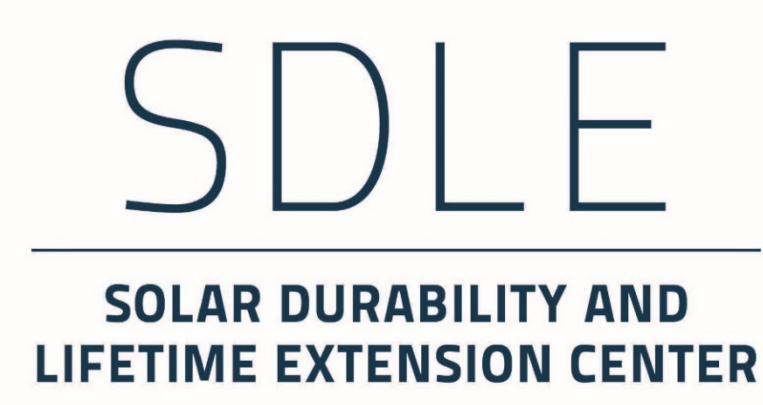


# Data-driven Design of PV Module Materials: Informed by a Non-relational Data Warehouse & Analytics Environment with > 3.4 GW of PV Plant Datasets



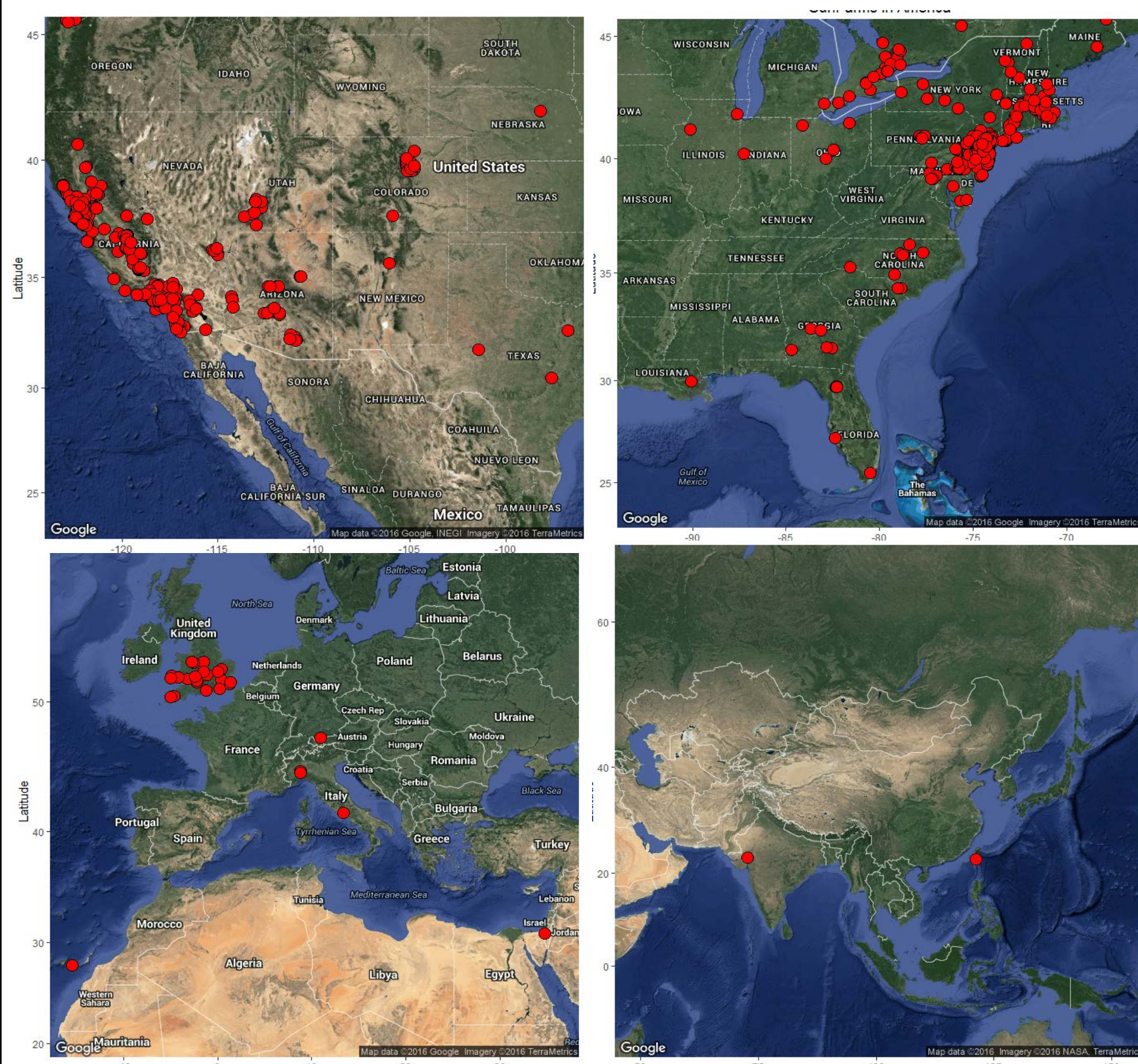
Roger H. French<sup>1</sup>, Laura S. Bruckman, and Timothy J. Peshek

