

Results: Preliminary Insights, averaged across several variables



- **SZ3, SPERR, and JPEG2000:** highest compression
 - JPEG2000 only has global error bounds (not pointwise)
- **Bit Rounding + PCO** is competitive (but is quite conservative)

Outlook: Next steps, publication, and future work

Q1-Q2

Preparing the small version of the benchmark

Initial subset of data, compressors, and metrics

Q3-Q4

Initial publication and recommendations

Per-variable compressor scoreboard with trade-offs

2026-

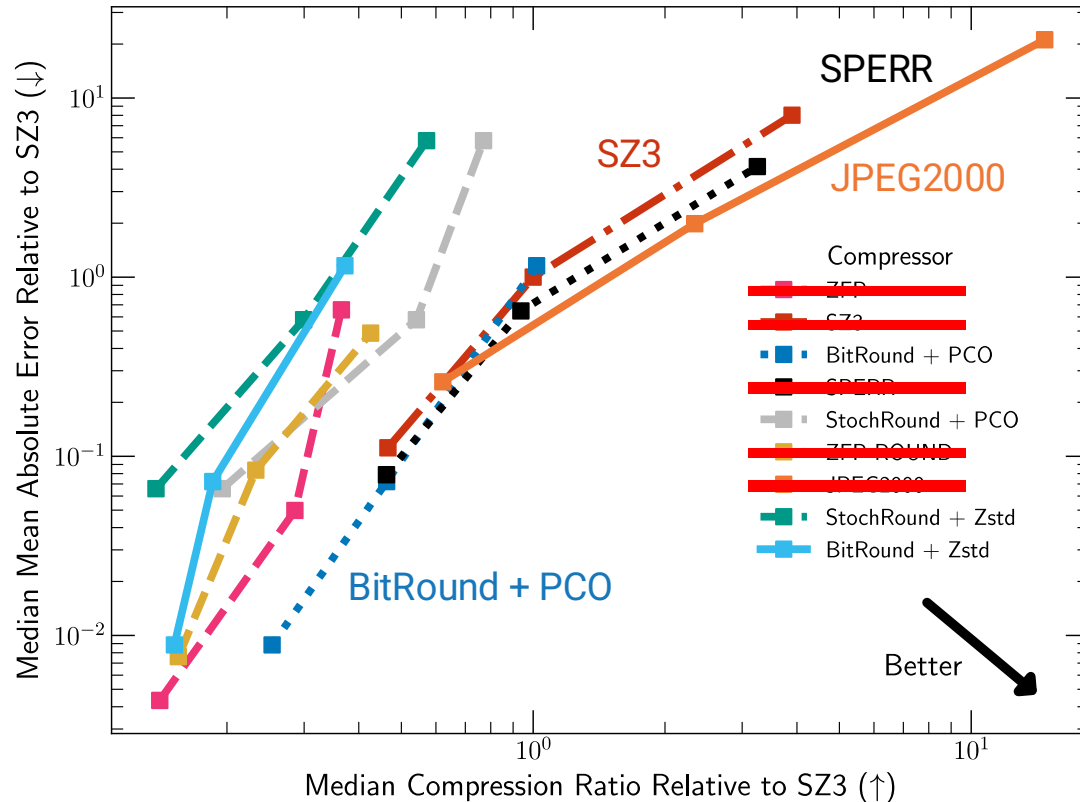
Expanded benchmark, renew recommendations

Larger datasets, more compressors (ML), more metrics



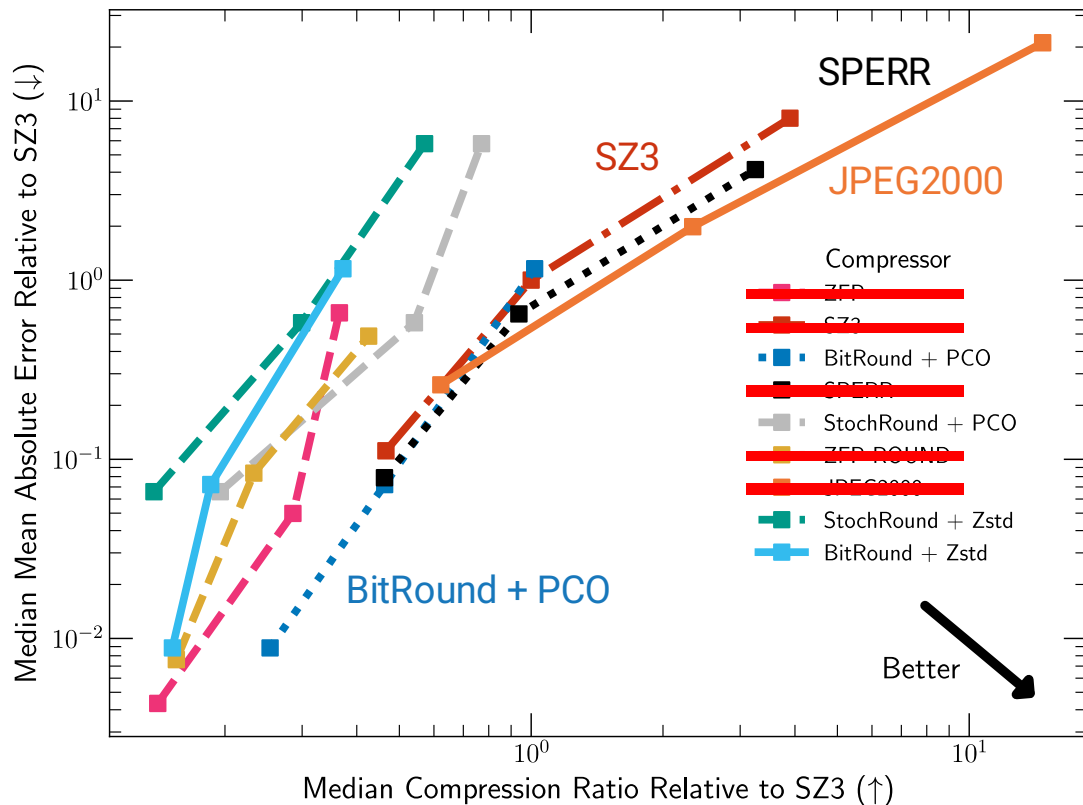
Spooky stories of what can go wrong with lossy compression

