



Tom Hoff President, Research and Consulting April 10, 2012



Acknowledgements



- Funding for this work has been provided by the California Solar Initiative (CSI) Grant Agreement titled "Advanced Modeling and Verification for High Penetration PV"
- Special thanks to Smita Gupta, Project Manager

Funding Approver	California Public Utilities Commission
Funding Distributor	California IOUs
Program Manager	Itron





Project Participants

Companies, Agencies & Organizations











Utilities



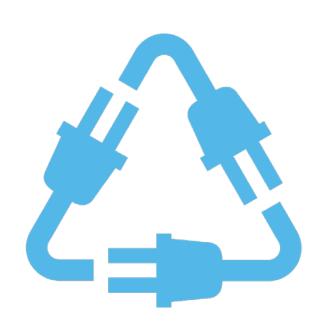






Clean Power Research

- Founded by Dr. Thomas E. Hoff in 1998
- Pioneers in economic analysis of PV and clean energy
- Software services powered by research and industry-leading data
- Research and consulting in Napa, CA Software development in Kirkland, WA





Company Structure

Research
Develops Analysis





Software
Delivers analysis via
software tools



Consulting
Applies analysis to
client-specific issues



CLEAN POWER RESEARCH PRODUCTS & SERVICES

DATA SOURCES

Solar Anywhere®

PowerTariffs™

Incentives

TMY (NREL)

SOLAR PREDICTION

Forecast, real-time and historical timeseries irradiance data and power simulation for planning, developing and operating solar installations

SolarAnywhere®

ECONOMIC VALUATION

Embeddable energy, economic and environmental analysis for streamlining sales of renewable energy, EV and energy efficiency projects.

CleanPower Estimator®

QuickQuotes™

PowerBill®

PROGRAM OPTIMIZATION

Automated incentive management and optimization services for renewable energy programs

PowerClerk®

RESEARCH AND CONSULTING



Clean Power Research®

Project Overview

PlannedTasks

- Enhance spatial and temporal resolution of SolarAnywhere[®] for selected CA locations
- Enhance PVSimulator to include output variability and validate
- Integrate PV modeling capabilities into a distribution system engineering and analysis tool
- Create initial Excel tool to calculate economic value of a specific PV fleet configuration

Actual Accomplishments

- Enhanced spatial and temporal resolution of SolarAnywhere[®] for the entire state of CA
- Developed, validated, and patented approach to simulate both output variability and output
- Developed web services to integrate PV modeling capabilities into distribution system tools
- Created full web-based tool to calculate economic value of a specific PV fleet configuration

Task 1

Enhanced Spatial and Temporal Resolution of SolarAnywhere for the Entire State of CA



What is SolarAnywhere?

Web-accessible solar irradiance data service

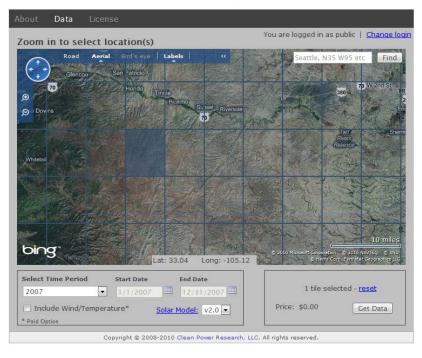
Innovative

- Historical values from 1998 through latest hour
- Hour-by-hour forecasts up to a week in advance

Accessible

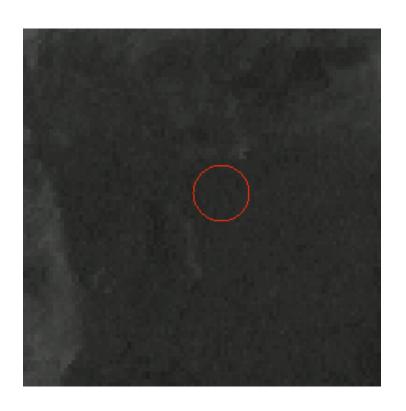
- Select locations via webbased map
- Programmatically accessible web service
- Integrate with system operation tools







