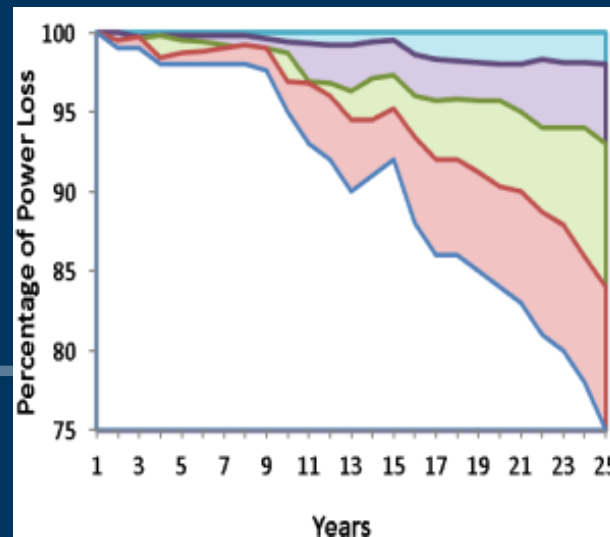
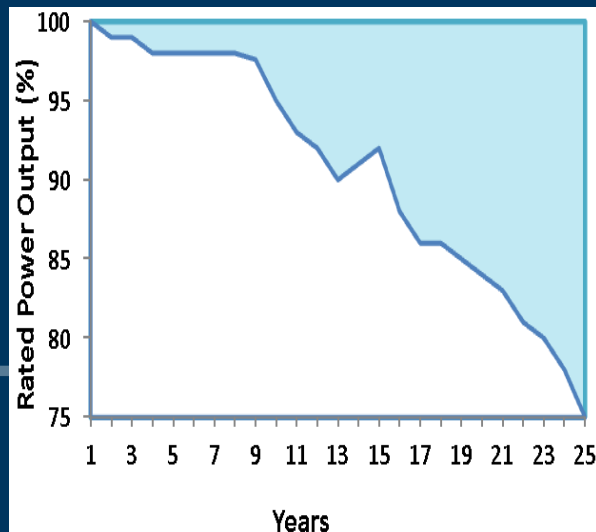


Lifetime & Degradation Science: *A Data Science Approach for Reliability*

Roger H. French

Solar Durability & Lifetime Extension Center
Case Western Reserve University
Cleveland OH 44106



SDLE Acknowledgements



Research Associates

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Funding



Outline

Lifetime and Degradation Science

- A Data Science approach to reliability

Photodegradation of Acrylic

- Domain Guided Statistical Analytics

PV Module Degradation

- Under 85C/85RH Damp Heat Exposure

Real World PV Module Performance

- Analytics of 20 Brands of PV Modules

Conclusions

The Lifetime & Degradation Science Approach: L&DS Triangle

Integrate reliability & lifetime performance throughout

- Entire Technology Development process.
- Applicable to High Cost of Failure systems

Develop Population-based Studies

- Real world operational studies
- With accelerated exposure studies
- Across multiple technology generations

Data Science Approach Using

- Accelerated Lab Exposures
- Real World Exposures
- Automated Data Acquisition
- Unbiased Analysis

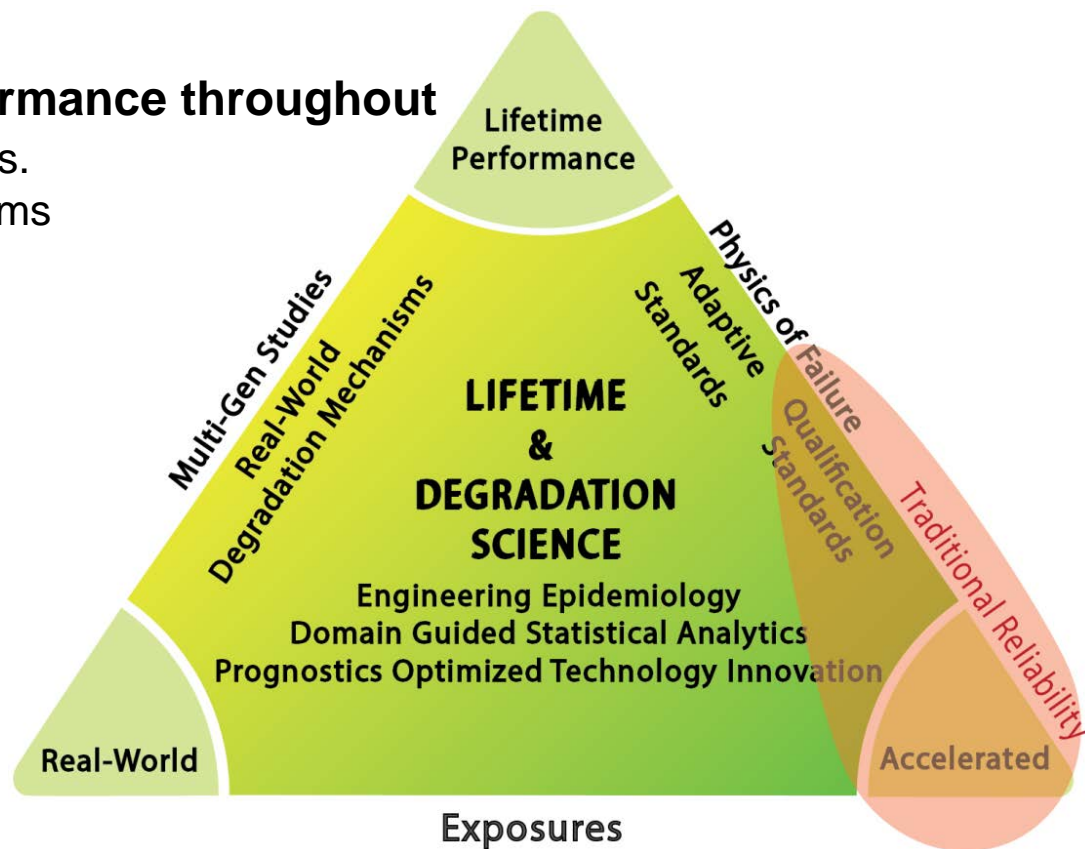
Develop domain-science guided, statistically informed data analytics

- Founded in real-world degradation mechanisms

Based on mechanistic pathway models

- Predict system's lifetime performance

Prognostics to guide next generation technology development



CWRU

Solar Durability & Lifetime Extension Center



US Energy Secretary Steven Chu
visits CWRU SDLE & MORE Centers

Focused on the
Societal Impact of
Technology Reliability Fiascos

Long Lived,
Multi-generational
Engineered Systems

Solar Durability & Lifetime Extension Center

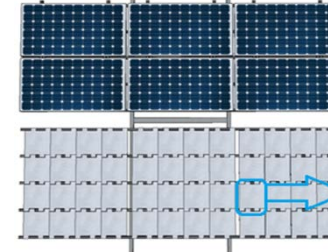
Lifetime and Degradation Science

- A stress-response framework
- Determine degradation mechanisms & Rates
- Efficacy and lifetime prediction & standards



Exposures: Outdoors, Real Time

- 14 2-axis trackers & fixed mount
- 142 Modules & 8000 Samples
- 1, 2, 4, 5X Suns Irradiance



Exposures: Indoors, Accelerated

- Solar Simulators: 1-50 & 50-1200X Suns
- QUV/Spray, Q-Sun, Q-Fog
- CSZ Environmental Test Chambers

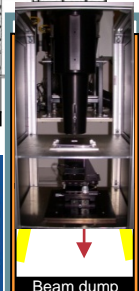
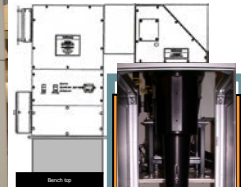
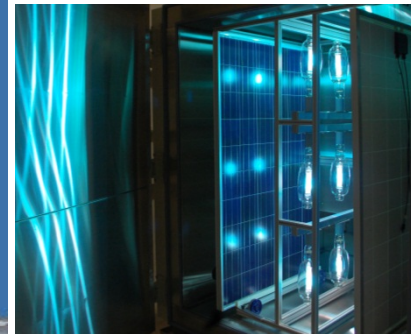


Evaluations:

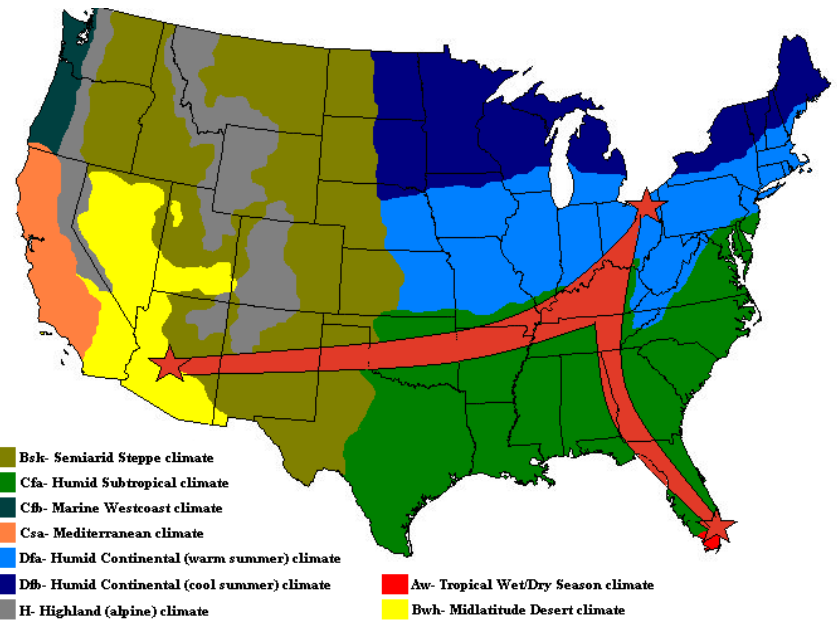
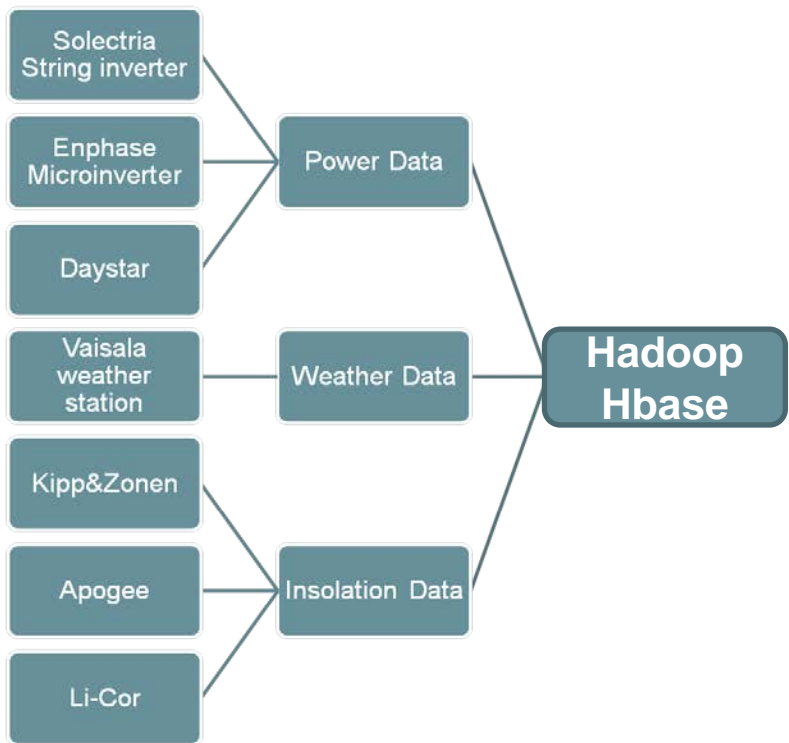
- Optical, Interfacial, Physical
- IR, Raman, UV/Vis VUV,
- Ellipsometry, Light scattering