Problem: Many popular compressors don't handle edge cases



- **ZFP**: error bias, no NaNs, no Inf's
- **ZFP-ROUND**: no NaNs, no Inf's
- **SZ3**: no NaNs (v3.2)
- **SPERR**: no NaNs or big values, loose bounds
- **JPEG2000**: no NaNs, no Inf's, no pointwise error bounds

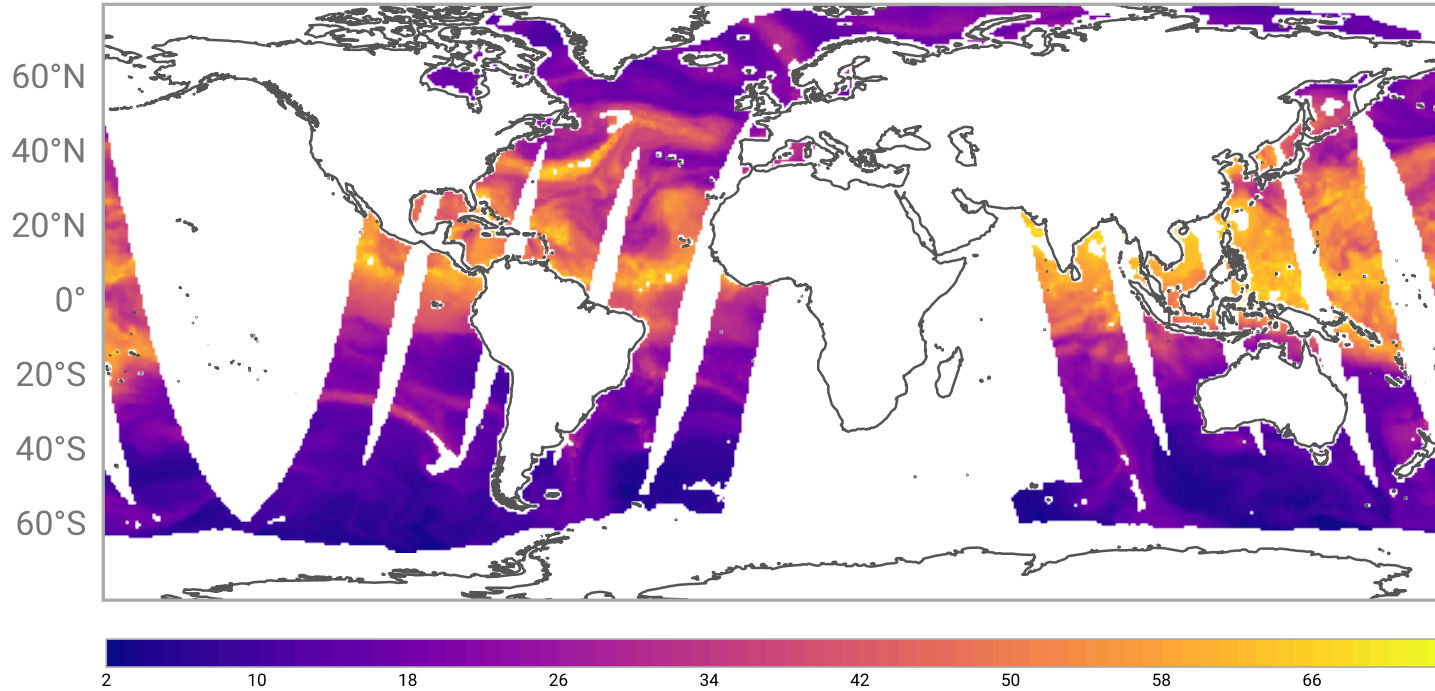
Problem: Many popular compressors don't handle edge cases