[1] Solar Durability & Lifetime Extension Center, Case Western Reserve University, Cleveland OH, USA

## Objective

- A. Bringing operating PV plants into research**
- As a critical epidemiological population
  - Non-relational data warehouse – Energy-CRADLE
  - Currently hosting data from over 3.4GW
  - From 780 field developed power plants
- B. Using data science methods**
- Such as machine learning, predictive network modeling
  - Aggregate and integrate DuraMat research results
  - Into common system-level models of PV modules
  - Exposed to real-world conditions and lab-based accelerated exposures

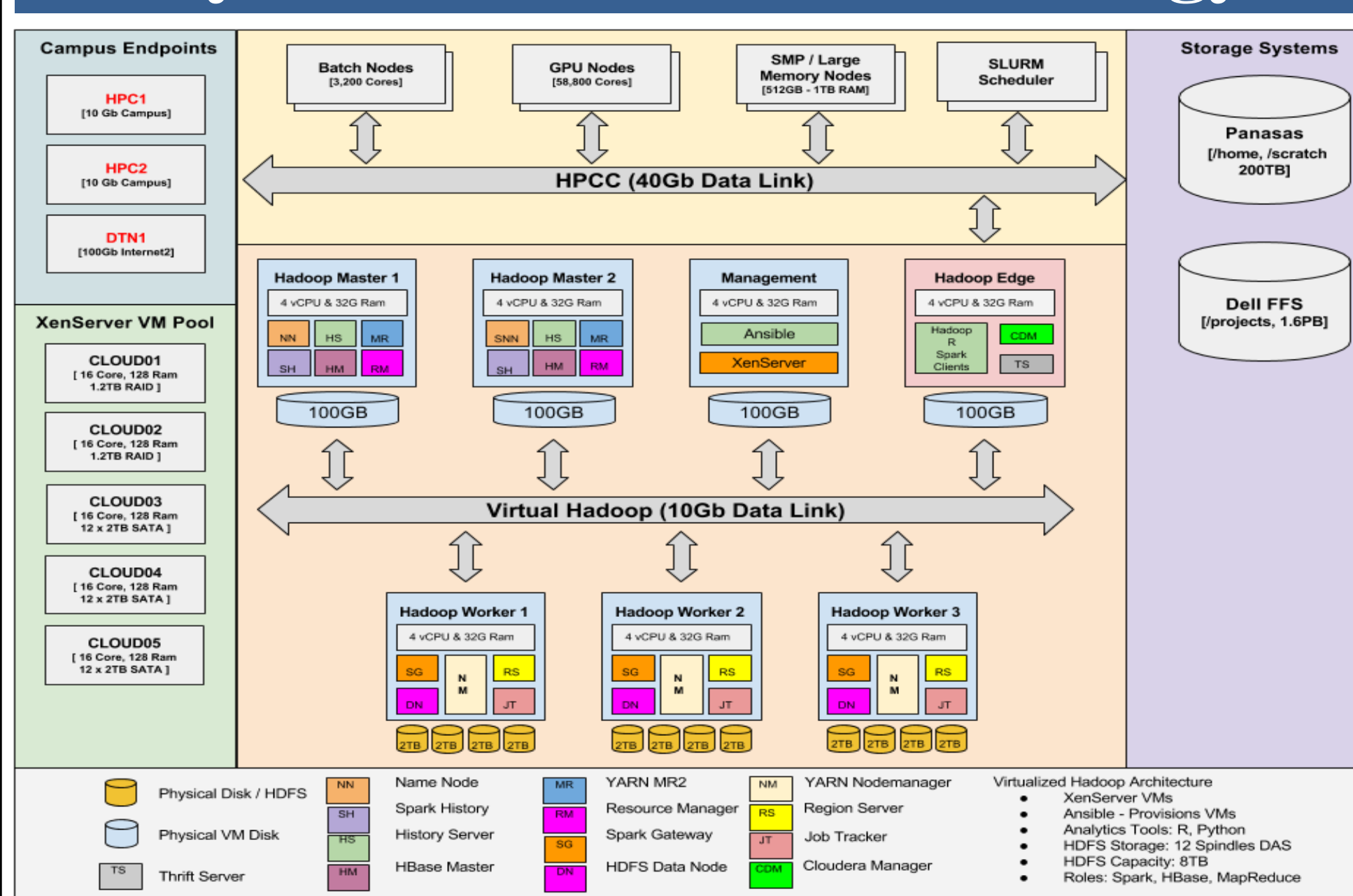
## Global SunFarm Network



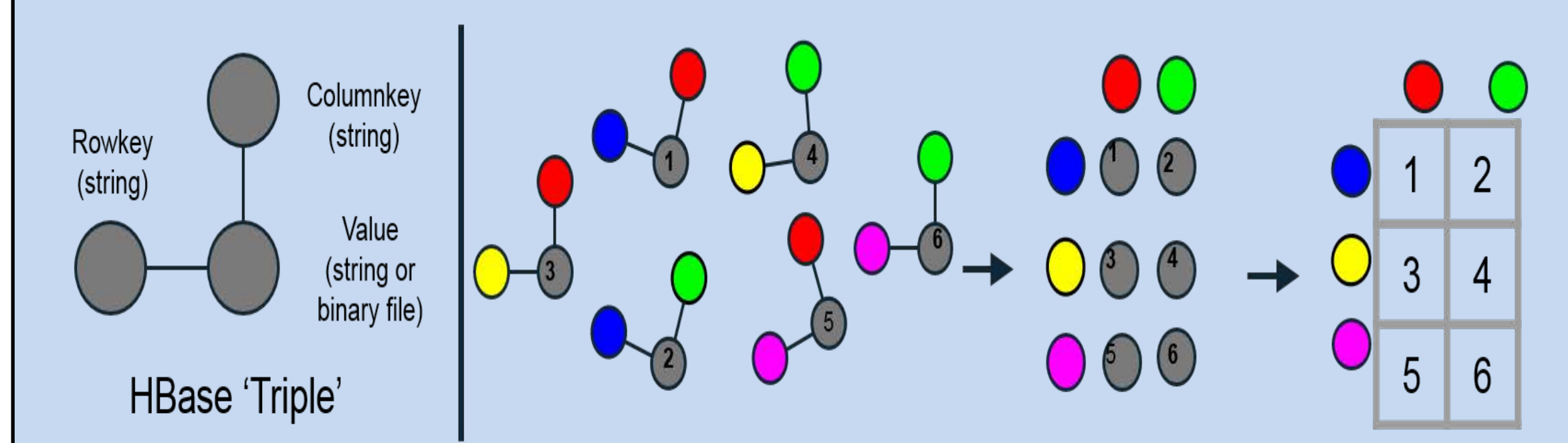
### SDLE PV Data Covers ~3.4 GW

- 787 PV Plant Sites encompasses 1.92% of Global PV Plant Generation
- 5638 PV Plants (Inv. & Modules)
- Distributed in 13 different climate zones
- 60 PV Module Brands/Models
- 38 PV Inverter Brands/Models
- Single Modules to 265 MW power plants
- Going Back Up To 15 years

## Common Research & Analytics Data Lifecycle Environment for Energy



## Hadoop/Hbase & NoSQL DB Abstraction [2]



Combines Lab data (Spectra, Images etc.) With Time-series Data (PV Power Plant Data)

