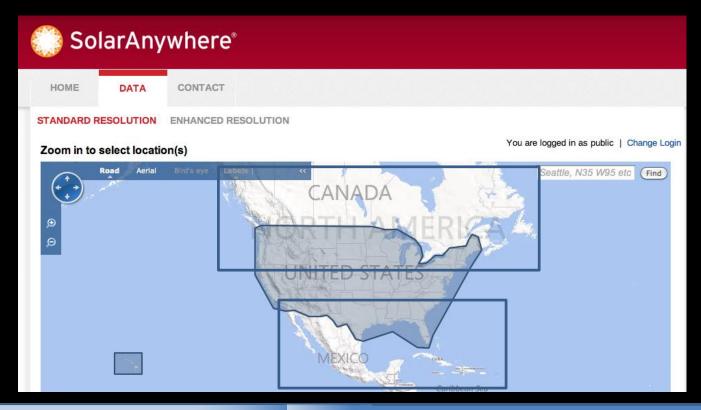
An Operational Solar Forecast Model For PV Fleet Simulation

Richard Perez & Skip Dise

Jim Schlemmer Sergey Kivalov Karl Hemker, Jr. Adam Kankiewicz

Agenda

- Historical and forecast platform
- Blended forecast approach
- Application in electric grid operation



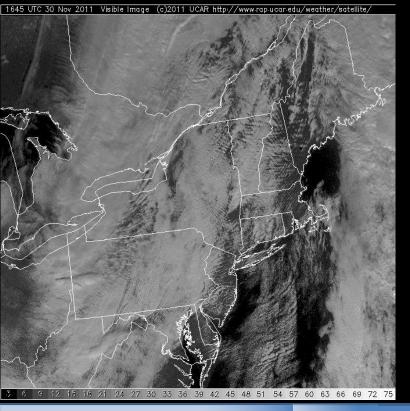
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NOW

FORECAST

1998

→ 6 days ahead



SATELLITE REMOTE SENSING MODEL

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1998

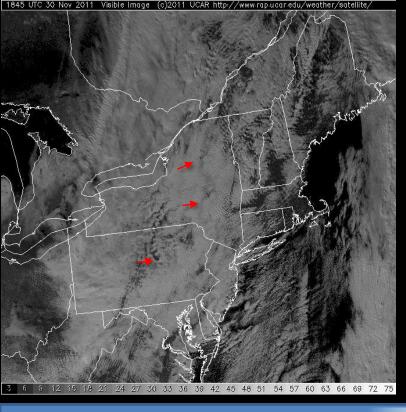
→ 6 days ahead

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Original Model



Cloud speed calculated individually for each image pixel, by minimizing pixel-surrounding scene difference between two consecutive images.

Cloud speed is applied to generate future image

With high resolution images, model can be applied to generate 1-minute resolution data forecasts

The procedure can also be applied to generate 1 minute historical data

CLOUD MOTION

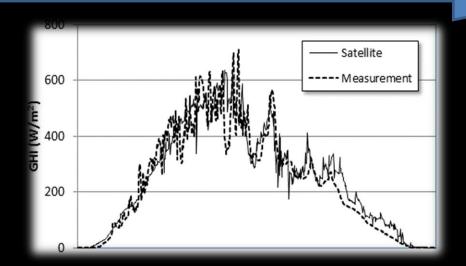
CLIMATOLOGY

NOW

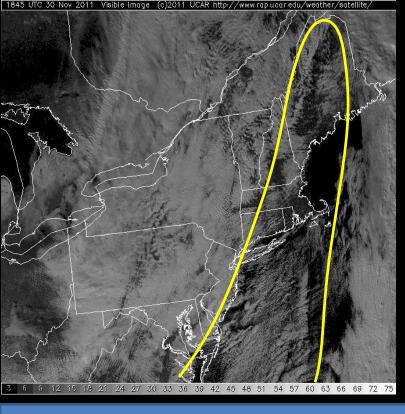
HOURS AHEAD

DAYS AHEAD

Original Model



© R. Perez & et al.





