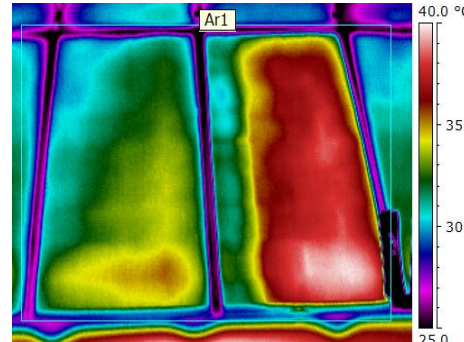


*Exceptional service in the national interest*



# Integration of PV-RPM into the System Advisor Model (SAM)

*8<sup>th</sup> PV Performance Modeling and Monitoring Workshop*  
*May 9, 2017*

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# Acknowledgements

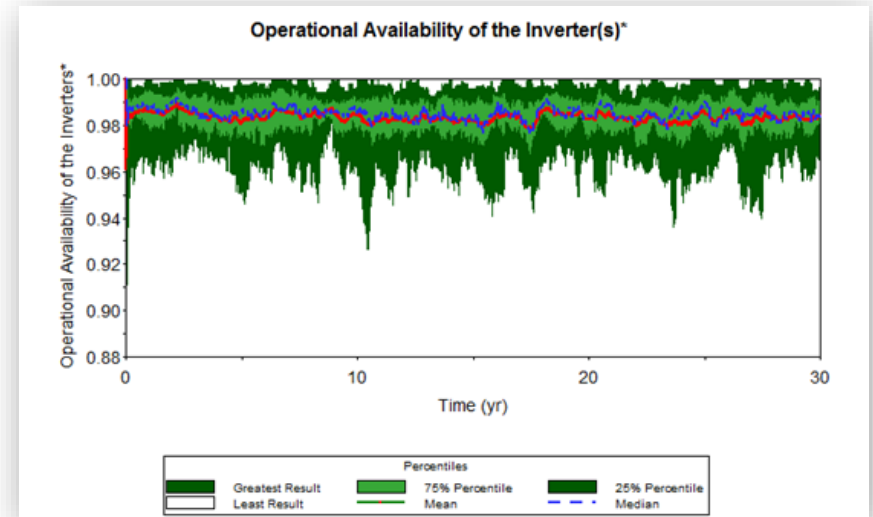
- DOE SunShot Photovoltaics Program and Soft Costs Program (through NREL) for providing funding
- Renee Gooding from SNL for data analysis
- Data partners

# Outline

- What is the PV Reliability Performance Model (PV-RPM)? Why is this important?
- Benefits of integrating into SAM
- Features
- Data Analysis and Results
  - Both older and newer system components
- Next Steps

# What is the PV Reliability Performance Model?

- Developed by SNL in 2010 as a proof-of-concept to evaluate PV performance impacts from probabilistic ‘events’ (faults/failures) impacting modules and inverters



- Goldsim player platform – limited evaluation capabilities (system configuration and failure mode types)

# Why include reliability in a PV performance model?

- PV fleets are aging with different failure modes that vary by manufacturer, age and location, resulting in additional lifetime power and energy production uncertainty

## Current Performance Model

- Energy and power output
- Maintenance/component repair intervals: No energy loss estimate
- Exceedance probabilities for solar resource variability
- Error estimates due to weather uncertainty and model inputs
- Single LCOE

## Performance Model with Reliability Elements

- Probabilistic power and energy production, and losses
- Probabilistic estimates of *when* and *how many* events per component
- Additional uncertainty around component fault/failure impacts
- Probabilistic representation of LCOE

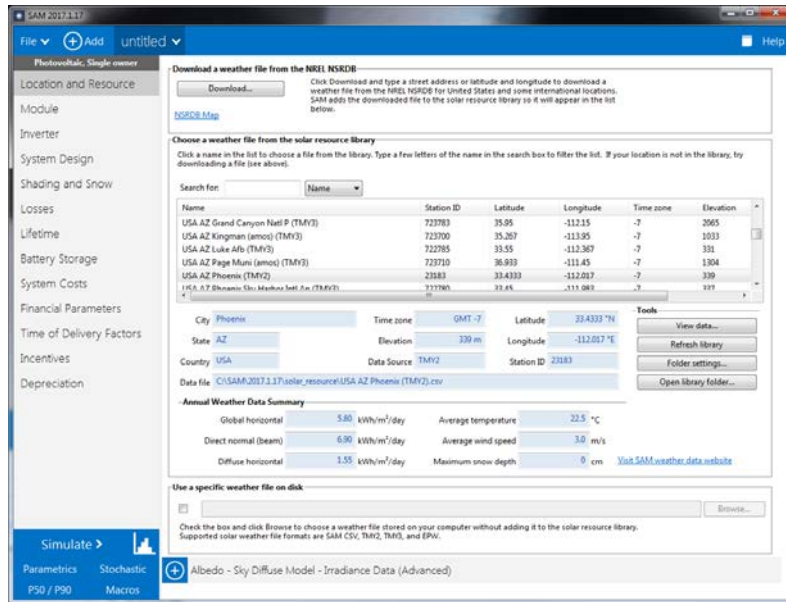
# Benefits of Integrating into SAM

- More flexibility for different system design options
- SAM already has Monte Carlo – Latin Hypercube Sampling capability integrated from SNL Dakota software
- Open-source in LK script will allow for user customization. This could include developing failure modes for batteries, for example