Integration/Fabrication

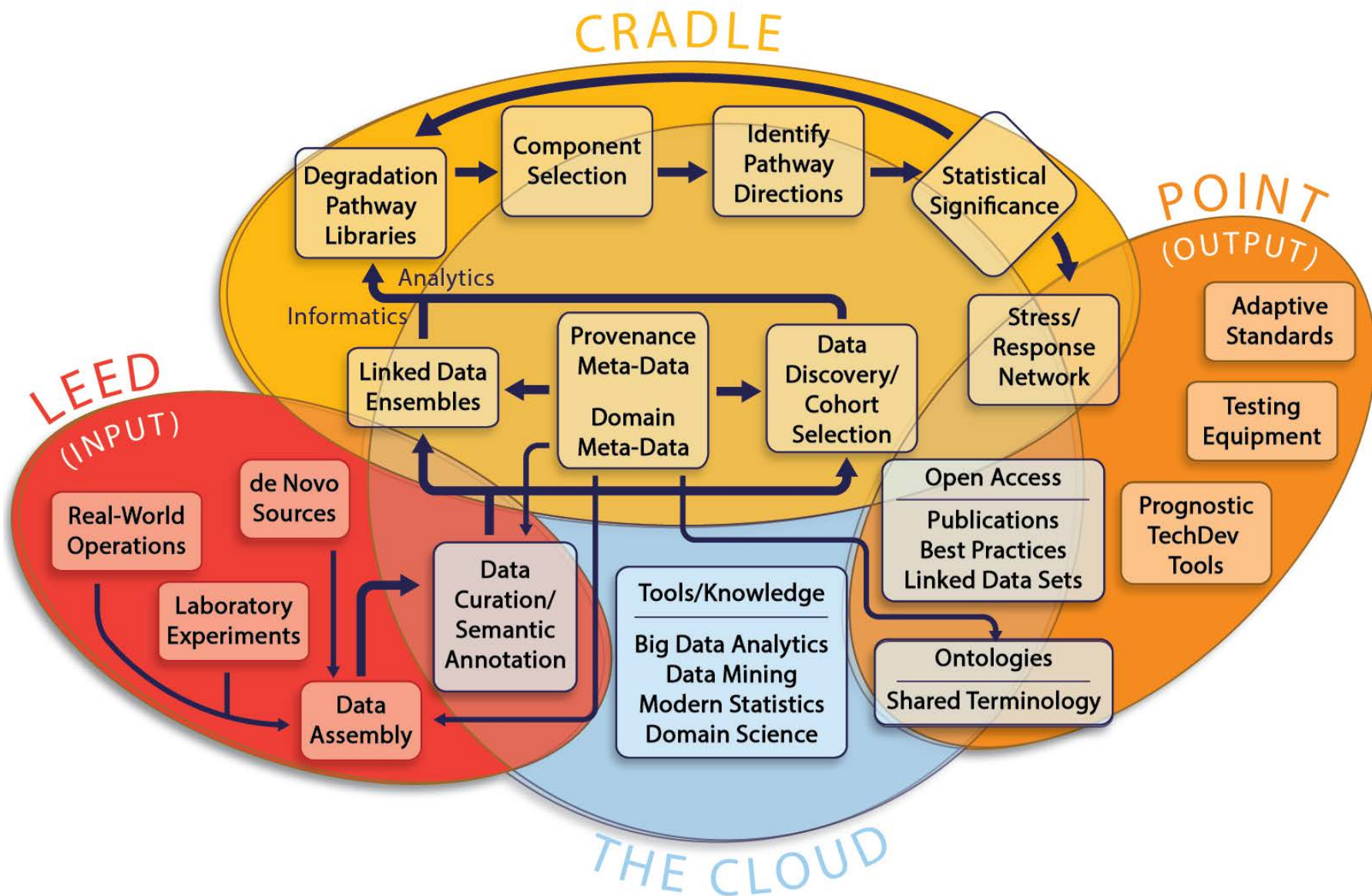
- Mini- & full size modules



Energy-CRADLE: data source



L&DS: System for Lifetime and Durability



Hadoop/Hbase Energy CRADLE Operational

Energy CRADLE

- Up and Running
- Processing Data
- Of 4 SunFarms

Store 7 Years Data

- Minute by Minute

Only at 10% Capacity

10 Node Cluster

- 20 Cores
- 100 Gb RAM
- 1 Tb Storage

Now Building

- Analytics Servers
- Point in Time Servers

And Scaling CRADLE

Energy CRADLE - Hadoop Interface DataProcessing Portal

CRADLE Hadoop Map/Reduce Interface

CRADLE Hadoop HDFS System

CRADLE HBase File System

master Hadoop Map/Reduce Administration

State: RUNNING
 Started: Mon Sep 23 11:24:37 EDT 2013
 Version: 1.1.2, r1440782
 Compiled: Thu Jan 31 02:03:24 UTC 2013 by hortonofo
 Identifier: 201309231124
 SafeMode: OFF

Cluster Summary (Heap Size is 289.44 MB/888.94 MB)

Running Map Tasks	Running Reduce Tasks	Total Submissions	Nodes	Occupied Map Slots	Occupied Reduce Slots	Reserved Map Slots	Reserved Reduce Slots	Map Task Capacity	Reduce Task Capacity	Avg. Tasks/Node	Blacklisted Nodes	Graylisted Nodes	Excluded Nodes
0	0	40	9	0	0	0	0	18	18	4.00	0	0	0

Scheduling Information

Queue Name	State	Scheduling Information
default	running	N/A

Filter (Jobid, Priority, User, Name)

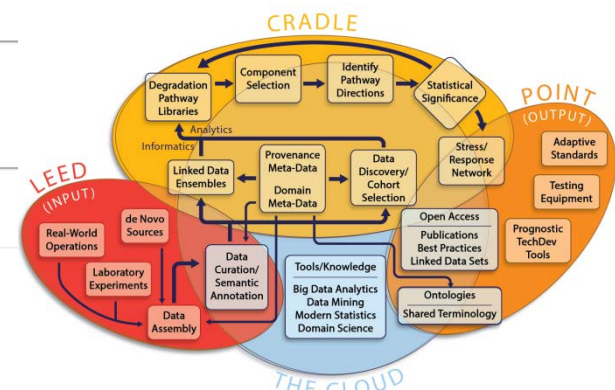
Example: 'user:smith 3200' will filter by 'smith' only in the user field and '3200' in all fields

Running Jobs

[None](#)

Completed Jobs

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L&DS: Domain-Science Guided, Statistical Analytics

Founded in Physical/Chemical Mechanisms

Exploratory Data Analysis

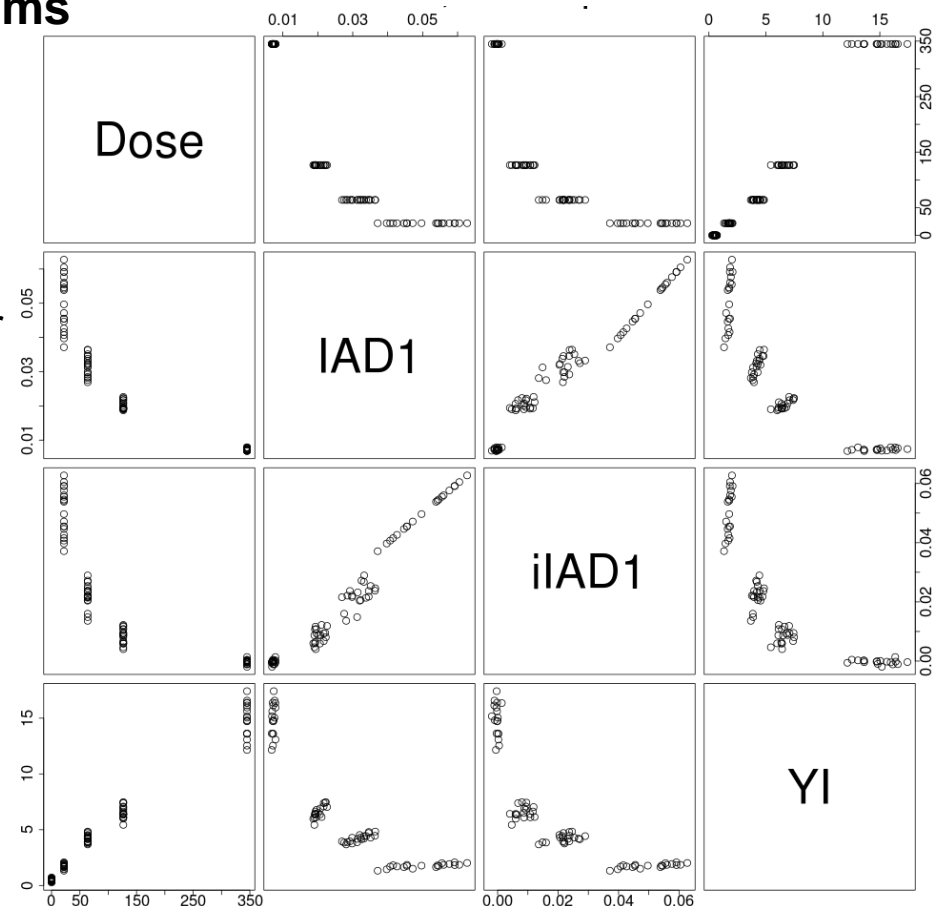
- Pairs Plots: Survey of Large Data Sets
- Explore relationships compared to time/dose
- Explore relationships compared to one another
- Information Content, Correlation, Significance