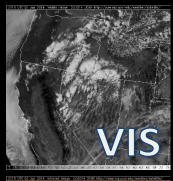
CLIMATOLOGY

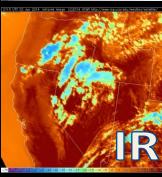
NOW

HOURS AHEAD

DAYS AHEAD

Original Model







GLOBAL SCALE MODELS

- NCEP GFS
- ECMWF

CONTINENTAL SCALE MODELS

NAM

HIGH RESOLUTION ASSIMILATION MODELS

- RAP
- HRRR

Cloud motion

NDFD

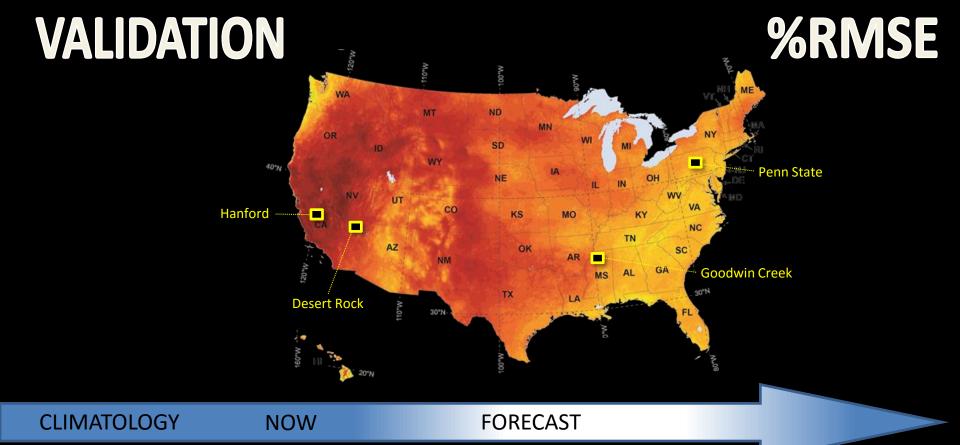
Operational NWP models

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NEW Model OPTIMUM MIX