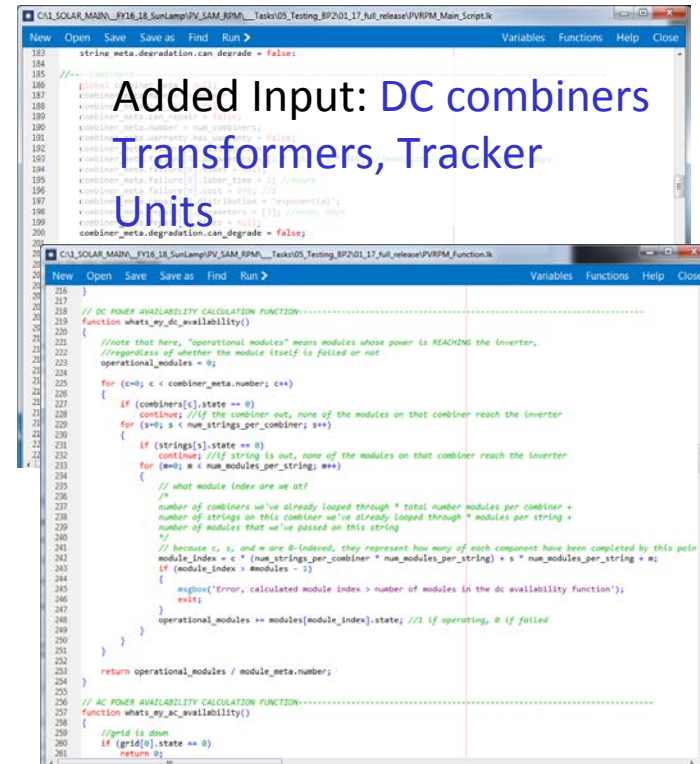
## *...and Challenges*

- Bottom-up model only allows for ‘even’ and similar component configurations. Only 1 inverter type per site and same number of modules per dc combiner, for example
- Simulation time for larger architectures

# SAM Implementation – LK Script



- Main Script  
*Where distributions are defined*
- Function Script  
*Not modified by user*



Added Input: DC combiners  
Transformers, Tracker  
Units

- System design window  
*Same as any other SAM model*

- Components for analysis

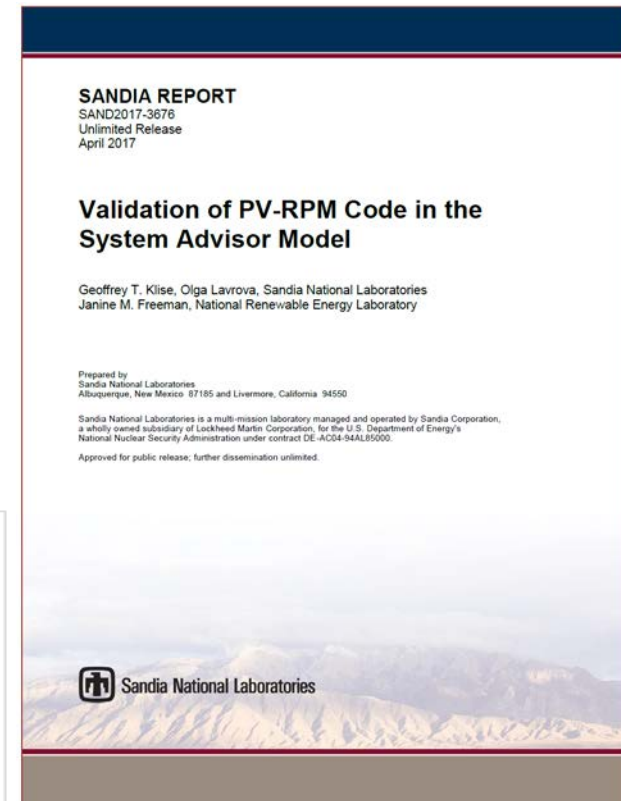
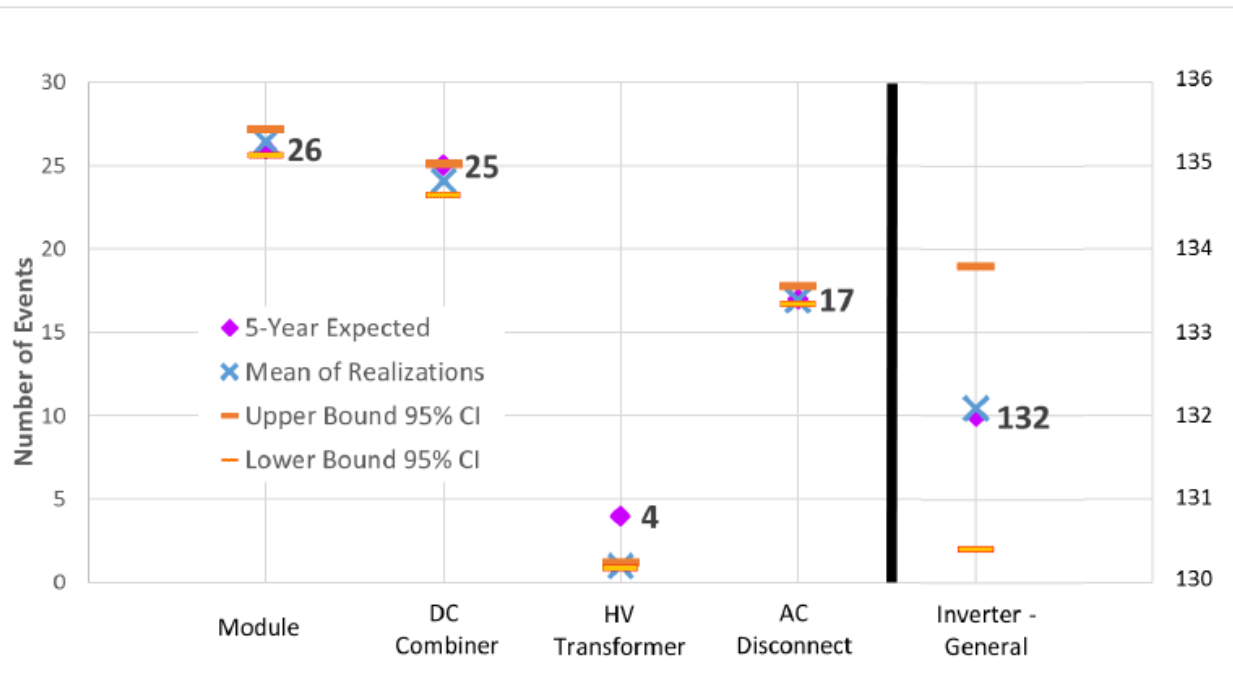
Modules  
Strings  
DC Combiners  
Inverters  
AC Disconnects  
Transformers  
Grid Impacts  
Trackers