Functional Forms among Variables

- Based on Fundamental Mechanisms
- Using realistic functional forms
- No over fitting, check goodness of fit

Number of Coincident Observations

- $n-2$
- Degrees of Freedom



Combination of Metrics for Statistically Significant Relationships

- R^2 , Adjusted- R^2 , Predictive- R^2
- p-values
- Akaike Information Content (AIC) (Information Entropy)

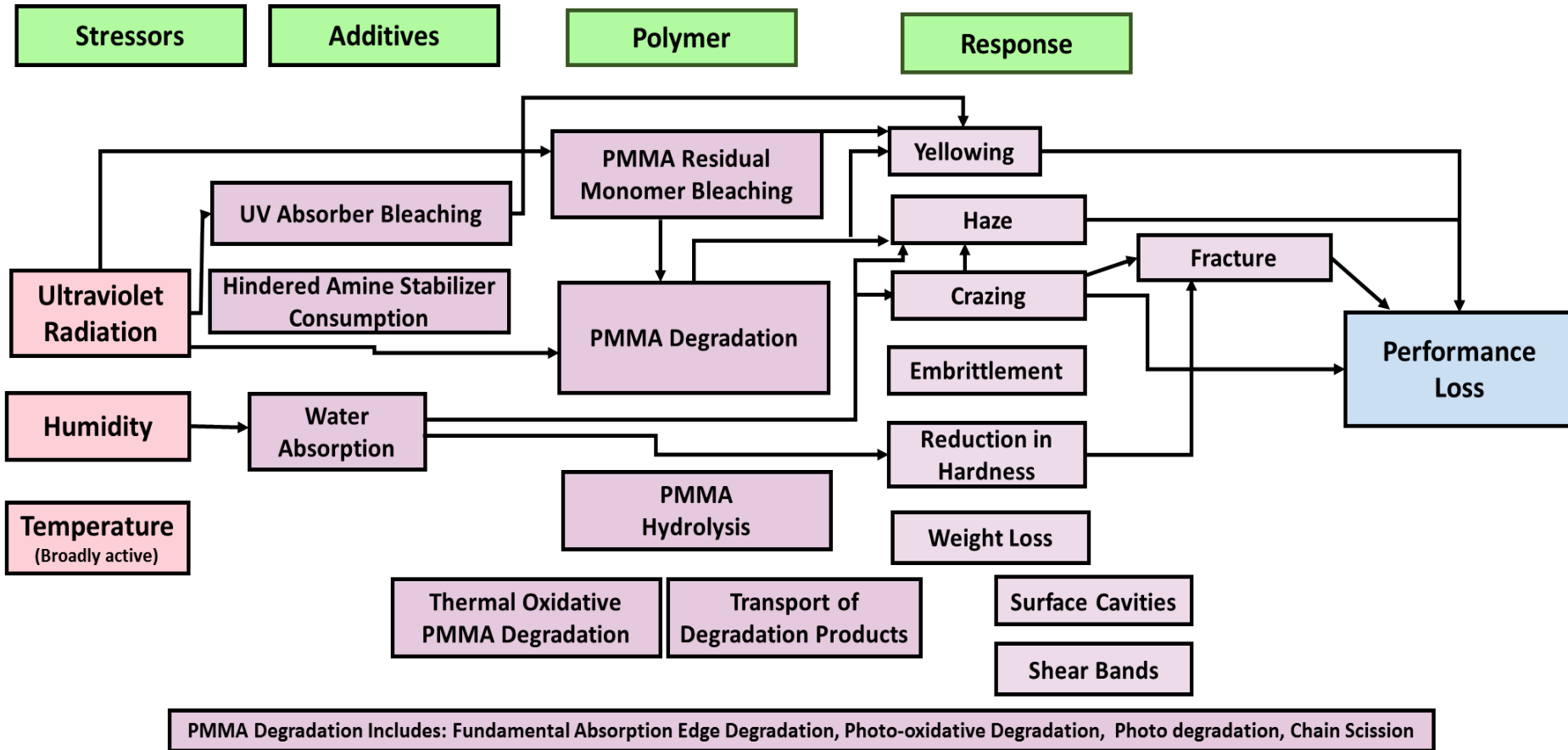
Analytics of Degradation Pathways of Solar Grade Acrylic

UVT

MPA

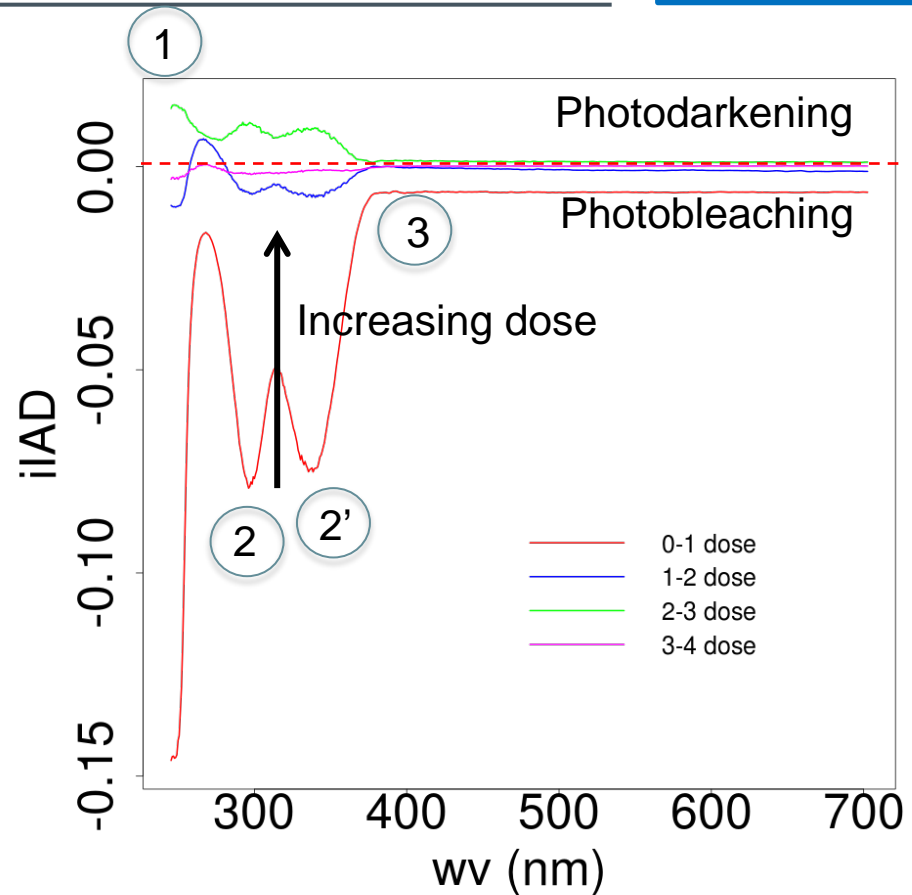
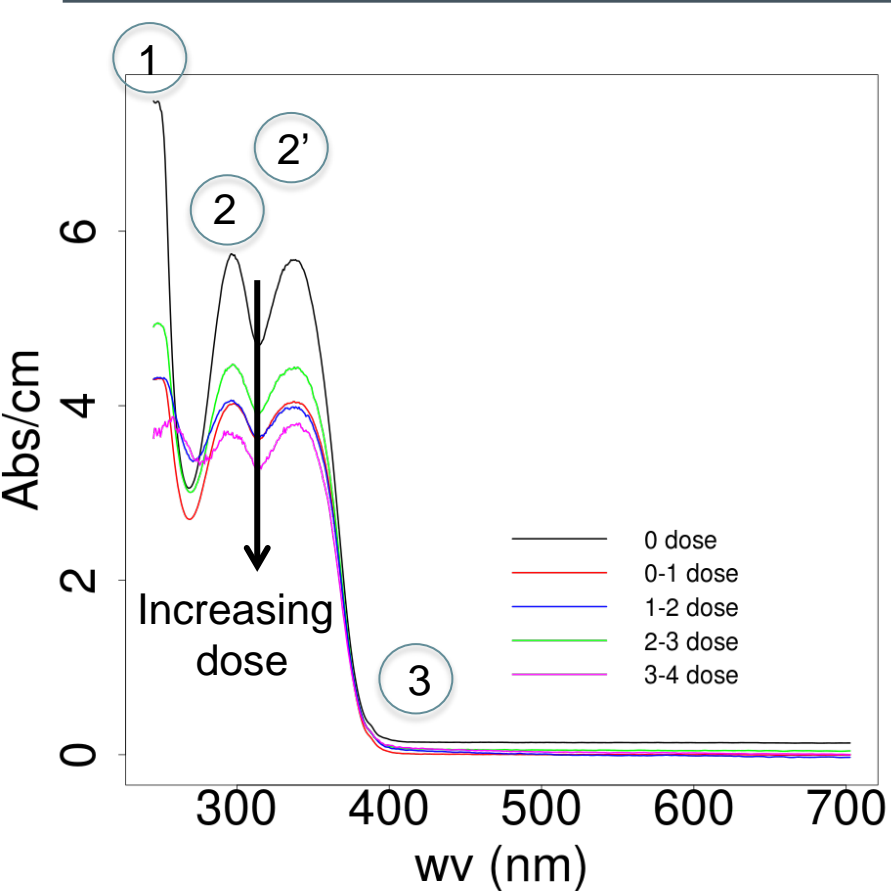
Myles Murray, Laura Bruckman

