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Performance Model Assessment for Multi-Junction Concentrating Photovoltaic Systems

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Sandia is a multiprogram laboratory operated by Sandia Corporation, a Lockheed Martin Company,
for the United States Department of Energy's National Nuclear Security Administration
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Relevance of Annual Energy Calculations

- PV Systems Increasingly Valued for Annual Energy Output
 - Power Purchase Agreements Common
 - Flat-Plate PV System Power Ratings Do Not Accurately Indicate Performance
 - Comparison with CPV System Ratings is Difficult
- System Performance Models are an Alternative to System Ratings
 - Provide Ability to Predict Annual Energy Production for Various Locations
- DOE/NREL Has Developed the Solar Advisor Model (SAM)
 - Key DOE Metrics are Annual Yield and Levelized Cost of Energy
 - Performance Models in SAM Include the Sandia PV Array Performance Model, the Sandia Inverter Model, and a Simpler CPV Model
 - Available Free at www.nrel.gov/analysis/sam



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Model Validation Approach

- Through laboratory or field testing, develop performance coefficients for components or system
- Concurrently collect irradiance, weather, and system performance data over a period of time, preferably one year
- Model system performance using performance coefficients and measured irradiance and weather data
- Apply system derate parameters
 - Sometimes calculated, but usually estimated from experience
- Compare modeled to measured performance
 - As a function of environmental parameters
 - Irradiance, air temperature, wind speed, air mass
 - Also examine internal parameters, when available
 - For example, V_{dc} , module temperature...



Models Evaluated

- Solar Advisor Model
 - Sandia Array Performance Model
 - Empirical Model – Inputs Include Irradiance, Air Temperature, Wind Speed, and Air Mass
 - Simple CPV Model
 - Applies Maximum Power Temperature Coefficient to Efficiency
 - Uses Sandia Model to Calculate Cell Temperature
 - Simple CPV Model Without Temperature Corrections
 - Sun-Hour Model (Eff • DNI)
 - Sandia Inverter Model
 - AC and DC Derate Factors
- ASTM E 2527 – 06 Translation Equation
 - $P = E \cdot (a_1 + a_2 \cdot E + a_3 \cdot T_a + a_4 \cdot v)$



Generation of Performance Coefficients

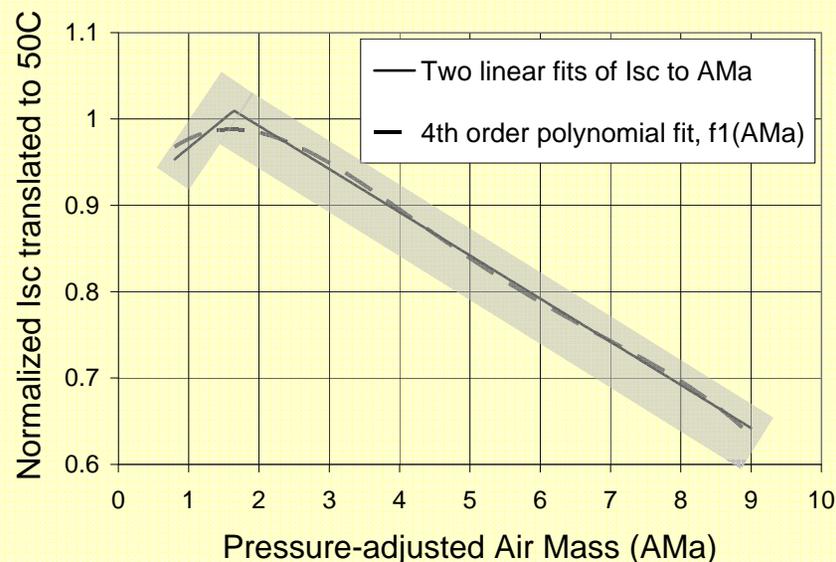
Sandia Array Performance Model – DC

- Data is Collected During All-Day Testing and a Temperature Test
- Regression Analysis Generates

- V_{OC} , I_{SC} , V_{mp} , I_{mp} and Temperature Coefficients for Each
- Additional Coefficients that Determine V_{OC} , I_{SC} , I_{mp} and V_{mp} as a Function of Cell Temperature and Effective Irradiance
- a , b , and ΔT Coefficients Relating Cell Temperature to Ambient Temperature, Wind Speed, and Irradiance

$$T_c = T_a + E \cdot \{e^{a+b \cdot WS}\} + E/E_o \cdot \Delta T$$

- An Air Mass Function That Determines Effective Irradiance:
 $DNI \cdot f(AM_a)$





