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A Standardized Approach to PV System Performance Model Validation

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**PV Performance Modeling Workshop
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Sandia is a multiprogram laboratory operated by Sandia Corporation, a Lockheed Martin Company,
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Introduction

- **PV performance models are used for prediction of expected energy production for project proposals**
 - Evaluation of different designs (e.g., tracking vs. fixed, module technology, inverter, BOS) and locations.
- **Many performance models available**
 - Klise and Stein (2009) surveys available models
- **Models are based on different conceptual approaches and implementations are not consistent.**
- **Results vary between models run for same system and weather.**



Goals

- **Develop a standard method for validating PV performance models in order to:**
 - Increase confidence and understanding in model results
 - Identify areas for model improvements, gaps in existing data, and sources of modeling error
 - Support consistent, well informed business decisions that will ultimately allow solar technology solutions to prosper.



PV Modeling Steps

- **Read inputs:**
 - Array design (module, string, inverter, mounting, tracking, ground cover, etc.)
 - Weather (irradiance, temperature, wind speed, etc.)
- **Translate irradiance to plane-of-array (POA)**
 - Sun position calculation, irradiance model
- **Evaluate 'effective' irradiance**
 - Angle on incidence effects
 - Spectral effects (air mass correlations or physics models)
- **Determine cell temperature**
- **Calculate I_{mp} , V_{mp} , and P_{mp}**
- **Estimate and apply derates (soiling, DC losses, mismatch, array utilization, etc)**
- **Model inverter performance (P_{ac})**



Model Validation Process

- **Develop data sets including system description, weather data and performance data for multiple technologies, applications, and climates.**
 - Understand and document data uncertainty
- **Provide the system description and weather data to modelers, who will model the system and provide results.**
 - Fully document model parameters and assumptions
- **Apply a unified mathematical/statistical approach for comparing measured and modeled quantities and document comparisons in a standardized reporting format.**
 - Propagate uncertainties, if possible
- **Identify opportunities for model improvement**



Mathematical/Statistical Approach

- **Identify quantities for validation**
 - DC + AC power, POA irradiance, module temperature, etc.
- **Calculate model residuals (Residual = modeled values – measured value)**
 - Calculate summary statistics (R^2 , RMSE, MBE, annual bias, etc.)
 - Plot residuals vs. time
 - Plot distribution of residuals
 - Test correlation between residuals and other variables
- **Residuals from a valid model will be as small as possible and randomly distributed**



Example Application of Validation Approach

- **1 kW DC, m-SI, fixed latitude tilt, photovoltaic system in Albuquerque, NM**
 - 1 year of hourly-averaged weather and performance data collected at site.
 - GHI, DNI, DHI, air temperature, wind speed (multiple instruments)
 - DC (and AC) current and voltage, module temperature
- **Run two performance models in Solar Advisor Model (SAM)**
 - Sandia PV Array Performance Model (SAPM)
 - CEC 5-Parameter Model (Univ. of Wisconsin)
- **Set derate factors to zero**