- Output File  
*Power & energy loss, costs, labor hours, LCOE, failures per component. Time series and annual results, per realization*

# Event Validation – Existing Dataset

## Details in recently published white paper

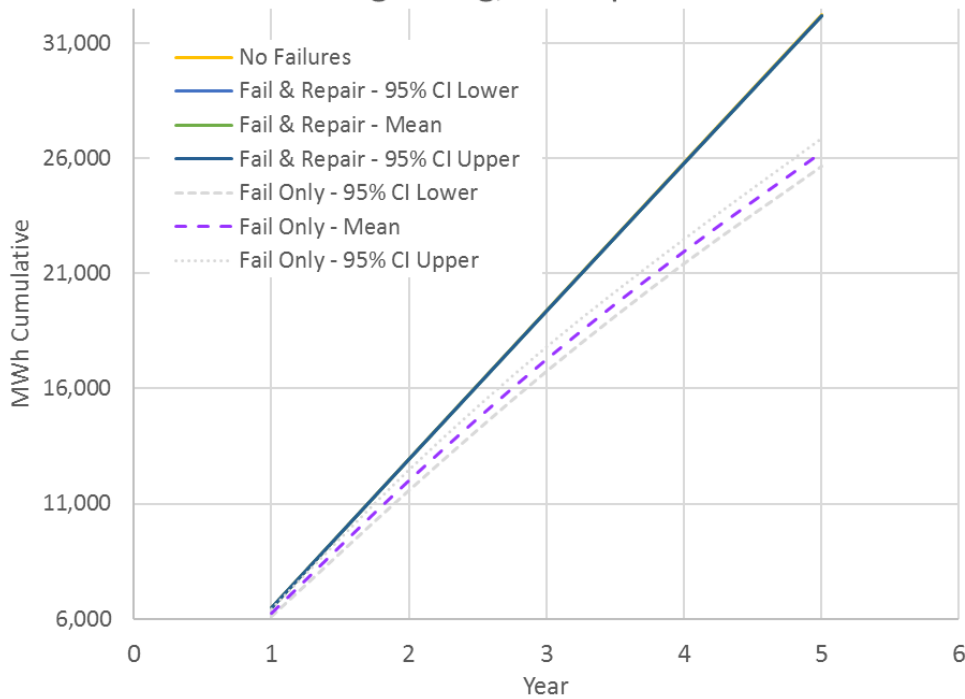
- Validated against proof-of-concept model results using same probability distributions (3.5 MW over 5-years)
- The mean of 100 SAM realizations revealed that for 4 out of 5 components, 95% of the sampled intervals contained the “5-year expected” value for Number of Events (Failures)
- This is not comparing against *actual*, just *expected* to evaluate how well models match





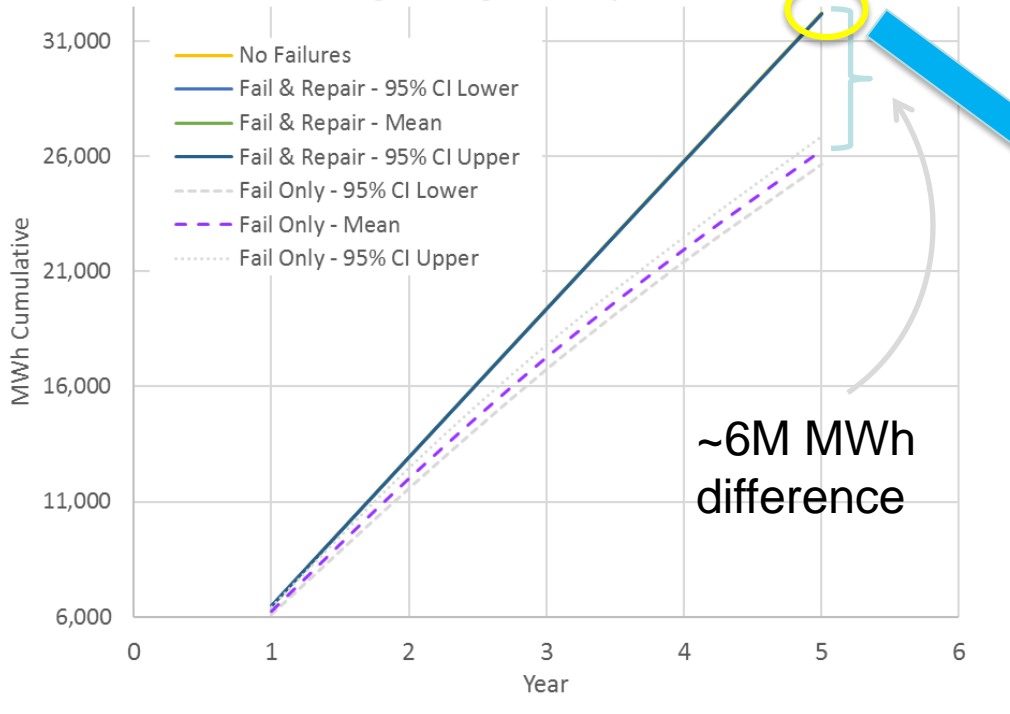
# SAM Analysis – Existing Dataset

5-Yr. Cumulative Energy Production:  
10 Realizations of Inverter Failure due to  
Lightning, No Repair