Generation of Performance Coefficients

SAM Simple CPV Model – DC

- Applies Maximum Power Temperature Coefficient
 - $P = P_o \cdot E/E_o \cdot (1-\gamma) \cdot (T_a - T_c)$
- Uses Sandia Model Cell Temperature Calculation with Typical Values of Coefficients
 - $a = -3.2, b = -0.09, \Delta T = 17$

Sandia Inverter Model

- Coefficients are Derived from California Energy Commission Test Data that Relate Efficiency to DC Input Power and Voltage

AC and DC Derate Factors

- Set to Match Measured and Modeled Power Over Total Measurement Period, ~30 days in fall 2009

ASTM E 2527 – 06 Translation Equation – AC

- Coefficients Generated by Regression Analysis over Few Clear Days



Residual Analysis

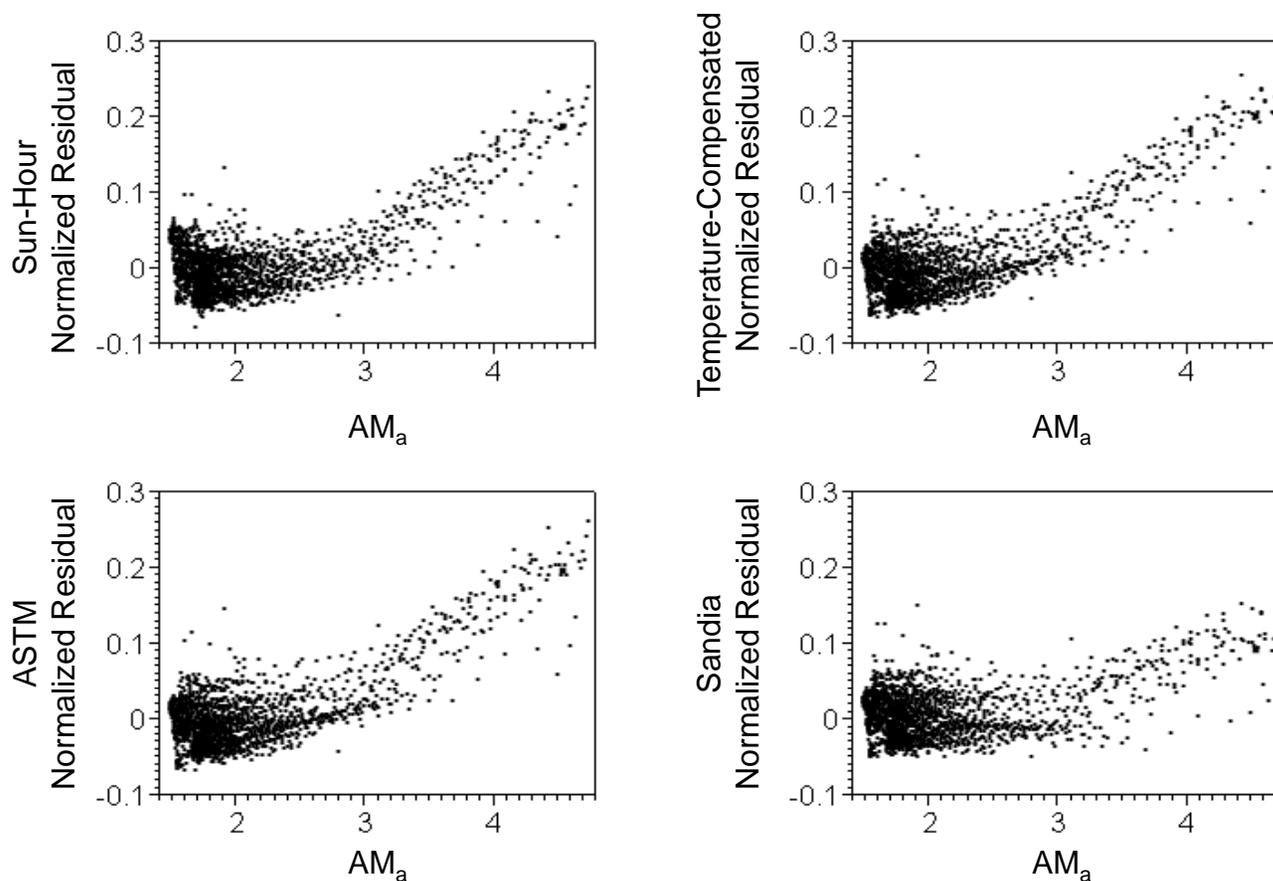
- Normalized residuals are the difference between modeled (bias removed) and measured results
 - $(P_{\text{mod}} - P_{\text{meas}}) / P_{\text{rating}}$ (W/W_p)
 - Since the systems are different sizes, we normalized to the nominal rating
- Model results were “derated” to remove bias error
- Standardized deviation of normalized residuals show all models work well over the short duration of the test