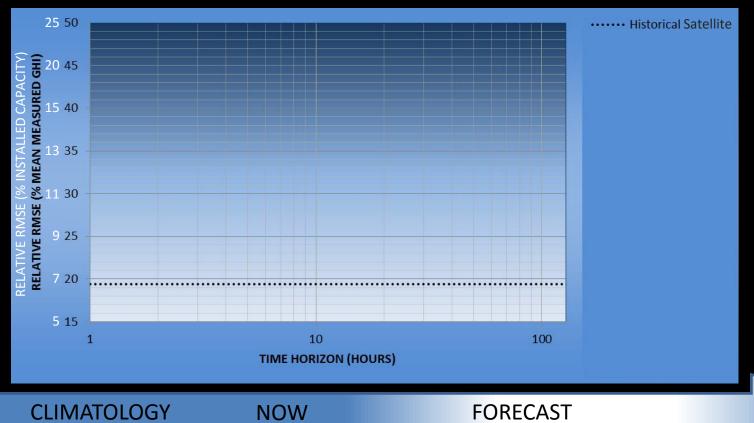


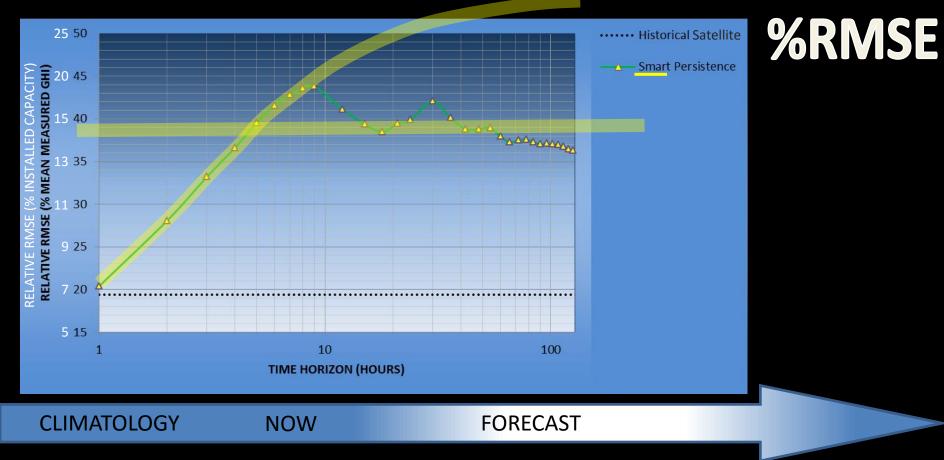


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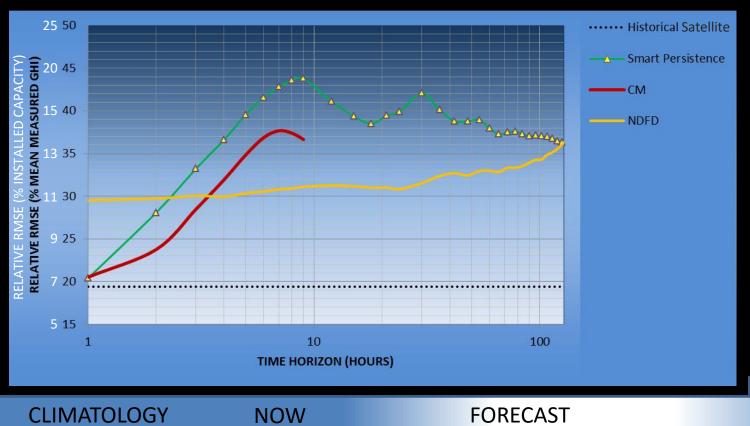


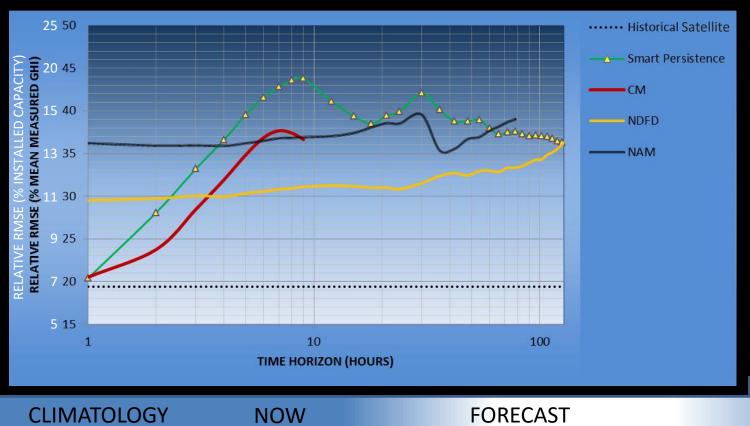


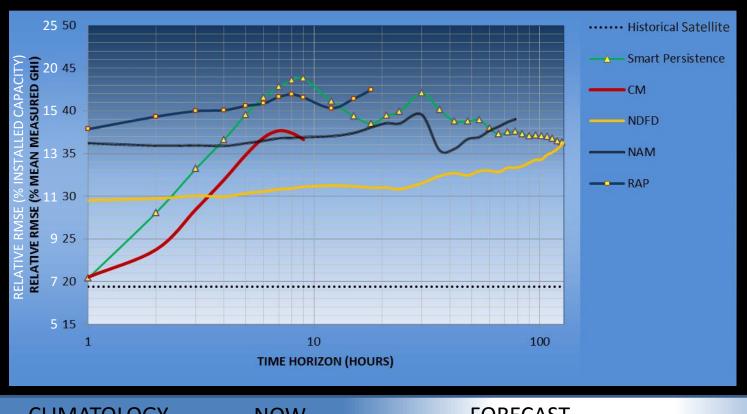
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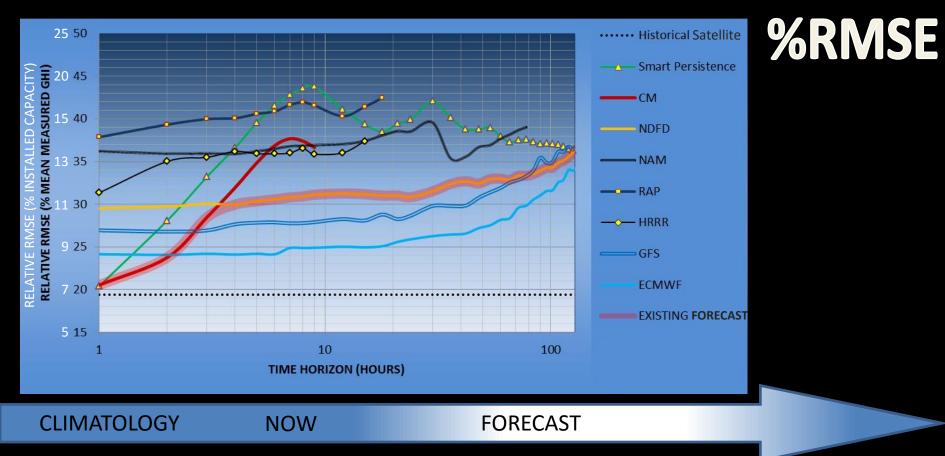
NOW

