



ProDROMOS: Managing Distribution Voltage using Photovoltaic Inverters

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ENERGISE ProDROMOS Project

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Programmable Distribution Resource Open Management Optimization System
(ProDROMOS)¹

Create an Advanced Distribution Management System (ADMS) that:

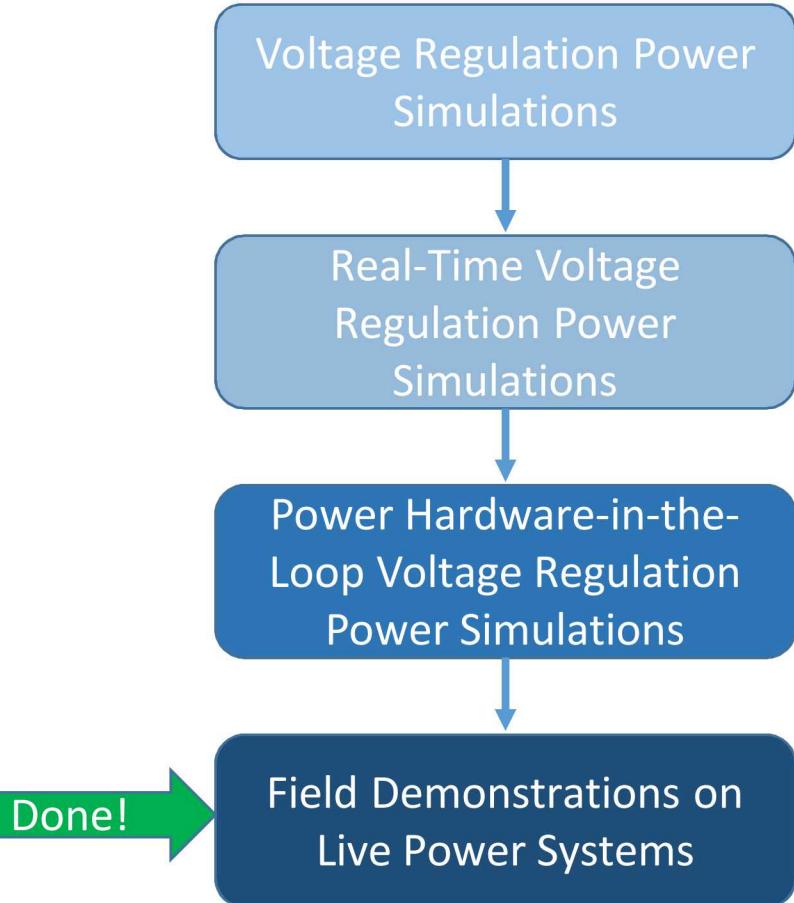
- captures distribution circuit telemetry
- performs state estimation, and
- issues optimal DER setpoints based on PV production forecasts.

Implemented on a live power system using 684kW PV system

Compared three control strategies: autonomous, central optimization, distributed optimization

Adopted by Connected Energy (ADMS vendor)

¹Prodromos is Greek for "forerunner" and the prodromoi were a light cavalry army unit in ancient Greece used for scouting missions.



