

# **Weather and Climate Intelligence for an Increasingly Weather Dependent Grid**

**A Summary of the ESIG Weather Inputs Task Force Report and Proposed Next Steps**



Justin Sharp Technical Leader

Avista DPAG December 3, 2024

Support and content provided by:GridL黄B **ESIG ENERGY SYSTEMS INTEGRATION GROUP** 



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# **EPRI Load Forecasting Initiative**



Improved load forecasts at operational and planning timescales\* will drive more efficient investment decisions and better grid performance.

EPRI launched a 24-month initiative (ending in Q4 2025) to address critical needs in load forecasting that will work across three areas:

> **Industry Coordination** Enable knowledge-sharing and collaboration among utilities, ISOs/RTOs, etc. **01**

> > **Long-Term Forecasting (Planning)**

Develop methodologies and guidance to incorporate new load drivers **02**

**Short-Term Forecasting (Operations)**

Develop methodologies and guidance to mitigate changes in forecast accuracy **03**



\*we are defining "planning timescales" as >1-year ahead

### **Acknowledgements/Disclaimer:**

The report and its companion documents were output from the ESIG Weather Inputs Task Force project. I'd like to acknowledge the help of all the members of the task force especially to those contributed to the writing and those who provided deep review and comments.

The project was convened and supported by ESIG. Additional funding was provided by GridLab and Sharply Focused. While largely objective, some of this



**Weather Input Datasets for** Power System Modeling **USING EXISTING DATASETS** 



A Report of the Energy Systems **Integration Group's Weather Datasets Project Team** 2023

#### [Report Landing Page](https://www.esig.energy/weather-data-for-power-system-planning/)

- **[Executive Summary](https://www.esig.energy/wp-content/uploads/2023/10/ESIG-Weather-Datasets-exec-summary-2023.pdf)**
- [Main Report](https://www.esig.energy/wp-content/uploads/2023/10/ESIG-Weather-Datasets-full-report-high-resolution-2023b.pdf)
- [Summary Report](https://www.esig.energy/wp-content/uploads/2023/10/ESIG-Weather-Datasets-summary-report-2023.pdf)
- [Meteorology 101](https://www.esig.energy/wp-content/uploads/2023/10/ESIG-Weather-Datasets-meteorology-101-2023.pdf)



**Scan for report landing page**

presentation represents my own views, some of which may not necessarily be the official views of task force members or member organizations.



### **Motivation: The Energy Transition**



THE ELECTRIC SYSTEM IS CHANGING

**…RADICALLY…**

AND IS FULL OF UNCERTAINTY THE SECTOR HAS TO EVOLVE ITS METHODS ACCORDINGLY





Findings included in seminal consensus-based reports from the ESIG Rethinking Resource Adequacy initiative The quality of power system studies becomes increasingly dependent on characterization of weather. Analysis must incorporate weather more comprehensively





The Evolving Role of Extreme Weather Events in the U.S. Power System with **High Levels of Variable Renewable** Energy

Josh Novacheck.<sup>1</sup> Justin Sharp.<sup>2</sup> Marty Schwarz. Paul Donohoo-Vallett 3 Zach Tzavelis 1 Grant Bust

National Renewable Foergy Labora Sharply Focused, LLC I.S. Department of Energ

EPRI



The Evolving Role of Extreme Weather Events in the U.S. Power System with High Levels of Variable Renewable Energy (Abstract:<https://www.osti.gov/biblio/1837959> | Full Report: [https://doi.org/10.2172/1837959\)](https://doi.org/10.2172/1837959)

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#### **Electricity System Weather-Dependence**



#### Typical magnitude is approximated by the thickness of the lines.

![](_page_5_Picture_3.jpeg)

While all environmental variables are interdependent, these are some of the strongest internal links.

Dependence of the electricity system on the climate system.

- Strength of dependence is highly variable and depends on asset type and location.
- Degree of dependence can be greatly amplified by specific weather and climate conditions.

