



Cornell University



*Development and Testing of New Tools*

# Tools for Secure Planning and Operations of Systems with Stochastic Sources, Energy Storage and Active Demand

Ray Zimmerman, Carlos Murillo-Sánchez, Bob Thomas

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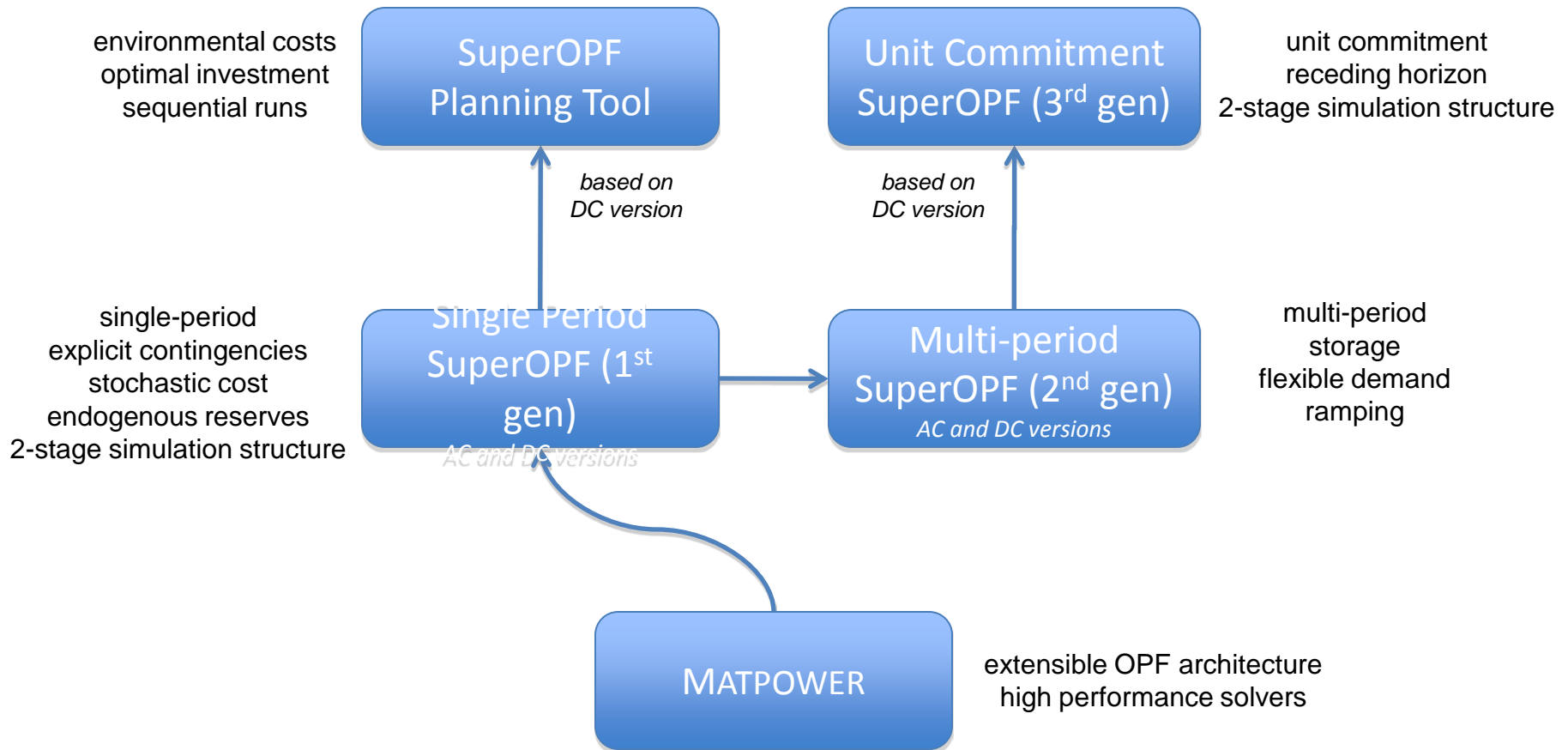
# Other Contributors