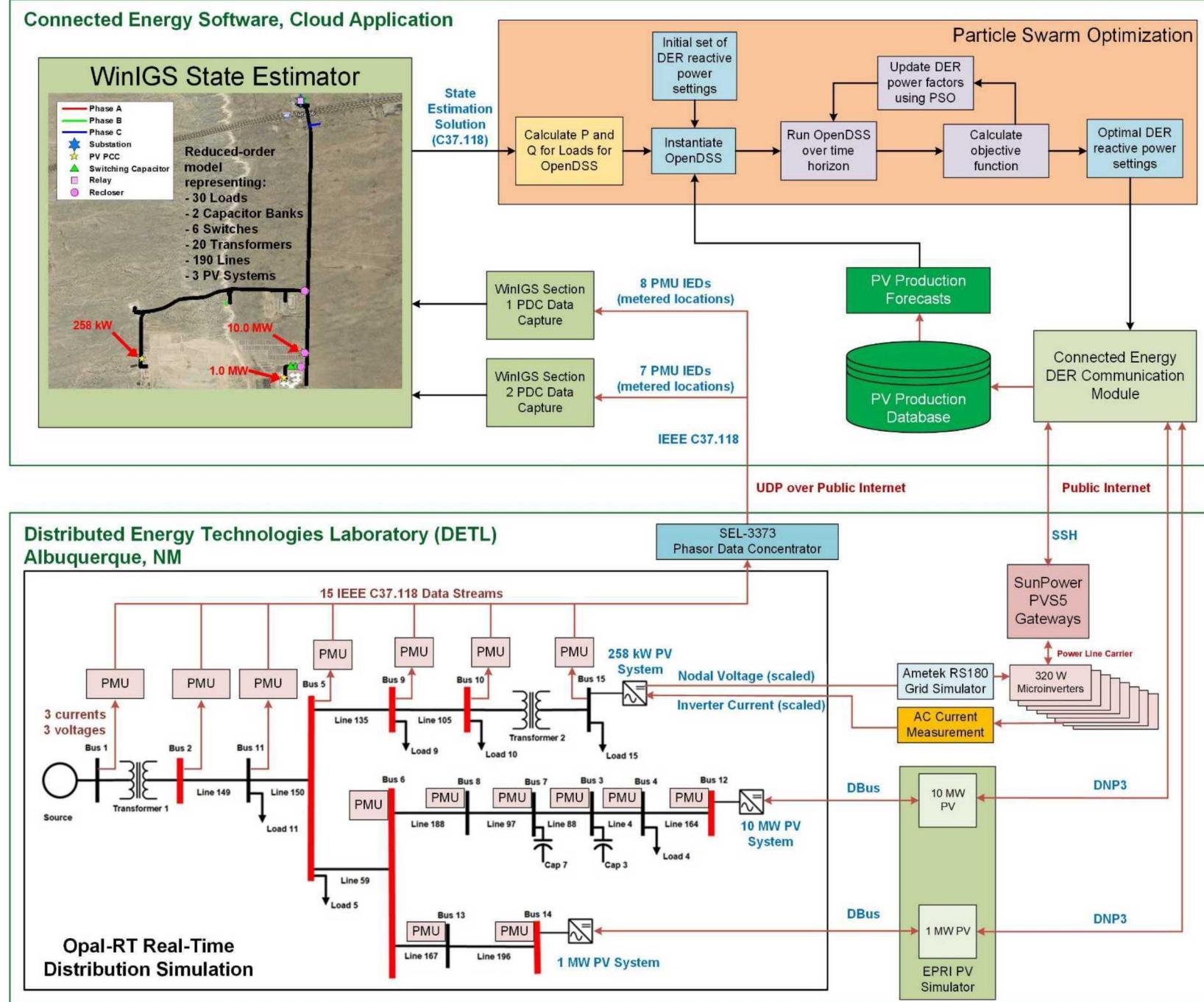
IMPLEMENTATION

Opal-RT Communication Interfaces

- PMU C37.118 to state estimator
- Opal DataBus Interface receives P/Q values for EPRI PV Simulators and transmits bus voltages and frequency

Information Flow

- The State Estimator ingests PMU data to produce current/voltage estimates for the distribution system
- State estimation data and PV generation forecasts populate an OpenDSS model.
- PSO wraps the OpenDSS model to calculate the optimal PF setpoints for each of the DER devices.
- DER PF settings are issued through proprietary SSH commands and IEEE 1815 (DNP3) commands