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 - **SPERR**: no NaNs or big values, loose bounds
 - **JPEG2000**: no NaNs, no Inf's, no pointwise error bounds
- Explicit failure to compress**

Problem: Hallucination in non-ML lossy compressors

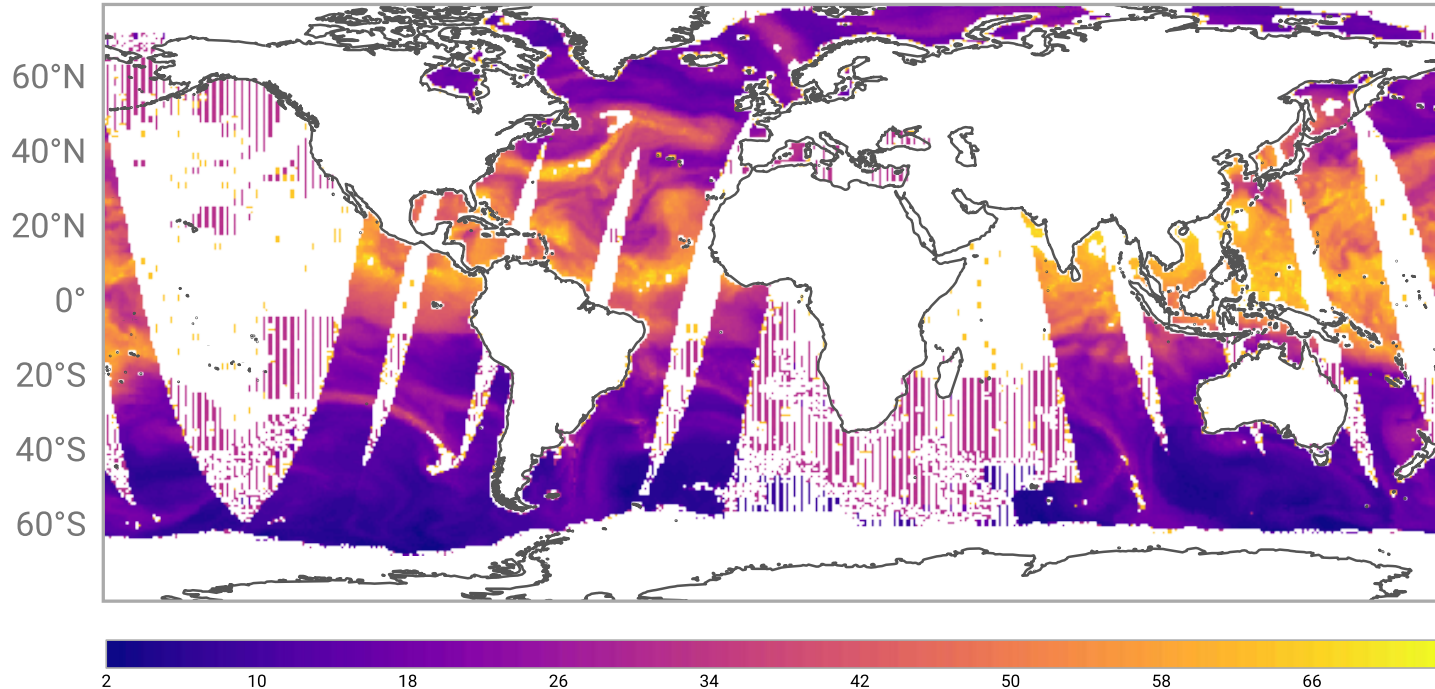
Original data, NaN values on land and from missing coverage



Water Vapor Path [$\text{kg} \cdot \text{m}^{-2}$], Satellite data at 06:00 on 02.08.2020