### High Performance PV Data Analytics: Petabyte Data Warehouse In A Petaflop HPC Environment

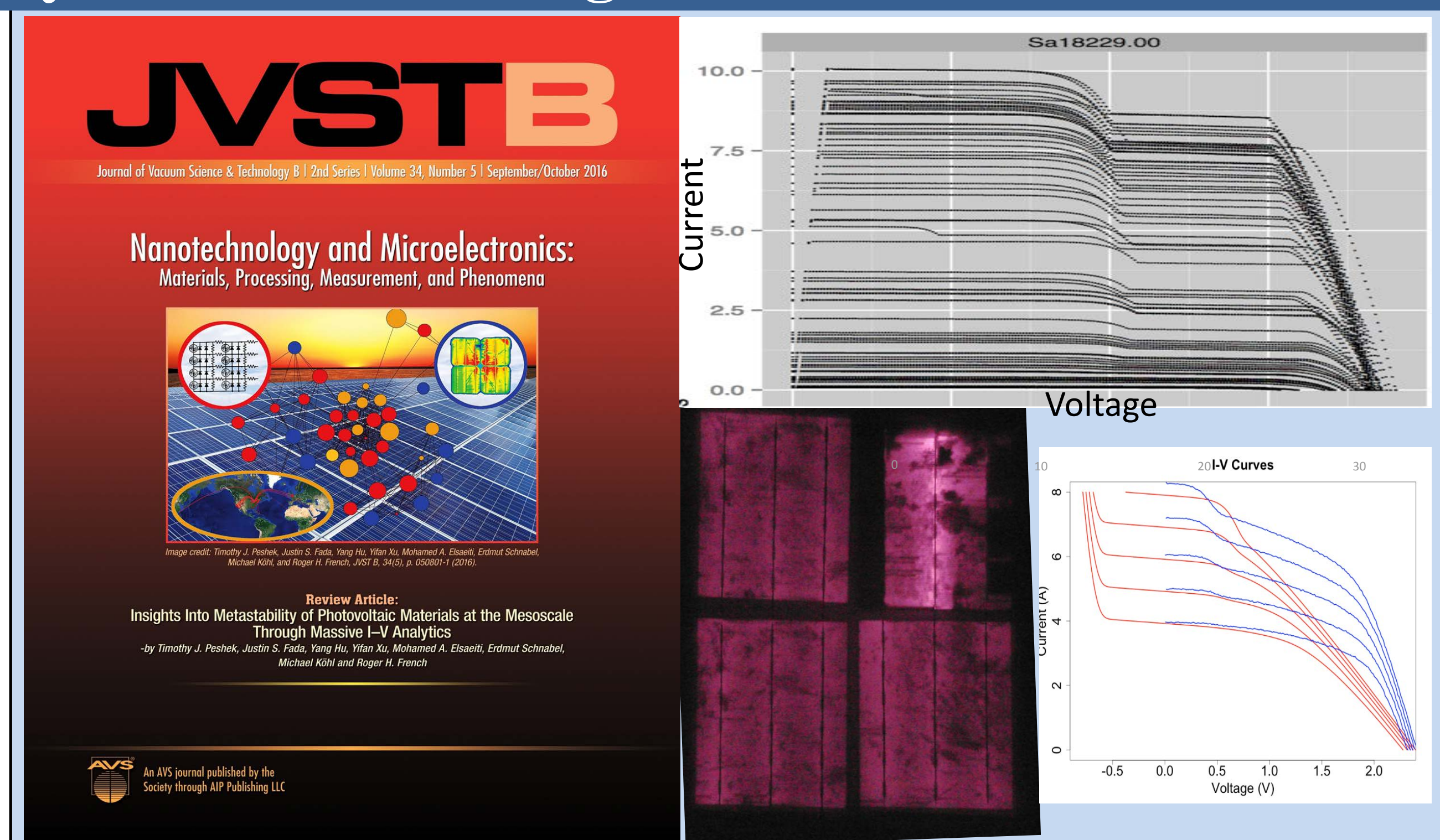
- In-place Analytics: Distributed R-analytics in Hadoop/HDFS
- In-memory Data Extraction: To Separate HPC Compute Nodes

A non-relational data warehouse for the analysis of field and laboratory data from multiple heterogeneous photovoltaic test sites