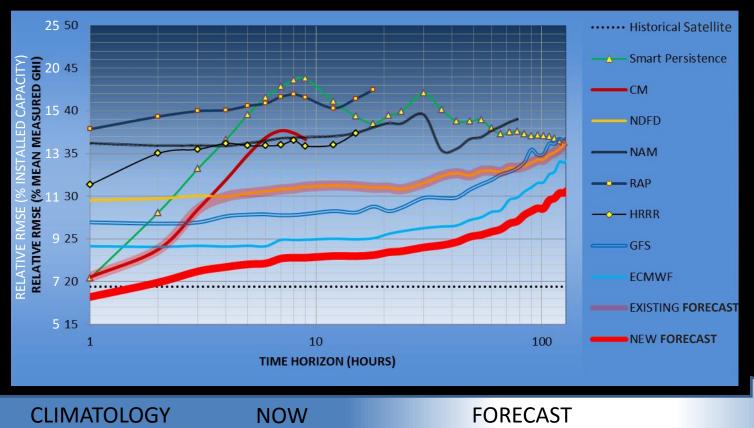
FORECAST

NEW Model



© R. Perez & et al.



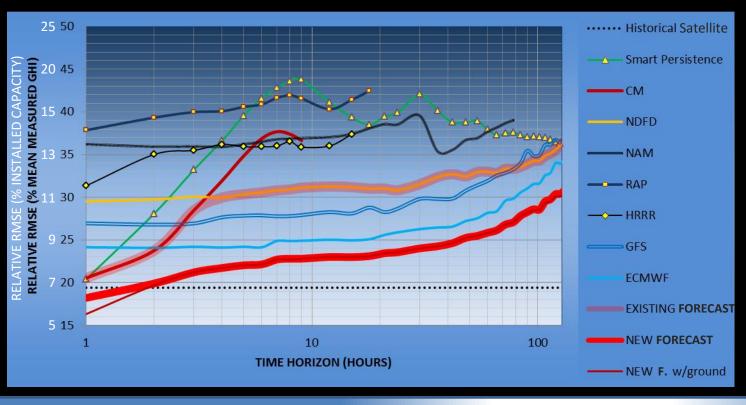




CLIMATOLOGY

NOW

FORECAST



CLIMATOLOGY

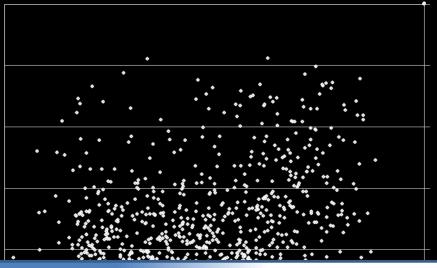
NOW

FORECAST

After SMUD'sTJ Vargas et al.

