

Using AI downscaling and physical models to anticipate & mitigate wind drought months

Seasonal wind forecasts: how to anticipate and mitigate wind drought impacts

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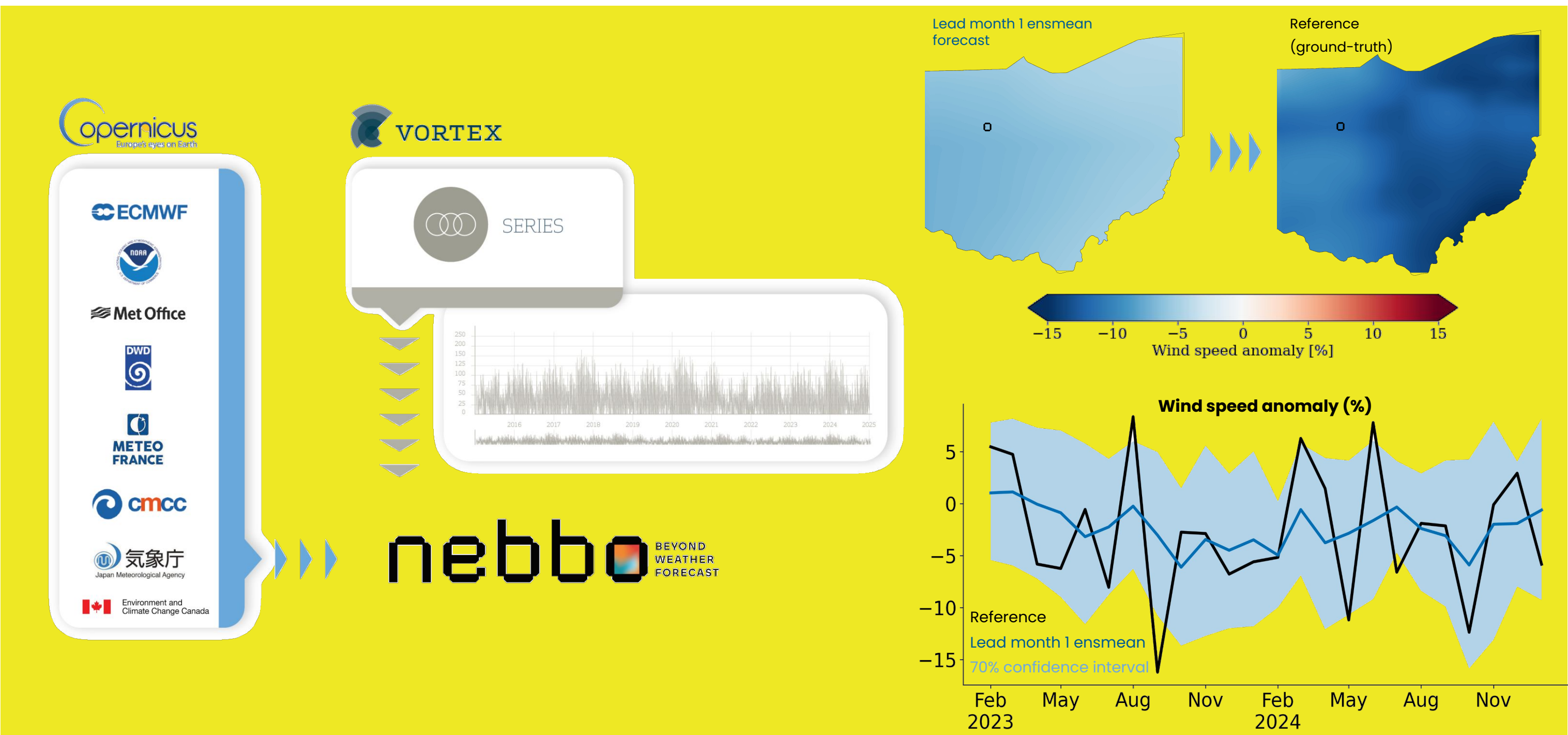


Fig. 1: Nebbo methodology represented at the left side. At the top right, the Hog Creek wind farm is displayed in an Ohio map (top) for both forecast & historical layers during the peak of its wind drought period (Oct 24). At the bottom right, the time series for the Nebbo multi-model forecast (ensemble mean) with lead month 1 and 70% coverage is shown for Feb 23 – Jan 25.

Introduction

Nebbo leverages **AI-enhanced seasonal forecasts** to **reliably** foresee **anomalous** episodes. In this poster, the **Hog Creek** park (40.79°, -83.730°; Dola, Ohio) is featured as a **use-case**. During **May – Nov 2024**, the site suffered from a **wind drought** which was **successfully foreseen months in advance**.