**JVSTB In Press**

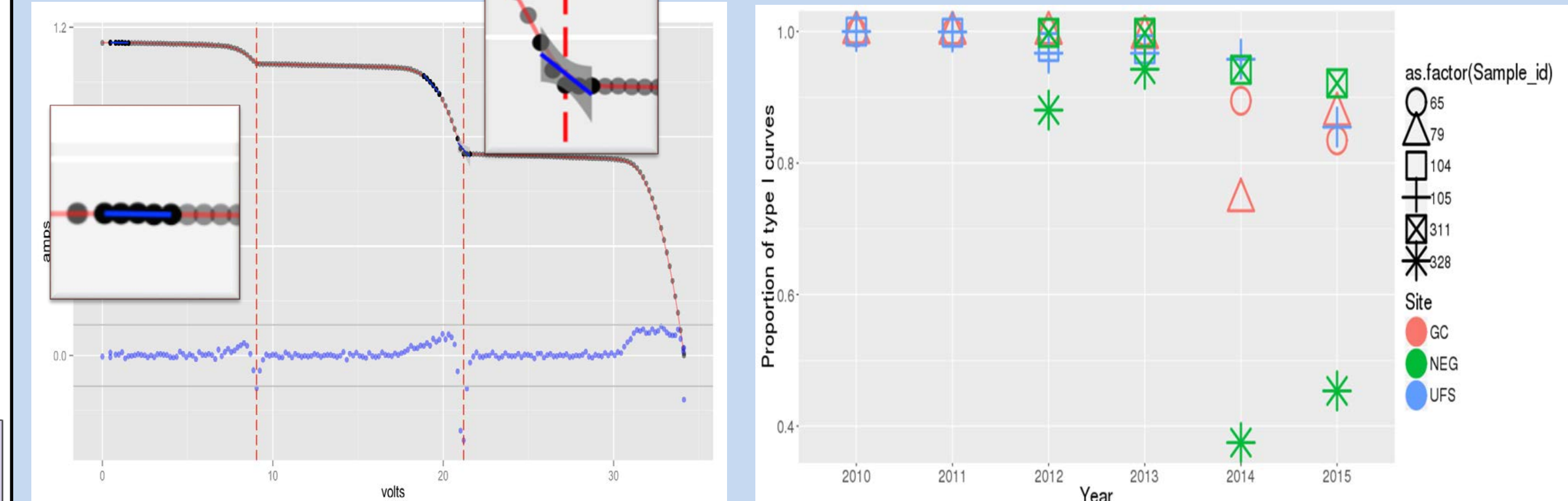
Yang Hu, Member, IEEE, Venkat Yashwanth Gunapati, Pei Zhao, Devin Gordon, Nicholas R. Wheeler, Mohammad A. Hossain, Member, IEEE, Timothy J. Peshek, Member, IEEE, Laura S. Bruckman, Guo-Qiang Zhang, Member, IEEE, and Roger H. French, Member, IEEE

## Metastability of PV Module Materials Identified by Machine Learning 2.2 million I-V curves[2][3]



### Insights into metastability of photovoltaic materials at the mesoscale through massive I-V analytics

**JVSTB** Timothy J. Peshek,<sup>1,\*)</sup> Justin S. Fada,<sup>1</sup> Yang Hu,<sup>1</sup> Yifan Xu,<sup>1</sup> Mohamed A. Elsaeti,<sup>1</sup> Erdmut Schnabel,<sup>2</sup> Michael Köhl,<sup>2</sup> and Roger H. French<sup>1</sup>



### Underlying Machine Learning Procedures:

Local linear regression fitting + Residual Thresholding

Classify I-V curve into five categories

- Type I** :  $V_{oc}$  only
- Type II** :  $V_{oc}$  + one bypass diode turns on
- Type III** :  $V_{oc}$  + two bypass diodes turn on

Detecting heterogeneity in PV modules from massive real-world "step" I-V curves: A machine learning approach