Sandia's Outdoor Test Facility

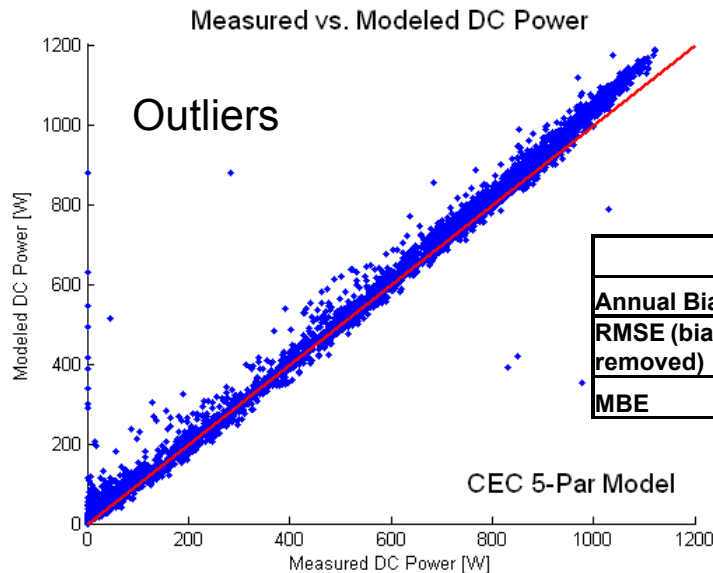
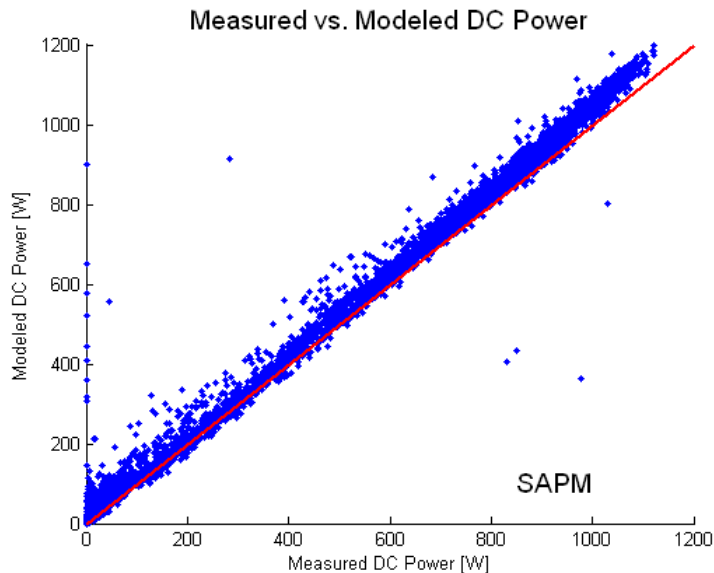


Inverter and DAS Configuration



Comparison of DC Power

- **Measured vs. Modeled looks nearly identical**
- **Slight difference in bias error**
 - Annual bias is same magnitude as typical derate factor
- **Is there a fundamental difference between the models???**