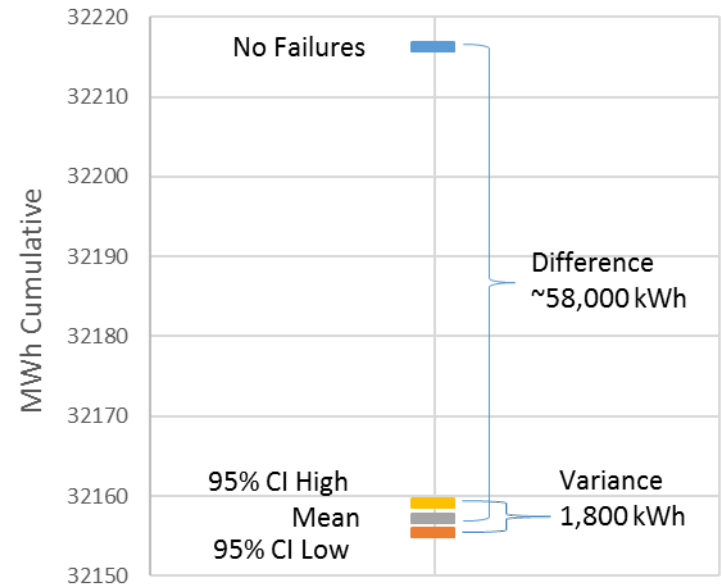
# SAM Analysis – Existing Dataset

5-Yr. Cumulative Energy Production:  
10 Realizations of Inverter Failure due to  
Lightning, No Repair



- Graph at bottom is the energy production range between No Failures & Fail and Repair