Standardized deviation of normalized residuals

	Sun-hour	Temp Comp	ASTM	SNL
System 1	0.0458	0.0508	0.0499	0.0343
System 2	0.0516	0.0538	0.0502	0.0476



Normalized Residuals vs. Air Mass



For a valid model, residuals should be randomly distributed



Residual Analysis

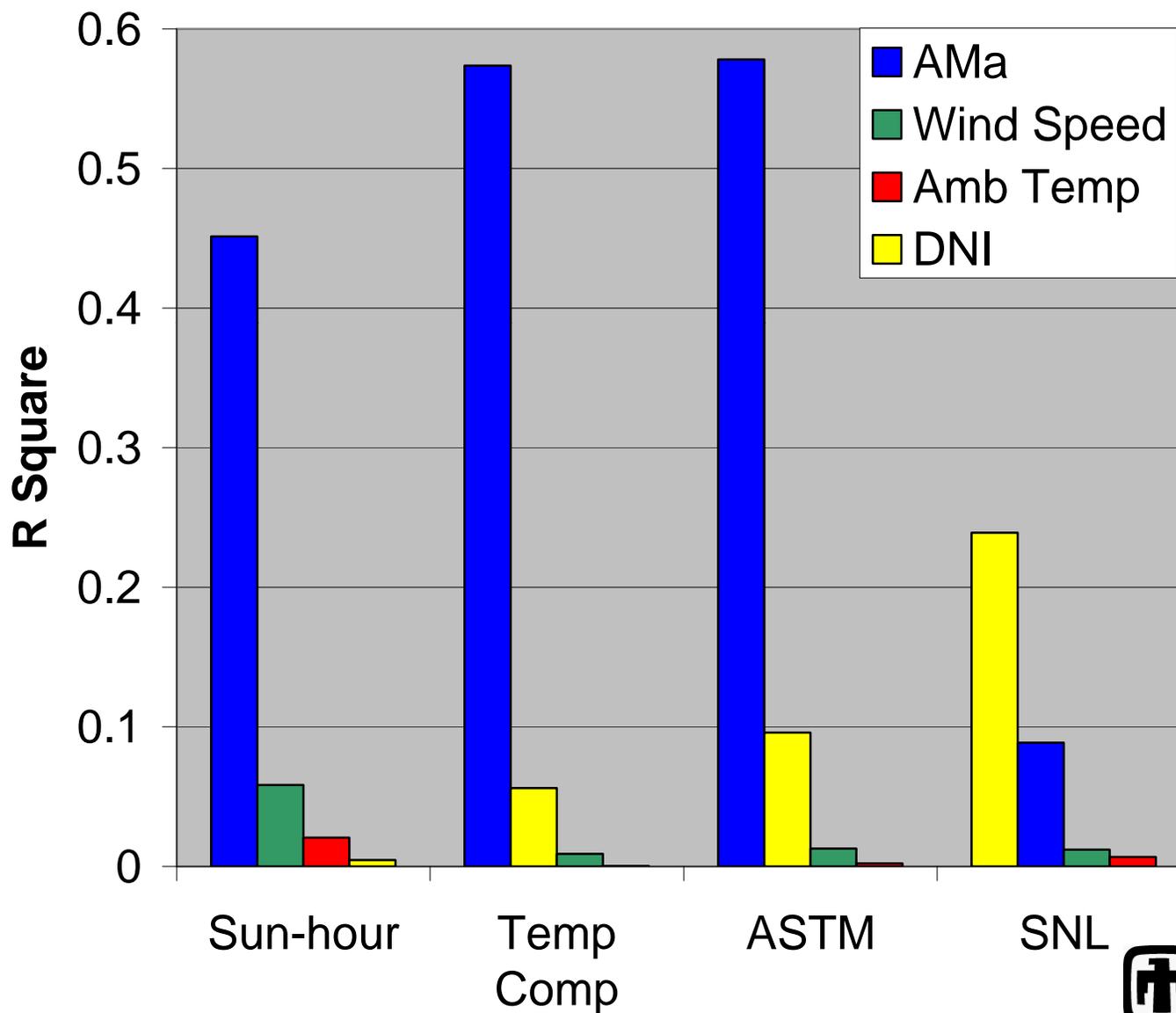
- Our objective – to see how models can be improved
 - Correlation between residuals and model inputs indicates opportunities for improvement
- Stepwise regression uses a stepped series of linear regressions:

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

- Y is the model residuals
- X includes DNI, Air Mass, Ambient Temperature, Wind Speed

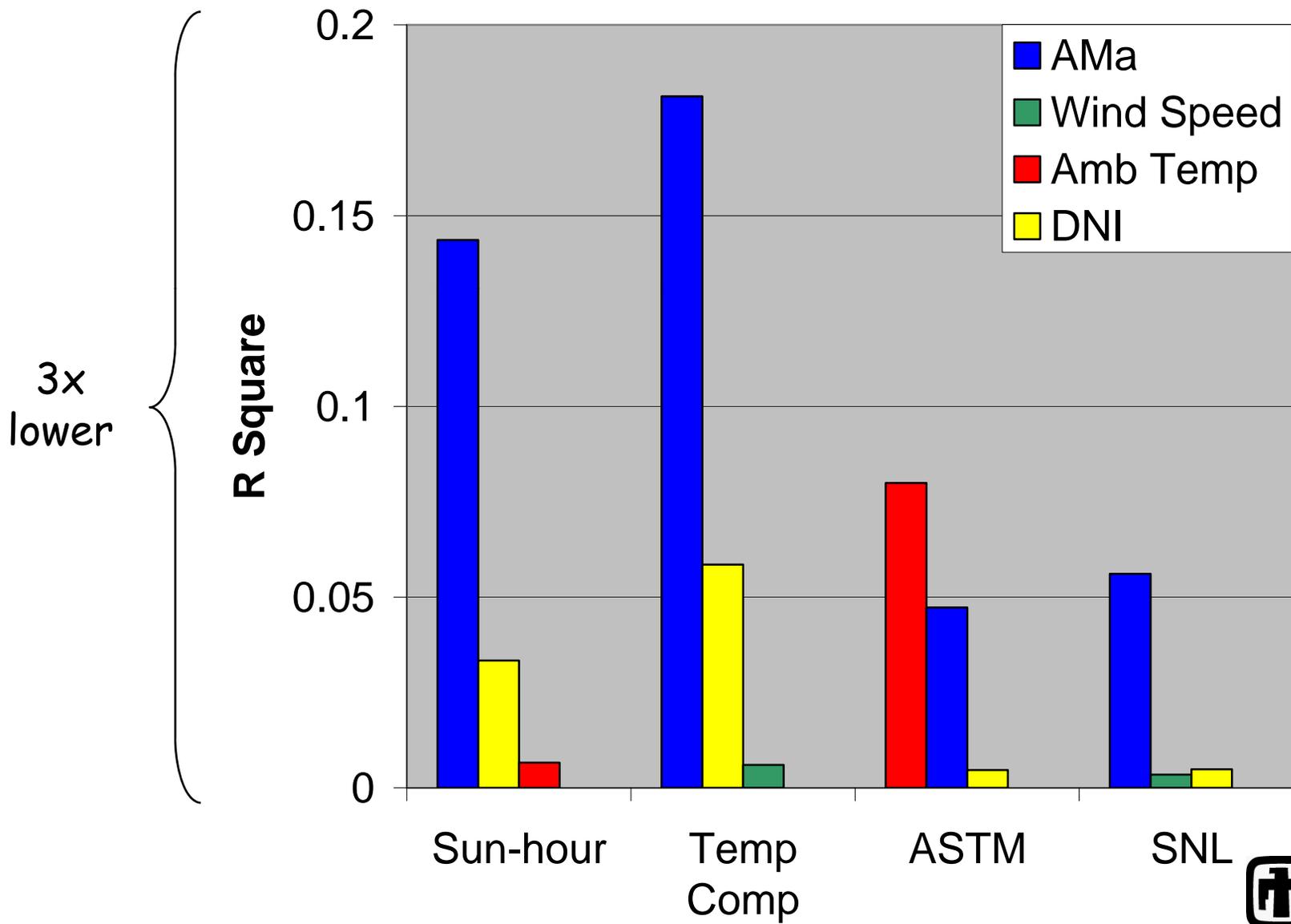


Results – System 1





Results – System 2





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Conclusions

- All models, even the simplest, worked well
 - A longer test (greater variations in input parameters) and multiple locations might produce different results
 - Only the Sandia model corrects for air mass
 - Step-wise analysis of the residuals of the other models showed the most sensitivity to air mass
 - The sun-hour model, having no temperature correction, was also sensitive to wind speed and temperature
- Future work
 - Analyze data for a longer test period (1 year)
 - Expand the number of systems and locations
 - Evaluate derate factors
 - Evaluate modeling of self-shading