Convert Satellite Images to Irradiance





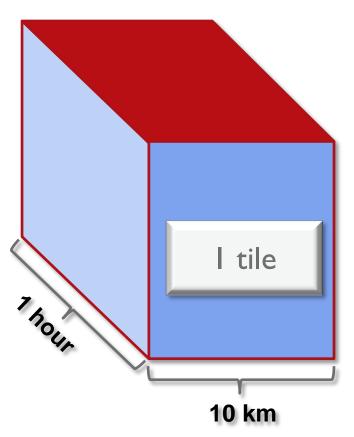
12 PM

4 PM

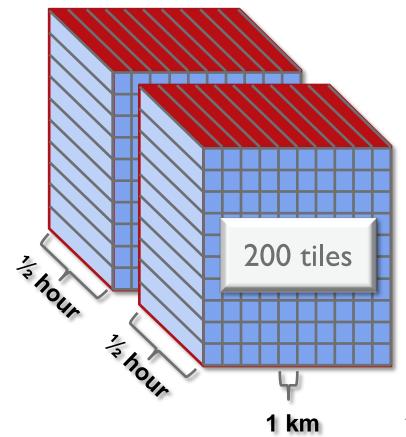
SolarAnywhere Enhanced Resolution



10 km grid, 1 hour data



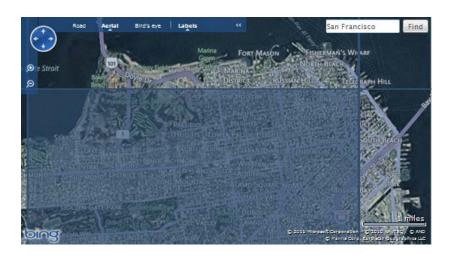
1 km grid, 1/2 hour data



Example: San Francisco, CA



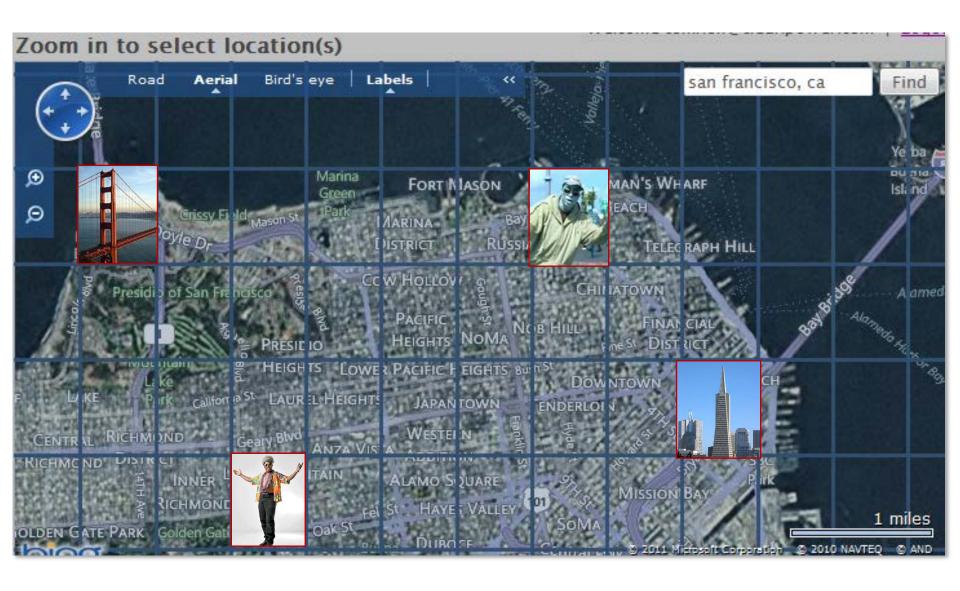




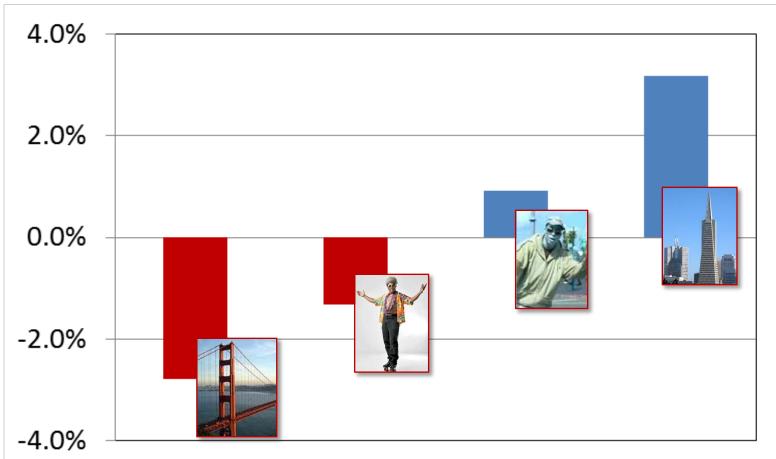




Example: San Francisco, CA



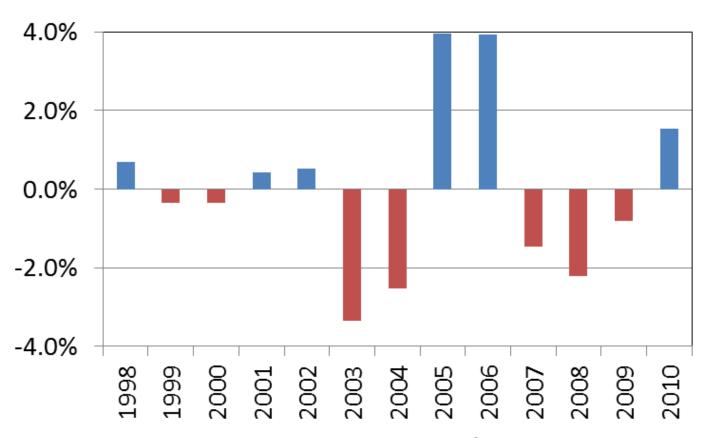
Annual Variation by Location



Result based on 2003-2010 SolarAnywhere Enhanced Resolution data



Variability by Year



Result based on 2003-2010 SolarAnywhere Enhanced Resolution and 1998 – 2002 SolarAnywhere Standard Resolution data



SolarAnywhere Status

- Standard resolution (10 km grid, 1 hour resolution) data from 1998 through 7 day forecast anywhere in U.S.
- Enhanced resolution (1 km grid, ½ hour resolution) data from 2003 through 7 day forecast anywhere in California
- Provide data to perform high penetration renewables studies (e.g., CAISO high penetration study) and multiple validation studies (UCSD)

www.SolarAnywhere.com

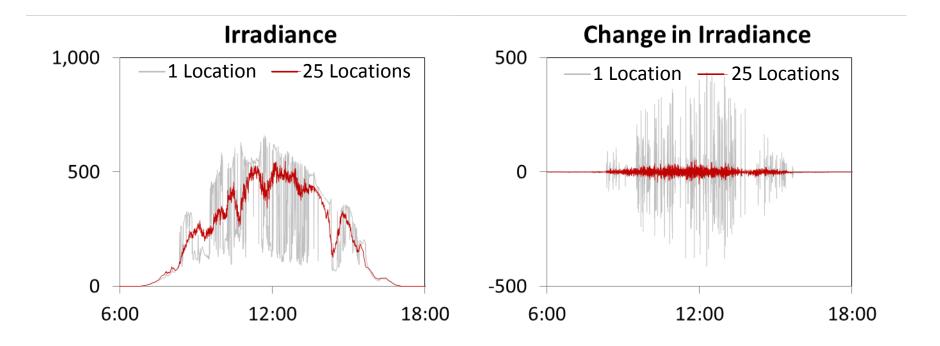


Task 2

Developed, Validated, and Patented Approach to Simulate Both Output Variability and Output



What is Output Variability?



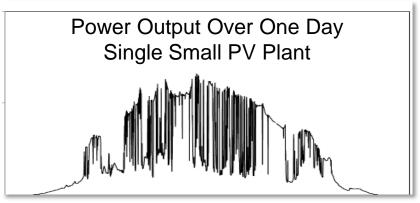


PV's Intermittency Challenges

- Who will face the challenges of PV intermittency?
 - Independent system operators
 - Load serving entity schedulers
 - Distribution system planners
 - Distribution system operators
- What do they need to address the challenges?
 - Aggregated fleet power and variability (real-time and forecasted)



Independent System Operators



If this is 1 PV plant what will I do with 100,000?