Year 5 - No Failures and Lightning Events  
Repaired after Failure



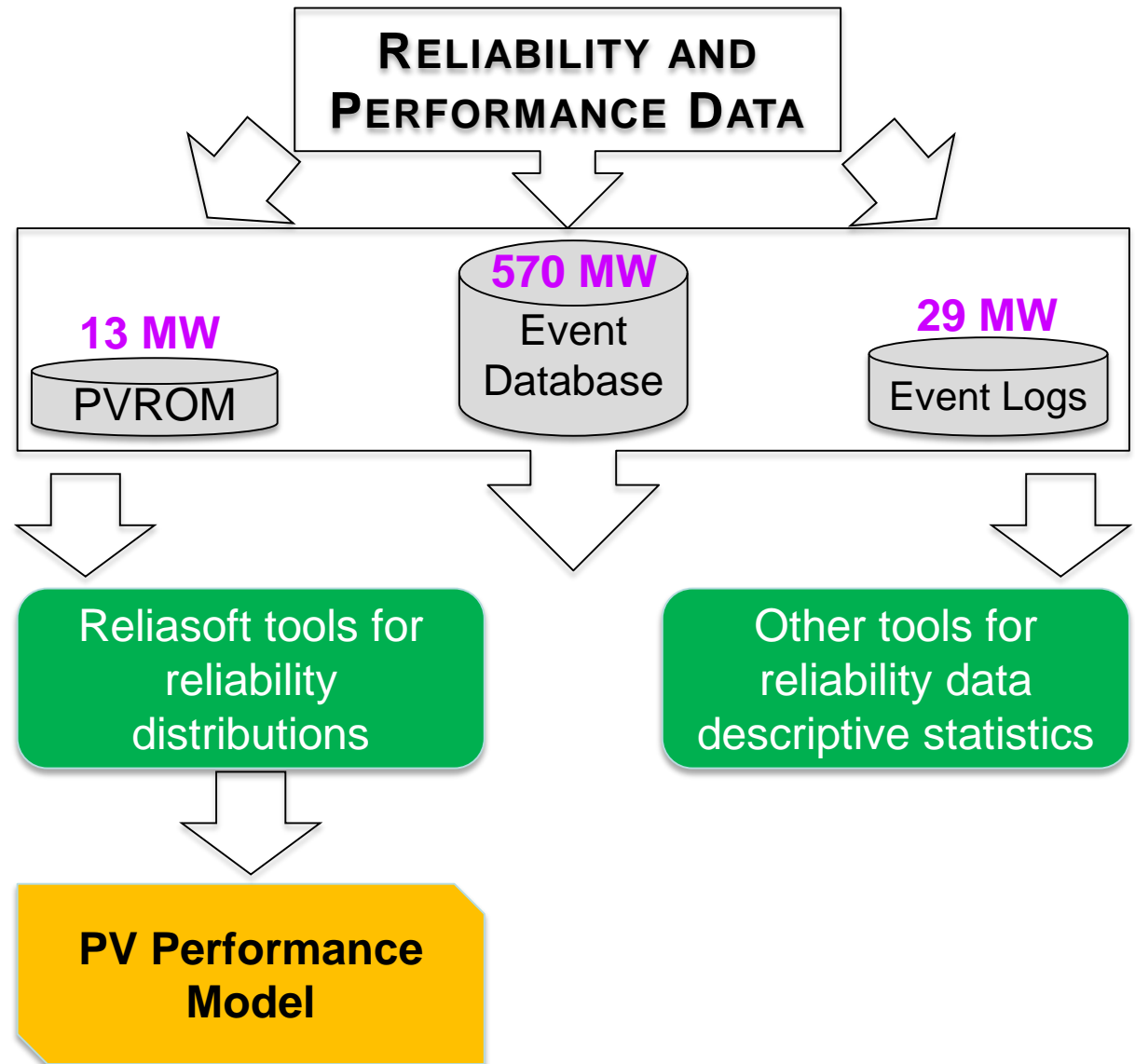
# Event Validation – Newer Dataset

Raw Data  
Generation

Data  
Collection &  
Storage

Data  
Analysis

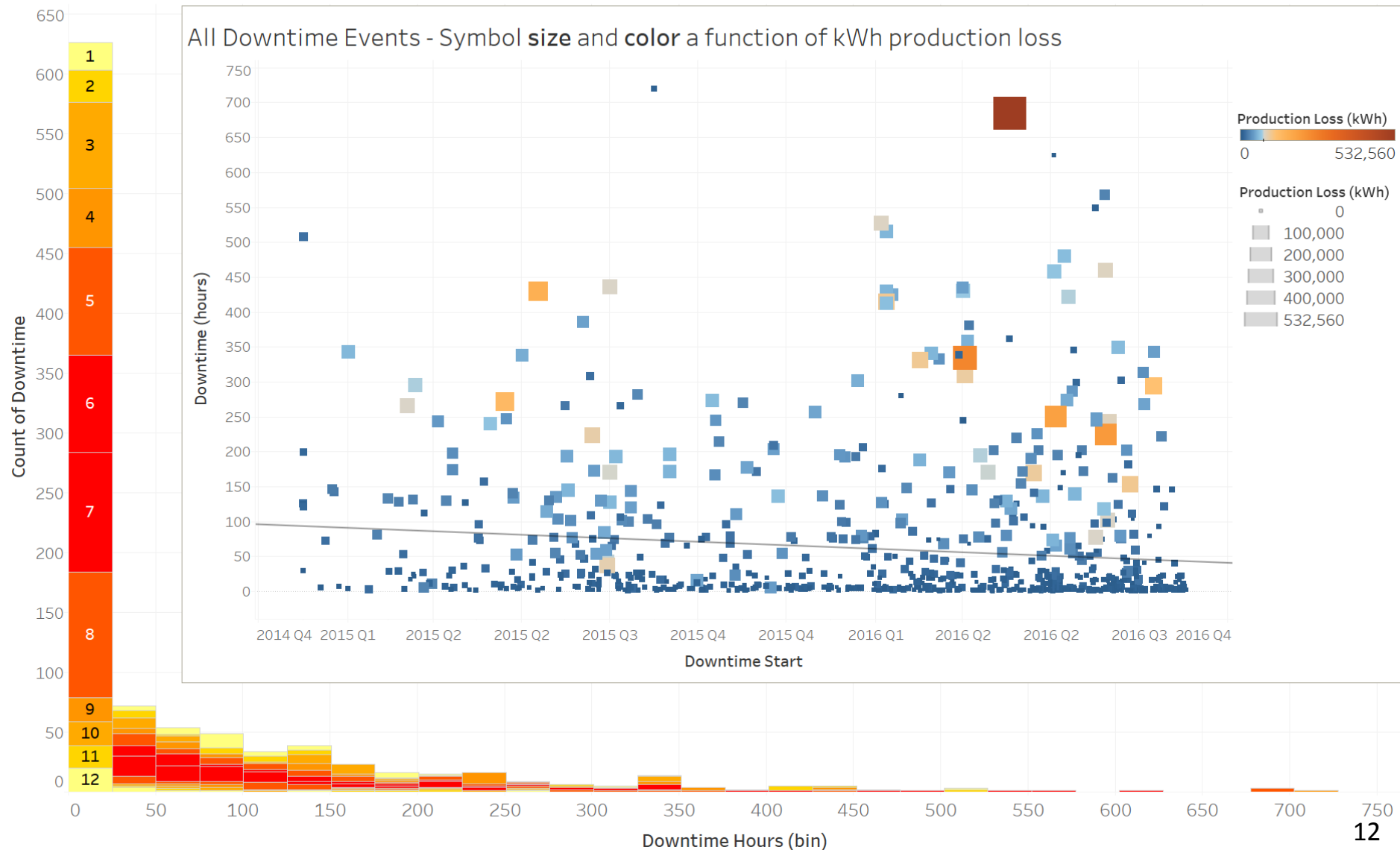
Data Application  
and Further  
Analysis