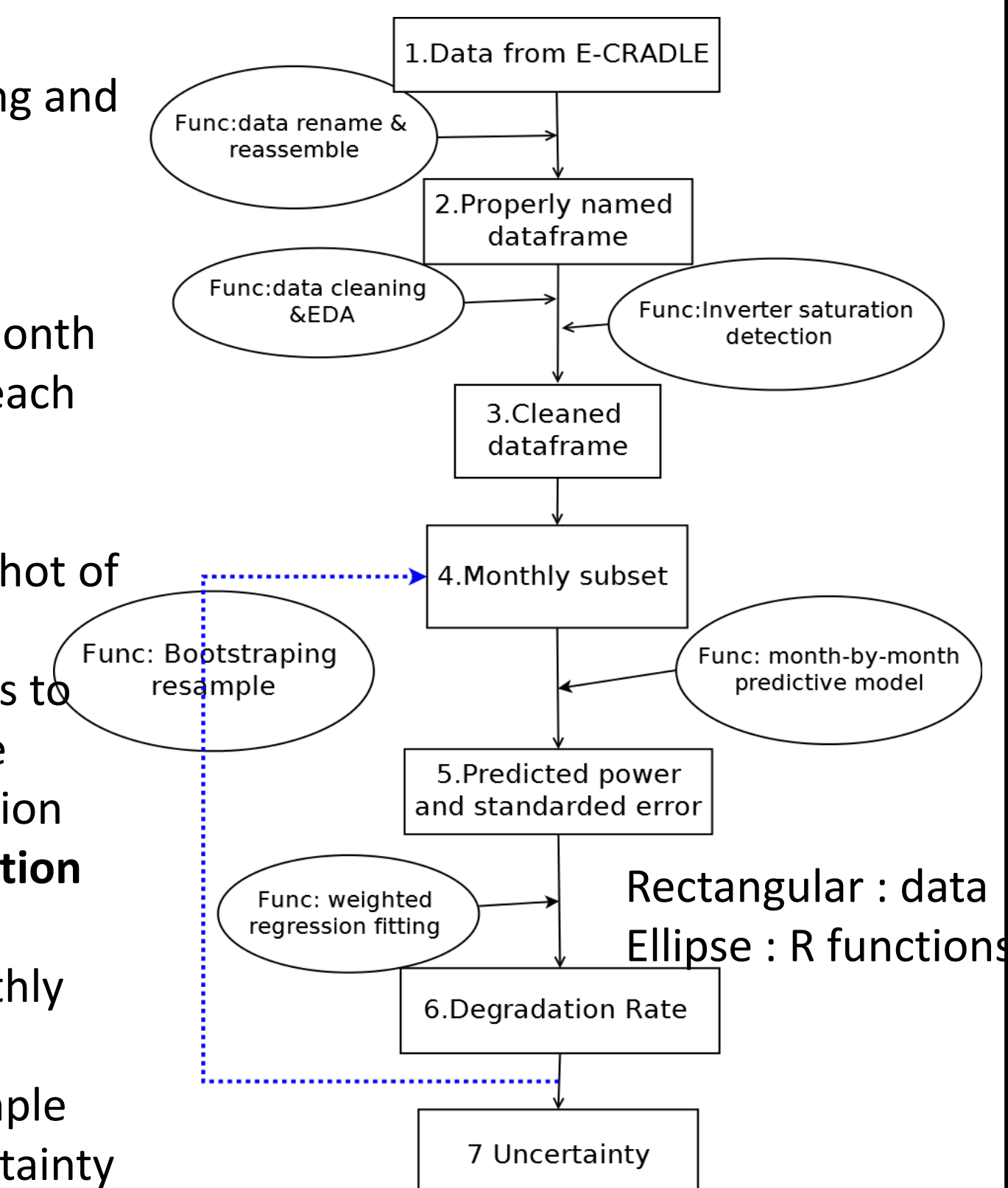
Yang Hu<sup>1</sup>, Erdmut Schnabel<sup>2</sup>, Michael Köhl<sup>2</sup>, Roger H. French<sup>1</sup>, and Timothy J. Peshek<sup>1</sup>