DIGITAL TWIN CONCEPT

Problem

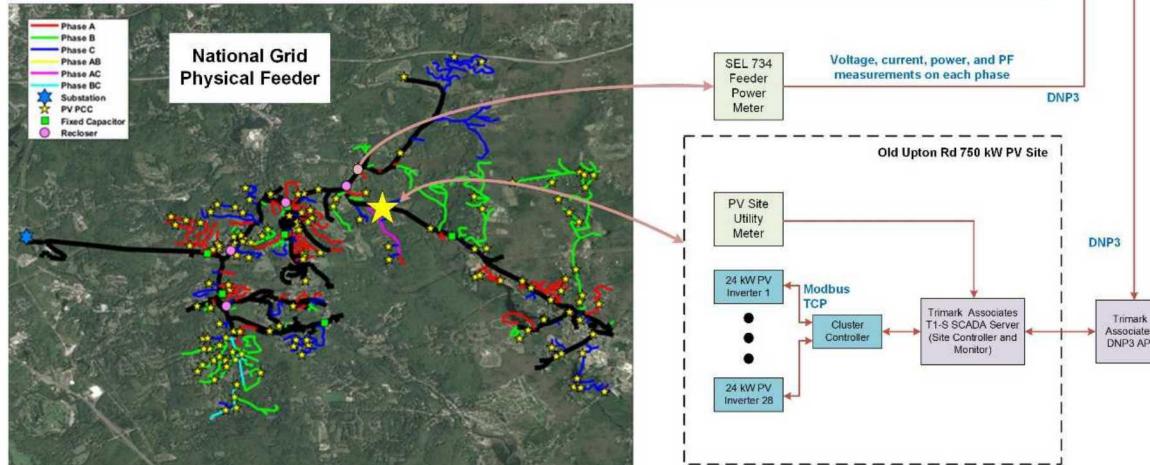
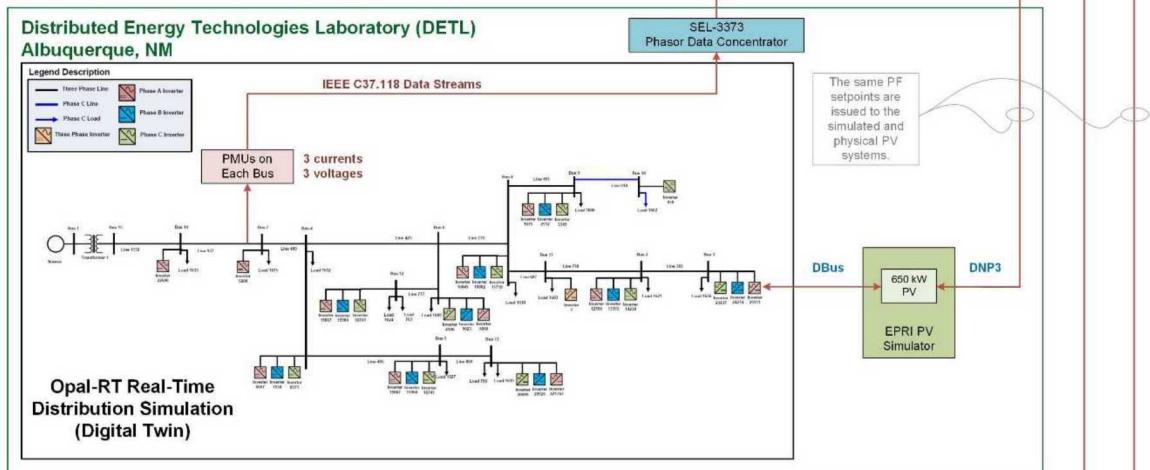
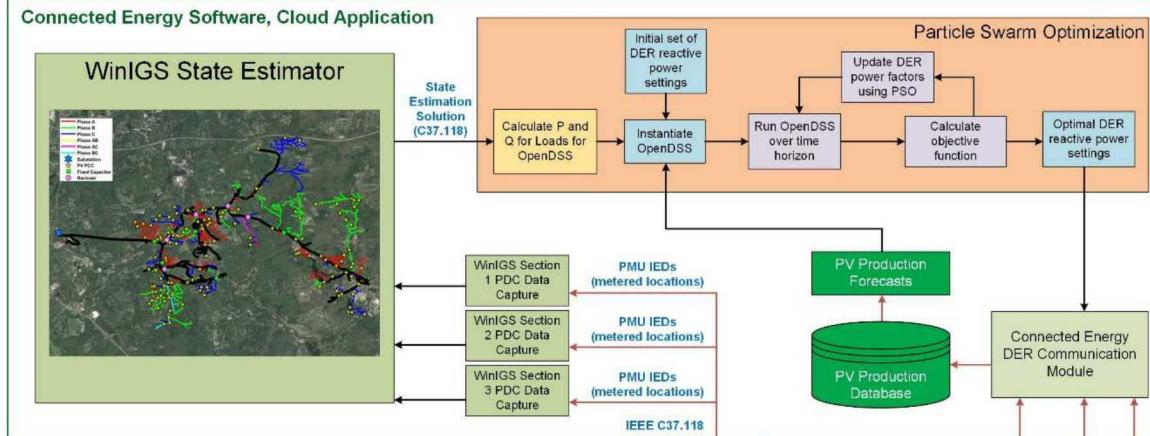
- Not enough Intelligent Electronic Devices (IEDs, i.e., PMUs, DERs, meters, etc.) to make state estimation observable for the field demonstration
- Short-term load forecasts or historical data is often used as “pseudo-measurements” to get a solution, but the team doesn’t have access to this data

Proposal

- Use a real-time digital twin of the feeder to estimate the system operations
 - If general behavior of digital twin is similar to the physical feeder, the “optimal” PF settings should support feeder voltages
- PV PF setpoints are sent to the physical and virtual PV system

Challenges

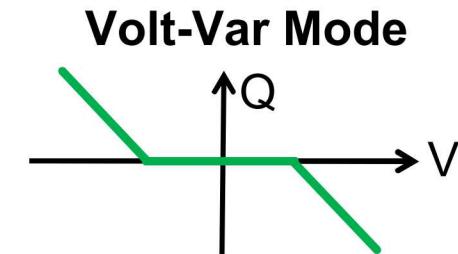
- This does not account for the current load (only pre-recorded versions)



▪ Distributed Autonomous Control

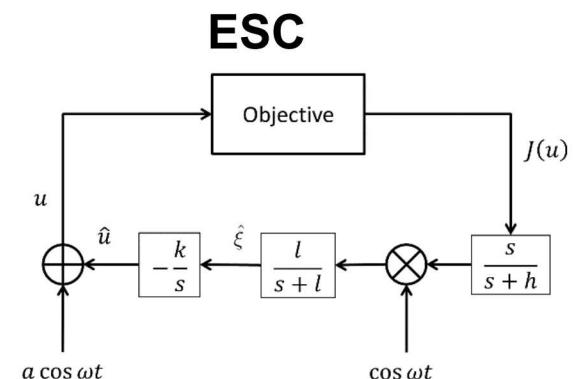
- Function: volt-var
- Pros: Simple, requires little or no communications, DER locations not needed
- Cons: does not reach global optimum

Note: rather 'gentle' volt-var profile in this evaluation



▪ Extremum Seeking Control (ESC)

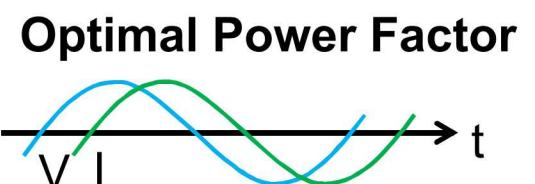
- Function: new grid-support function
- Pros: can achieve global optimum
- Cons: requires fitness function broadcast (with new inverter function), careful selection of parameters