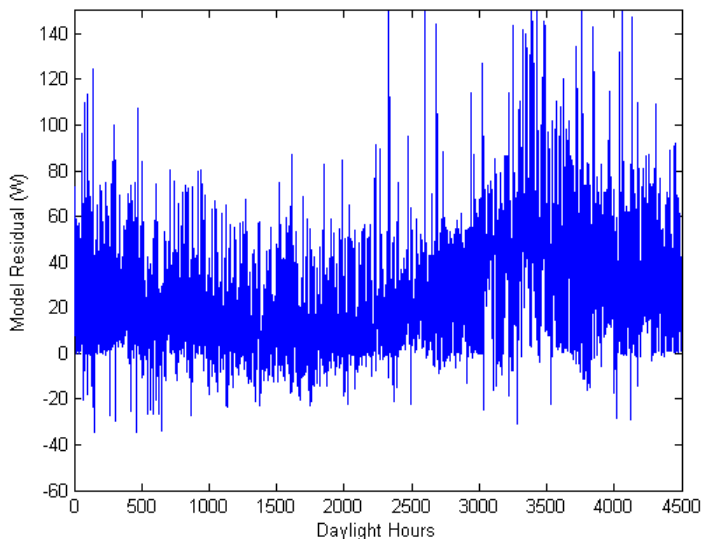


	SAPM	CEC 5 Par
Annual Bias	5.6%	3.3%
RMSE (bias removed)	26 W	23 W
MBE	27 W	16 W

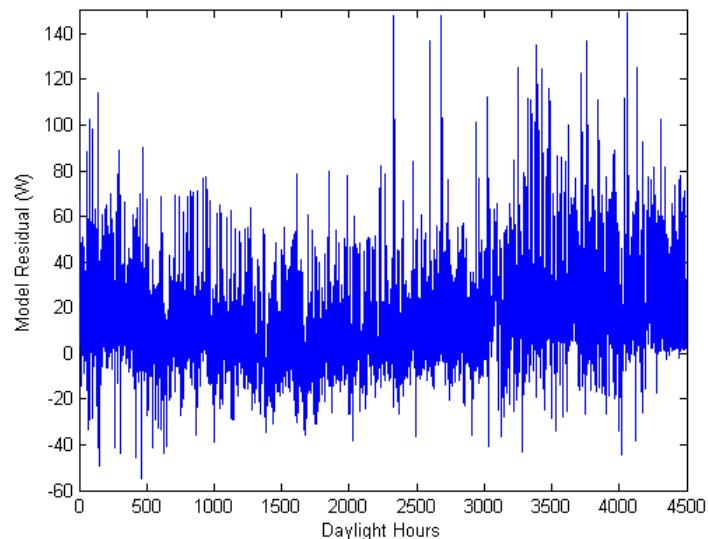
Residual vs. Time

- **Period is from April 2007 to March 2008**
- **Outlier ($-150 < R < 150$ W) and night time data are removed**
 - Outliers due to snow on sensor and array
- **Sustained jumps in residuals may indicate soiling/cleaning cycles**
- **Differences between the model begin to appear.**

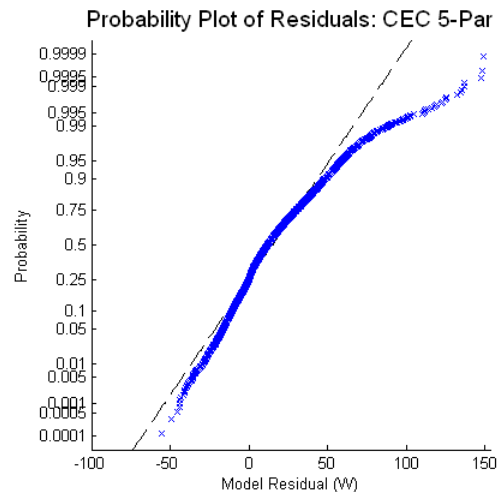
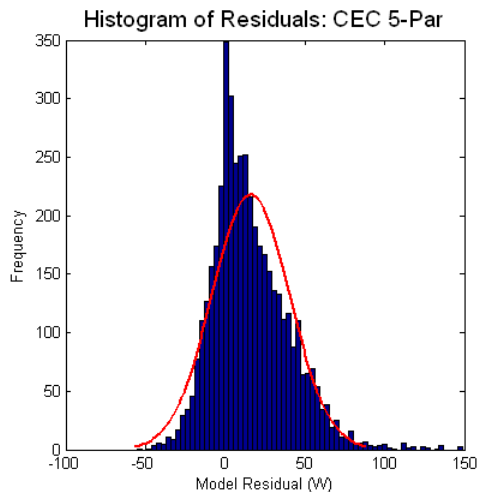
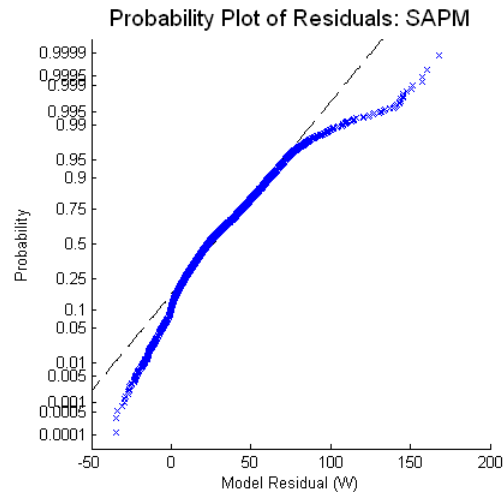
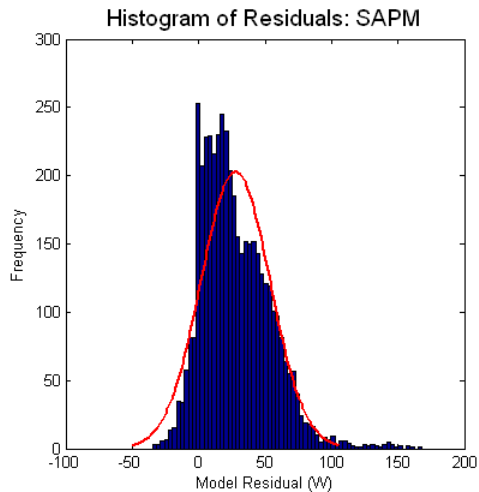
Residual Run Plot: SAPM