Source: ESIG Weather Data for Power System Planning [https://www.esig.energy/weather](https://www.esig.energy/weather-data-for-power-system-planning/)-data-for-powersystem [-planning/](https://www.esig.energy/weather-data-for-power-system-planning/)

![](_page_5_Picture_11.jpeg)

# Active High-Quality Public Weather Observations

![](_page_6_Figure_1.jpeg)

- Power system models have always incorporated weather
	- Treatment mostly concerned with impact of temperature on load, and sampling of hydro years
	- Data needed from urban areas (with plenty of observations), and existing streamflow measurements
- Obs. of weather impacting wind and solar output are not widely available and MUST BE SYNTHESIZED
	- Fields vary rapidly across short distances and times, and are needed for remote areas
	- Observations are sparse, have a short history, and where they exist they are mostly **proprietary**
- The complex interaction between variables impacting load, wind, and solar MUST now be considered, and must be coincident and physically consistent (in time and space),and chronological.
- The interconnectivity in time and space yields complex, yet organized, multi-dimensional probability distributions that must be realistic for accurate power system analysis.
- DERs/storage and other weather impacts on G,T, and D add more layers of complexity.

![](_page_7_Figure_9.jpeg)

#### The Main Attributes of Time Series Data Necessary to Meet General Power System Modeling Needs

Source: ESIG Weather Data for Power System Planning [https://www.esig.energy/weather-data-for-power](https://www.esig.energy/weather-data-for-power-system-planning/)[system-planning/](https://www.esig.energy/weather-data-for-power-system-planning/)

![](_page_8_Picture_24.jpeg)

## **Existing Weather and Climate Data is Inadequate for Comprehensive Analysis of Modern Power Systems**

![](_page_9_Picture_1.jpeg)

The biggest issues are one or more of the following:

- **EXTED Insufficient validation and uncertainty** quantification (true for all datasets)
- Insufficient spatial or temporal resolution
- **EXEDENCE IN SURGION INCOCOLEY** Insufficient time history/lack of ongoing extension
- **EXEDE:** Distributions that don't match reality, especially for extreme events
- Data use from non-static modeling platforms
- **A lack of knowledge of the limitations of current datasets and the downstream impact of biases and inaccuracies**

#### Why does this matter?

- You can't correctly predict supply and demand if the weather data isn't good. Sometimes, you'll be WAY off.
- Coincident weather data is need to assess compound risks to generation, transmission and distribution.

![](_page_10_Picture_207.jpeg)

#### The Data We Have Today

The data currently available to the sector (on left) is not adequate for the tasks at hand. No single dataset meets all the needs. Mixing and matching causes physical consistency issues.

#### **TABLE 2**

Summary of Current Power System Modeling Weather Input Data Sources

Summary of the most applicable datasets globally that are (or can be) used to provide weather inputs for power system analysis tasks, especially for providing estimate of site-level generation, and concurrent weather-driven load and generation outage risks. The degree to which the needs of each column heading are met is estimated with color coding. See documentation for each dataset for all details. Footnotes on next page. P76, main report.

Source: Energy Systems Integration Group

### **How Bad is it Really? A Use Case Specific Validation**

- We must validate according to the use case. E.g. For RA, the distributions, and especially the tails, matter more than the averages
- **The distribution of coincident tail events MUST be close to reality**
- Example here:
	- WINDTK data in the BPA area.
	- Wind resource in BPA BA is notoriously difficult to predict with NWP => WFIP2 Project
		- Complex terrain in the region needs a minimum of 1.33 km resolution to resolve. More on this later.
		- Stable boundary layer issues in the wintertime. => Low wind AND high load

These biased low wind speed events frequently coincide with high load events due to regional mesoscale meteorology

![](_page_11_Figure_9.jpeg)

![](_page_11_Picture_12.jpeg)

# **Don't We Produce This Data and Successfully Use it in Operations?**

Yes, we do. Which leads to the radical statement that:

![](_page_12_Picture_2.jpeg)

Historical supply and demand estimations used in power system planning analysis are often less accurate than forecasts used in operations!

Why? Proprietary plant datasets are available and used for training/validation of operational forecasts, more attention is paid to them, and we only need data for next few days, versus for the last few decades.