Problem: Hallucination in non-ML lossy compressors

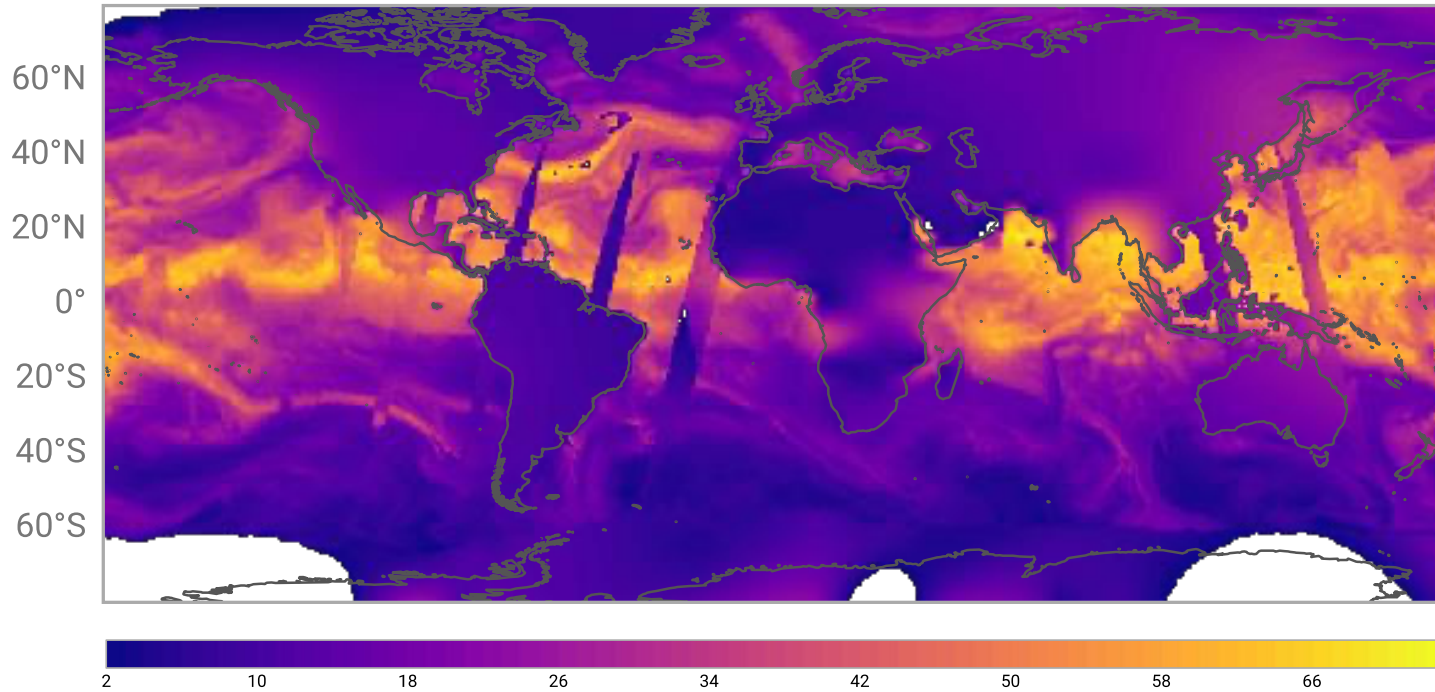
ZFP, absolute error bound $\epsilon_{\text{abs}} = 0.1 \text{ kg} \cdot \text{m}^{-2}$



Water Vapor Path [$\text{kg} \cdot \text{m}^{-2}$], Satellite data at 06:00 on 02.08.2020

Problem: Hallucination in non-ML lossy compressors

SZ3, absolute error bound $\epsilon_{\text{abs}} = 0.1 \text{ kg} \cdot \text{m}^{-2}$



Water Vapor Path [$\text{kg} \cdot \text{m}^{-2}$], Satellite data at 06:00 on 02.08.2020



Compression Safeguards: Fearless Lossy Compression with declarative Safety

Juniper Tyree^[1,4], Milan Klöwer^[2], Clément Bouvier^[1,4], Daniel Köhler^[1], Robert Underwood^[3],
Heikki J. Järvinen^[1,4], et al.*

[1] INAR, University of Helsinki, [2] University of Oxford, [3] Argonne National Laboratory, [4] ESIWACE3, * pending authors

How can the compression safeguards help **You**?

Declare your safety requirements, the safeguards guarantee them

Preserve **special values** (NaN, Inf, 0, missing value, etc.)

Regional pointwise error bounds (abs, rel, noa, ratio)

Preserve properties, e.g. **signs & monotonic sequences**

Error bounds over **quantities of interest** (x^2 , ∂X , ...)

Logical **combinations** over all of the above



Outlook: Next steps, publication, and future work

-Q3

Compression Safeguards Software

Initial integration with numcodecs, possibly Zarr

Q4-

Initial publication and evaluation

Examples of preserving complex safety requirements

2026-

Expansion and Operationalization

Further safeguards, CDF integration, perf optimizations



Convince Yourself through Outreach and an Open Science Laboratory



Trying something out for yourself
is especially important if you
have **doubts**



To **convince yourself** that lossy compression is safe, you need to **try it out yourself**

Trying something out should be easy
to **engage earlier** and **delay setup costs**
after **convincing** users

Solution for the entire community

i.e. not just for our project (lossy compression) but also for your project, documentation, demos, ...

The Online Laboratory for Climate Science and Meteorology

What's easier than opening a URL in your browser?