- Maintain balancing area reliability
- Plan for resources to accommodate PV
- Procure capacity and ramping resources based on PV fleet forecasts

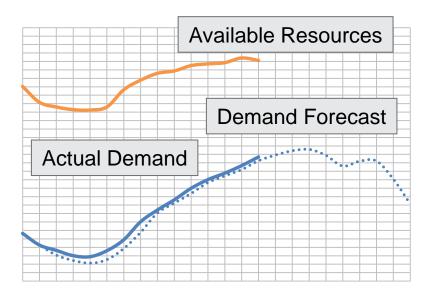




Load Serving Entities

- Submit demand forecasts to system operator
- Distributed PV hampers ability to forecast demand
- Inaccurate forecasts result in financial penalties
- Need to adjust demand forecasts based on distributed
 PV fleet forecasts







Distribution System Planners

- Ensure customers receive reliable power
- Distributed PV complicates distribution planning
- Customers will complain if power is unreliable
- Planners need to understand
 PV fleet output on feeders to keep customers happy







Distribution System Operators

- Storage will be deployed within smart grids
- Storage needs to be intelligently dispatched to maximize value
- Operators need PV fleet forecasts within distribution system to intelligently dispatch storage



A123 batteries in distribution system



Perception: PV is Highly Variable and Unpredictable

How can high-penetration PV resources be integrated without incurring large backup generation and other equipment costs?



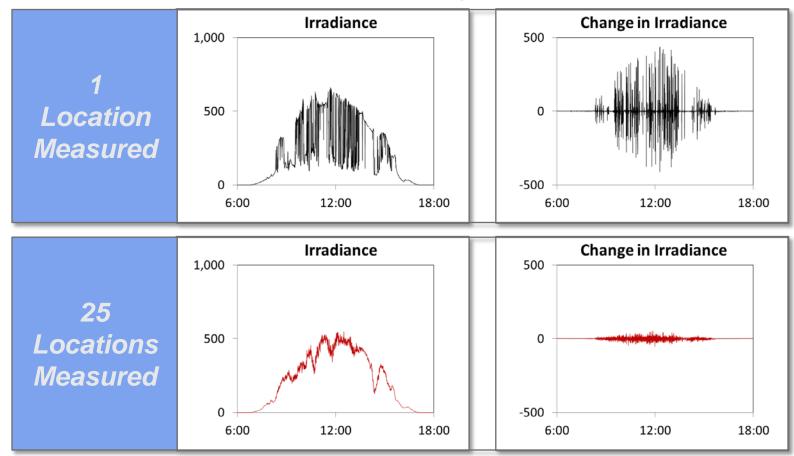
The story begins with assessing variability of an actual PV fleet...





Geographic Diversity Leverages Weather System Inertia

10 second irradiance data from 4 x 4 km grid in Napa on Nov. 21, 2010

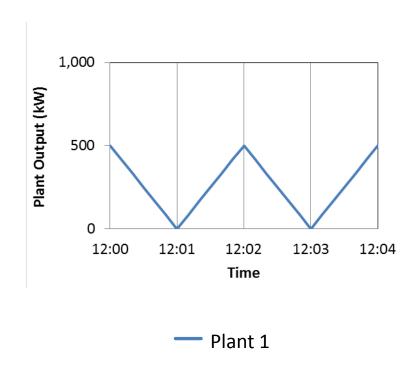


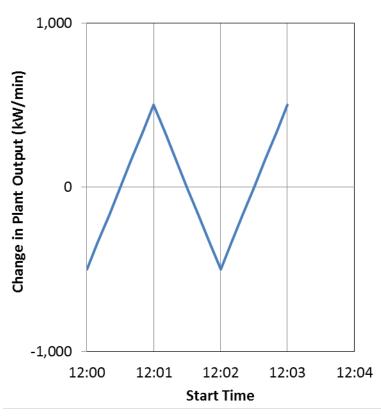


Variability Can Be Quantified Using Correlation Coefficients



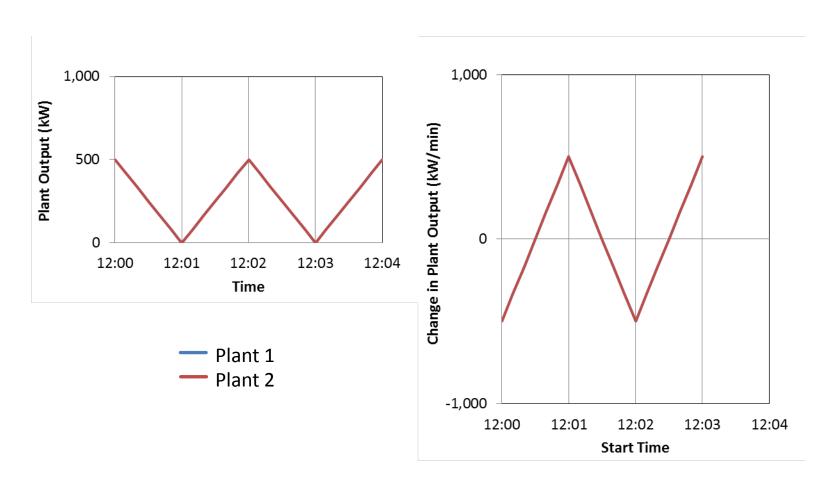
Hypothetical Example of Two 500-kW Plants: Perfect Positive Correlation (100%)





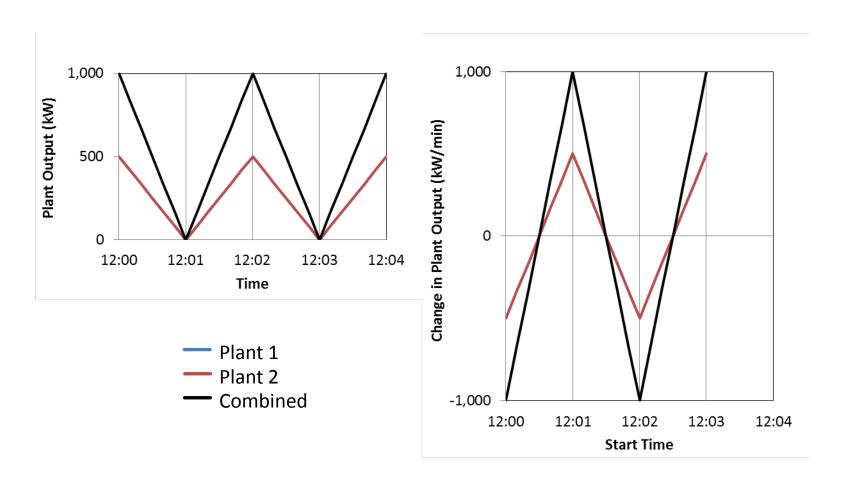


Hypothetical Example of Two 500-kW Plants: Perfect Positive Correlation (100%)