- Input scenario creation and analysis
  - **C. Lindsay Anderson** (Cornell)
  - **Amandeep Gupta** (Cornell)
  - **Jung Mo** (Cornell)
  - Tim Mount (Cornell)
  - Laurie Tupper (Cornell)
- Simulation environment and testing
  - **Daniel Muñoz-Álvarez** (Cornell)
  - Wooyoung Jeon (Cornell)
  - **Alberto Lamadrid** (Lehigh)
- MATPOWER
  - Shrirang Abhyankar (Argonne)
  - Alexander J. Flueck (IIT)
  - Daniel Molzahn (U Wisc/Madison)
  - MATPOWER user base (worldwide)

# Outline

- Overview of Tools
- MATPOWER
- SuperOPF Planning Tool
- Multiperiod SuperOPF (*2<sup>nd</sup> gen*)
  - *especially storage model*
- Simulation Environment
- Unit Commitment SuperOPF (*3<sup>rd</sup> gen*)
- Discussion of UC SuperOPF Implementation  
Preliminary Results

# Project Overview



# MATOWER

Free, open-source power system simulation environment with extensible OPF and interfaces to state-of-the-art solvers.

- bug fixes, performance enhancements, general maintenance
- used worldwide in teaching, research, consulting
- momentum growing
- 37,000 downloads of version 4.x
- 12,000 of those in the last year
- growing user support needs
- serves as foundation for all tools in this project

# MATPOWER

Near term (few months) plans, new release including ...

- accumulated enhancements, fixes since v4.1
- contributed code:
  - continuation power flow
    - contributed by Shrirang Abhyankar (Argonne), Alex Flueck, (IIT)
  - applications of SDP (semi-definite programming) relaxations to the OPF
    - contributed by Dan Molzahn (U of Wisc/Madison)
    - solver for SDP relaxation of OPF problem
    - sufficient condition for global optimality of specified OPF solution
    - sufficient conditions for insolvability of the power flow equations

# MATPOWER

Longer term (over next year) ...

- integrate 3<sup>rd</sup> generation SuperOPF into a new MATPOWER release
- assured wide distribution
- significant boost for other researchers
- increased visibility and opportunities for feedback
- LOTS of cleanup and documentation work required to make this work

# SuperOPF Planning Tool

- used extensively by R & M Project 2E:  
“Mapping Energy Futures: SuperOPF Planning Tool” (*Bill Schulze*)
- based on 1<sup>st</sup> gen (single-period) SuperOPF
  - coupled DC OPF of multiple scenarios
  - tied together by capacities that reflect investment/retirement
  - additional constraints, e.g. regional build limits