Acrylic Degradation Mechanism Pathway Library: Survey



Each Node Represents a Degradation Mechanism

Sample of Possible Degradation Mechanism Pathways



1-Fundamental Absorption Edge

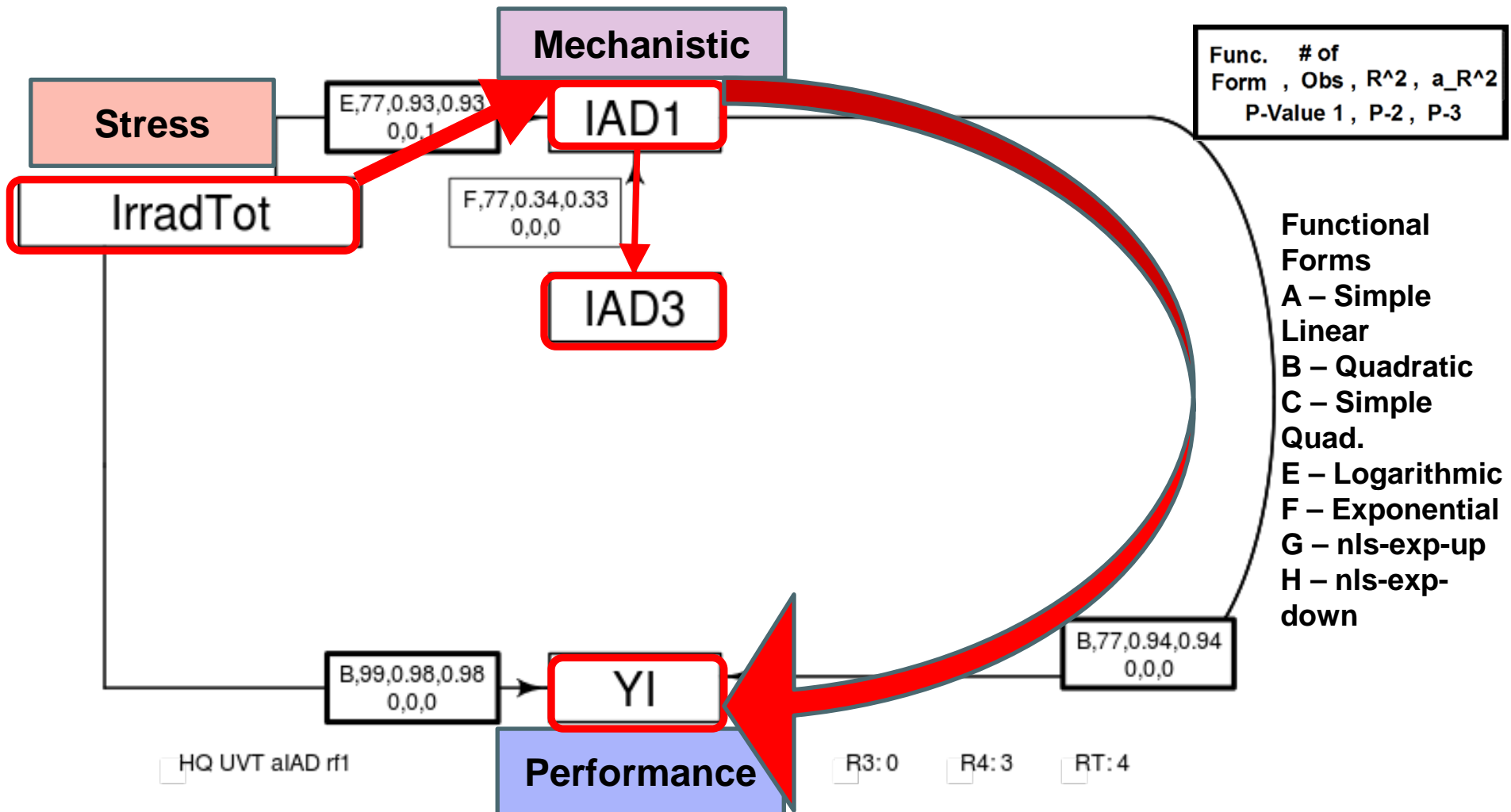
2 and 2' (297 & 339 nm)-Tinuvin Type Stabilizer Package

- Initial High Rate of Photobleaching

3-Yellowing Region

Statistical Pathway Diagram: UVT-Base Resin

UVT

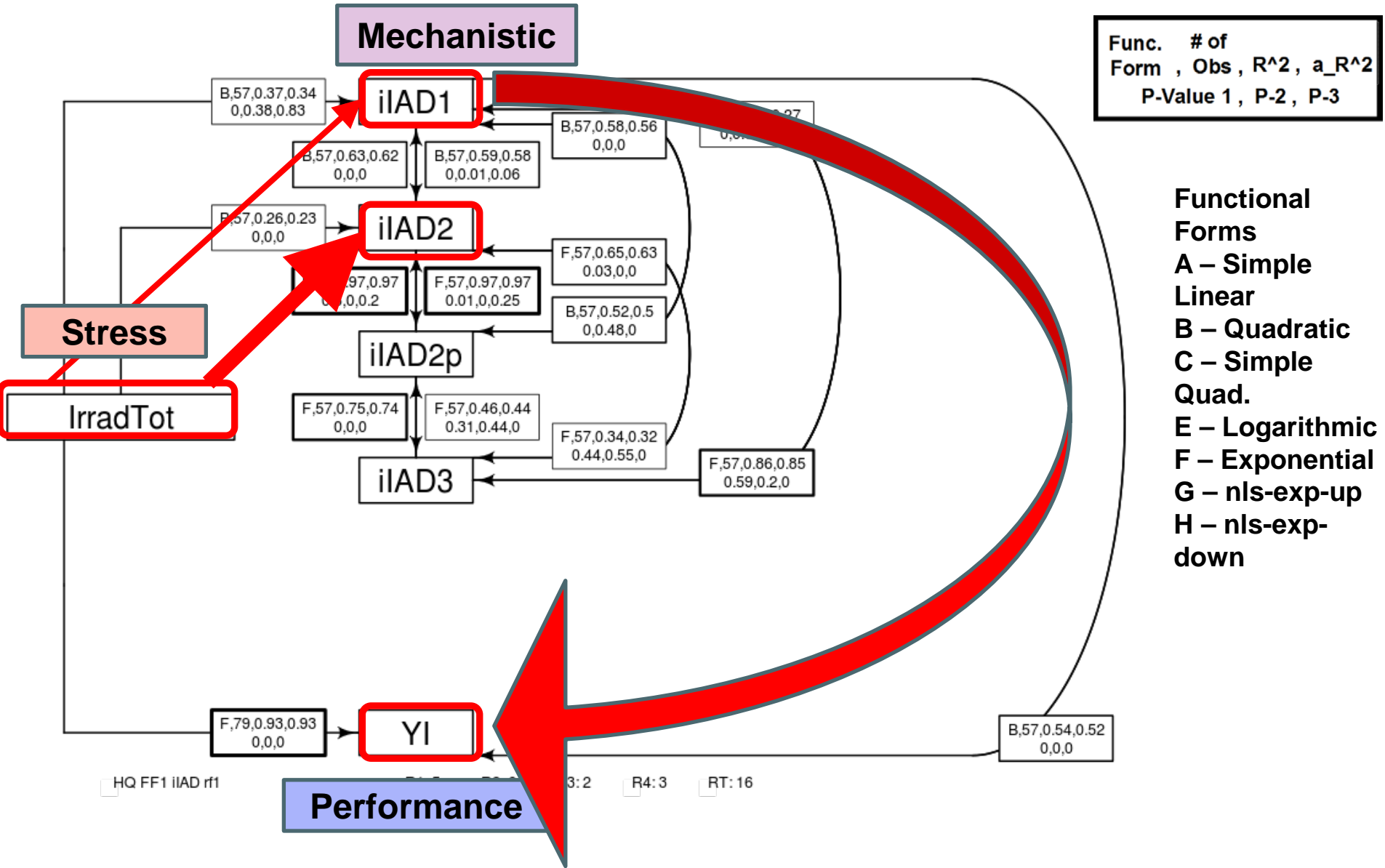


IAD1 (fundamental absorption edge) degrades-proceeds to yellowing.

Weak Relationship between IAD1 degradation and IAD3 (one wavelength)

Statistical Pathway Diagram: MPA Stabilizer

MPA



Domain-Science Guided, Statistical Analytics: PV Module Study

Statistical and Domain Analytics Applied to PV Module Lifetime and Degradation Science

LAURA S. BRUCKMAN¹, NICHOLAS R. WHEELER², JUNHENG MA³, ETHAN WANG⁴,
CARL K. WANG⁴, IVAN CHOU⁵, JIAYANG SUN³, AND ROGER H. FRENCH⁶ (Member, IEEE)

Semi-gSEM Methodology

Semi-Supervised by Domain Knowledge,
Generalized Structural Equation Modeling

Statistical and Domain Analytics Applied to PV Module Lifetime and Degradation Science

Laura S. Bruckman, Nicholas R. Wheeler, Junheng Ma, Ethan Wang, Carl K. Wang, Ivan Chou, Jiayang Sun and Roger H. French, *Member, IEEE*,

Abstract—A better understanding of the degradation modes and rates for photovoltaic (PV) modules is necessary to optimize and extend the lifetime of these modules. Lifetime and degradation science (L&DS) is used to understand degradation modes, mechanisms and rates of materials, components and systems in order to predict lifetime of PV modules. A PV module lifetime and degradation science (PVM L&DS) model is an essential component to predict lifetime and mitigate degradation of PV modules. Previously published accelerated testing data from Underwriter Laboratories on PV modules with fluorinated polyester (FPE) backsheets which included eight modules that were exposed to 4000 hours of damp heat (85% relative humidity at 85°C) and eight exposed to 4000 hours of ultraviolet light (80 W/m² of 280-400 nm wavelengths at 60°C) (UV preconditioning) were used to determine statistically significant relationships between the applied stresses and measured responses. There were fifteen different variables tracking aspects of system performance, degradation mechanisms, component metrics and time. Modules were analyzed for three system performance metrics (fill factor, peak power, and wet insulation). The results were statistically analyzed to identify variable transformations, statistically significant relationships, and to develop the PVM L&DS model informed by structural equation modeling techniques. The statistically significant relationships and significant model coefficients obtained in the first level above, combined with domain analytics,

A Recent U.S. Department of Energy workshop on Science for Energy Technologies [1] identified photovoltaics (PV) lifetime and degradation science (L&DS) [2], [3], [4] as a critical scientific challenge for robust adoption of PV. The PVQA Task Force was developed as an international task force to work towards defining what is needed for lifetime qualification standards and tests [5]. Developing and defining useful lifetime qualification standards and tests is complicated since even single degradation modes, mechanisms and rates are not clearly understood and two factor effects are even more complex. Therefore, a methodical domain and statistical approach is necessary to cross-correlate stressors, degrees of stress and degradation modes, mechanisms and rates for materials, components and systems. This cross-correlation can help provide a better understanding of degradation and lifetime performance in order to guarantee the minimal 25-year lifetime performance of PV modules [6].