Residual Run Plot: CEC 5-Par



Residual Distributions



Both models have residuals that appear quite normal

Slight left skewness due to concentration of near zero residuals and a positive mean residual (no derate)



Residual Correlations

- **Residuals are differences (model – measured)**
- **Residuals from a ‘Perfect’ model will be randomly distributed and uncorrelated with input variables.**
- **Residual analysis identifies any correlations if they exist.**
 - These represent potential ‘flaws’ in the model and/or parameters.
- **Stepwise regression allows variables which affect residuals to be identified and ranked.**

$$Y = b_0 + \sum_{j=1}^P b_j X_j$$

Y = dependent variables

$X = P$ vectors of independent variables

b = linear regression coefficients



Stepwise Results

- Stepwise regression was run for each model
 - Variables examined include incident beam, diffuse, and total radiation, air temperature, wind speed, sun zenith and azimuth angles, angle of incidence, and air mass
- Incremental R^2 value is the fraction of the residual variance explained by the correlation with the variable identified (in order of influence)

SAPM residuals most correlated with air temperature (18% of variance)
CEC 5-Par residuals most correlated with incident beam radiation (12% of variance)

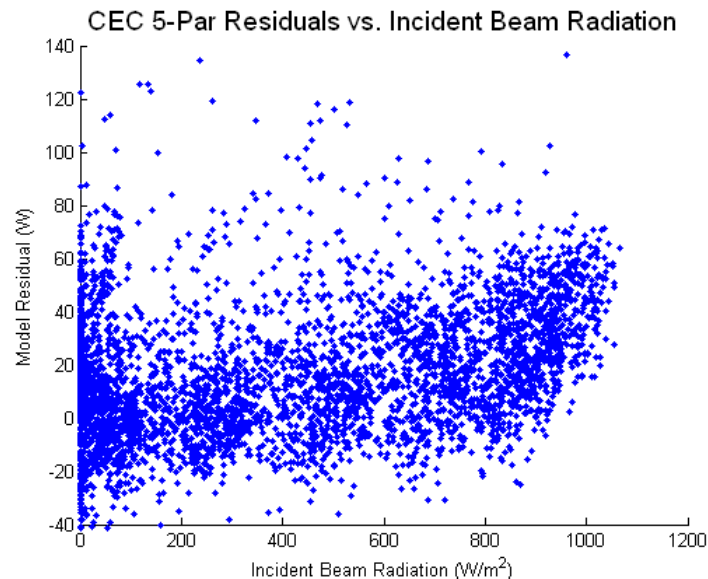
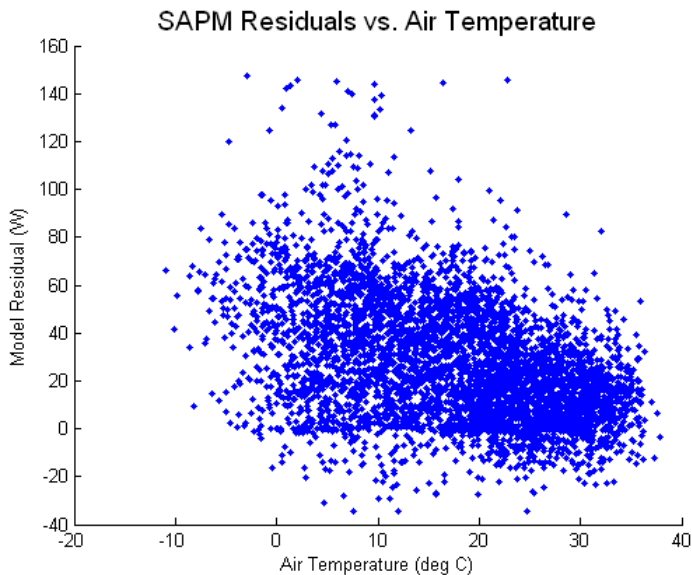
SAPM			
Order	Variable	R^2	Incremental R^2
1	Temp	0.18	0.18
2	Incident Tot	0.35	0.17
3	Azimuth	0.37	0.02
4	Zenith	0.39	0.02
CEC 5-Par			
Order	Variable	R^2	Incremental R^2
1	Incident beam	0.12	0.12
2	Temp	0.22	0.10
3	WS	0.27	0.05
4	Azimuth	0.28	0.01