# Our Weather "Intelligence" is Inadequate

Producer(s) Create initial and ongoing gridded archives Bias correction

Ongoing generic R&D

#### Gridded Weather Data

- Physically consistent weather variables
- Multi-decadal, historical and future
- **Not coordinated with sector needs**
- **Insufficient resolution for general power systems use**

We are transitioning to a much more weather dependent electric system:

- Demand is becoming much more weather dependent
- Wind and solar are instantaneously defined by weather
- Other infrastructure is at increasing risk from weather Yet, our weather intelligence isn't even close to adequate
- Uncoordinated, lacking vision and leadership
- Not created with sector needs in mind

Users End-use application of data

# We Need Vision, Investment & Leadership

Holistic View of a Weather Intelligence Support Framework For The Electric System

#### Producer(s)

Create initial and ongoing gridded archives Bias correction

Coordinate with curators on access

#### Gridded Weather Data

- Physically consistent weather variables
- Multi-decadal historical with ongoing consistent extension, and multiple futures
- Periodically refreshed
- At a fidelity that can represent actual grid conditions (supply, demand, T&D)

#### Validator(s)

QA/QC of validation data Validation and uncertainty quantification of gridded data

Coordinate with producers/curators

#### **Curator**

Facilitate data access Provide uncertainty information Document, guide, and educate

#### Ground Truth Data

- Weather and power data from RE fleet
- Dedicated power system field environmental data

#### Users

End-use application of data Provision of fleet data as appropriate

Ongoing Sector Specific R&D

Methodological improvements Refresh Recommendations

#### **Ongoing Oversight:**

- **Requirements** gathering/update
- Trans-disciplinary coordination
- Feedback facilitation
- R2O Coordination

**Weather Dependence and Complexity are Increasing Rapidly**

**Weather/Climate Are Becoming Central Yet We Are Largely Flying Blind**

There is an Imperative for Dedicated, Accurate, and Expertly Curated Weather Information to Support the Energy Transition!

The risks resulting from inaction:

- Reliability issues tied to renewables
- Inefficient system design and planning

16 © 2024 Electric Power Research Institute, Inc. All rights reserved. **Risk \$\$\$'s are orders of magnitude higher than task investment \$'s**

![](_page_15_Picture_7.jpeg)

Weather Input Datasets for **Power System Modeling** A NEEDS ASSESSMENT AND GUIDANCE FOR **USING EXISTING DATASETS** 

![](_page_15_Picture_9.jpeg)

A Report of the Energy Systems **Integration Group's Weather Datasets Project Team** 2023

![](_page_15_Picture_11.jpeg)

![](_page_16_Picture_0.jpeg)

**Analysis of our increasingly weather dependent system must be data driven**

**How do we mitigate the current shortcomings of available data?**

![](_page_16_Picture_3.jpeg)

**One solution is to fly blind and largely ignore the problems and hope they wash out in the analysis. This is often the current practice.**

### **Garbage In = Garbage Out**

Learn more in the four-part video series Weather and Climate Intelligence for the Energy Transition produced by Sharply Focused for GridLab and hosted on GridLab's LinkedIn Page

![](_page_16_Picture_7.jpeg)

**There are achievable, better options. I'll take you through a (methodologically agnostic) proposed approach.**

#### Requirements Gathering and Selection Process

![](_page_17_Figure_1.jpeg)

<sup>1</sup>May all be the same organization. <sup>2</sup> Should *not* be the same organization; creates a conflict of interest.

#### Production, Validation, and Curation Processes

![](_page_18_Figure_1.jpeg)

<sup>1</sup>May all be the same organization. <sup>2</sup> Should *not* be the same organization; creates a conflict of interest.

#### Production, Validation and Curation Processes, with Periodic Refresh

![](_page_19_Figure_1.jpeg)

<sup>1</sup>May all be the same organization. <sup>2</sup> Should *not* be the same organization; creates a conflict of interest.

### **How Much Will It Cost?**

**Rough** figures based on costs for high volume NWP work for high-cost case (1-km CONUS NWP) back to 1990, extended continually to 2035 with extensive validation.

- Selection process with comprehensive validation and comparison to existing datasets: \$2-3M
- Initial dataset production: \$8-15 M. Ongoing \$1-2 M/yr. Includes all storage
- Initial dissemination and curation tasks: \$1 M. Ongoing: \$400-700K/yr. Management: \$200K/yr
- Validation and uncertainty quantification: \$500K/yr + Cost to Acquire Measurements
	- LEVERAGING THE RE BUILDOUT IS IMPERATIVE (as is standardization)
	- Consider cost sharing the physical assets to incentivize cooperation
	- Industry support level and validation thoroughness ultimately sets the cost
- **The value of an observational network to support data production and validation** needs detailed cost-benefit analysis. In the AI world, quality ground truth data is king.