CLOUDY

Cloudy forecast Clear conditions

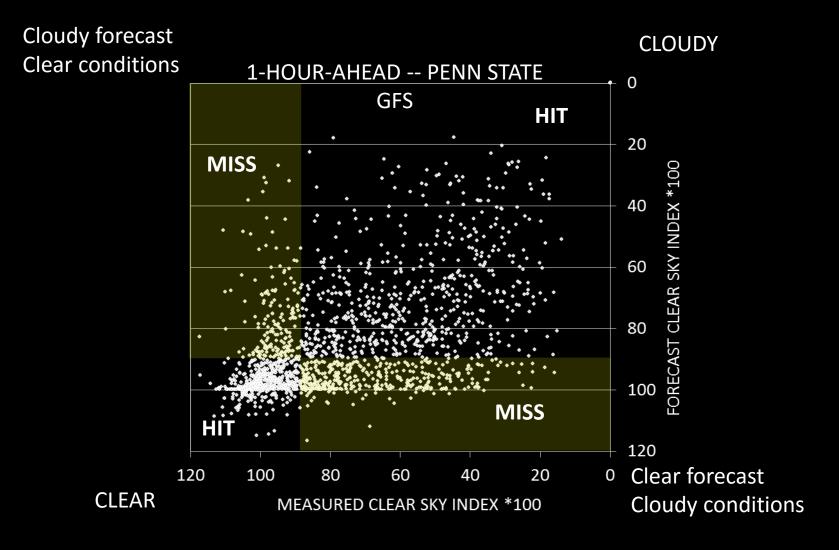


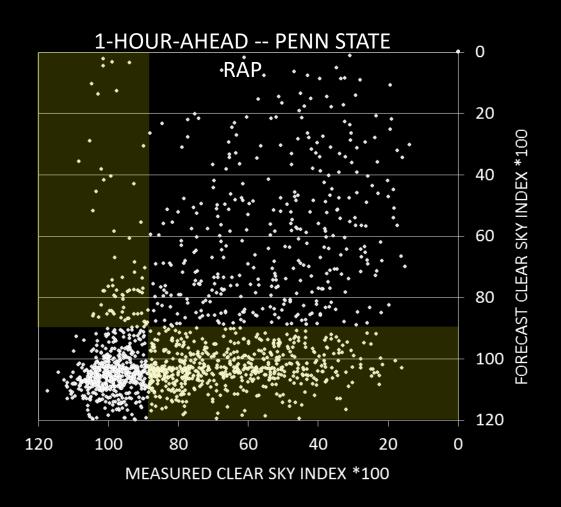
CLIMATOLOGY NOW FORECAST

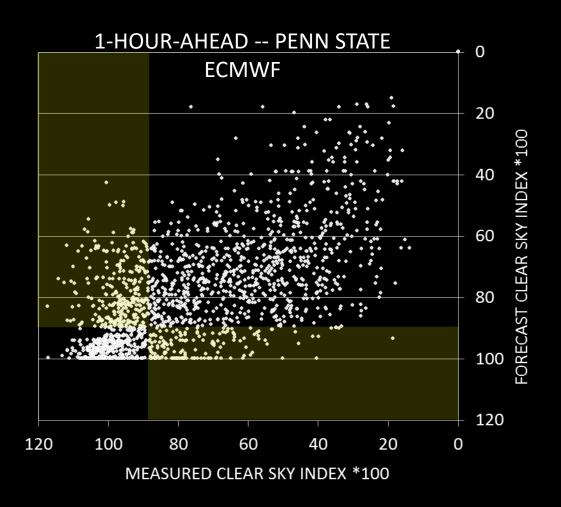
NEW Model