Methods

Our technology consists of 3 main steps:

- Forecast calibration through **statistical downscaling** using:
 - 3km mesoscale ERA5-simulated 30yrs time series as ground truth (Vortex SERIES)
 - 8 different state-of-the-art physical models from renowned weather centres (C3S hub)
- Multi-model** definition
- Predictive interval** construction
 - Used to avoid reliability issues and obtain statistically valid results from the forecasted probability distribution

Results

Based on our methodology, we conclude that:

- Forecast bias is corrected but variance is still slightly large compared to reference (Fig. 4).
- The multi-model approach is generally more robust than individual models.
- Predictive intervals are effective:
 - They cover the ground truth approximately 70% of the time, as expected (Fig. 2).
 - They remain informative with reasonable width (Fig. 2).
 - They offer a better alternative to an imperfect forecast distribution.

Discussion

Our **methodology** successfully provides **reliable forecasts** across different **lead months**, particularly the first ones. The **multi-model** approach and **predictive intervals** enhance accuracy, ensuring **statistically valid** and **informative** results. Thus, **stakeholders** could **use** this **technology** to **anticipate** and **mitigate** the **impact** of **wind drought** periods, such as the observed in Hog Creek during May-Nov 2024.

Extra tables & figures

| Forecast metrics | | | | |
|-----------------------------|------|------|------|--|
| for 1, 2 and 3 months ahead | | | | |
| | L1 | L2 | L3 | |
| MAE (m/s) | 0.33 | 0.35 | 0.37 | |
| Interval width (m/s) | 1.08 | 1.27 | 1.24 | |
| Coverage (%) | 75 | 79 | 75 | |
| Anomaly sign hit rate (%) | 79 | 79 | 71 | |

Fig. 2: Re-forecast metrics for Nebbo multi-model & Feb 2023 – Jan 2025 period. The ensemble mean of lead months 1 (L1), 2 (L2) & 3 (L3) is compared with Vortex SERIES ground-truth.

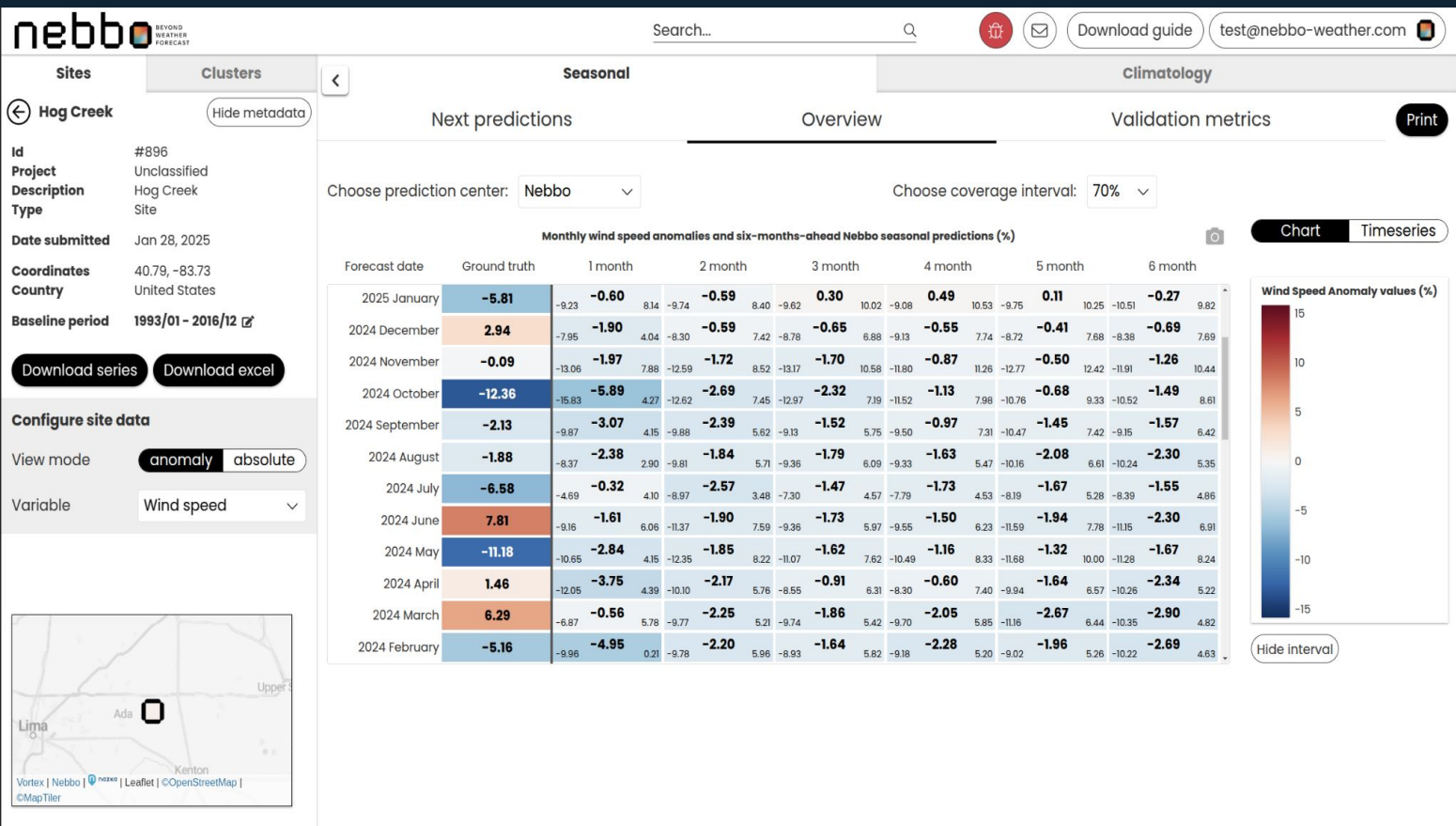


Fig. 3: Overview tab of the Nebbo interface. It showcases the different predictions (with intervals) for different forecast months and lead months.