**PVSC 2016 proceedings**

## Pipelined Time-Series Analysis Studies

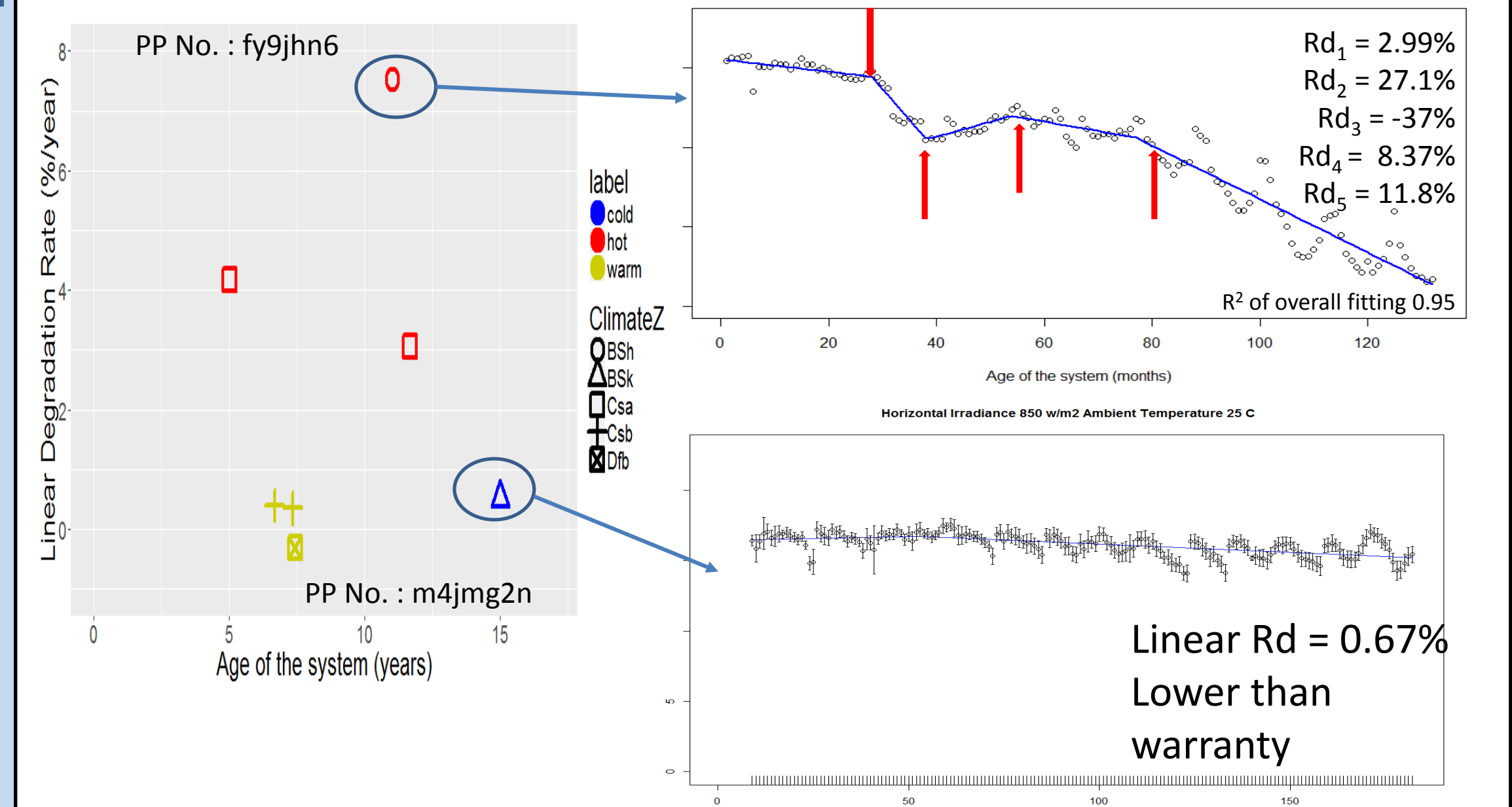
### Month-by-Month Rd Model

- Use all data noon time, morning and afternoon
- Categorize data by age
- Every 30 days from the first operation date considered a month
- Develop predictive model for each month
- Linear regression models
- These models serve as a snapshot of the system status
- Use monthly regression models to normalize system performance under the same climate condition
- Do not assume linear degradation rate**
- Look at the profile of the monthly predicted value
- Use bootstrap approach resample the data to estimate the uncertainty