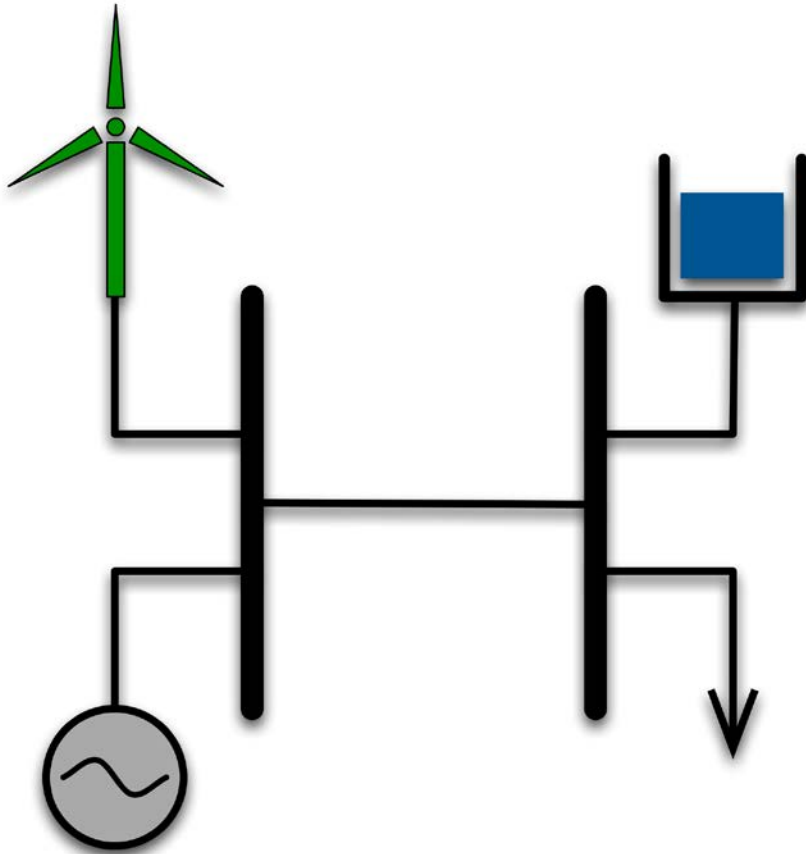
# SuperOPF Planning Tool

- modified formulation
  - added ability to specify scenario-specific availability factors
  - improved ability to model wind and solar
- greatly improved performance
  - via techniques to improve problem robustness enabling us to exploit the speed and scalability of interior point solvers
- current problem size
  - Eastern Interconnect with 73 representative hours
  - over 7 million variables, almost 19 million constraints
- looking toward integration of binary variables

# Multi-period SuperOPF (2<sup>nd</sup> gen)

- used extensively by R & M Project 2A: Evaluating Effects of Managing Controllable Demand & DER (*Tim Mount*)
- coupled OPF scenarios (wind and outage scenarios for multi-period horizon)
- linked within a period by reserve and redispatch vars/costs/constraints
- linked through time by storage and ramping vars/costs/constraints
- transitions from period-to-period, state-to-state governed by transition probability matrices
- implications
  - not tracking individual trajectories, only bounds on a “central path” (e.g. load following ramp)
  - not tracking actual amounts of stored energy, only storage state bounds for “central path”

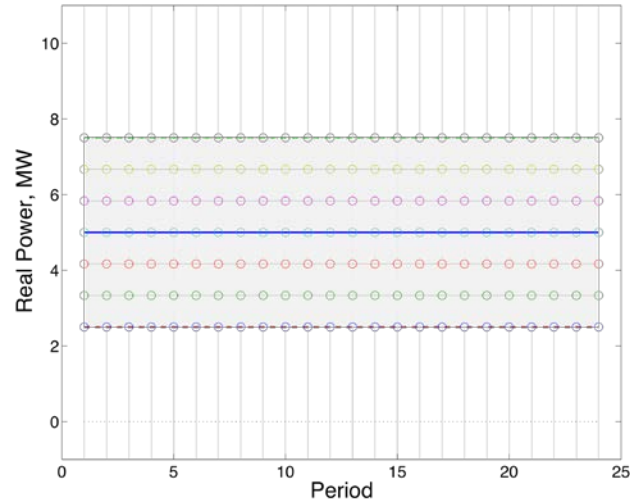
# Toy System for Illustration



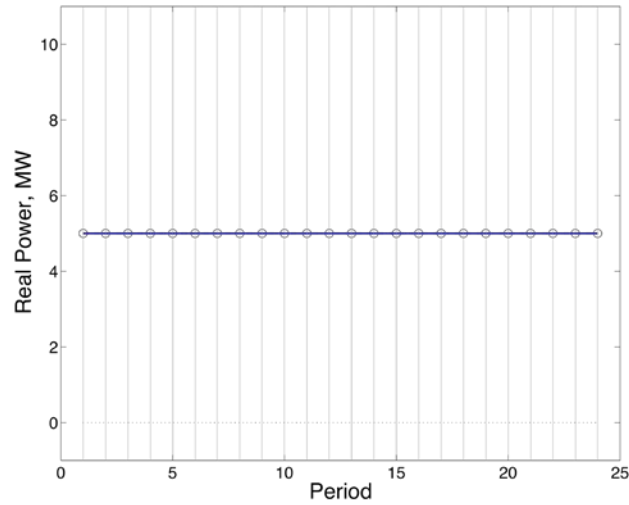
- conventional gen with quadratic cost
- constant deterministic load
- 100% efficient grid-level storage unit
- wind generator with normally distributed output, with 2 parameters:
  - variability
  - uncertainty