Forecast distributions for 1, 2 and 3 months ahead

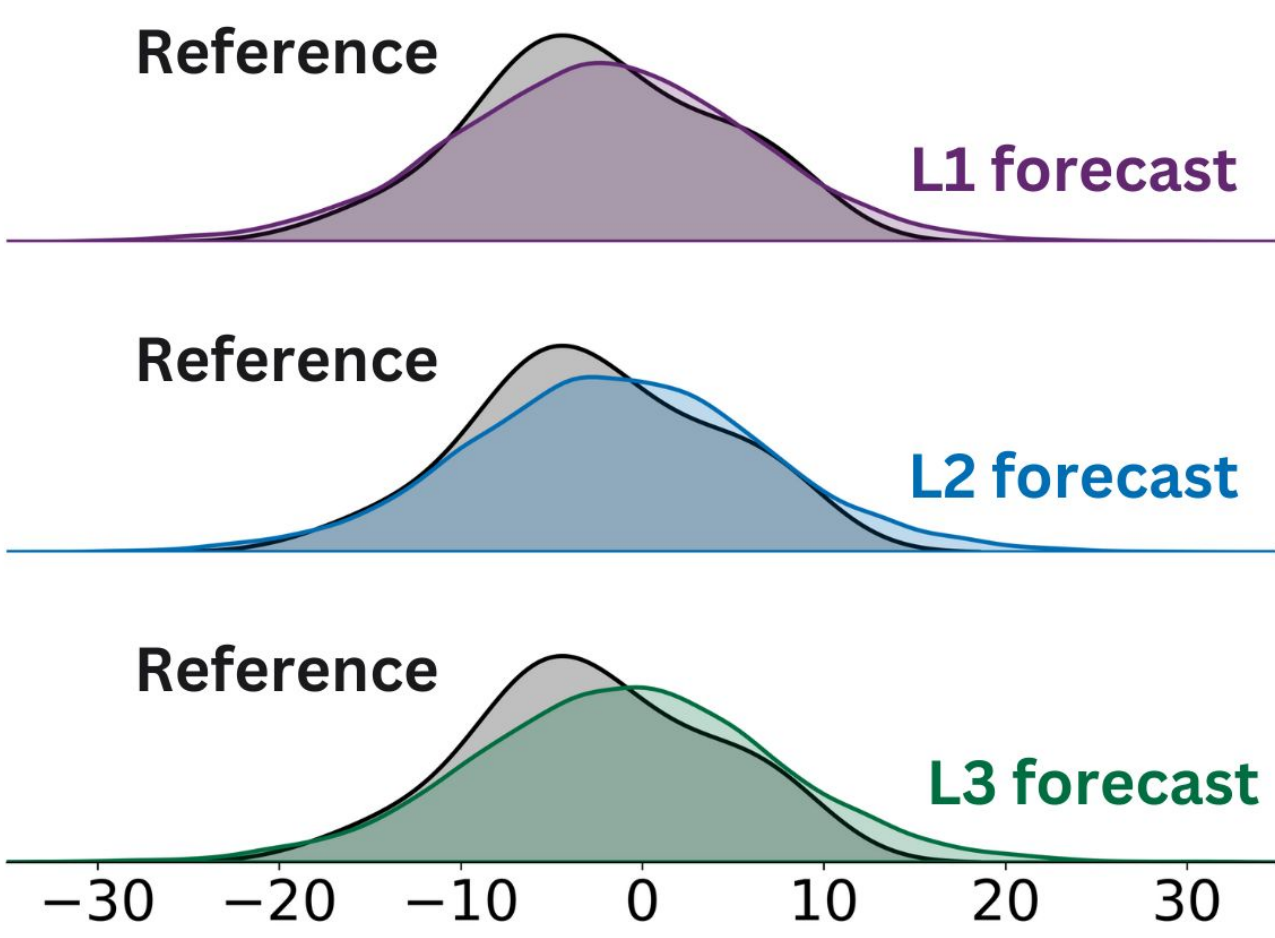


Fig. 4: Re-forecast distributions for Nebbo multi-model & Feb 2023 – Jan 2025 period. The ensemble forecast distributions of lead months 1 (L1), 2 (L2) & 3 (L3) is compared with Vortex SERIES ground-truth (Reference).

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Try our interface!