**Custom 1990-2035 Climate Dataset for Electrification: \$35-70M + validation hardware costs Expected grid decarbonization investment by 2035: \$330-740B<sup>1</sup>**

# PEN **XSINE**

OR

![](_page_20_Picture_14.jpeg)

# **Getting into the Weeds**

### **All Data Is Not Created Equal The Impact of Grid Spacing**

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 $^{\sim}$ 2 km

**Computer Resources Forecast Equations** Coordinate System Parameterization Scheme Resolution of Features Weather stations forecast points (3) Numerical grid forecast points (7x7x6)

### **Representation at 30 km Grid Spacing (About the same as ERA5)**

![](_page_22_Figure_1.jpeg)

![](_page_22_Figure_2.jpeg)

## **Representation at 1 km Grid Spacing (The WIND Toolkit has 2 km Resolution)**

![](_page_23_Figure_1.jpeg)

![](_page_23_Figure_2.jpeg)

At 9 km, narrow peaks and valleys are lost and the cre is lower. The complexity

behind the crest is lost and<br>becomes a wide valley.

![](_page_23_Figure_4.jpeg)

he top plot shows a cross-section of hypothetical complex topography represented at 3 km grid pacing. The middle plot uses the average of sets of three 3 km points for each 9 km point. In the plot, three 9 km points were averaged to get to each 27 km point

![](_page_23_Figure_6.jpeg)

# **An Example of Resolution Impacts**

![](_page_24_Figure_1.jpeg)

![](_page_24_Figure_3.jpeg)

- …have spatial and temporal scales relevant to the system being modeled. See example on right.
	- Accurately capture the resource drivers and their variability
	- Capture the uncertainty in estimates of the resource drivers
	- Do the same for drivers of system load

![](_page_25_Figure_5.jpeg)

### **Not All Data Is Created Equal. Model Data ≠ Observations**

▪ …

- …have spatial and temporal scales relevant to the system being modeled. See example on right.
	- Accurately capture the resource drivers and their variability
	- Capture the uncertainty in forecasts of the resource drivers
	- Do the same for drivers of system load/other weather risks
- …be concurrent and physically consistency
	- All variables represent the same time chronology and are from dynamically consistent sources
	- Mixing and matching should be avoided if possible and if done the consequences should be analyzed
		- Mixing examples: NSRDB and WINDTK; different NWP models on right; statistical models and observations

![](_page_26_Figure_9.jpeg)

# **Not All Data Is Created Equal. Model Data ≠ Observations**

▪ …

- …have spatial and temporal scales relevant to the system being modeled. See example on right.
	- Accurately capture the resource drivers and their variability
	- Capture the uncertainty in forecasts of the resource drivers
	- Do the same for drivers of system load
- …be concurrent and physically consistency
	- All variables represent the same time chronology and are from dynamically consistent sources
	- Mixing and matching should be avoided if possible and if done the consequences should be analyzed
		- Mixing examples: NSRDB and WINDTK; different NWP models on right; statistical models and observations
- …provide a 30+yr, updated time history with a consistent methodology that minimize biases, trends, and "artifacts"

![](_page_27_Figure_10.jpeg)

### **Not All Data Is Created Equal. Model Data ≠ Observations**

- …be validated and have uncertainty quantified.
- Which of these representations is right? Most useful?
	- They can't all be right!
- Have you validated the data you use?

![](_page_28_Figure_5.jpeg)

![](_page_28_Figure_6.jpeg)

▪ 50's, 60's, 70's or 80's?

- Light, moderate, strong wind?
- Cloudy or sunny?