# Uncertainty & Variability

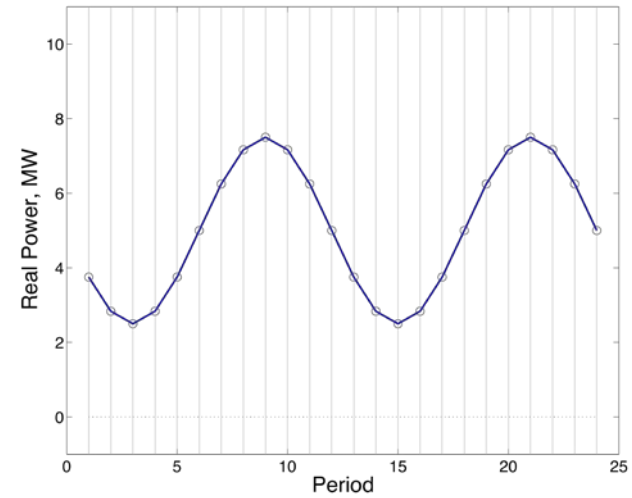
Uncertain



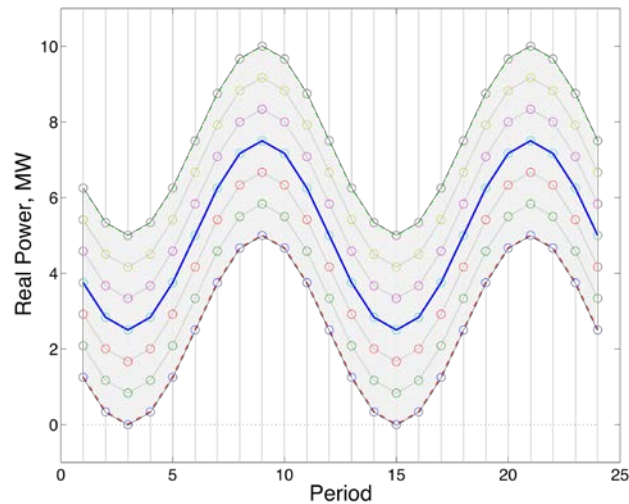