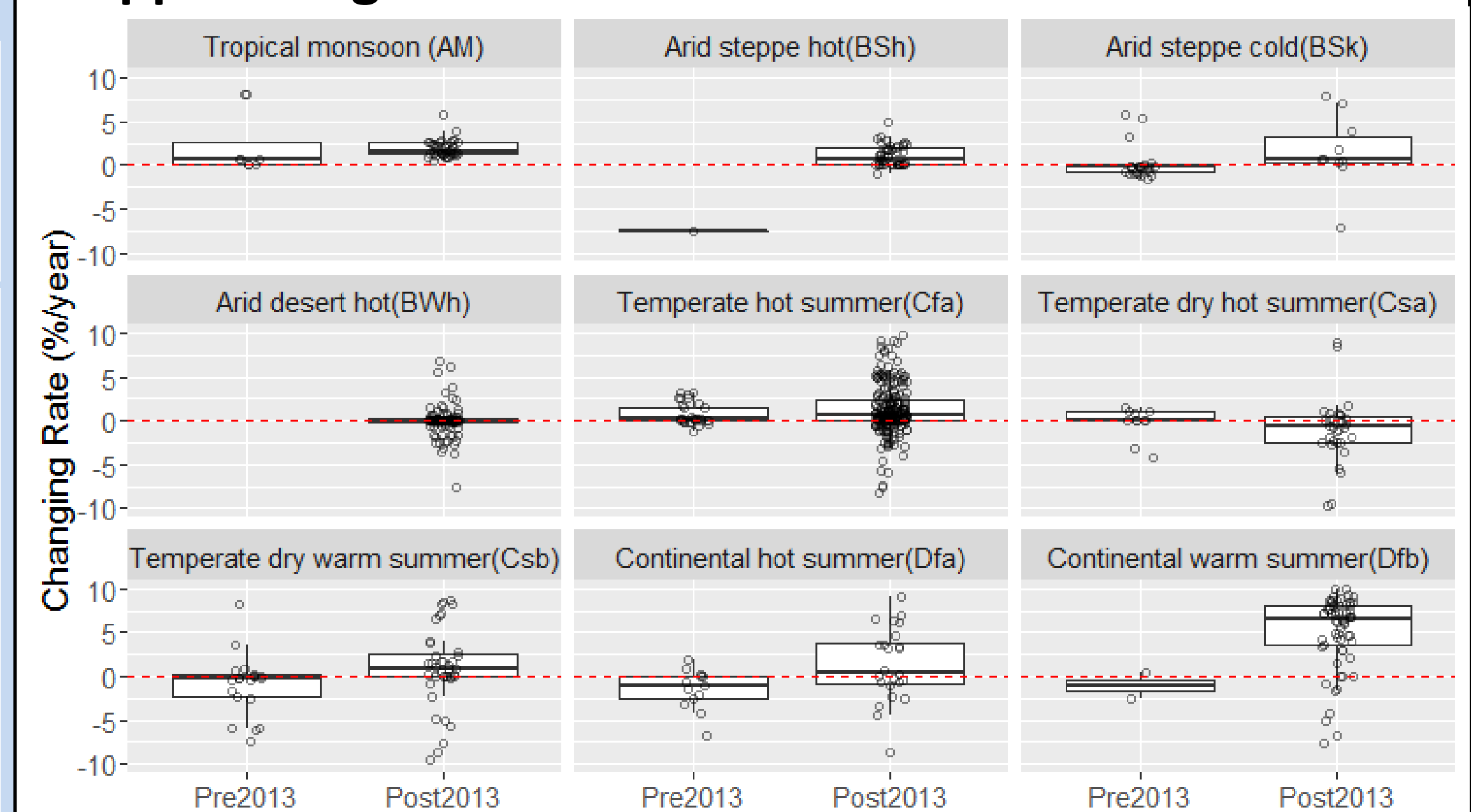


### PV systems degrade faster than warranty in hot climate

- Seven PV systems
- Identical PV modules



### Changing rate distribution of 804 PV systems at 9 Koppen-Geiger climate zones



## Acknowledgements

- This work is supported by the Bay Area Photovoltaic Consortium Prime Award No. DE-EE0004946, Subaward Agreement No. 60220829-51077-T.
- The SDLE center was established through funding through the Ohio Third Frontier, Wright Project Program Award Tech 12-004.

## References

- Hu, Yang et al. "A non-relational data warehouse for the analysis of field and laboratory data from multiple heterogeneous photovoltaic test sites" *Journal of Photovoltaics*, IEEE in press
- Peshek, Timothy J., et al. "Insights into metastability of photovoltaic materials at the mesoscale through massive I-V analytics." *Journal of Vacuum Science & Technology B* 34.5 (2016): 050801.
- Hu, Yang et al. "Detecting heterogeneity in PV modules from massive real-world "step" I-V curves: A machine learning approach." *PVSC 2016 proceedings*