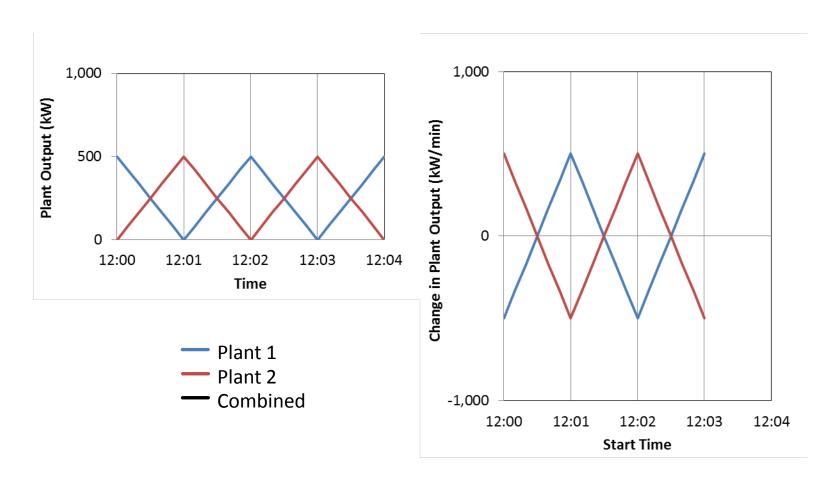


Hypothetical Example of Two 500-kW Plants: Perfect Positive Correlation (100%)



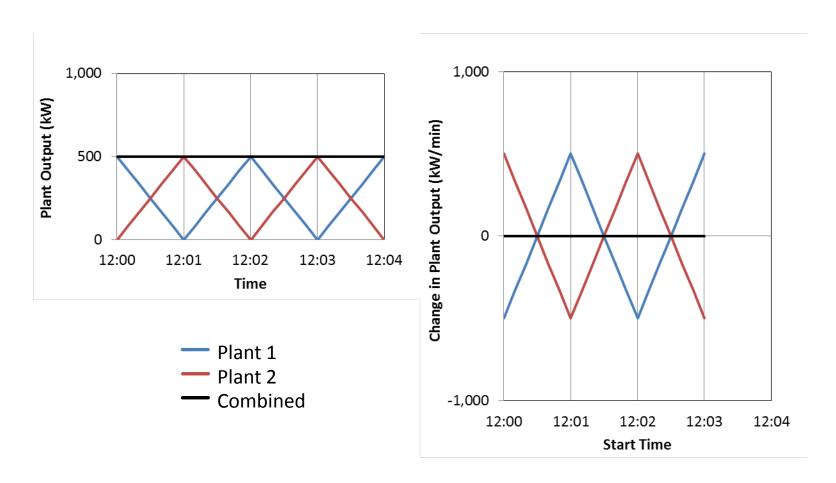


Hypothetical Example of Two 500-kW Plants: Perfect Negative Correlation (-100%)



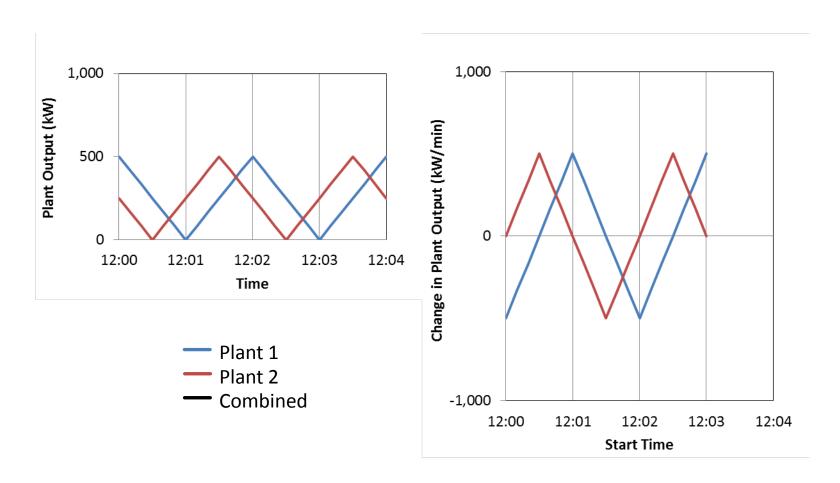


Hypothetical Example of Two 500-kW Plants: Perfect Negative Correlation (-100%)



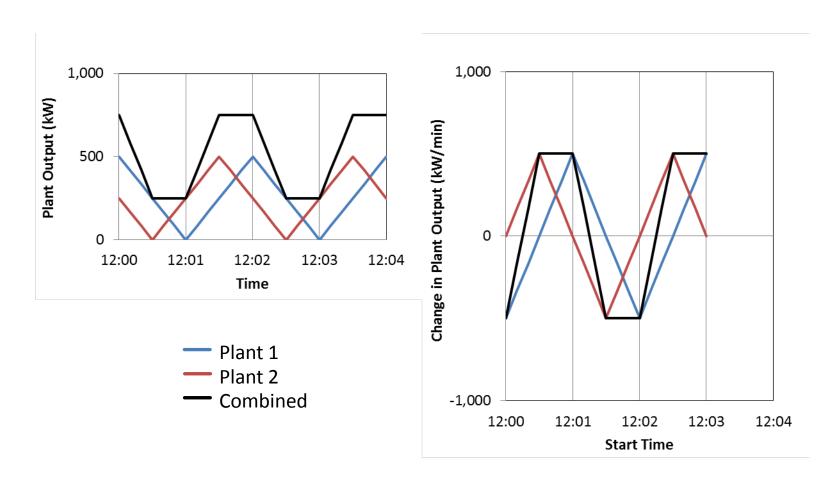


Hypothetical Example of Two 500-kW Plants: Uncorrelated (0%)