A. Reliability and Prognostic Approaches

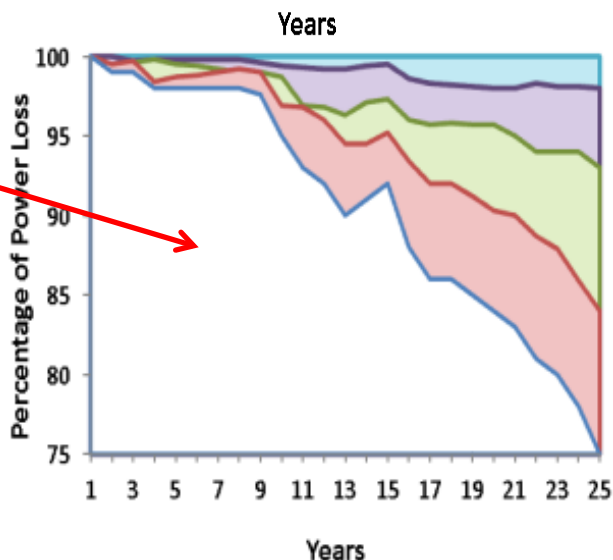
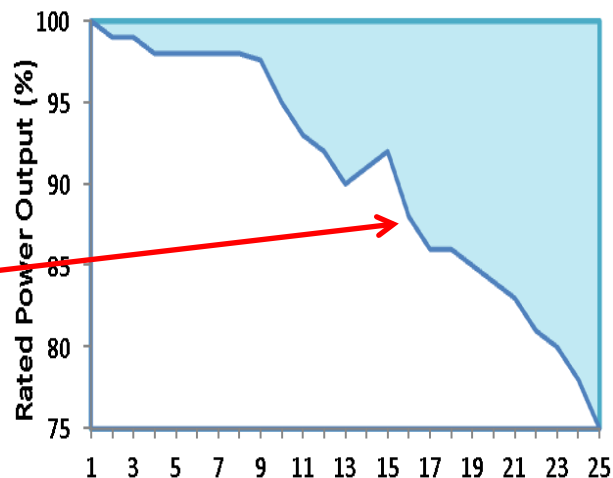
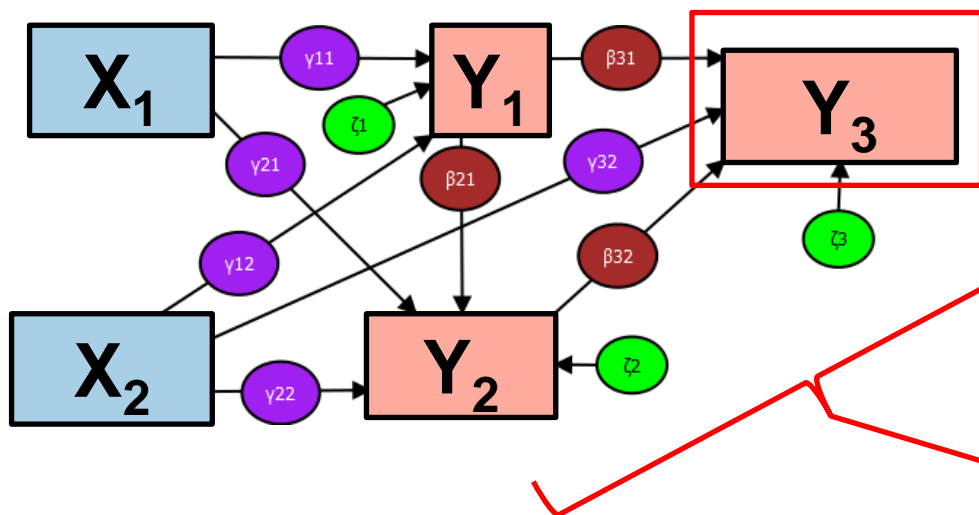
The original method to determine the reliability of a system or component was to collect failure data and use a single,

Laura S. Bruckman, Nicholas R. Wheeler, Junheng Ma, Ethan Wang, Carl K. Wang, Ivan Chou, Jiayang Sun, and Roger H. French, “Statistical and Domain Analytics Applied to PV Module Lifetime and Degradation Science”, IEEE Access, invited, 2013. DOI:[10.1109/ACCESS.2013.2267611](https://doi.org/10.1109/ACCESS.2013.2267611)

L&DS: Semi-gSEM Methodology

A methodology for determining degradation mechanisms and rates

- Connects materials and components
- Bulk material properties and surfaces, interfaces
- To predict and improve system lifetime performance



Inspiration: Structural Equation Modeling

$$Y_{1i} = Y_{10} + \gamma_{11}X_{1i} + \gamma_{12}X_{2i} + \zeta_{1i}$$

$$Y_{2i} = Y_{20} + \gamma_{21}X_{1i} + \gamma_{22}X_{2i} + \beta_{21}Y_{1i} + \zeta_{2i}$$

$$Y_{3i} = Y_{30} + \gamma_{32}X_{2i} + \beta_{31}Y_{1i} + \beta_{32}Y_{2i} + \zeta_{3i}$$

Generalized for Non-Linear Relationships

UL PV Module Experiments – Developed Semi-gSEM Methodology

16 Modules Measured While Exposed:

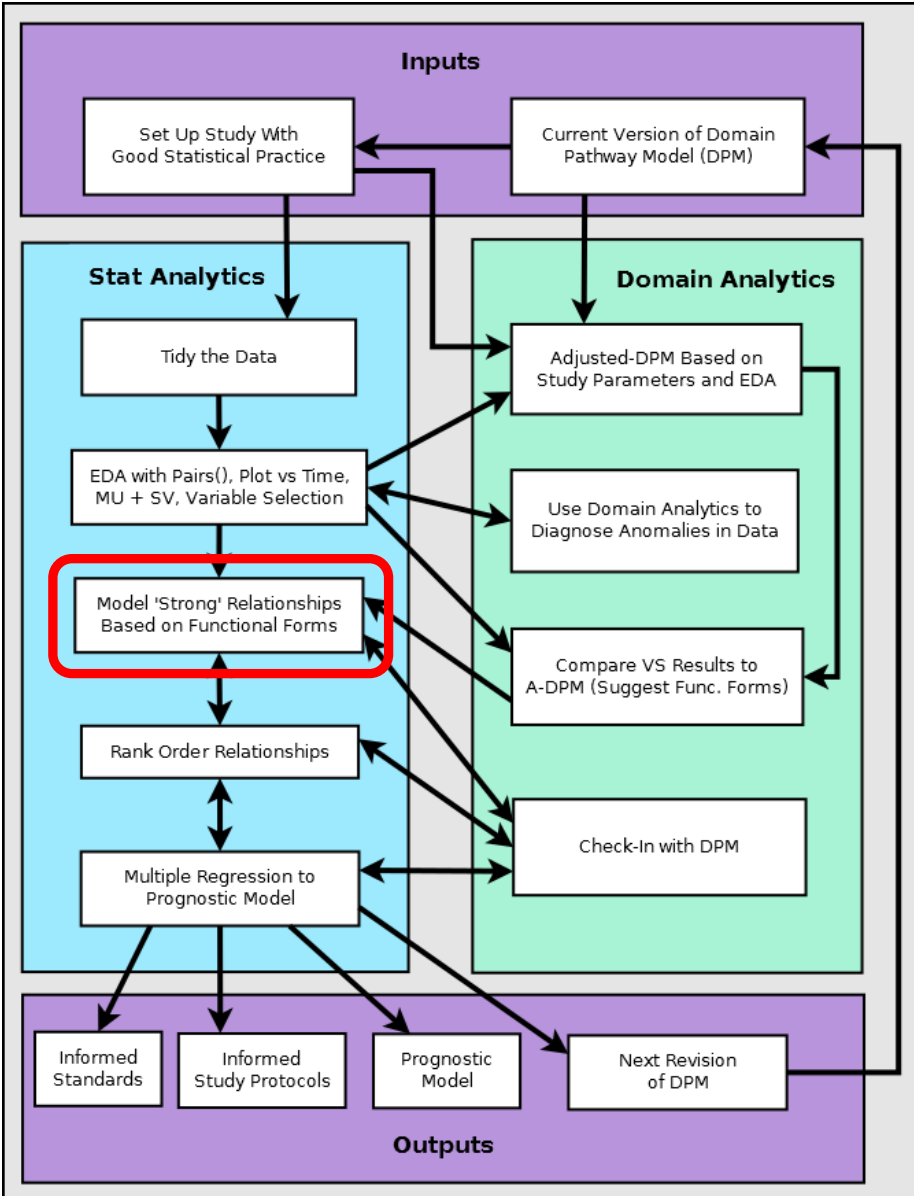
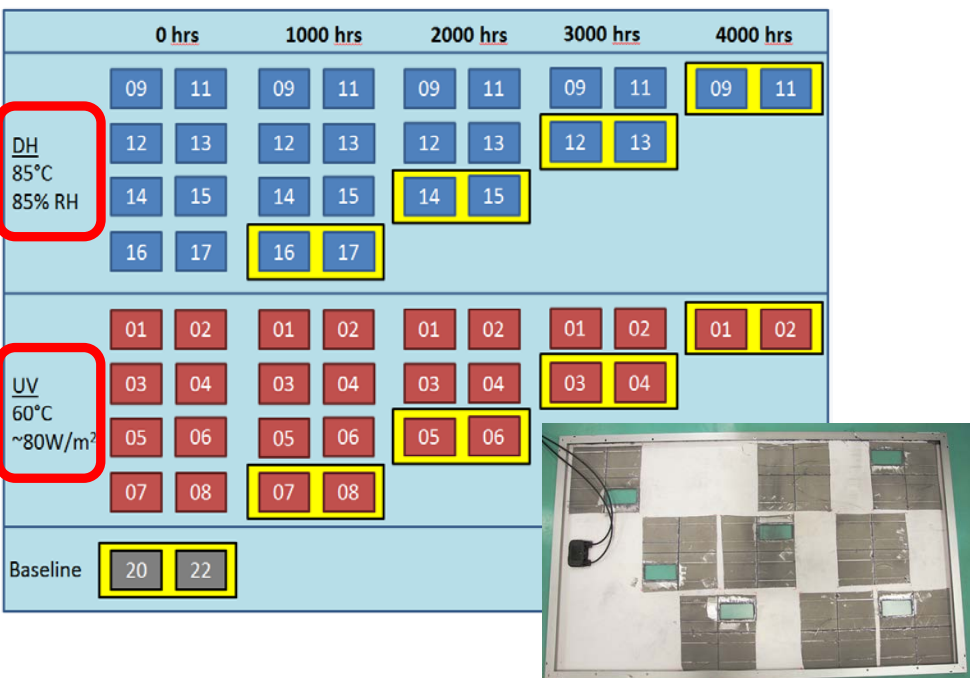
- Damp Heat – 85°C, 85%RH
- UV – 60°C, ~80W/m²

Every 1000 Hours, 2 Modules:

- Destructively Disassembled and Measured

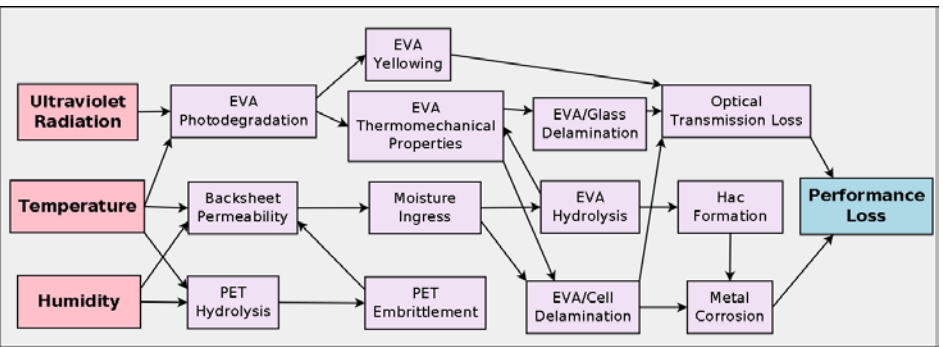
Final Dataset:

- 15 Variables Total
- 2 Different Stress Conditions

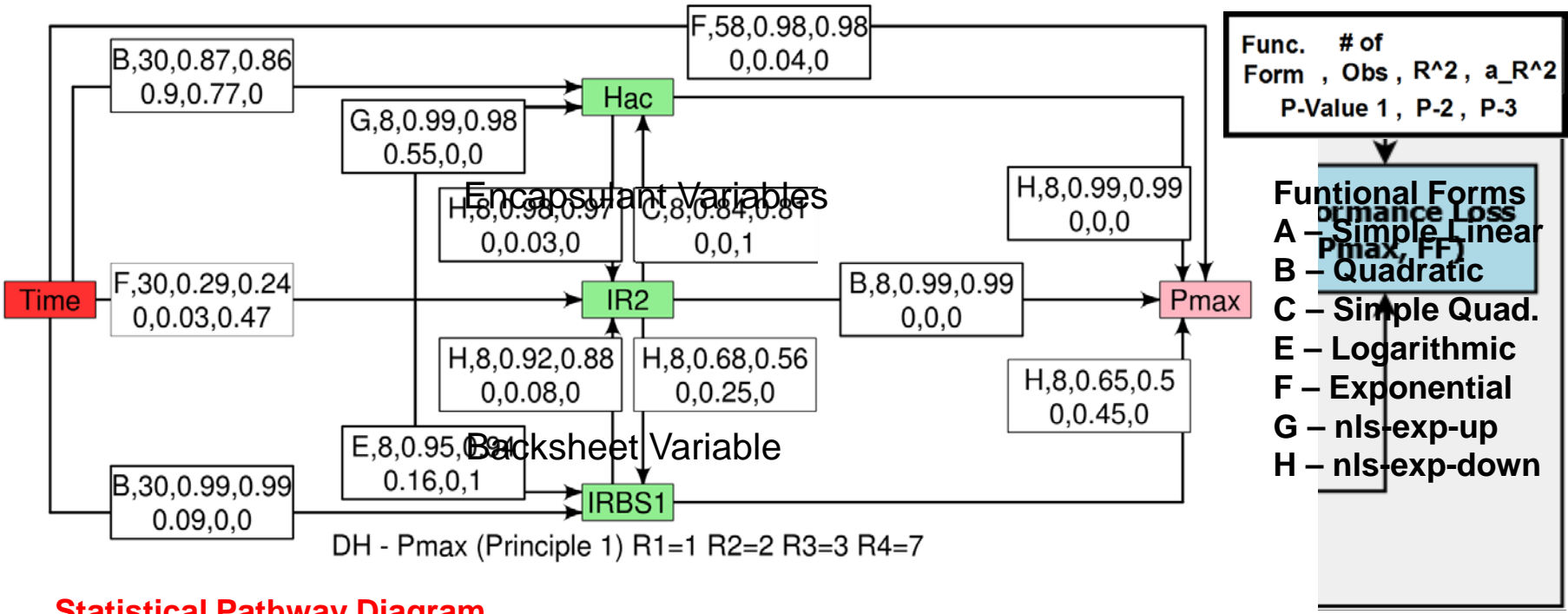
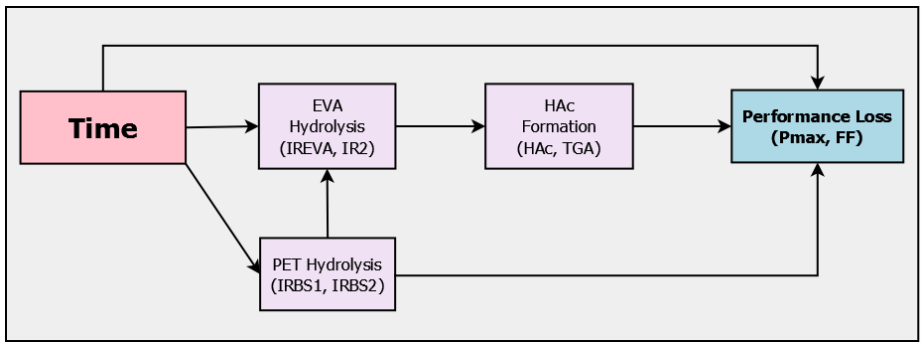


PV Module L&DS Model Development with Domain Semi-gSEM

Domain Pathway Library

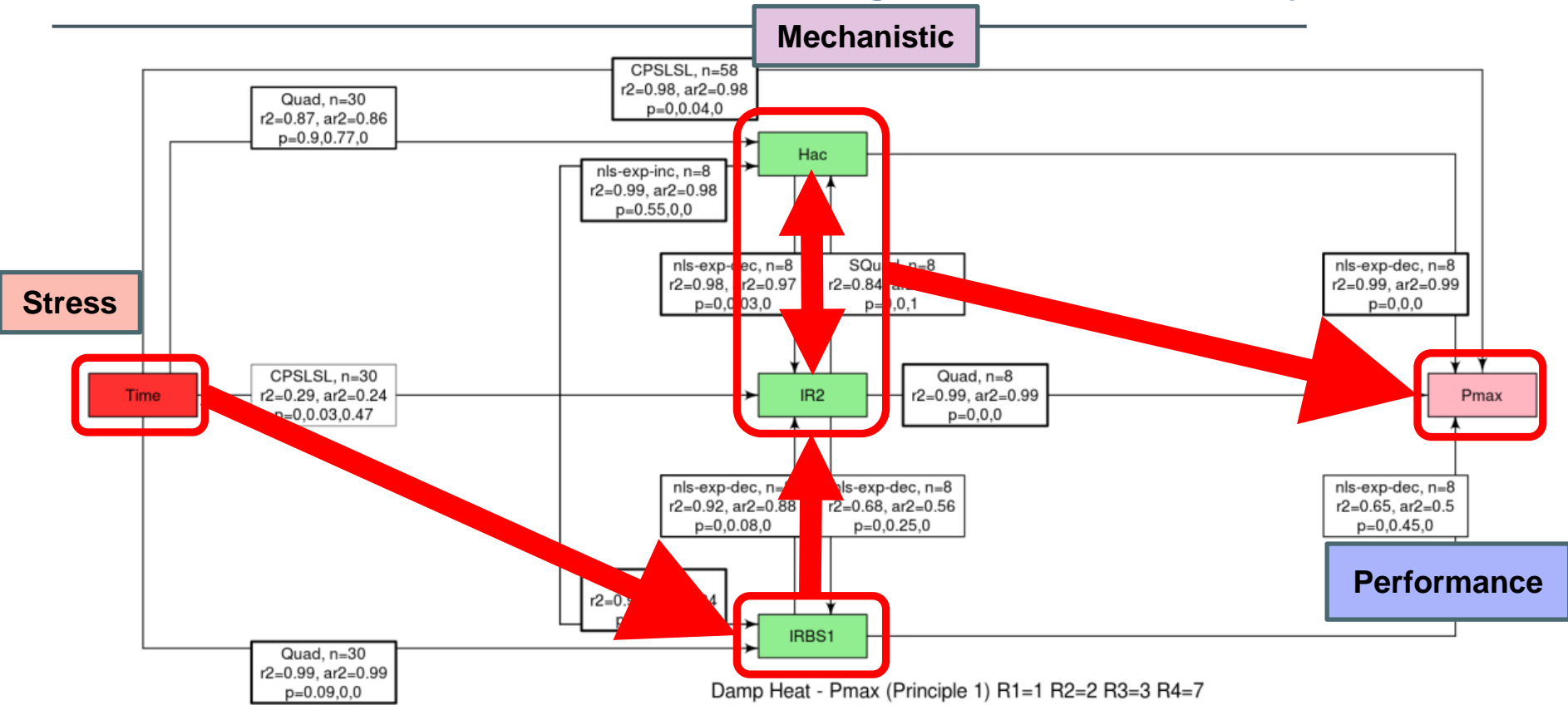


Study Pathway Diagram

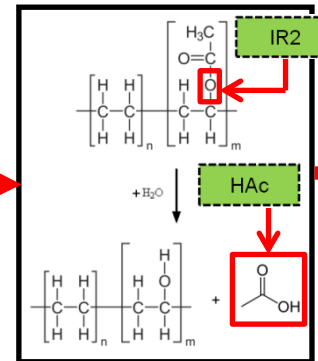
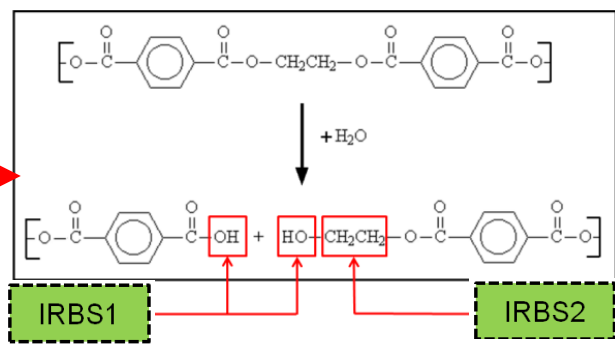