# Event Validation – Newer Dataset

## All Inverter Downtime - Frequency and Trend

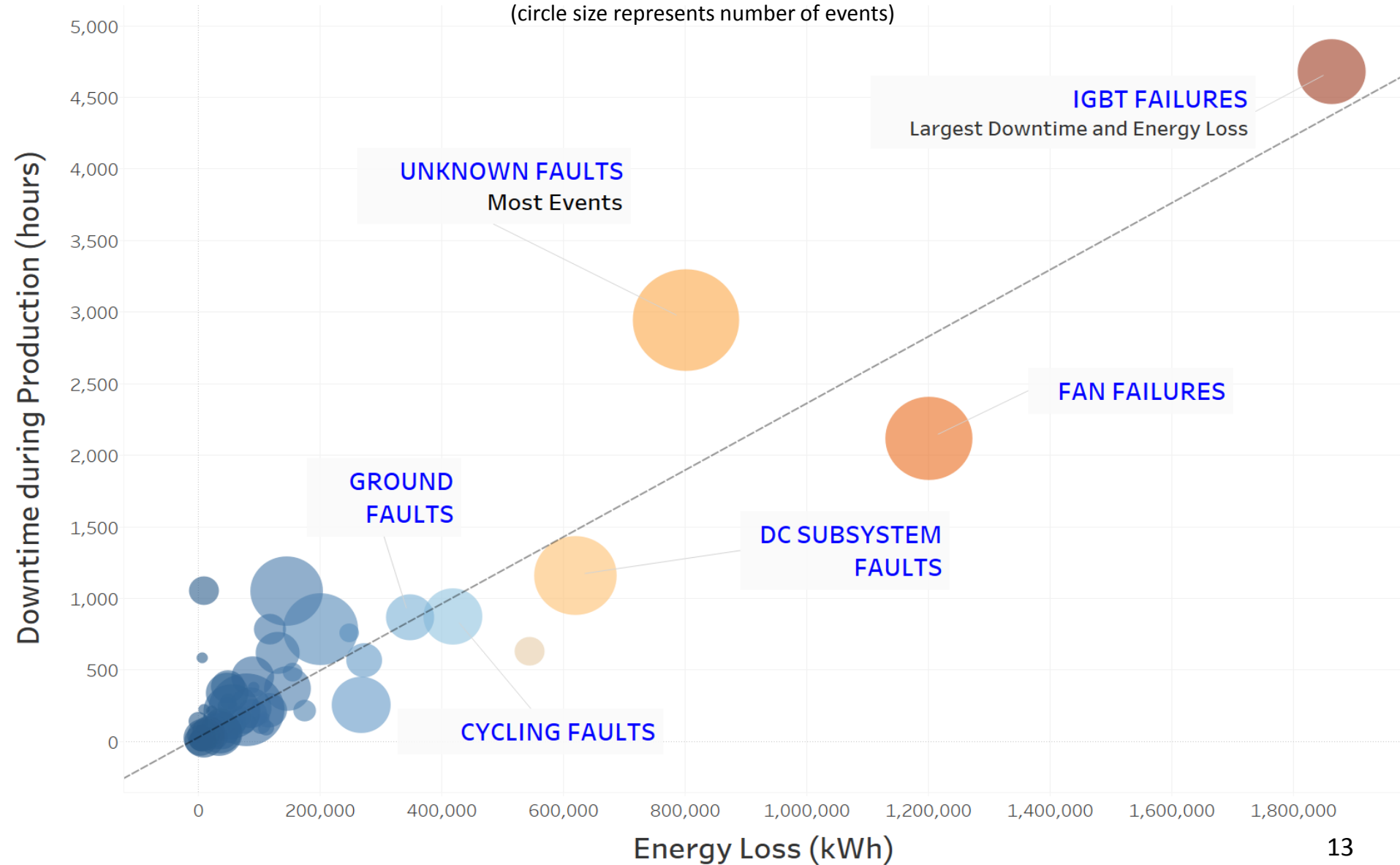
Distribution of Downtime Events - Red colors represent peak production months



# Event Validation – Newer Dataset

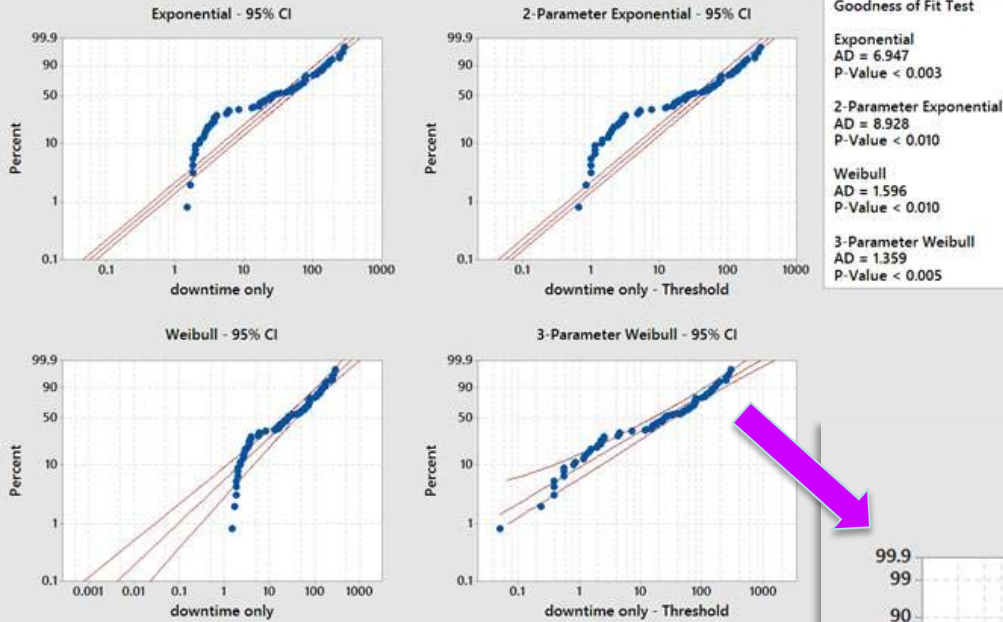
## Inverter Downtime vs. Energy Loss - All Events

(circle size represents number of events)