Serverless In-Browser
Interactive Computing



Code and data stay *local*



No setup, **no installation**, <1min to start

Reproducible and version-locked
Ensure your examples keep working

Supports many compiled scientific + especially Earth
Science Python packages

Extra support for **accessing large datasets**

Ease of use: same code, same results

Example: Software documentation is non-interactive

37a / 44

The screenshot shows a web browser window displaying the documentation for earthkit-plots. The browser's address bar shows the URL `earthkit-plots.readthedocs.io`. The page layout includes a dark blue sidebar on the left with the following navigation links: **earthkit-plots**, **stable**, **Search docs**, **EXAMPLES** (Examples, Gallery), **DOCUMENTATION** (User guide, API Reference, Development), **INSTALLATION** (Installation, License), and **PROJECTS** (earthkit). The main content area has a white background and features a header with the text "Welcome to the earthkit-plots documentation" and a row of tags: ESEE, Foundation, Maturity, Incubating, License, and Apache 2.0. The main text describes the project's purpose and lists key features: Concise, high-level API; Intelligent formatting; Customisable style libraries; Automatic data styling; and Complex grids supported out-of-the-box. A "stable" version selector is located in the bottom right corner of the main content area.

earthkit-plots

stable

Search docs

Welcome to the earthkit-plots documentation

ESEE Foundation Maturity Incubating License Apache 2.0

earthkit-plots leverages the power of the earthkit ecosystem to make producing publication-quality scientific graphics as simple and convenient as possible.

It is built on top of popular data science and visualisation tools like **numpy**, **xarray**, **matplotlib** and **cartopy**, but provides a very high-level API enriched with domain-specific knowledge, making it exceptionally easy to use.

Key features include:

- Concise, high-level API** Generate high-quality visualisations with minimal code.
- Intelligent formatting** Titles and labels automatically adapt based on common metadata standards.
- Customisable style libraries** Easily swap styles to match your organisation, project, or personal preferences.
- Automatic data styling** Detects metadata like variables and units to optionally apply appropriate formatting and styling.
- Complex grids supported out-of-the-box** Visualise grids like HEALPix and reduced gaussian without any extra legwork.

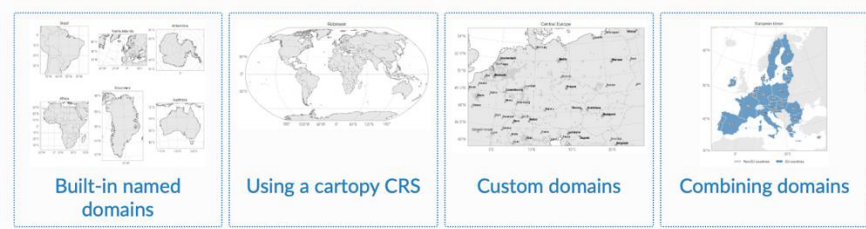
stable

Example: Software documentation is non-interactive

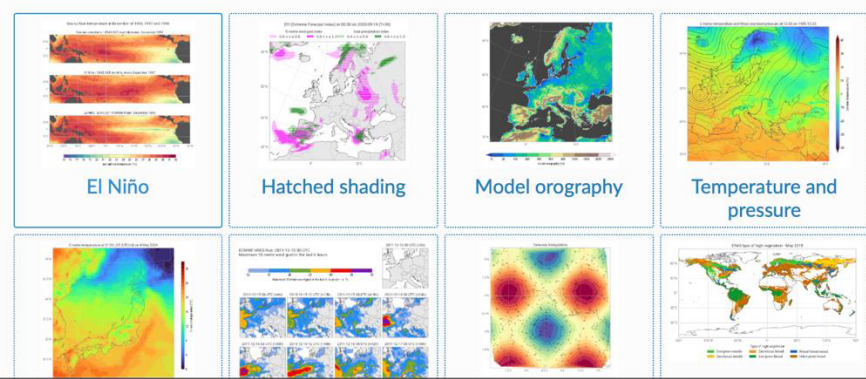
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Gallery