Constant and Deterministic



Variable

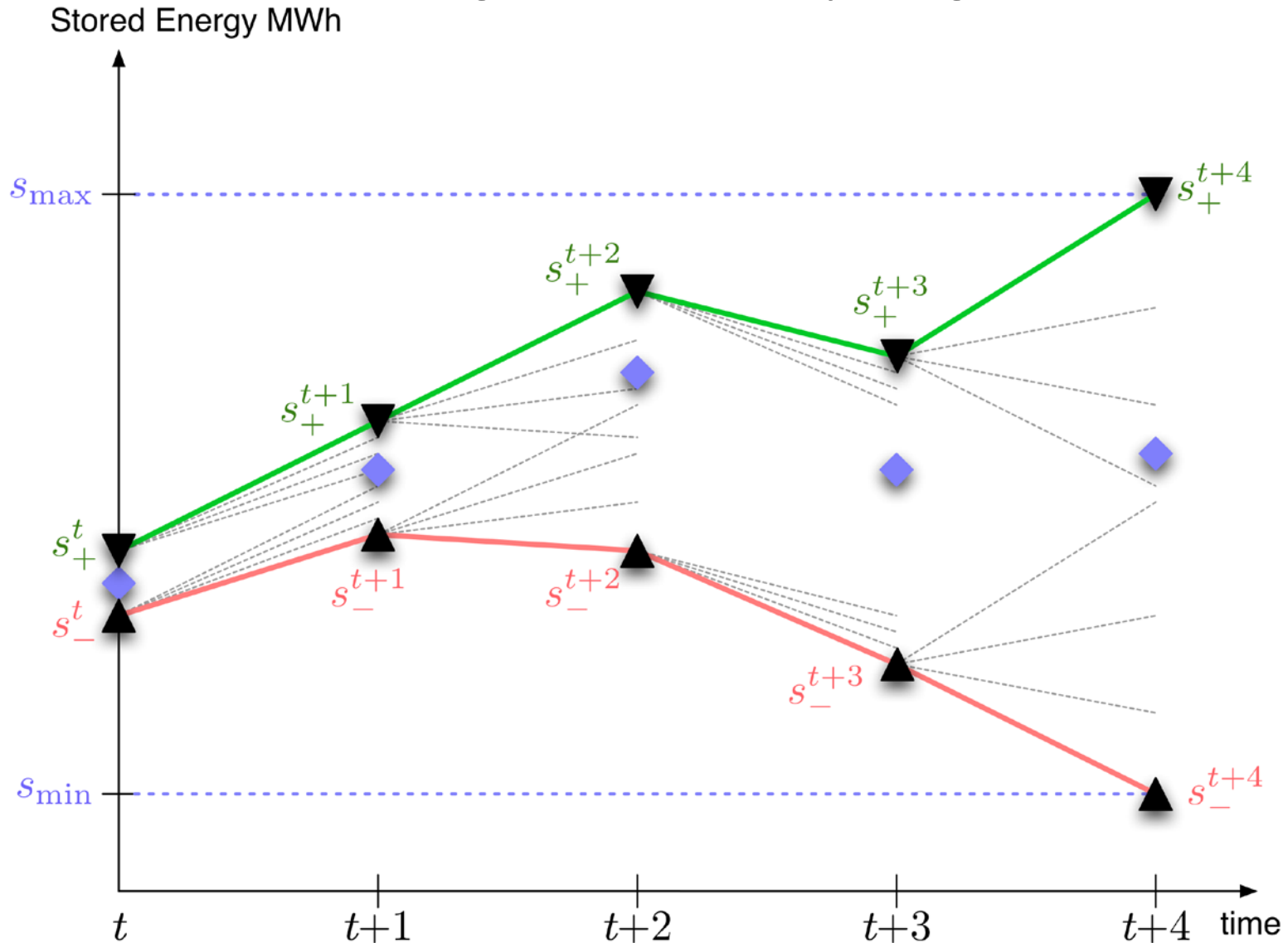


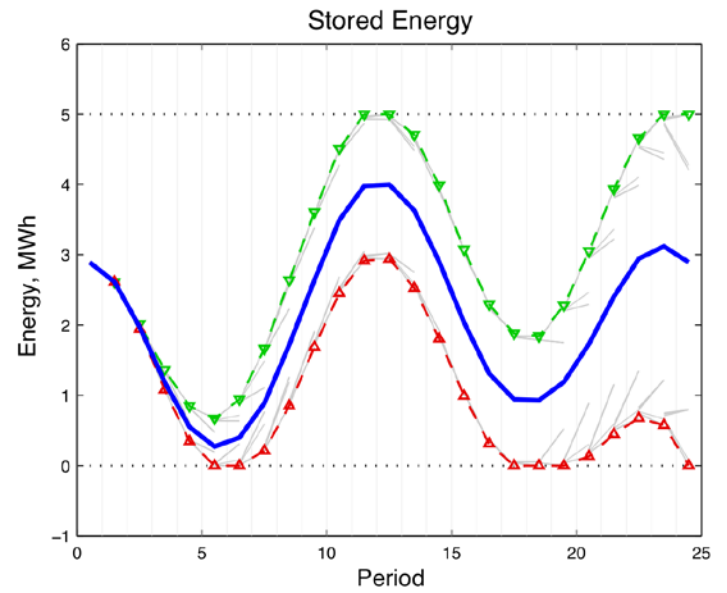