# Event Validation – Newer Dataset

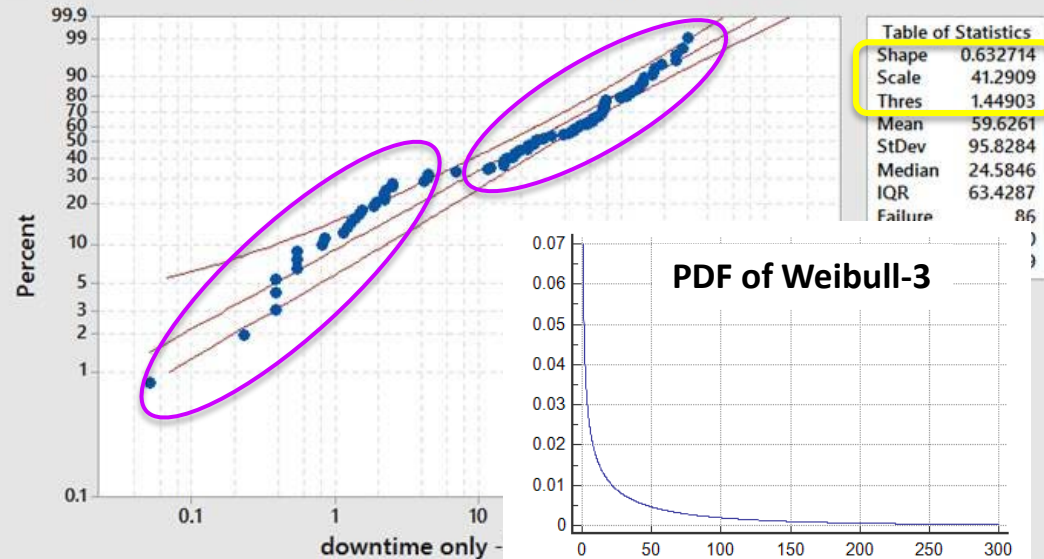
Probability Plot for downtime only



- Example showing ALL fan outage data across all sites
- Data fitting is done to find the distribution with a P-value and AD statistic that explains how well underlying data fits the selected distribution type

- Confidence intervals are narrower around the upper part of the distribution
- In this case, we have what appears to be a bi-modal distribution of data, owing to different downtime durations and likely, different failure modes

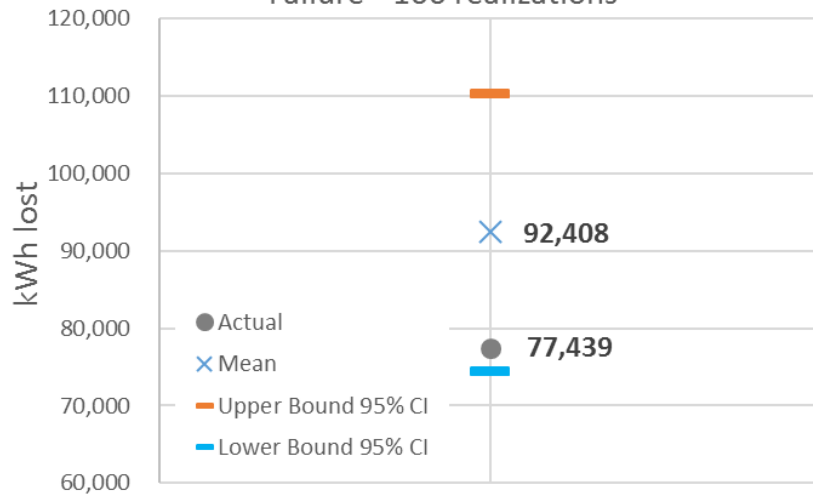
Probability Plot for downtime only  
3-Parameter Weibull - 95% CI  
Complete Data - ML Estimates



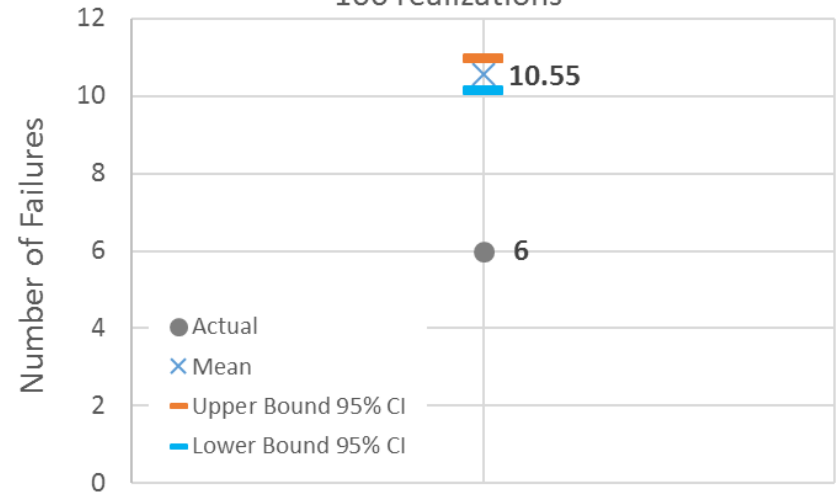
# Failure and Energy Loss Validation

- Inverter dataset – Coolant failures for one 24 MW system with 20 inverters (Impacts to 4)
- 1.2 years for analysis period. 1 year for model (can only model annually)
- 100 realizations
- **Energy Loss:** 95% of the sampled intervals contained the “actual” value for energy loss
- **Failures:** 5% of the sampled intervals did not contain the “actual” value for number of failures
- Failure and repair distributions only had 6 sample points. Greater potential for results with sampled intervals outside of the 95% confidence interval
  - Longer dataset collection period will result in better distribution fit