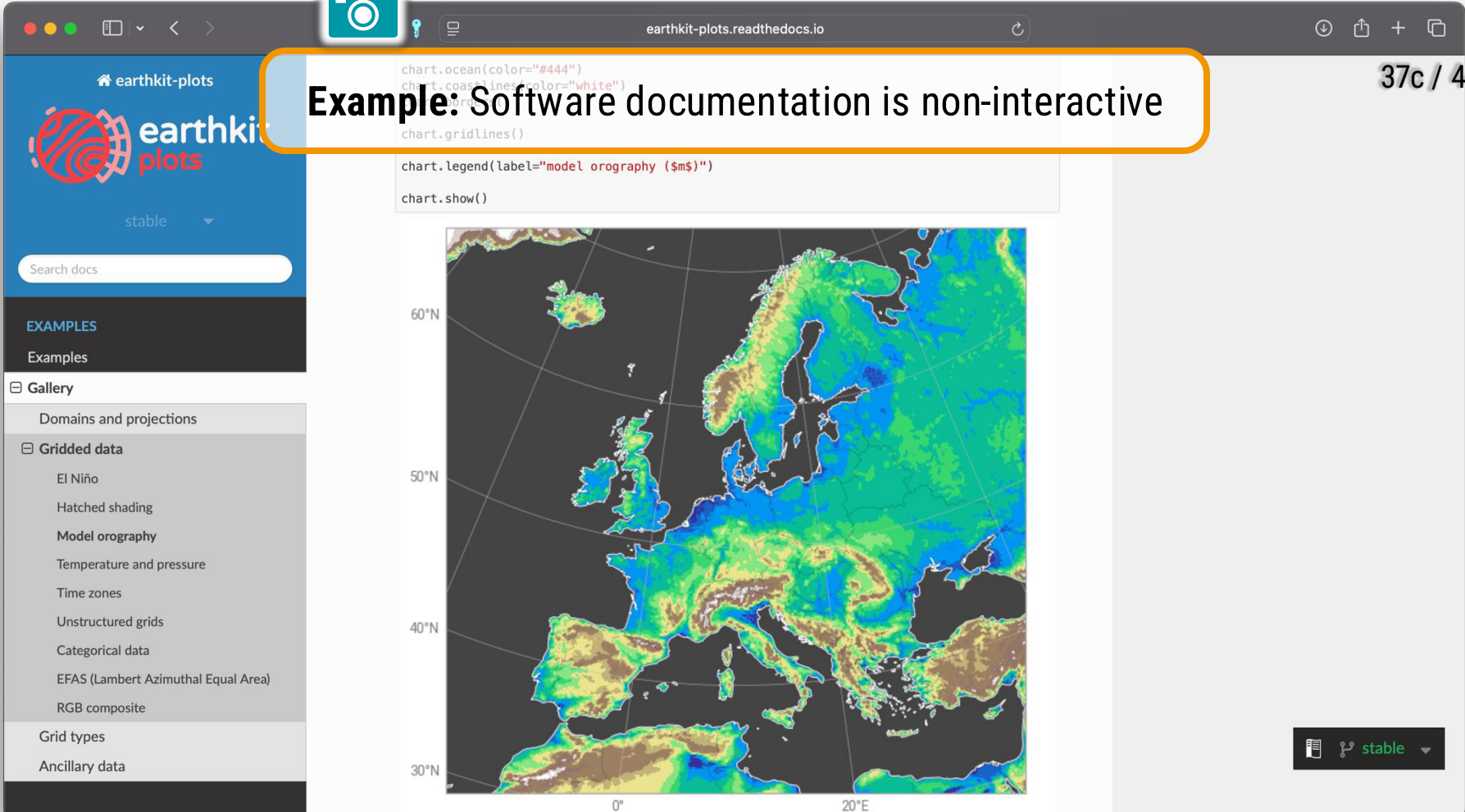
Domains and projections



Gridded data



Example: Software documentation is non-interactive



The screenshot shows a web browser window with the URL `earthkit-plots.readthedocs.io`. The page displays the `earthkit-plots` logo and a sidebar with navigation links. The main content area shows a map of Europe with a color-coded overlay, likely representing a weather or climate variable. The map is titled "model orography" and includes a legend. The map shows a color-coded overlay of Europe, with colors ranging from blue (low values) to red (high values). The map is titled "model orography" and includes a legend. The map shows a color-coded overlay of Europe, with colors ranging from blue (low values) to red (high values). The map is titled "model orography" and includes a legend.

earthkit-plots

stable

Search docs

EXAMPLES

Examples

Gallery

- Domains and projections
- Gridded data
 - El Niño
 - Hatched shading
 - Model orography
 - Temperature and pressure
 - Time zones
 - Unstructured grids
 - Categorical data
 - EFAS (Lambert Azimuthal Equal Area)
 - RGB composite
- Grid types
- Ancillary data

```
chart.ocean(color="#444")
chart.coastlines(color="white")
chart.gridlines()

chart.legend(label="model orography ($m$)")

chart.show()
```

60°N

50°N

40°N

30°N

0°

20°E

stable

37c / 44

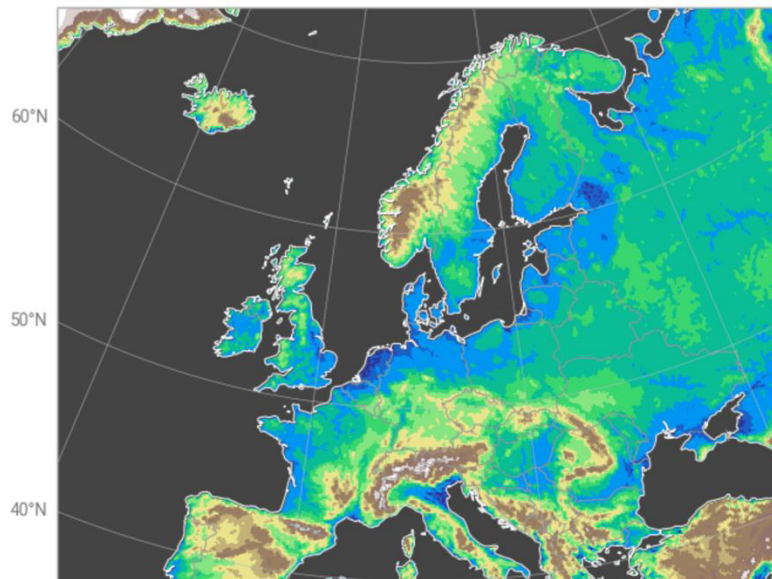
Demo: Interactive Documentation in the Online Laboratory