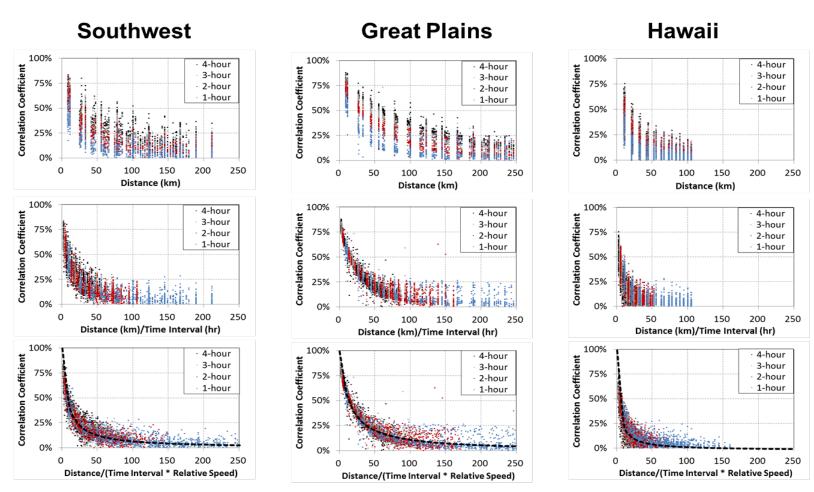


Hypothetical Example of Two 500-kW Plants: Uncorrelated (0%)





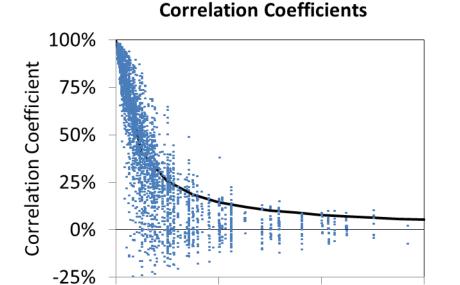
Correlation in Change in Clearness



Note: Distance / (Time Interval * Relative Speed) is related to Dispersion Factor



Correlation at Cordelia Junction

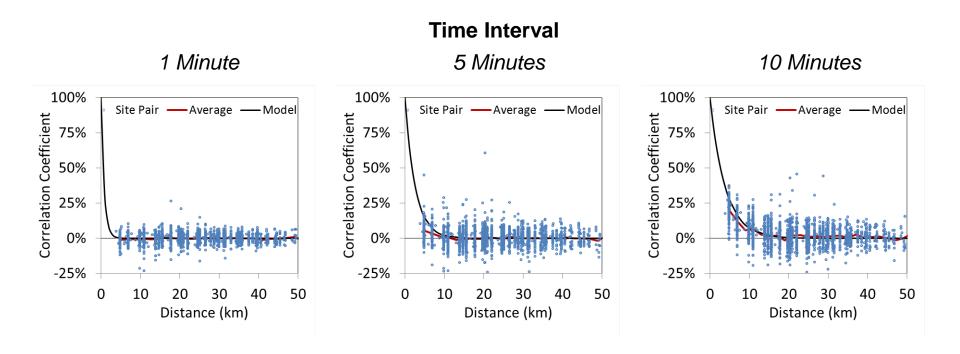


10 sec. data from 400m x 400m grid at Cordelia Junction, CA (Nov. 7, 2010)

Distance/Time Increment (m/s)



CPR's Model Required to Simulate Variability Confirmed Using SMUD's Network of Irradiance Sensors



Bing, et. al., "Solar Monitoring, Forecasting, and Variability Assessment at SMUD," forthcoming at ASES 2012



Variability Can Be Minimal for a Large PV Fleet

- PV fleet variability is much lower than one would expect based on single system performance
- The physical inertia of rotating machines in the current utility system is what minimizes short-term variability
- The weather system inertia that is leveraged by the geographic diversity associated with a large fleet of PV systems is what will minimize short-term variability for PV



...the story continues with determining how to predict PV fleet power generation



Direct Approach: Measure Everything

- For installed PV systems
 - Install meters on every PV system
 - Retrieve power output at the desired time interval
 - Sum output across all systems in the fleet
- When studying impacts of future systems
 - Collect irradiance data ...
 - ... from dense network of weather monitoring stations
 - ... covering all anticipated locations of interest
 - ... at the desired time interval
 - Simulate output for each PV system individually
 - Sum results for all systems

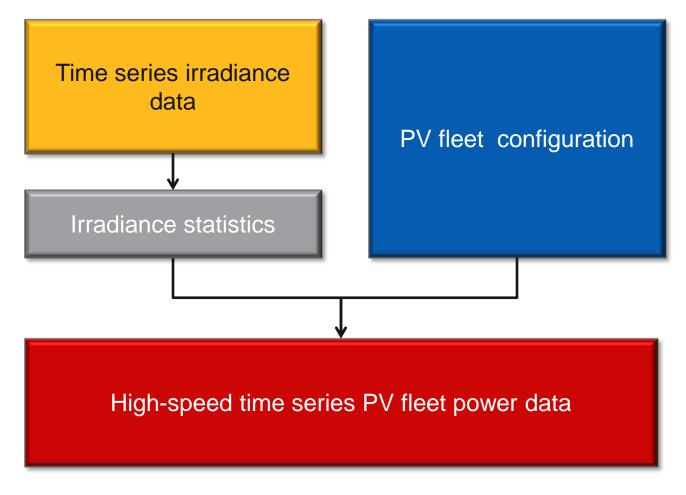


Problems with Direct Approach

- Expensive instrumentation and communications
- Requires measuring equipment
- Data must be recorded at desired time resolution
- Few dense irradiance networks exist
- Not applicable to forecasting



Indirect Approach





But do tools exist to implement this?

