

# Quantifying the Scientific Skill of High-Resolution Models with *pyhanami*

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## The Package: *pyhanami*

An open-source Python package for evaluating the **scientific skill** and **replicability** of Earth System Models (ESMs) through visualization tools, objective scalar-based metrics, and statistical testing.



Check our repo:



Easy input handling

## SimulationData class

1. Load climate simulation ensembles from a **NetCDF** file or an **xarray.Dataset** object.
2. Automatically verify the integrity and consistency of the input datasets.

Intuitive syntax:

```
1 import pyhanami
2 sim_data = pyhanami.SimulationData('sim_data_path', name='sim name')
```

Model evaluation

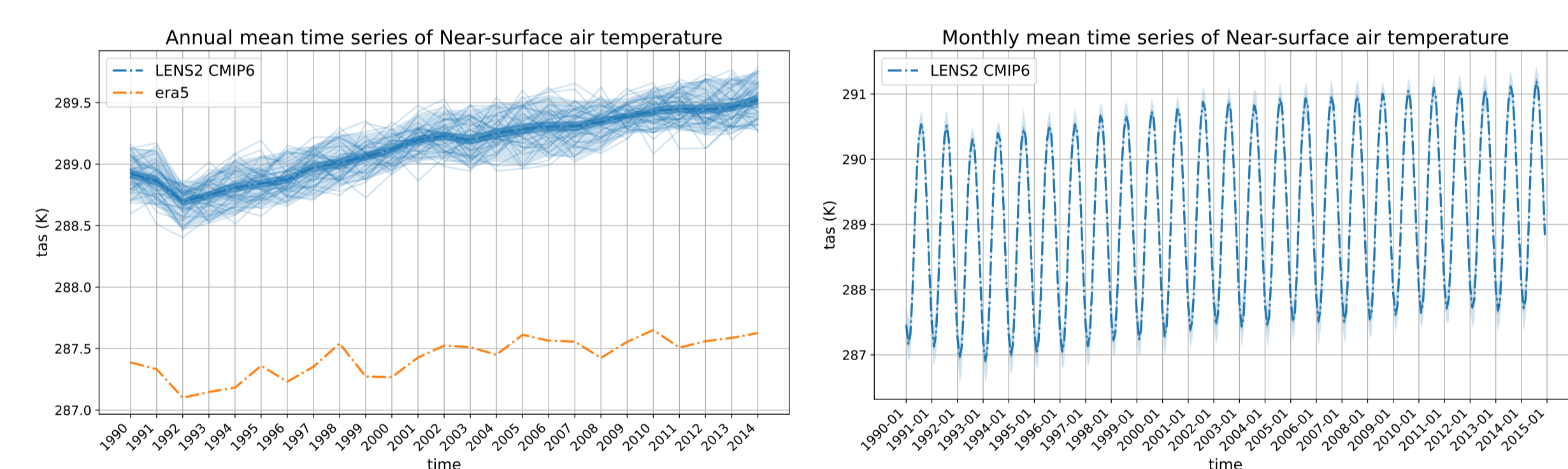
## DataDiagnostics class

Diagnostics of **ensemble** simulations for a given variable:

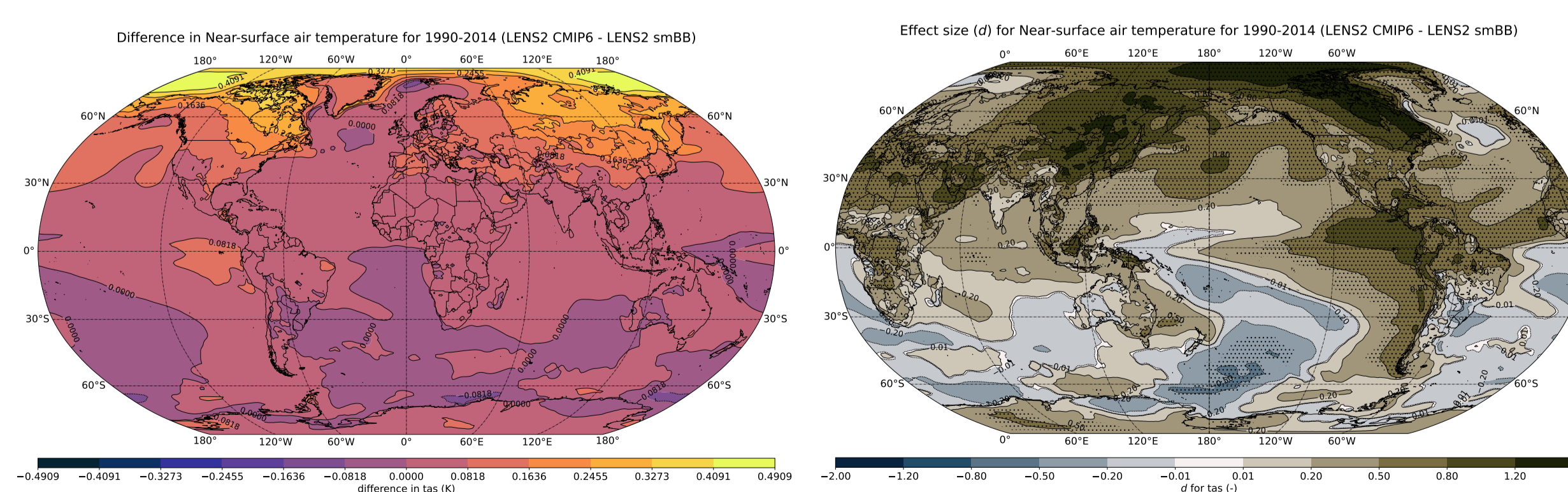
1. **Time series**: multiple simulations + observations
2. **Spatial plots**: absolute difference + effect size (Cohen's *d*)

Usage examples employing CESM2 LENS2 data:

```
1 diags = pyhanami.DataDiagnostics(lens2_cimp6)
2 diags.time_series_plot('tas', start_year=1990, end_year=2014, obs=True, plot_ens=True)
3 diags.time_series_plot('tas', start_year=1990, end_year=2014, time_freq='monthly')
```



```
1 diags.add_dataset(lens2_smbb)
2 diags.abs_diff_plot('tas', start_year=1990, end_year=2014)
3 diags.eff_size_plot('tas', start_year=1990, end_year=2014, clon=180)
```



## ScientificEvaluation class

Plots and **scalar scores** for evaluating scientific model skill. Examples using IFS-NEMO simulations at different resolutions:

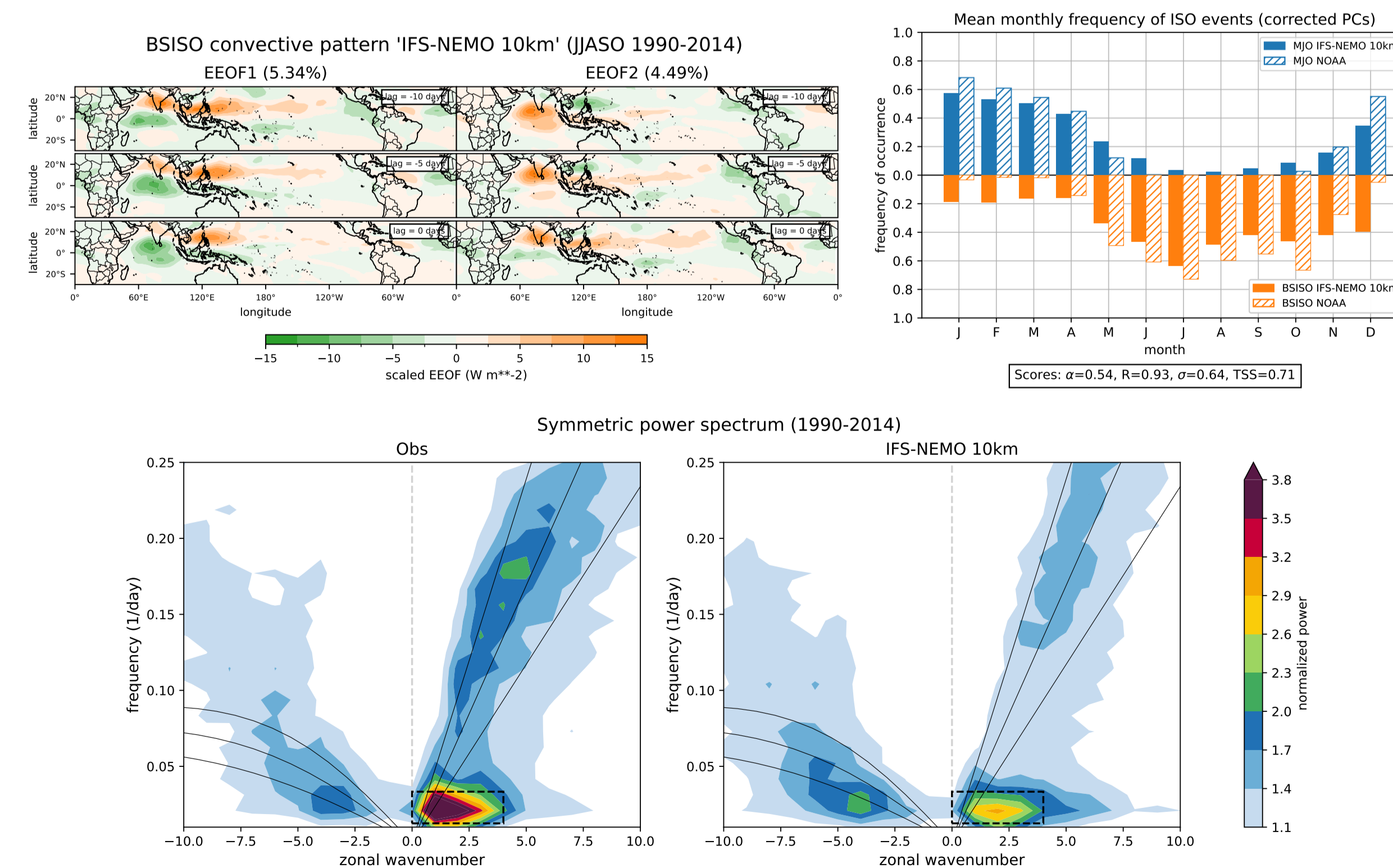
```
1 sciskill = pyhanami.ScientificEvaluation([ifs_nemo_5km, ifs_nemo_10km, ifs_nemo_25km])
2 x_analysis = sciskill.compute_x_scores('IFS-NEMO 10km', start_year=1990, end_year=2014)
```

### 1. Tropical Intraseasonal Oscillation (ISO)

Predominant phenomenon in the tropics throughout the year:

- In boreal winter: **Madden-Julian Oscillation (MJO)**
- In boreal summer: **Boreal Summer ISO (BSISO)**

Evaluation: Empirical Orthogonal Function (EOF) analyses + Wavenumber-frequency power spectra



### 2. Tropical Cyclones (TCs)

Warm-core, rapidly rotating storm systems characterized by strong winds and heavy precipitation that begin over the tropical oceans. Evaluation:

Track TC trajectories ⇒ Compute TC metrics (duration, intensity, latitude) ⇒ Compute TC scores (bias, correlations)

Examples of usage employing several reanalyses and NICAM16-7S data (with ~ 56 km horizontal resolution):

Global climatological mean bias (1990-2010)

5° x 5°	$\bar{b}_{clim, count}$ (#)	$\bar{b}_{clim, tcd}$ (days)	$\bar{b}_{clim, ace}$ (10 <sup>-4</sup> kn <sup>2</sup> )	$\bar{b}_{clim, pace}$ (10 <sup>-4</sup> kn <sup>2</sup> )	$\bar{b}_{clim, lmi}$ (° lat.)
IBTrACS	107.7	851.9	772.3	743.7	18.6
ERA5	-26.9	-329.0	-499.7	-347.1	5.0
JRA-55	-36.5	-377.3	-543.9	-458.8	4.9
NICAM16-7S	-32.3	-282.8	-408.1	-219.2	3.3

Negative bias      No bias      Positive bias

### Extra and Future Features

- ✓ Spatial bias plots (simulations - observations)
- ✓ General scalar scores (bias, RMSE, correlation)
- ✓ Replicability test (following (K.Keller et al., 2025))
- ✓ Comprehensive documentation in ReadTheDocs
- 🔗 Additional climate phenomena (e.g., precipitation)
- 🔗 Intake data catalogue (standardized data interface)

Documentation:



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