One-year Energy Loss due to Coolant System Failure - 100 realizations



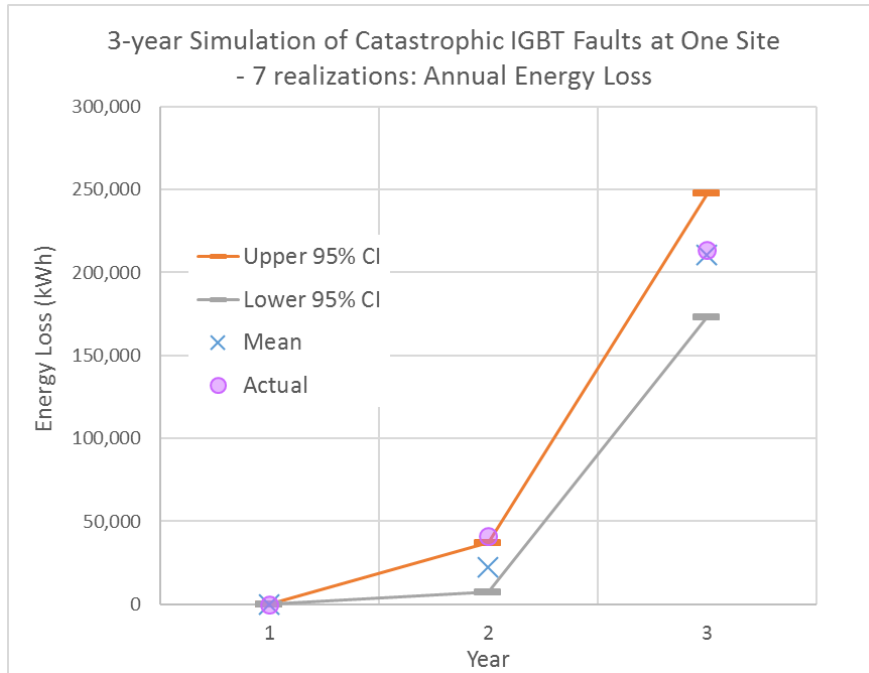
Failures due to Coolant System Failure - 100 realizations



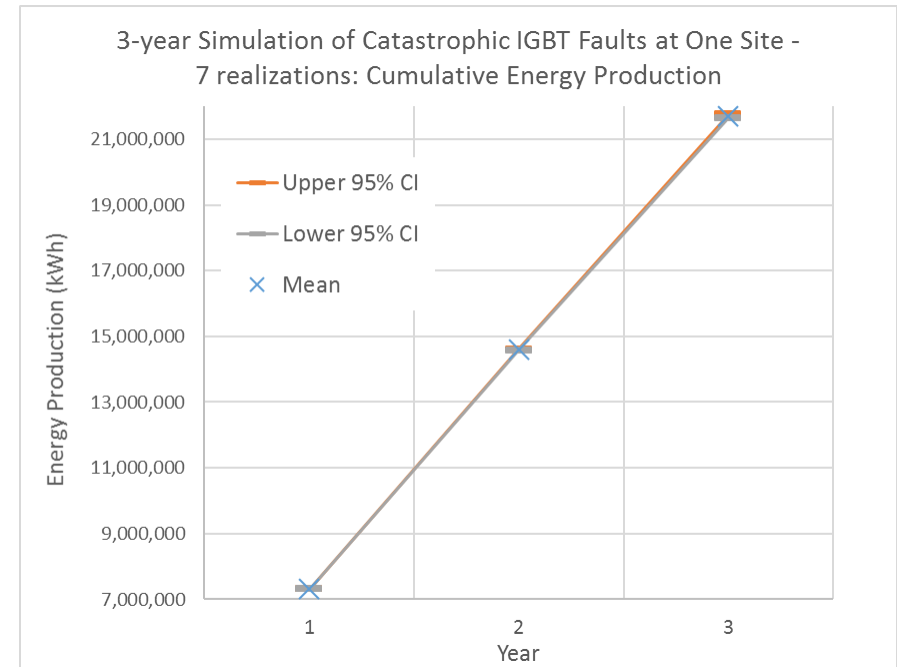
# Failure and Energy Loss Validation

- Inverter dataset – IGBT failures for one 5 MW system with 8-500 kW inverters (impacts to 7)
- 2.8 years for analysis period. 3 years for model (can only model annually)
- 7 realizations
- **Energy Loss:** Year 2 –5% of the sampled intervals did not contain the “actual” value for energy loss. Year 3 – 95% of the sampled intervals did contain the “actual” value for energy loss
- **Failures:** Actual failures = 7, Mean of modeled failures = 8

## Annual Energy Loss (ea. of 3 years)



## Cumulative Energy Production (3 years)



Difficult to see spread in first 3 years

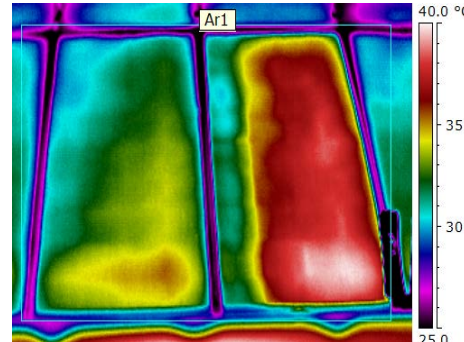
No data to compare “actual” energy production



# Next Steps

- Beta review by industry
  - Looking for additional test users for new reliability feature within SAM. Beta version will be distributed to 33 volunteers late May
- Improving realization speed
- Incorporating beta test feedback
- User manual development
- Use case analysis (FY 18)

*Exceptional service in the national interest*



# Thank You

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