▪ Particle Swarm Optimization (PSO)

- Function: power factor or reactive power commands
- Pros: direct influence over DER equipment to achieve objective
- Cons: requires telemetry, knowledge of DER locations, and state estimator/feeder model

Note: Forecasting tool estimates PV power production



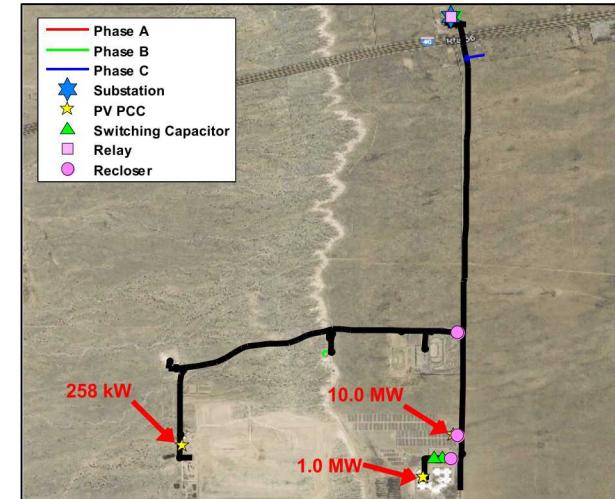
DISTRIBUTION CIRCUIT FEEDER MODEL

OpenDSS models were converted to reduced-order RT-Lab models

The PNM feeder has ~440% PV penetration because of large utility-scale PV systems.

Lines #	Transformers #	Loads #	Buses #	Voltage (V)	Load Power		PV Power (kVA)
					Active (kW)	Reactive (kVAR)	
12	2	14	15	7200/277	2568.63	1418.71	11258.00

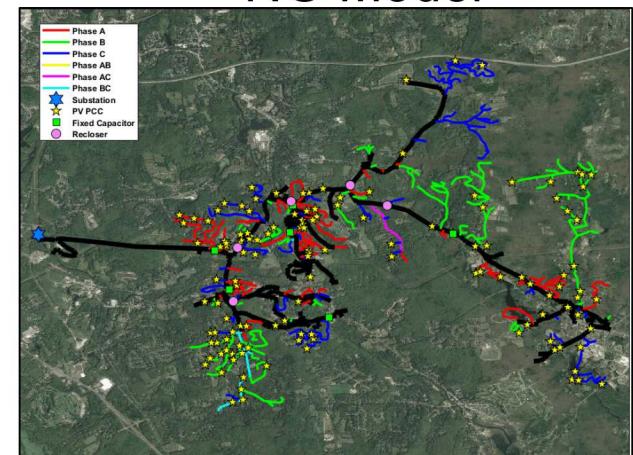
PNM Model



The NG feeder has 50% penetration chiefly as distributed PV.

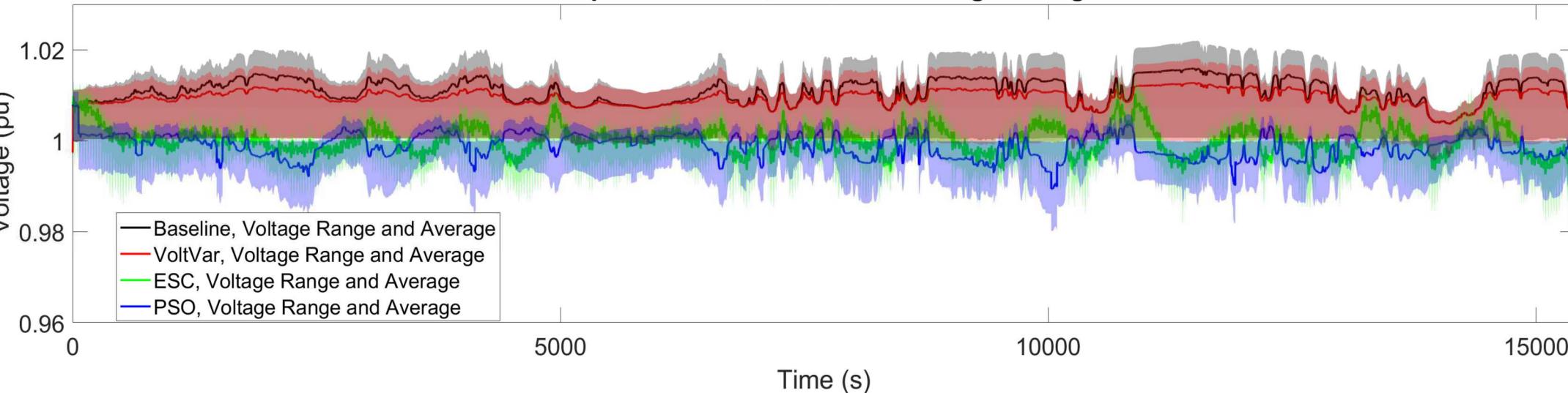
Lines #	Transformers #	Loads #	Buses #	Voltage (V)	Load Power		PV Power (kVA)
					Active (kW)	Reactive (kVAR)	
13	3	43	15	8000	9494.76	318.10	5495.36