Statistical Pathway Diagram

Damp Heat – Indicated Active Degradation Pathways



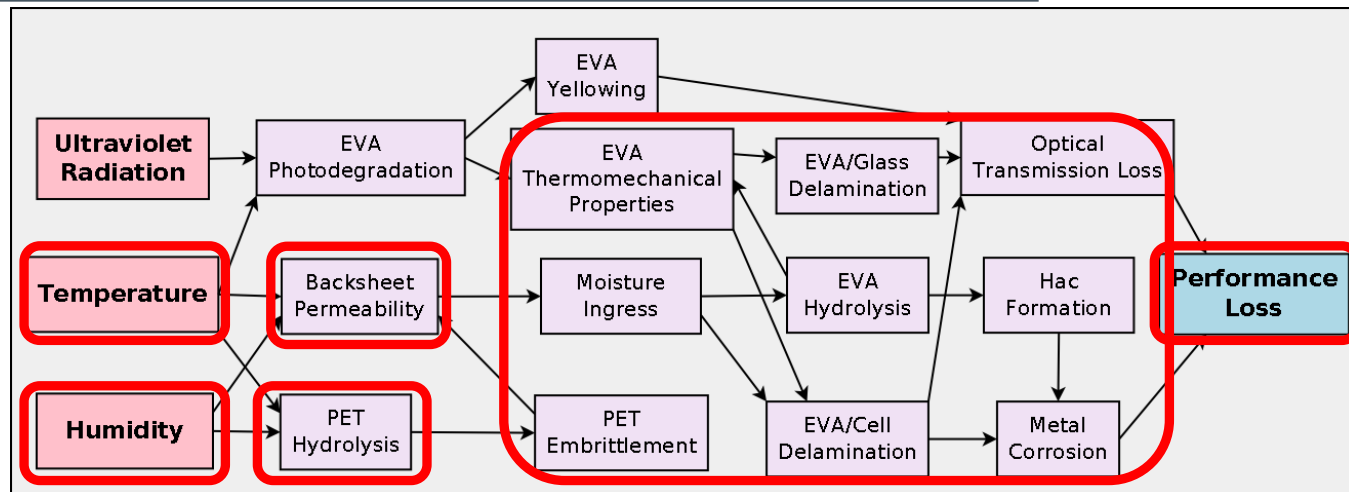
Damp Heat
85°C
85%RH



Performance Loss

Damp Heat Results - Expected vs Observed

Damp Heat
85°C
85%RH

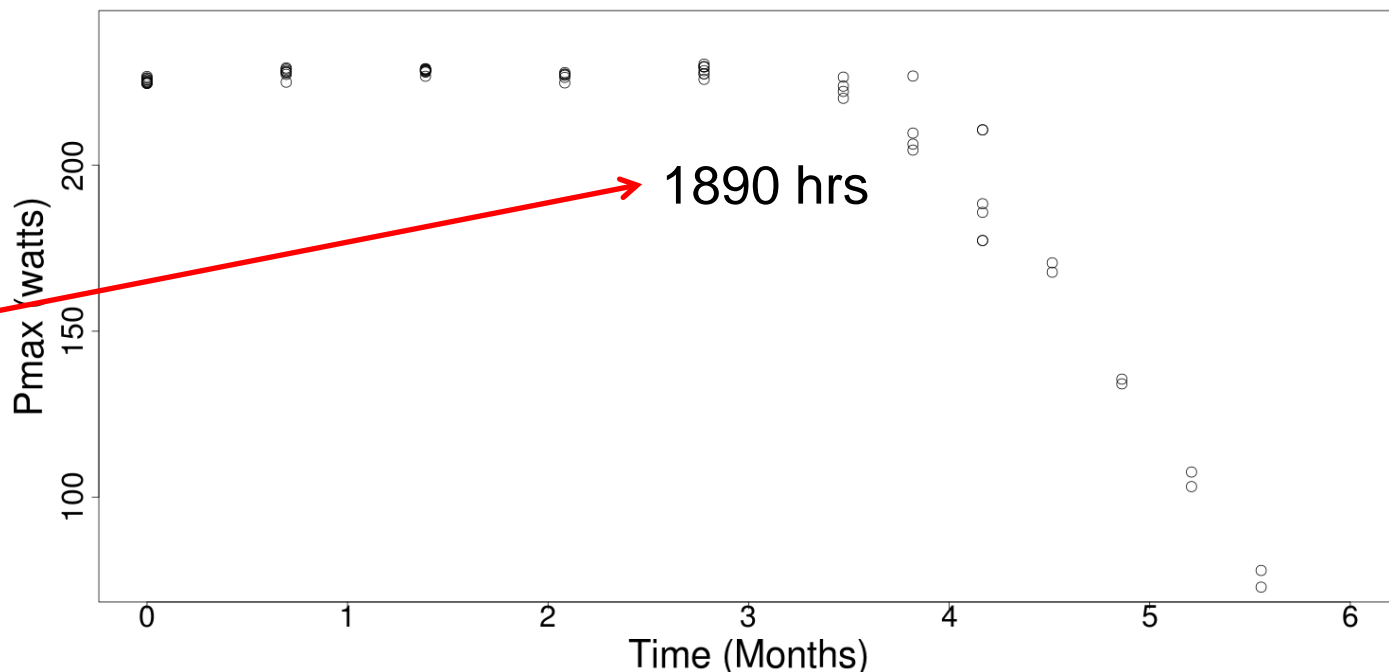


Expected:

- Moisture Related Responses

Observed:

- Dramatic Degradation
- Change Point Evident
- Model Shows Many and Strong Relationships



Case Study of 60 Crystalline Silicon Modules of 20 Brands

Cross-sectional comparison of initial performance



Yang Hu, Mohammad Hossain,
Tim Peshek, Yifan Xu,
Jiayang Sun

Outdoor Test Facilities

SDLE Center SunFarm

- 14 high precision dual axis trackers
- 2 sites adjustable tilt racking
- 148 PV modules from 24 manufacturers
- Over 8000 samples

Power Degradation Rates of c-Si PV

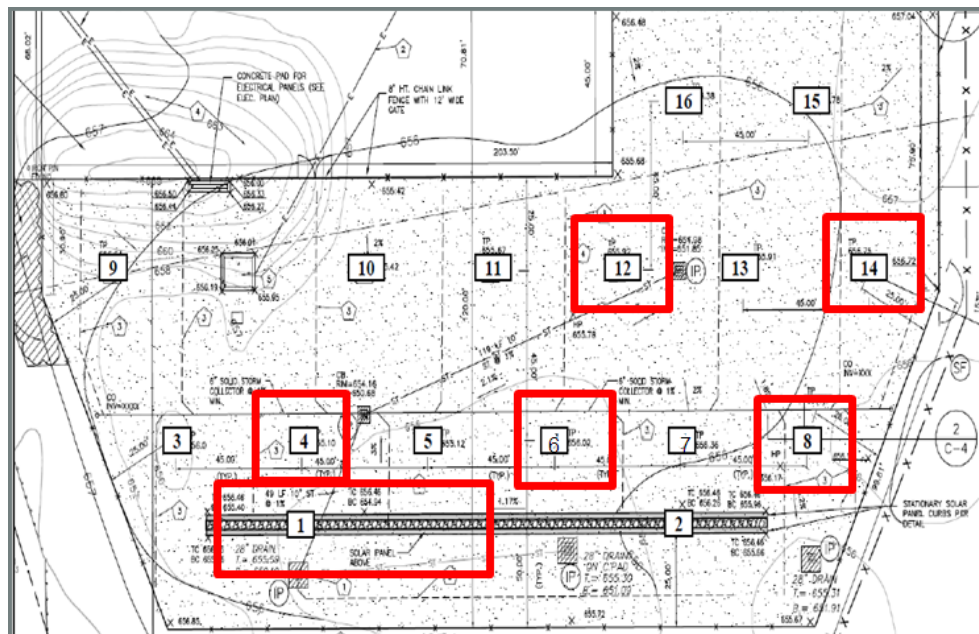
- Using time series analysis
- Across many climatic zones

Lifetime and Degradation Studies of

- Systems, Components, Materials

Data Acquisition

- Minute by minute time series
- Pyranometer: horizontal and Plane of Array
- Weather Stations
- Power Data: Daystar, Enphase, Solectria



Exploratory Data Analysis: Correlation of 60 Modules Over Time

Manufacturer	Model	Power
Astronergy	CHSM 6610P 235 watt	235 watt
AUO	240-P Silver Frame	240 watt
Bosch	c-si M60-225-16	225 watt
Conergy	PowerPlus 220W	220 watt
CSI	CS6P	220watt
EcoSolargy	EC O230S156P-60	230 watt
ET Solar	ET-P660235BZ Zero Rack BLK	235 watt
Helios	6T240	240 watt
Hyundai	HiS-M230MG	230 watt
Kyocera	KD240GX-LPB	240 watt
LG	LG220P1C	220 watt
Mage	Powertec Plus 230/PH	230 watt
MX Solar	MX60 230	230 watt
Perlight	PLM-250M-60 Mono	240 watt
REC	REC230PEBLK	230 watt
Sanyo	HIT-N220A01	220 watt
Schott	POLY230	230 watt
Schuco	MPE240-PS-09	240 watt
Sharp	ND-U235Q1	235 watt
Siliken	SLK60P6L SLV/WHT 235W	235watt
Solar World	SW230Poly 2.0 Frame	230 watt
Trina	TSM-230-PA05	230 watt
UpSolar	UP-M240P	240 watt
Yingli	YL230P-29b	230 watt