### **Not All Data Is Created Equal. Model Data ≠ Observations**

NAM 2m Tmp (F), 10m Wind (kt), MSLP (mb)

#### bias, MAE, RMSE, and correlation between ERA5 Ava:9.306103 idar Avg:10.145639 measured and simulated data R2:0904928 Rias: -0.83954

**More Nuance: Offshore Validation Example**

**EXALED FIGUARE 160 m wind speed data evaluated with** hourly average (+/- 30 mins) "Hudson North" Lidar measurements over a ~ 2-yr period

▪ Typical analyses look at standard metrics such as

- Analysis indicates fairly good performance
	- $-$  R<sup>2</sup> of about 0.90
	- $-$  Bias (ERA5 a bit low): -0.83 m/s
	- MAE: 1.34 m/s
	- Similar performance results at other nearby measurement sites

### permission<br>Hudson North Hourly Average Wind Speed

Slide adapted from content from Dr. John Zack, MESO

Inc. and used with

![](_page_29_Figure_8.jpeg)

#### **So ERA5 should be fine for offshore wind integration studies?**

![](_page_29_Picture_12.jpeg)

# **The Impact of Data Assimilation Cycles**

- **The ERA5 data is created with a DA scheme where** there are two major data assimilation cycles per day
- If we examine the average absolute value of the 1-hr wind speed change by month and time of day we see larger average changes occur at the transition time between data assimilation periods (0500 EST and 1700 EST)
- **The Impact varies substantially by time of year**
- **Therefore, ramp event analysis using the ERA5 data is** suspect for this location…and likely others

Slide adapted from content from Dr. John Zack, MESO Inc. and used with permission

![](_page_30_Figure_6.jpeg)

#### This is among the list of known problems listed on the ERA5 web page…but who has time to read the manual or footnotes?<br>B. ERA5 diurnal cycle for near surface winds: the hourly data reveals a mismatch in the analysed near surface wind speed between the end of one assimilation cycle and the

beginning of the next (which occurs at 9:00 - 10:00 and 21:00 - 22:00 UTC). This problem mostly occurs in low latitude oceanic regions, though it can also be seen over Europe and the USA. We cannot rectify this problem in the analyses. The forecast near surface winds show much better agreement between the assimilation cycles, at

![](_page_30_Picture_11.jpeg)

### **Low Hanging Fruit for Validation and Uncertainty Quantification**

- Comprehensive industry wide data transparency and sharing is required: Met., generation, and availability data
	- Little proprietary value per site but a tremendous untapped asset if made public across all generators
- This will enable validation and UQ of synthetic datasets which is imperative for valid application. Ground truth data is also key to the model improvement process
- ERCOT is leading the way. Others should follow ASAP
	- This might require legislation/regulation.

![](_page_31_Picture_6.jpeg)

![](_page_31_Picture_7.jpeg)

![](_page_31_Picture_8.jpeg)

![](_page_31_Picture_9.jpeg)

## **User Knowledge/Education**

#### **Overview of the Current World of Datasets for Power System Planning**

- $\Box$  Wide Range of Methods to Construct Datasets
	- A few fundamental types of approaches
	- Enormous number of significant variations within types  $\circ$

#### $\Box$  Therefore: Wide Range of Datasets Exist

- **Typically have very different attributes depending**  $\circ$ on how they were constructed
- Consistency of data attributes (e.g. spatial/temporal  $\circ$ correlations) between datasets should not be assumed
- Critical need to evaluate comparative performance  $\circ$ on parameters/scales important to specific applications

![](_page_32_Figure_9.jpeg)

Let's examine the key attributes of the fundamental types of approaches...

#### Used with permission.

![](_page_32_Picture_12.jpeg)

# **Summary**

- Weather impact on the electric system is increasing significantly in complex, interdependent ways that need to be considered in planning and operations
- **EX Current weather datasets and practices for their use are inadequate and** urgently need updating. Key issues:
	- Lack of attention to validation and uncertainty quantification
	- Lack of understand of the inaccuracies, biases, and limitations of available synthetic weather data. They should not be treated like observations.
	- Lack of holistic vision, sector coordination, and funding to fix the problem
	- Lack of sector weather and power data sharing to improve modeling and validation
- Producing quality fit-for-purpose weather datasets is achievable, and an imperative ongoing need for the sector. Not a once and done research project.
- **Industry momentum to move forward is building**
- **EPRI's has the trans-disciplinary skills to coordinate this with multiple partners** 
	- A strong match for Public-Private Partnership
	- Looking for utility partners to team with to push forward the process

# **Sharply** Focused

![](_page_34_Picture_1.jpeg)

# GridLAB

CI