NG Model

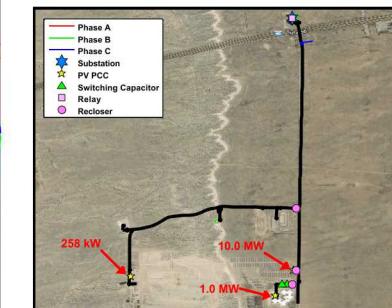


COMPARISON OF VOLTAGE REGULATION APPROACHES

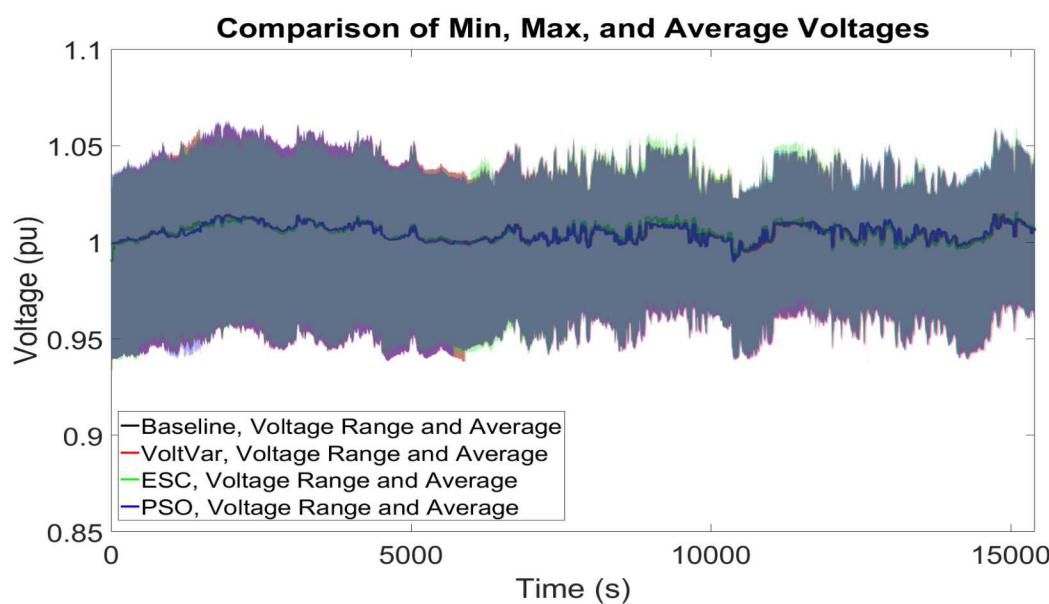
Comparison of Min, Max, and Average Voltages



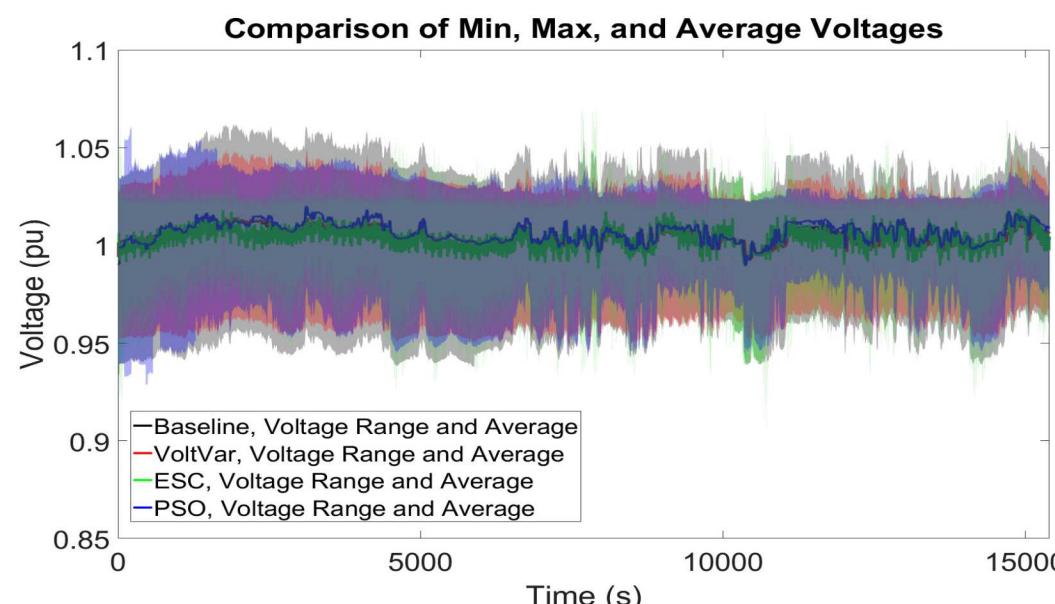
PNM Model



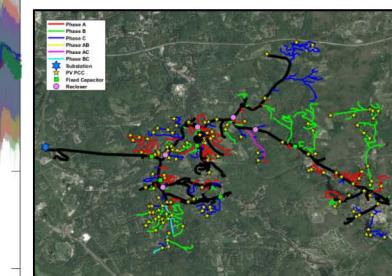
Single PV Control



All PV Control



NG Model



IMPACT OF EACH VOLTAGE REGULATION TECHNIQUE

A scoring approach is used to measure the effectiveness of each voltage regulation technique:

$$score = \frac{1}{T} \int_{t=0}^{t_{end}} \sum_{b=1}^N (|v_{bl} - v_{nom}| - |v_{reg} - v_{nom}|) dt$$

where,

v_{bl} : Baseline Voltage

v_{nom} : Target Voltage

v_{reg} : Voltage with control applied

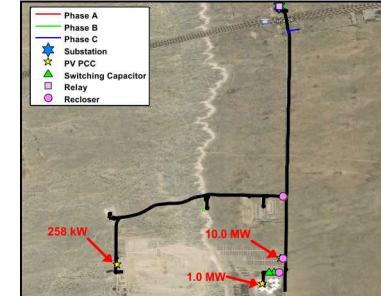
T : Time Period

b : bus

t : time

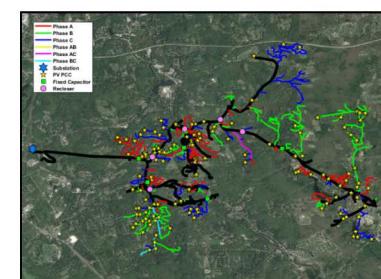
West Feeder Score					
	Phase A	Phase B	Phase C	Average	Improvement (%)
VV	0.024	0.024	0.024	0.071	12.9%
ESC	0.140	0.140	0.132	0.412	74.5%
PSO	0.139	0.139	0.130	0.408	73.7%
Best Score	0.186	0.188	0.179	0.553	

PNM Model



East Feeder Score Controlling a Single PV					
	Phase A	Phase B	Phase C	Average	Improvement (%)
VV	0.000	0.000	0.000	-0.001	0.0%
ESC	0.012	0.000	0.031	0.043	3.2%
PSO	-0.001	0.000	0.004	0.002	0.2%
Best Score	0.194	0.635	0.507	1.336	

NG Model



East Feeder Score Controlling All PV					
	Phase A	Phase B	Phase C	Average	Improvement (%)
VV	-0.004	0.122	0.085	0.203	15.2%
ESC	-0.023	0.328	0.202	0.508	38.0%
PSO	-0.023	0.124	0.137	0.238	17.8%
Best Score	0.194	0.635	0.507	1.336	

- **Incremental development approach was effective** (simulation to real time to PHIL to field)
 - Communications between measurement equipment, ADMS controllers, and DER devices can be verified.
 - Build confidence in controls before field deployment.
- **Digital twin was necessary during development** to overcome sparse measurements for state estimation
- **Observations about control options**
 - **Volt-var** functionality provides some DER voltage regulation without communications.
 - In low communication environments, **extremum seeking control** is a viable means to control a fleet of DER devices to track toward optimal PF setpoints, but it is relatively slow and the system must be tolerant of probing signal ripple.
 - State estimation-fed, model-based **DER optimization** is a viable control strategy with sufficient telemetry.
- **Open question, and observations:**
 - How well could negative-sequence inverters regulate voltage on unbalanced feeders?
 - Available telemetry and communications will rarely supply what is assumed during ADMS development
 - Software interoperability continues to be challenging

Thank You!