39% of SAPM variance explained

28% of CEC 5-Par variance explained

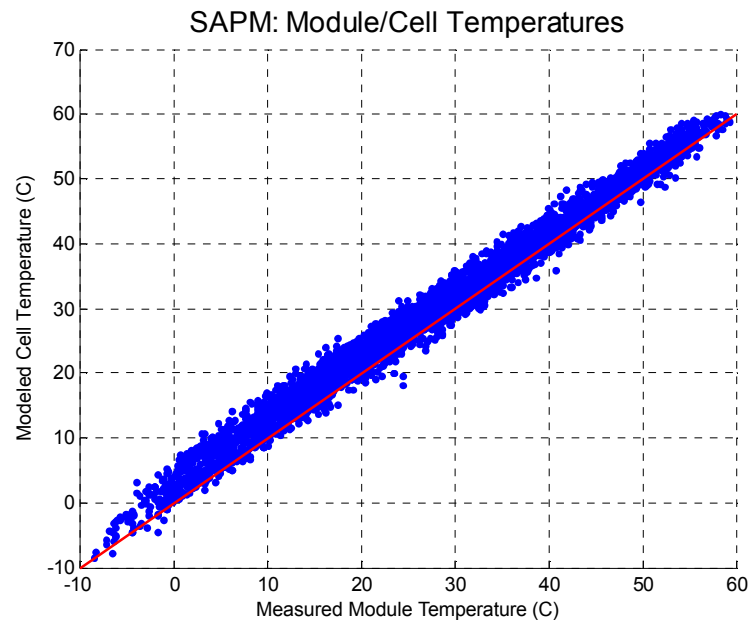
Primary Variable Correlations

- **SAPM residual correlation with air temperature suggests:**
 - Module temperature coefficients need to be adjusted or cell temperature model needs to be improved.
- **CEC 5-Par residual correlation with incident beam radiation**
 - Still investigating this correlation



Module Temperature Model

- Module temperature model appears to work well for this rack-mounted system.
- Module temperature coefficients likely need to be adjusted.





Ongoing Work

- **Collection of performance and weather data from more systems is needed.**
 - **Selection of different technologies**
 - **Diverse locations**
 - **Multiple configurations**
- **Side-by-side comparisons are important because weather data is similar and measurement accuracy is consistent across systems.**
- **Sandia National Laboratories will publish reference data sets for validation.**
- **Sponsor workshop this fall/winter on PV performance modeling**
 - **Participants simulate a reference system**
 - **Comparison of results from various models**



Summary

- **A standardized model validation approach has been developed with input from industry partners.**
 - Based on residual analysis
 - Provides valuable information for model developers
- **Provided an example application of the approach**
- **Next steps include:**
 - collection of data from a representative range of technologies, climates, and designs
 - Model validation report (template?)
- **PV modeling workshop being planned for end of 2010.**



THANK YOU