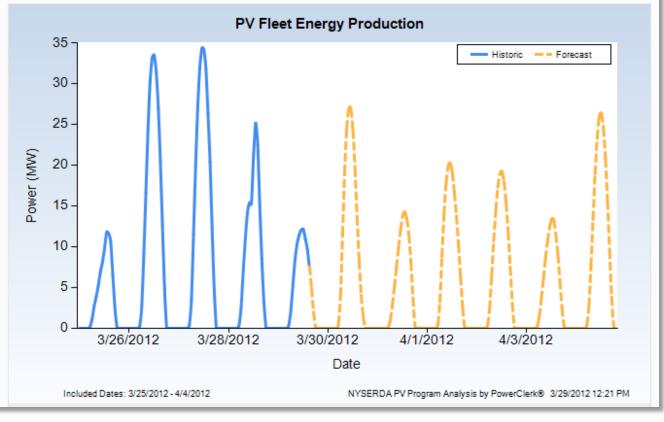


NYSERDA PV Fleet Output: In Use for 1+ Year

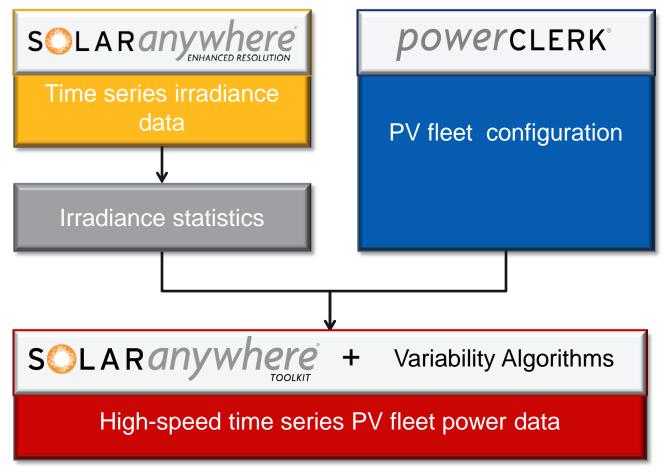




Available Reports Map PV Fleet Energy Production Program Summary Installation Status Applications by Status Cumulative Production Cumulative Capacity Capacity by Manufacturer Price by County Price Deviation by County Price Distribution Capacity Distribution Incentive Distribution



Use of Software Services to Implement Approach

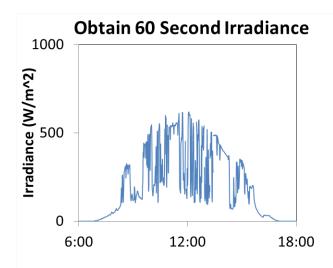


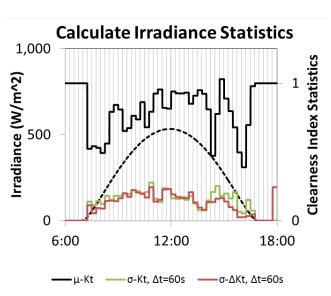


How do the pieces fit together?

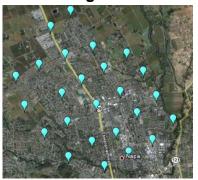


Generate High-Speed Time Series PV Fleet Power Data





Combine with Fleet Configuration



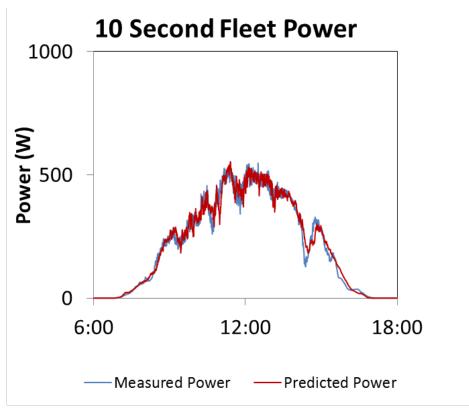
Patent Pending:

- 1. Computer-Implemented System and Method for Estimating Power Data for a Photovoltaic Power Generation Fleet
 - 2. Computer-Implemented System and Method for Determining Point-to-Point Correlation of Sky Clearness for Photovoltaic Power Generation Fleet Output Estimation
 - 3. Computer-Implemented System and Method for Efficiently Performing Area-to-Point Conversion of Satellite Imagery for Photovoltaic Power Generation Fleet Output Estimation



Result: High-Speed Time Series PV Fleet Data

Measured and simulated 10 second data from 25 locations in a 4 km x 4 km grid in Napa on Nov. 21, 2010





Task 2 Summary

- Utility system planners and operators need tools to predict PV fleet performance
- CPR has been developing such tools
 - Produced enhanced resolution solar resource data
 - ✓ Obtained PV system specifications
 - Designed simulation algorithms
- Work underway (CEC & CSI Phase 3 contract)
 - Generate high resolution solar resource data
 - Apply high speed PV fleet simulation algorithms
 - Validate PV fleet simulation using measured PV plant data



Task 3

Developed Web Services to Integrate PV Modeling Capabilities into Distribution System Tools



Grid Services Rationale

Solar Services (belongs at CPR)

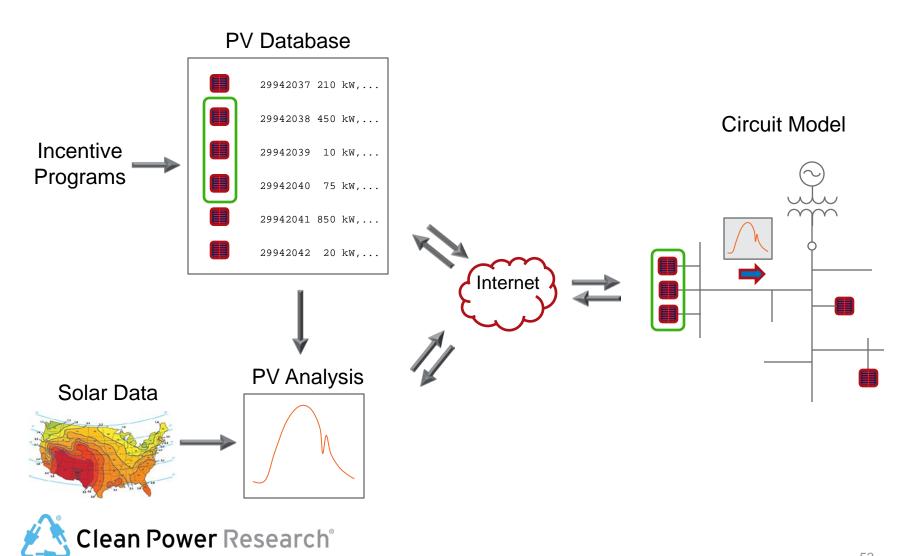
- PV modeling expertise
 - Solar irradiance
 - Cloud effects & shading
 - Sun angles
 - Forecasting
 - Variability
- PV equipment database
- Nationally recognized analytics