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```
[5]: # Compress the geopotential ( $m^2/s^2$ ) using ZFP
```

```
z = CodecStack(  
    Zfp(mode="fixed-accuracy", tolerance=100) ← strict error bound  
) .encode_decode_data_array(data[0].to_xarray().z)
```

```
chart = earthkit.plots.Map(domain=[-15, 35, 32, 72]) ...
```



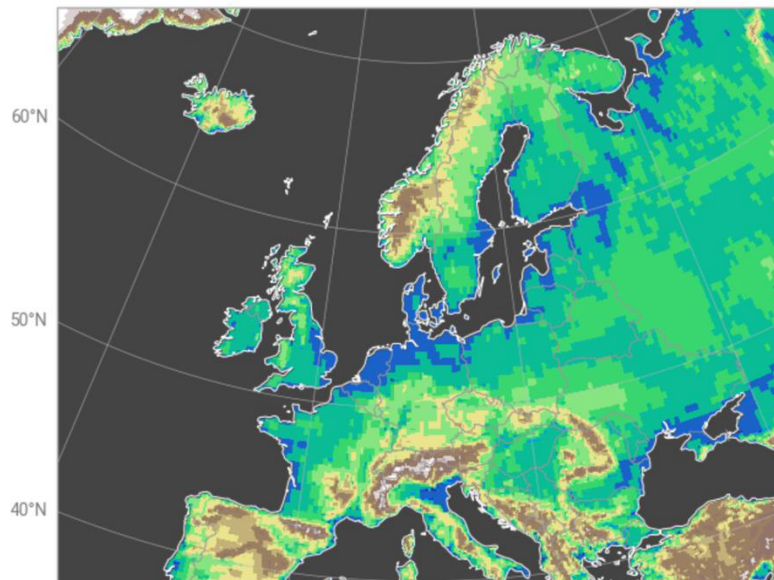
Demo: Interactive Documentation in the Online Laboratory

38b / 44

```
[7]: # Compress the geopotential ( $m^2/s^2$ ) using ZFP
```

```
z = CodecStack(  
    Zfp(mode="fixed-accuracy", tolerance=10000) ← loose error bound  
) .encode_decode_data_array(data[0].to_xarray().z)
```

```
chart = earthkit.plots.Map(domain=[-15, 35, 32, 72]) ...
```



Demo: ClimateBenchPress

To summarize the performance of the two compressors across all the error bounds we can make a rate-distortion plot. Interestingly, for this dataset BitRounding seems to perform better than SZ3 because BitRounding produces higher compression ratios for a similar error level.

```
[15]: plot_variable_rd_curve(
      results,
      distortion_metric="MAE",
      outfile=plotspath / "rd_curve.png"
    )
Image((plotspath / "rd_curve.png").read_bytes(), width=512)
```

Figure 15 is a log-log plot showing the Mean Absolute Error (MAE) on the y-axis versus the Compression Ratio [raw B / enc B] on the x-axis. The x-axis ranges from 6×10^0 to 2×10^1 , and the y-axis ranges from 10^{-2} to 10^{-1} . Two compressors are compared: BitRound + PCO (blue squares) and SZ3 (red squares). Both compressors show a linear relationship on the log-log scale, indicating a power-law relationship between MAE and Compression Ratio. BitRound + PCO consistently achieves a lower MAE than SZ3 for the same compression ratio.

Compressor	Compression Ratio [raw B / enc B]	MAE
BitRound + PCO	~5.5	~0.003
BitRound + PCO	~11	~0.03
BitRound + PCO	~20	~0.2
SZ3	~5.5	~0.004
SZ3	~8	~0.03
SZ3	~11	~0.2





Demo: Online Compression Laboratory

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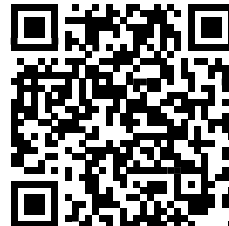
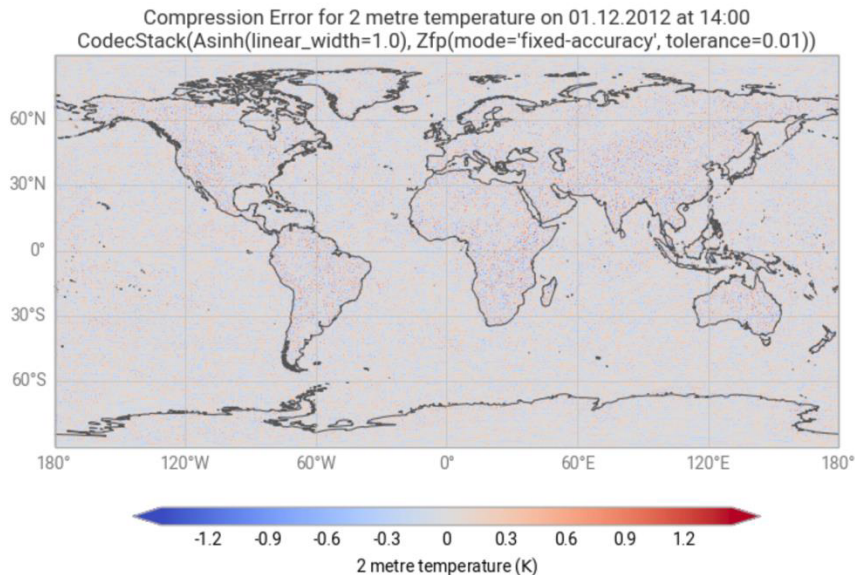
01-intro.ipynb

File Edit View Run Kernel Tabs Settings Help

Name	Modified
02-data-sources	25s ago
03-examples	24s ago
04-example-dat...	22s ago
data	21s ago
01-intro.ipynb	20s ago
99-vision.md	4s ago
LICENSE.txt	3s ago
README.md	2s ago
utils.py	now