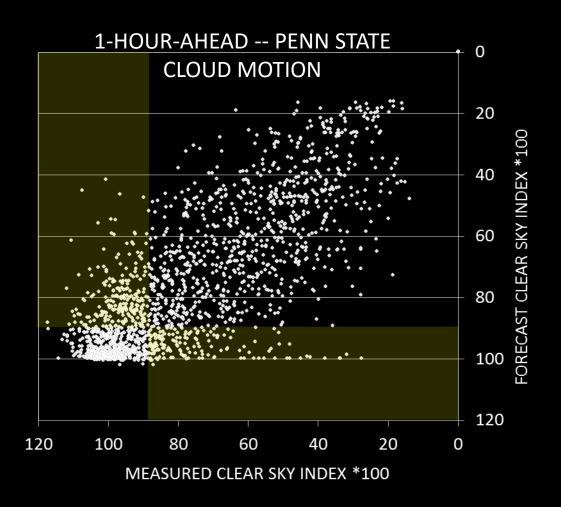
CLEAR

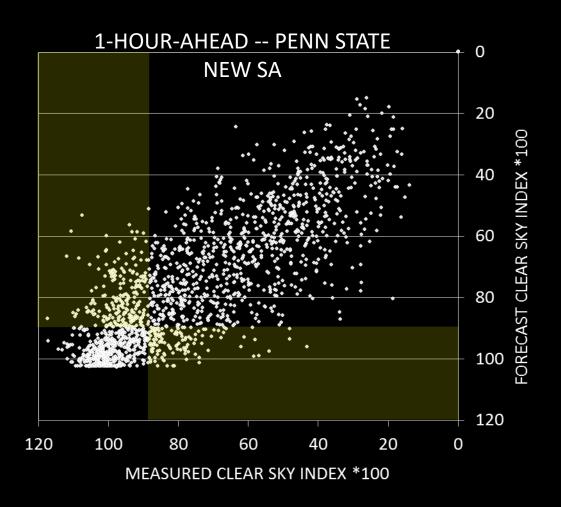
Clear forecast Cloudy conditions



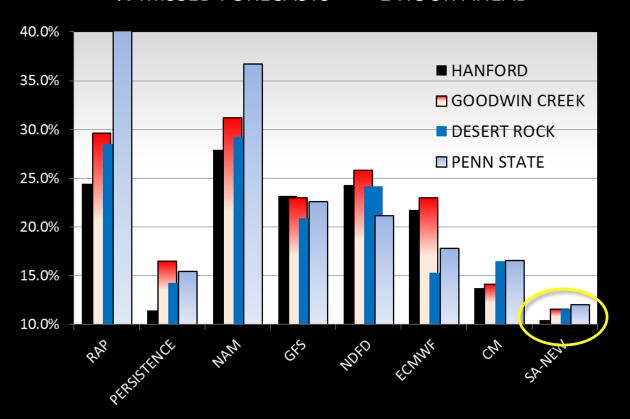




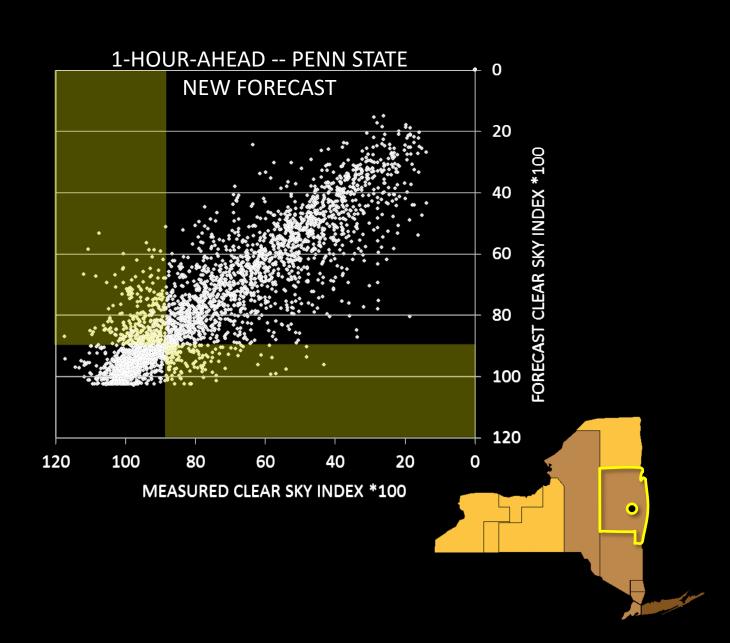