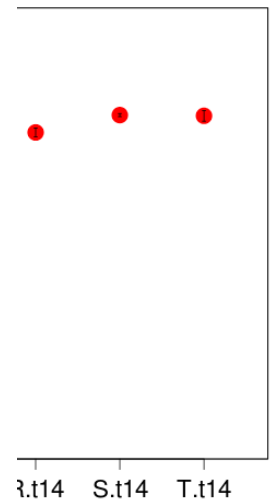
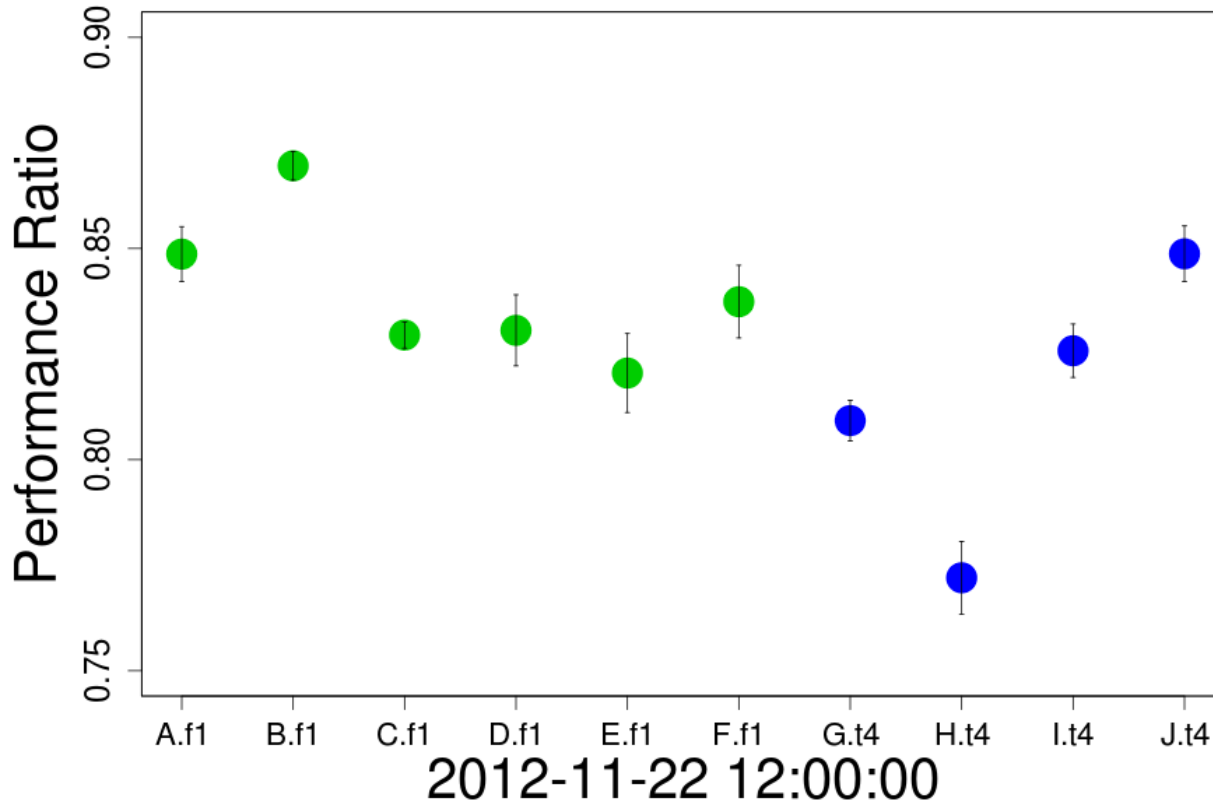
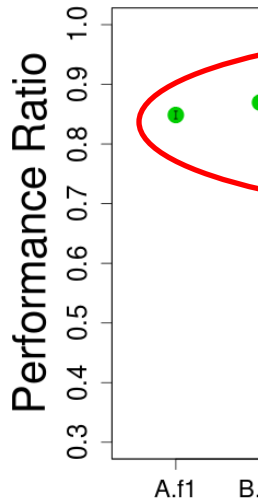
Case Study 2: Cross-sectional comparison of 20 manufacturers

SDLE SunFarm

- Brand names shown as A through T.
- Letter f or t are for fixed tilt rack and tracker.
- The performance ratio of each brand is shown as a dot with a standard error bar.
- The dots show the performance ratio of each brand.
- The standard error bar shows the standard error of the performance ratio.

At noon of a clear sunny day
Performance Ratio (PR)

• —
• where



Similarities/Differences Among “Equivalent” Modules

Time Series Data

- 15 minute Averaged,
- Over 82 Days

Hierarchical clustering of 60 modules

- SubSolNoonMean , Pearson, complete
- Dynamic Time Warping

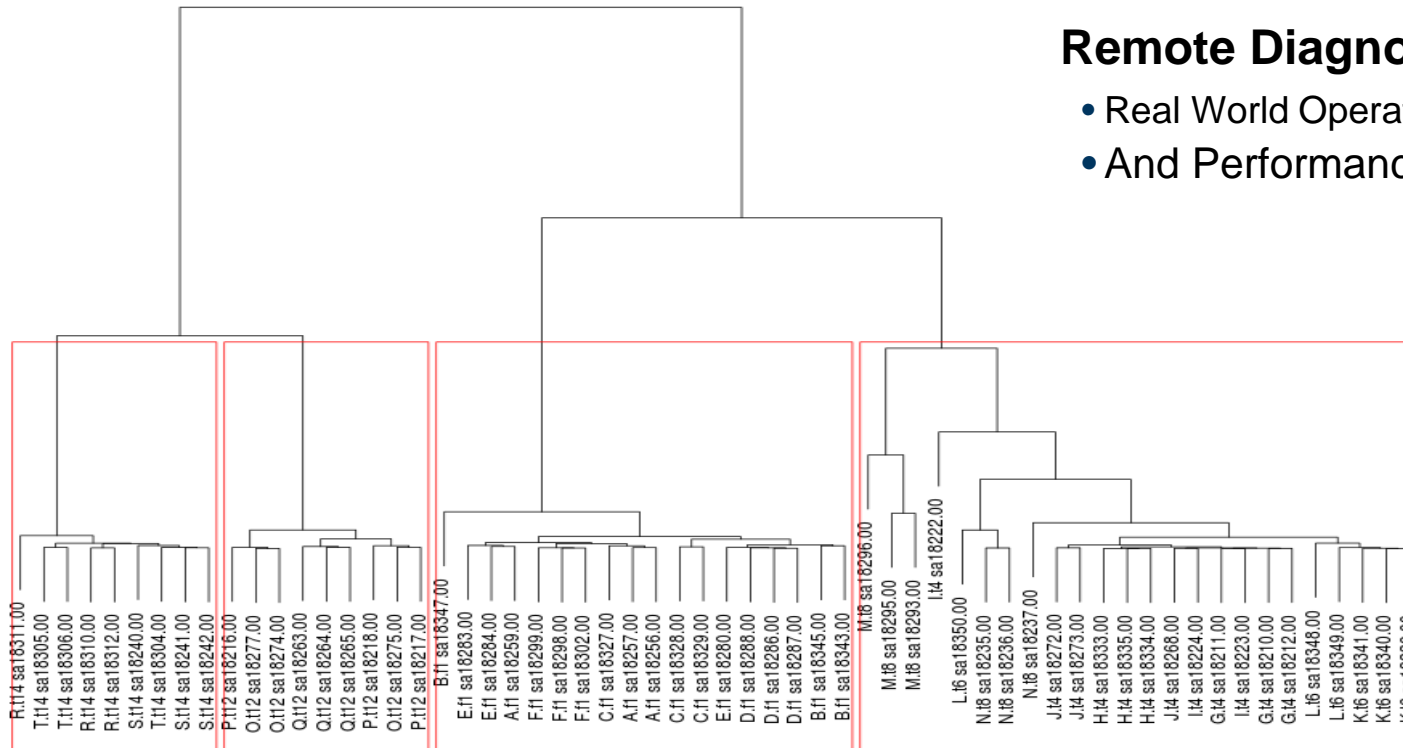
Clustering Identifies

- Fixed Rack vs. Trackers
- Tracker Reliability
- Module Characteristics
- Other Effects

Remote Diagnostics Of

- Real World Operations
- And Performance

Cluster Dendrogram



L&DS: Mesoscale Science Across Length and Time Scales

Applied to Photovoltaic Modules and Solar Mirrors

- Real World and Accelerated Lab Studies

Broadly applicable to

- Multi-generational technologies
- Exposed to harsh environments
- Long lived, High Cost

Combining

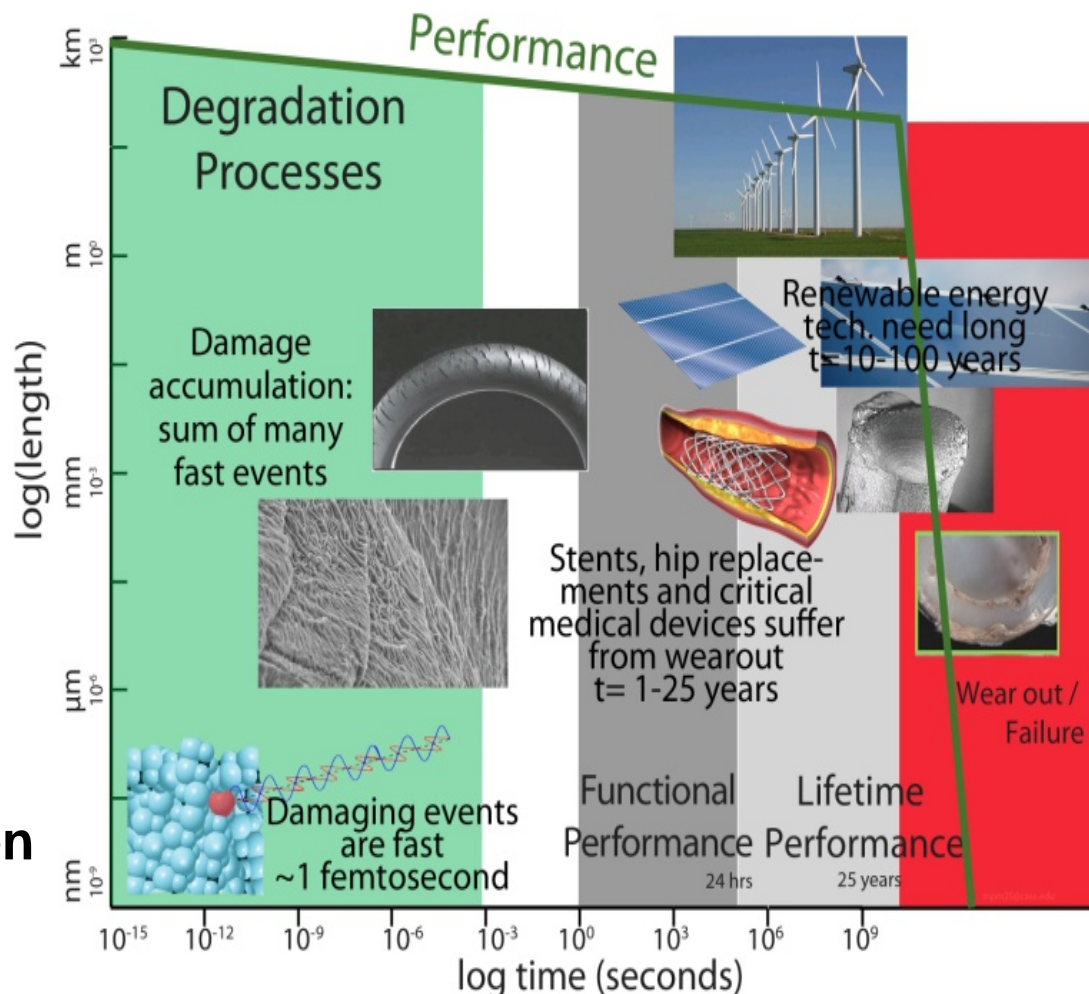
- Population based studies
- Analytics, Prognostics

Data Science Approach

- Data Sources \leftrightarrow Informatics \leftrightarrow
- Analytics \leftrightarrow Actionable Intelligence

For New Technology Insertion

- Optimized Lifetime Performance
- Without Lifetime Penalties





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OF ENGINEERING

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