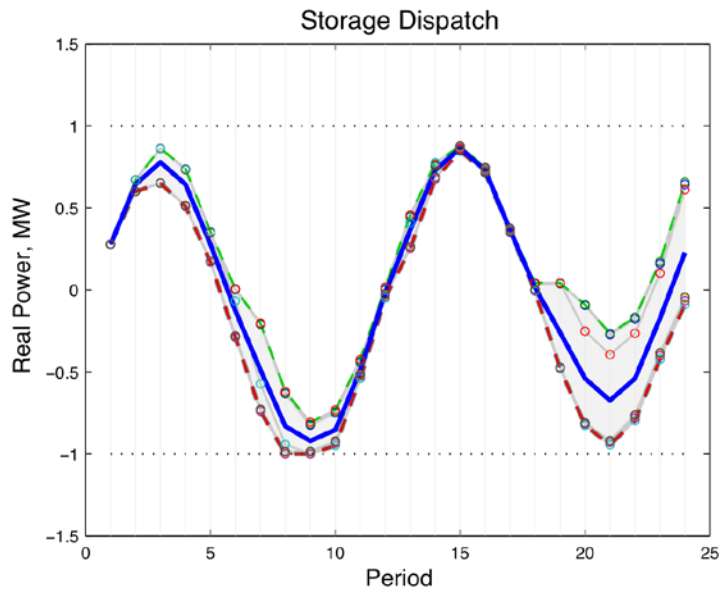
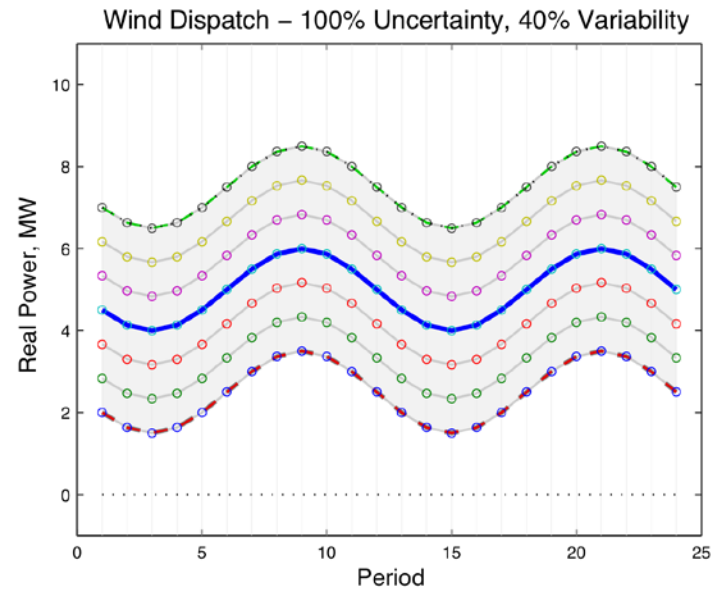
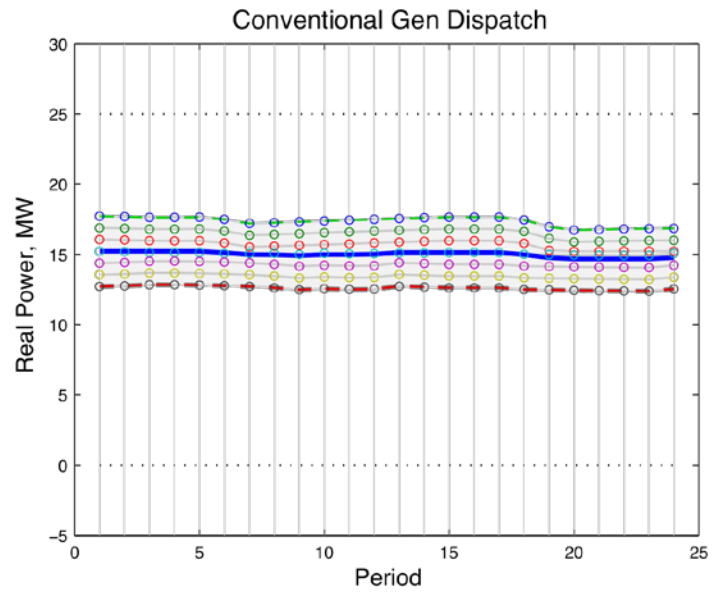
Variable and Uncertain