% MISSED FORECASTS -- 1 HOUR AHEAD



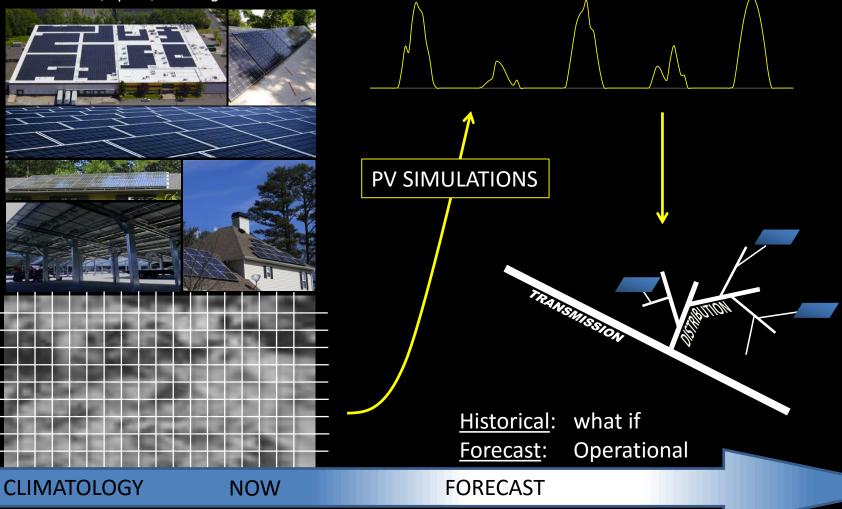
SMASLE REGION IEA HIT/MISS METRIC



APPLICATION: PV FLEET MANAGEMENT

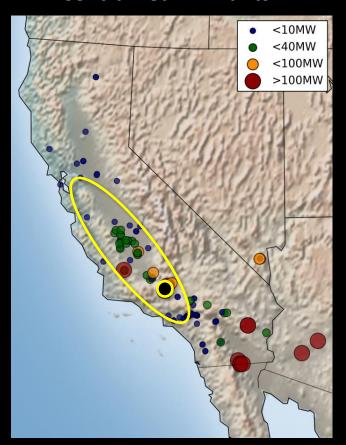
PV FLEET DATA:

Individual system size, orientation, location, specs, shading



APPLICATION: PV FLEET MANAGEMENT

Centralized PV Plants



Distributed PV



APPLICATION: PV FLEET MANAGEMENT



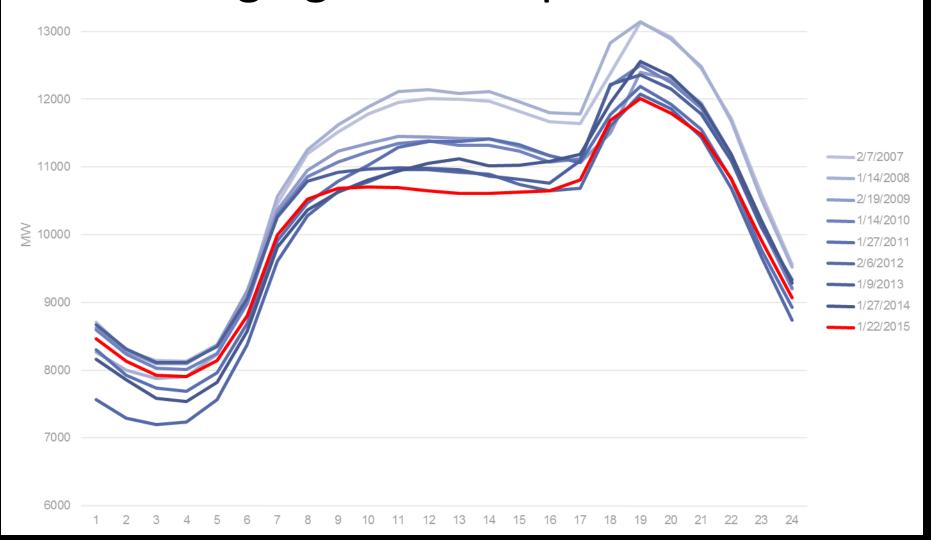
Distributed PV





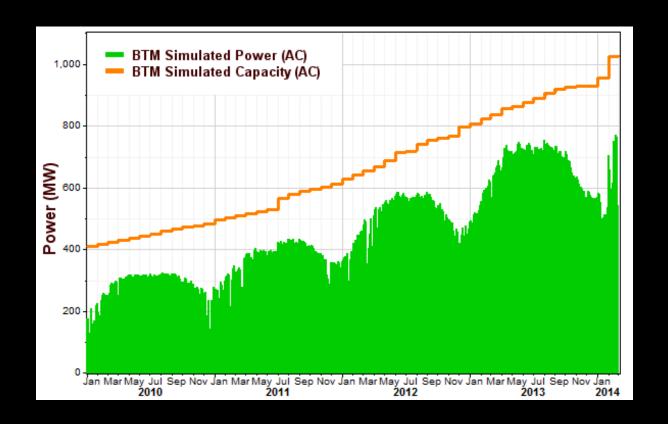
Detailed Individual PV system specs (location, modules, inverter, layout, etc. details)

Changing Load Shapes 2007-2015



Validation in Load Forecasting Models

- Itron and CAISO evaluated SolarAnywhere PV fleet historical production as training input into CAISO's Automated Load Forecast System (ALFS)
- Regions modeled from Jan 2010 through Feb 2014

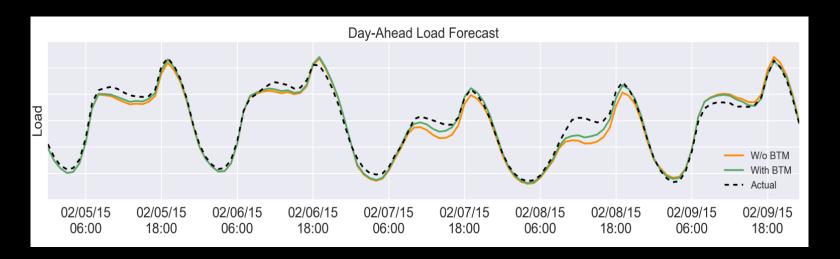


Preliminary Results (Historical Load Training)

Hour of Day Ahead Forecast	9 am		12 pm		3 pm	
CPR BTM Dataset?	No	Yes	No	Yes	No	Yes
Load Forecast Error (MW)	122.3	121.7	136.7	130.1	149.3	142.7
Load Forecast Error (%)	1.11%	1.10%	1.18%	1.12%	1.27%	1.21%
BTM Coefficient	-	-0.21	-	-0.92	-	-0.94
T-test (significant if < -1.64)	-	-1.8	-	-10.1	-	-8.9

- A BTM coefficient of -1 is the ideal result (i.e., for every predicted MW of BTM generation, one MW of load is shed)
- BTM production datasets offer a statistically significant improvement for midday and late-afternoon load forecasting
 - results lie quite close to the theoretical value of -1
- BTM production datasets offer a smaller impact for morning load forecasting
 - Believed to result from higher morning load variability

Preliminary Results (Load Forecasting)



- Historical training used as the base for inputting operational forecasting to the ALFS
- BTM forecasts significant improvement for mid-day and late-afternoon load forecasting
 - 9-months of results, up to 18% improvement on accuracy (at 2 PM DA prediction)

Preliminary Results (Load Forecasting)

	% Forecasts		
Minutes Ahead	Improved		
15	2.90%		
30	15.70%		
45	2.90%		
60	27.10%		
90	32.90%		
120	41.40%		
180	41.40%		
240	55.70%		
300	77.10%		
360	82.90%		

Conclusions

- Irradiance forecasts can be optimized for the North American climate
- Application to grid operation is needed
- Historical training vital to the neural net load forecast model
- PV fleet forecasts generate improved load forecasts for dayahead and 2-6 hour <u>ahead horizons</u>

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Thank You





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