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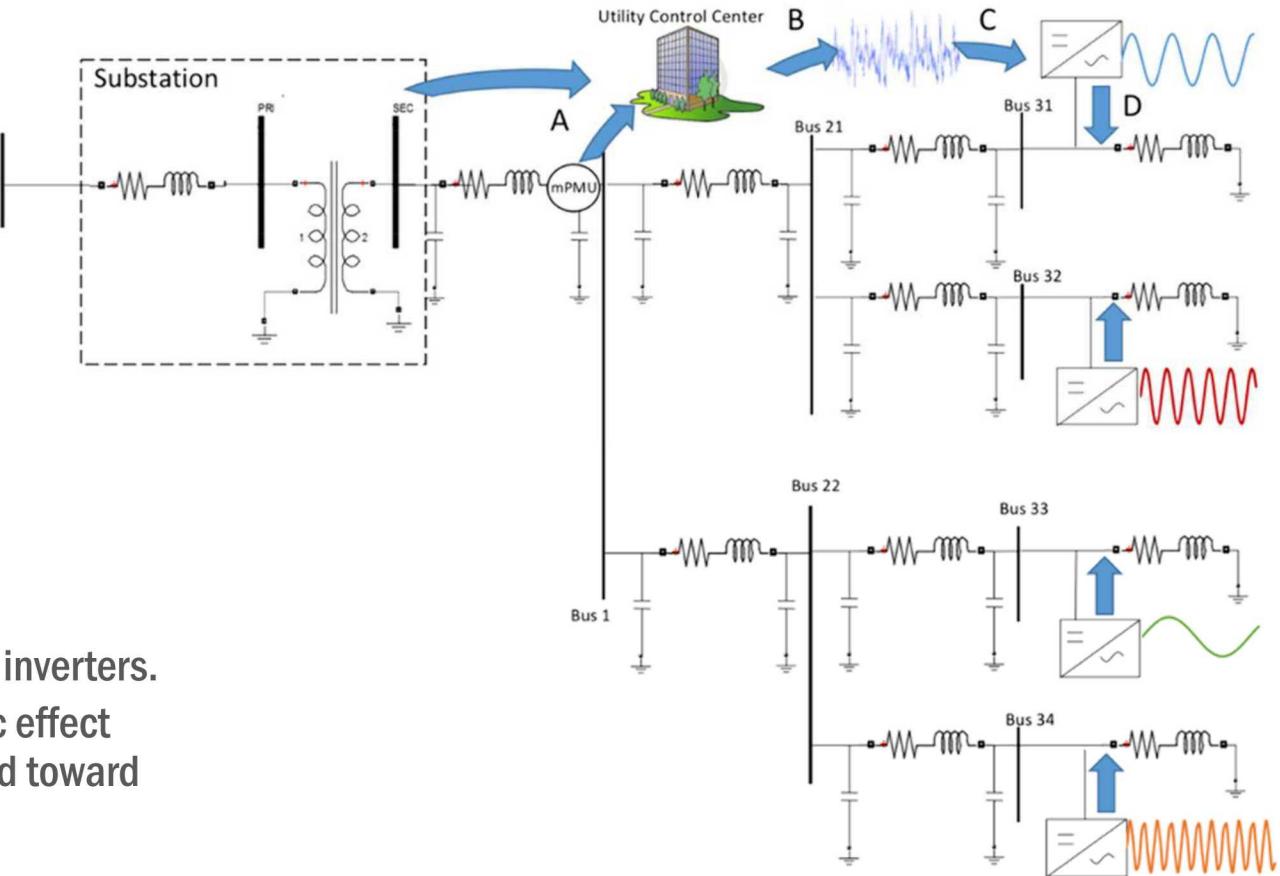


ESC

Extremum Seeking Control will be used as a comparison to the PF optimization technique

Steps in ESC:

- A. Centralized control center collects data from the power system
- B. Control center calculates the objective function, e.g., $J = 1/n * \sum [(V_i - V_n)/V_n]^2$
- C. Control center broadcasts objective function to all inverters.
- D. Individual inverters extract their frequency-specific effect on the objective function and adjust output to trend toward the global optimum.



- D. B. Arnold, M. Negrete-Pincetic, M. D. Sankur, D. M. Auslander and D. S. Callaway, "Model-Free Optimal Control of VAR Resources in Distribution Systems: An Extremum Seeking Approach," IEEE Transactions on Power Systems, vol. 31, no. 5, pp. 3583-3593, Sept. 2016.
- J. Johnson, R. Darbali, J. Hernandez-Alvidrez, A. Summers, J. Quiroz, D. Arnold, J. Anandan, "Distribution Voltage Regulation using Extremum Seeking Control with Power Hardware-in-the-Loop," IEEE Journal of Photovoltaics, vol. 8, no. 6, pp. 1824-1832, 2018.
- J. Johnson, S. Gonzalez, and D.B. Arnold, "Experimental Distribution Circuit Voltage Regulation using DER Power Factor, Volt-Var, and Extremum Seeking Control Methods," IEEE PVSC, Washington, DC, 25-30 June, 2017.
- D. B. Arnold, M. D. Sankur, M. Negrete-Pincetic and D. Callaway, "Model-Free Optimal Coordination of Distributed Energy Resources for Provisioning Transmission-Level Services," in IEEE Transactions on Power Systems, vol. 33, no. 1, pp. 817-828, 2017.
- Code: https://github.com/sunspec/prodromos/blob/master/optimization/extremum_seeking_control.py

PF OPTIMIZATION

- Optimization occurs every minute over a 15-min horizon
- OpenDSS simulation is instantiated with PV production forecast and current feeder status (which is assumed to persist)
 - State-estimation determines current feeder loads
 - Forecasting tool estimates PV power production
- Particle Swarm Optimization (PSO) is used to determine the optimal PF settings for the DER devices because of nonconvex fitness landscape

Objective Function:

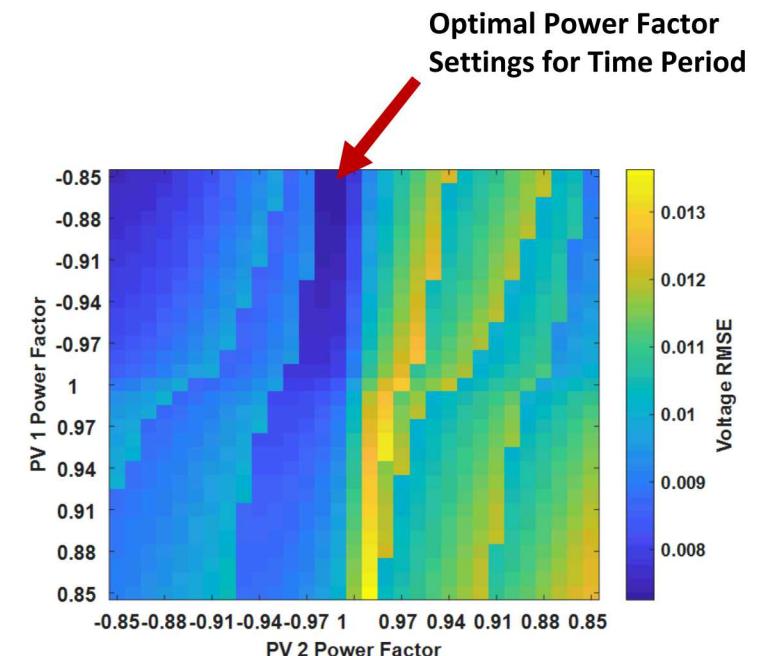
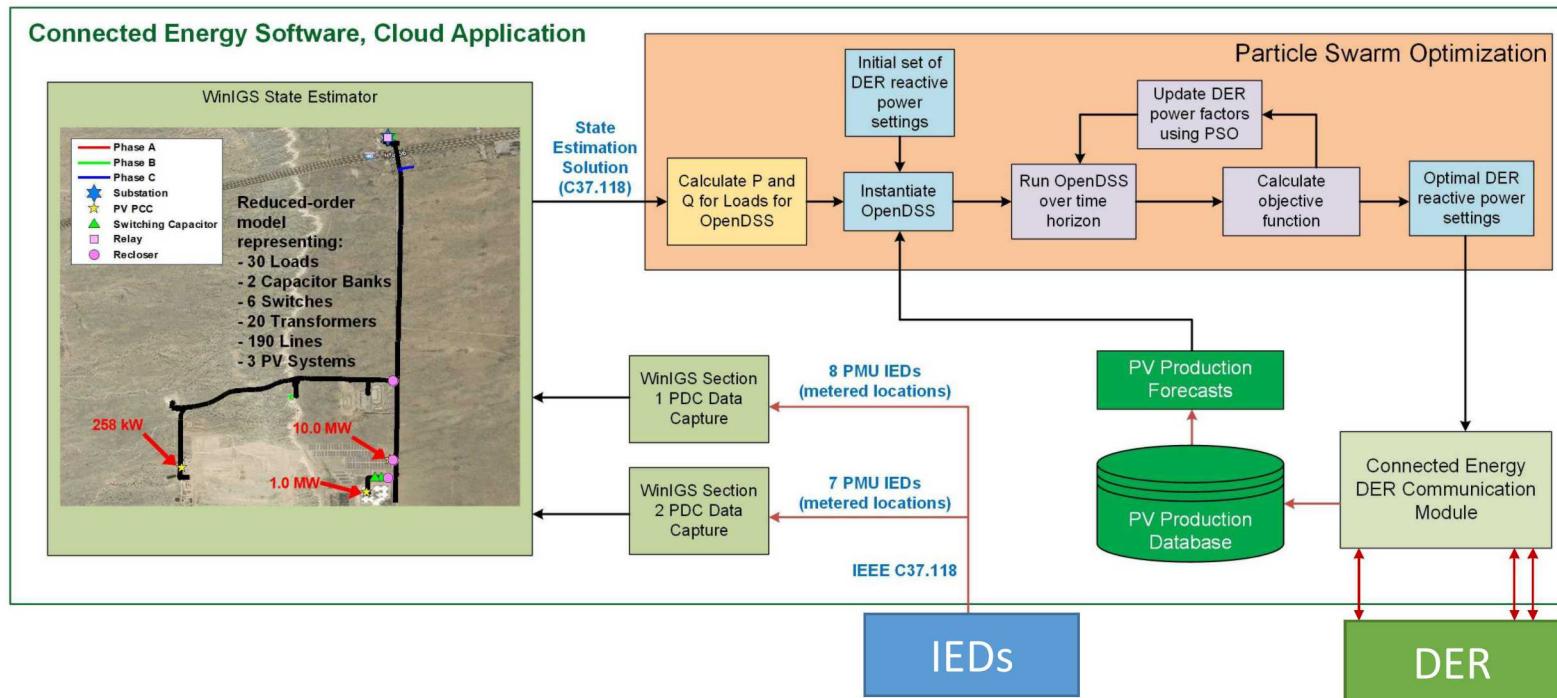
$$\min_{PF} w_0 \delta_{violation}(V) + w_1 \sigma(V - V_{base}) + w_2 C(PF)$$

$$\delta_{violation}(V) = 1 \text{ if any } |V| > V_{lim}$$

$$\sigma(V - V_{base}) \text{ is standard deviation of } V - V_{base}$$

$$C(PF) = \sum 1 - |PF|$$

Cost minimized when voltage = V_{base} and $PF=1$



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