Circuit Modeling Software (belongs at developer/utility)

- Circuit analysis expertise
- Developed and tested
- Accepted by utilities



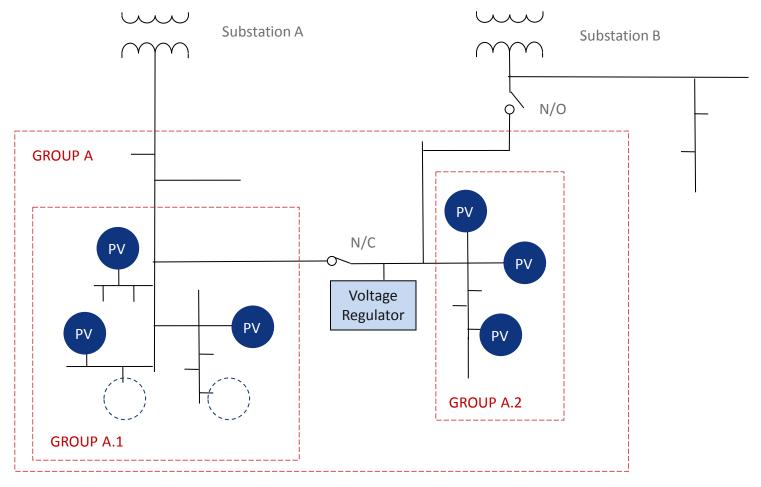
Accessing CPR Grid Services



Uses of Software Services

Web service	Result	Uses
PV System Power Statistics (Complete)	PV System: Min, max power by hour (10 year stats)	Load flow studies (voltages, currents)CVR
	PV System: Average, standard deviation of power by hour (10 year stats)	Estimate lifetime of devices (regulators, capacitors)Monte Carlo simulations?
PV System Power (Complete)	PV System: (real-time & hours-ahead forecast)	Fault locationLoad transfer decisions
PV Fleet Power and Variability (In Development)	PV Fleet: Variability (real time)	 Determine expected voltage swings Load transfer decisions Schedule reg. reserves (ISO)
	PV Fleet: Power (forecast)	 Estimate summer peak overloading Estimate day-ahead "net" loads for scheduling generation Schedule balancing energy (ISO)

New Concepts: Groups & Grid Status





Groups and Grid Status G4 G6 G3 G5 G8 G2 G7 G1 "study" system "online" system "offline" system



Task 4

Created Full Web-based Tool to Calculate Economic Value of a Specific PV Fleet Configuration



Distributed PV Calculator Project Objective

- Build a software tool to calculate the location-specific economic value of PV to the utility grid from the utility perspective
- Targeted for use by utility planners and policy analysts



Sample Study Questions

- How should a value-based Feed-in-Tariff (FIT) be designed?
- Is it worthwhile to target specific distribution areas?
 Which ones?
- How much PV is needed to delay a distribution upgrade by 1 year?
- How much load reduction is obtained from various PV system configurations?



Calculator Benefits

Full input control

PV specifications
Utility data

Economic assumptions

State-of-the-art valuation

Fast

Repeatable

Scenario analysis

Uses

PV incentive design

RPS planning

Screening

Saves cost and provides consistency



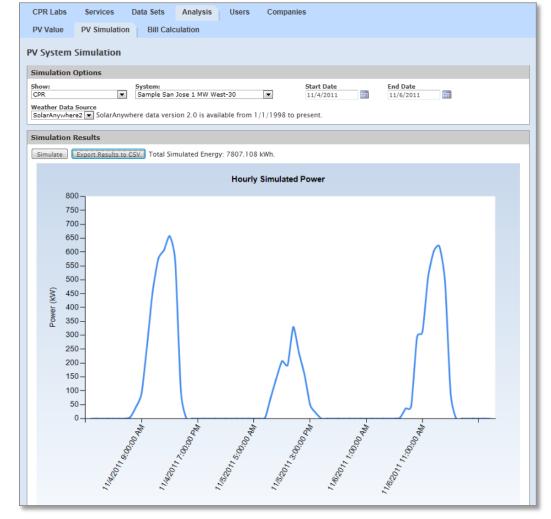
CPR's Distributed PV Value Analysis History

- Grid-Support PV: Evaluation of Criteria and Methods to Assess Local and System Benefits to Electric Utilities (1993)
- Final Results from the PVUSA Project at Kerman (1994, 1995, 1996)
- Photovoltaic Economics and Markets: The Sacramento Municipal Utility District as a Case Study (1996)
- A Preliminary Analysis of Block Island Power Company's Use of Clean Distributed Resources to Provide Power to Its Customers (1998)
- Distributed Resources: A Potentially Economically Attractive Option to Satisfy Increased Demand on Okanogan County Electric Cooperative's Mazama Feeder Line (1998)
- Final Results Report with a Determination of Stacked Benefits of Both Utility-Owned and Customer-Owned PV Systems to SMUD (2002)
- Determination of Photovoltaic Effective Capacity for Nevada Power (2003)
- Determination of Photovoltaic Effective Capacity for New Jersey (2004)
- Moving Towards a More Comprehensive Framework to Evaluate Distributed Photovoltaics (2005)
- The Value of Distributed Photovoltaics to Austin Energy and the City of Austin (2006)
- Shining on the Big Apple: Satisfying New York City's Peak Electrical Needs with PV (2008)
- PV Value Analysis for We Energies (2009)



PV Simulations

- Simulations for any defined systems in company
- Uses SolarAnywhere
- Historical, real time or forecast
- Downloadable results in CSV format





Loss Savings

- Distributed resources reduce system losses by producing power in the same location where it is used
- Loss savings increase the value of other benefits across generation, transmission, and distribution systems
- Correct loss savings depend upon the benefit category
- Loss savings are calculated on the margin



Energy Value

Definition

 Benefit from distributed PV generation's offset of wholesale energy purchases

Methodology

- Energy value equals PV output plus loss savings times marginal energy cost
- Marginal energy costs are based on fuel and O&M costs of generator operating on the margin (CCGT)