```
[16]: with xr.set_options(keep_attrs=True):
      da_zfp_error = (da_zfp - da).compute()

      plot_data(
          da_zfp_error, title_prefix="Compression Error for ",
          title_postfix=f"\n{zfp_compressor}", error=True,
      )
```



Simple 1 Python (Pyodide) | Idle

Mode: Edit Ln 1, Col 1 01-intro.ipynb 2

How can **You** benefit from the Online Laboratory today?

docs.climet.eu



Interactive and Reproducible Open Science Laboratory

Share your existing **notebooks** from repos / gists / URLs

Customize the provided **packages** and **versions**

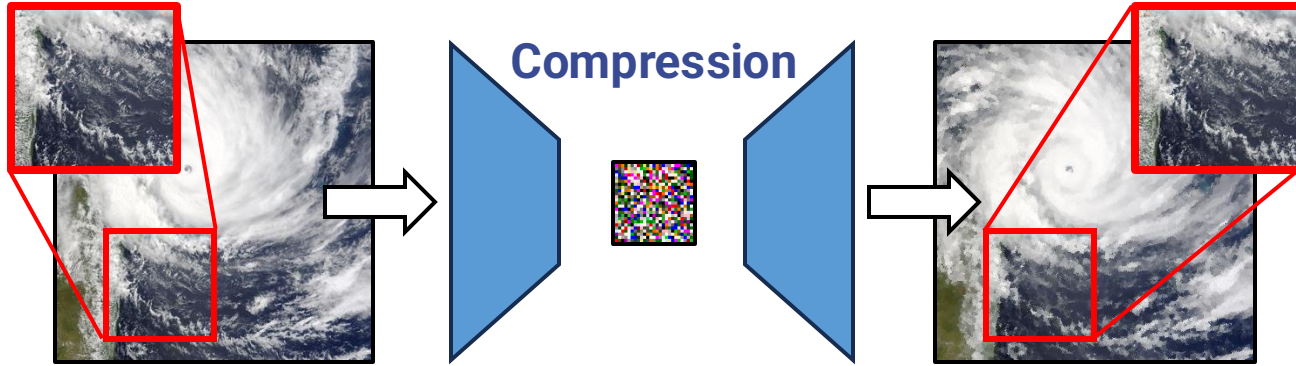
Ongoing development and support and new packages

Explore lossy data compression on climate model data



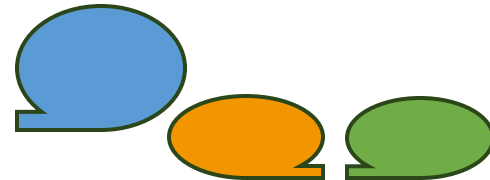
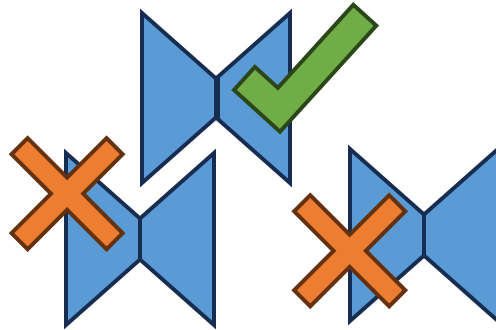
In summary ...

Weather & Climate Prediction will soon produce **too much data**



Paralyzing fear of compression artifacts or loss of details

Safe and Fearless Lossy Data Compression in Climate Science



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Towards Safe and Fearless Lossy Compression of Weather and Climate Model Data

Thank You for Your attention!

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 juntyr

Disclaimer: Funded by the European Union. Views and opinions expressed are however those of the author(s) only and do not necessarily reflect those of the European Union or the European High Performance Computing Joint Undertaking (JU). Neither the European Union nor the granting authority can be held responsible for them.

CONTINENTS Webinar, 28.05.2025



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