Generation Capacity Value

Definition

 Benefit from added capacity provided to the generation system by distributed PV

Methodology

 Generation capacity value equals generation capacity cost times PV's effective load carrying capability (ELCC), tal

carrying capability (ELCC), taking into account loss savings

Generation capacity cost is based on capital cost of CCGT



Fuel Price Hedge Value

Definition

 Benefit that distributed PV generation has no fuel price uncertainty

Methodology

 Fuel price hedge value is calculated by determining how much it would cost to eliminate the fuel price uncertainty associated with natural gas generation







T&D Capacity Value

Definition

 Benefit that distributed PV generation provides in reducing the burden on the T&D system and thus delaying the need for capital investments in the T&D system

Methodology

 T&D capacity value equals the expected long-term T&D system capacity upgrade cost, divided by load growth, times financial term, times a factor that represents match between PV system output (adjusted for losses) and T&D system load







Environmental Value

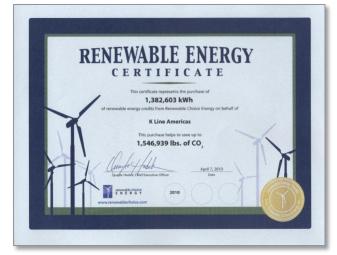
Definition

 Benefit that the environmental footprint of PV is considerably smaller than that of fossil-based generation

Methodology

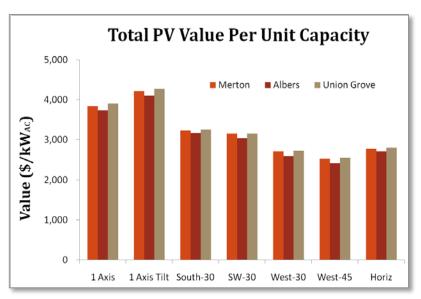
 Environmental value equals PV output times REC price

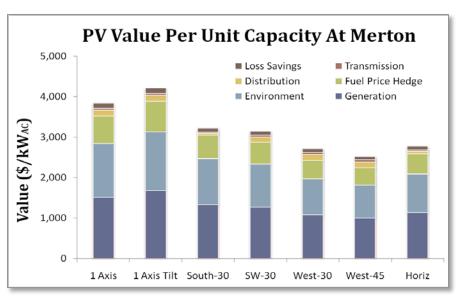


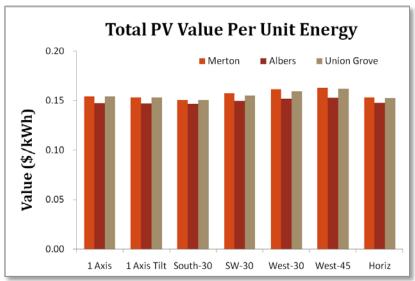




Graphing Results (in Excel)







On-line Tutorial



www.youtube.com/watch?v=3IKi0UjD008



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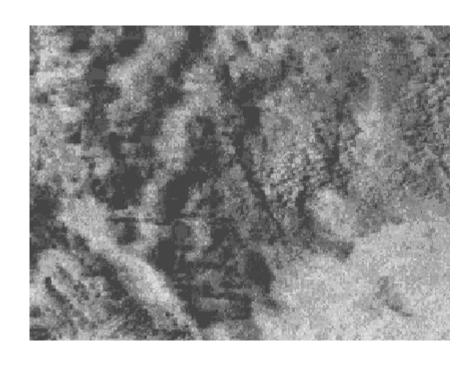


Sneak Previews

Cool Stuff Under Development



Creating 1-Minute Data Sets Using Cloud Motion Vector Approach

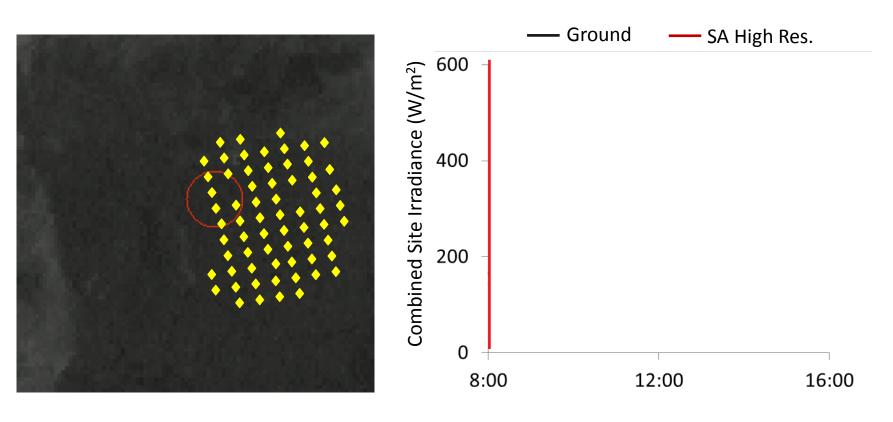


0 10 20 30 Minutes

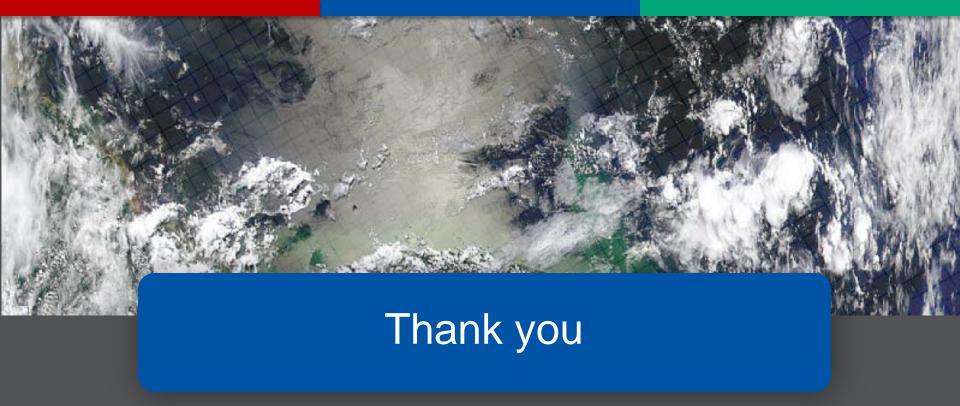


Measured Data From SMUD's 66 Sensor Network vs. Simulated Data on Highly Variable Day

Nov. 18, 2011







Clean Power Research®

Questions?
Contact: Tom Hoff
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