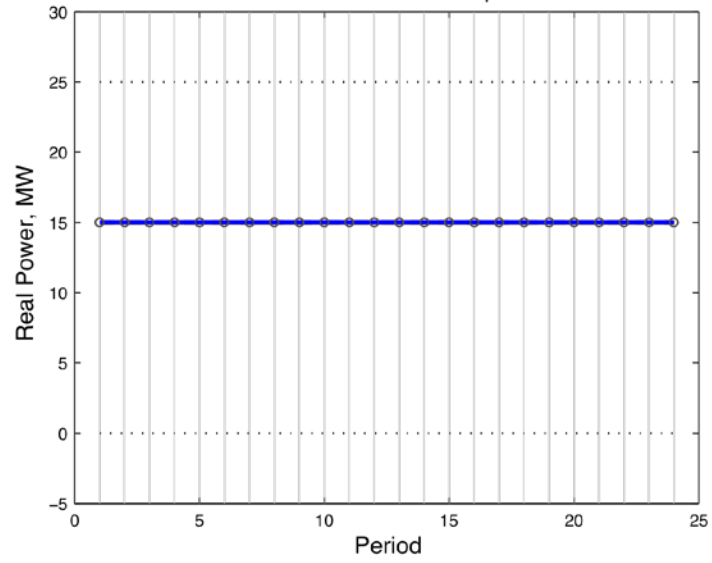
# Example of Tradeoff of Storage Usage

*time arbitrage vs. uncertainty mitigation*

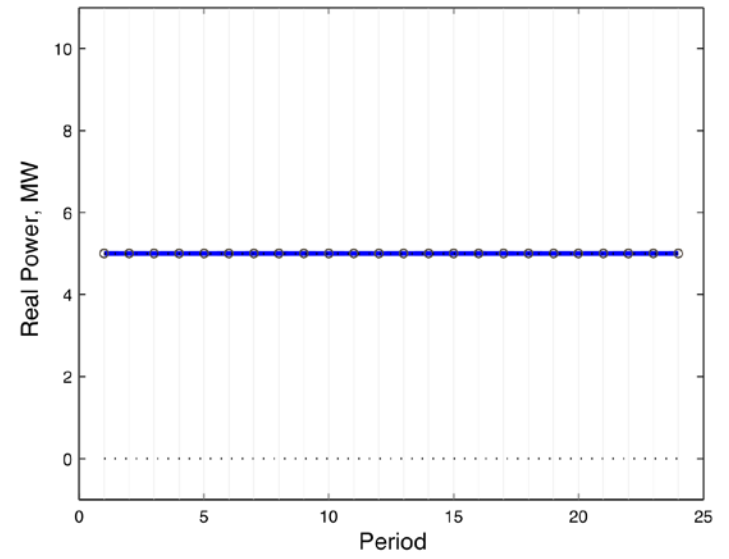




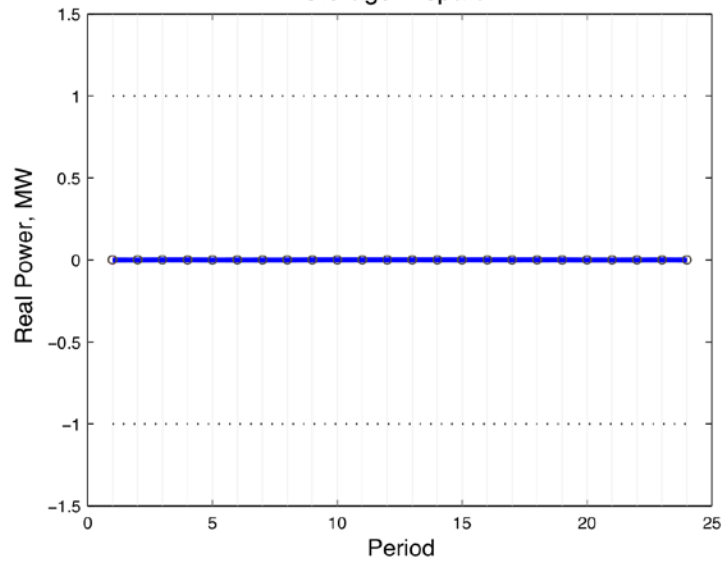
Conventional Gen Dispatch



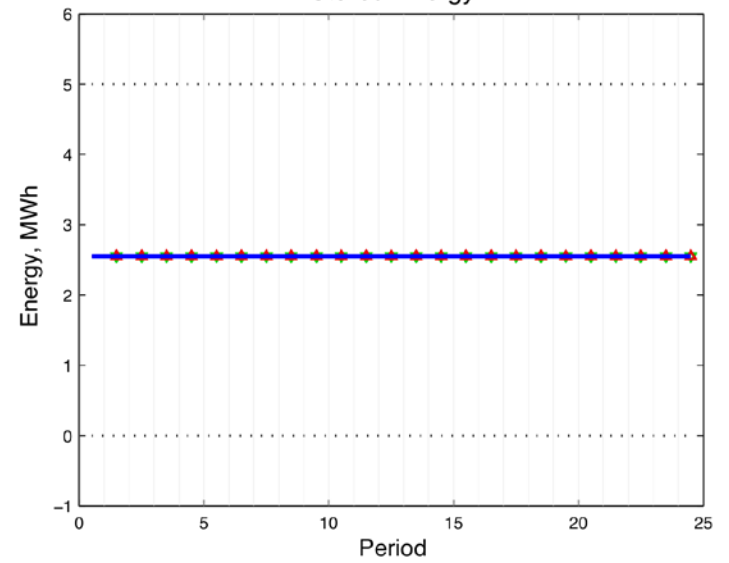
Wind Dispatch – 0% Uncertainty, 0% Variability

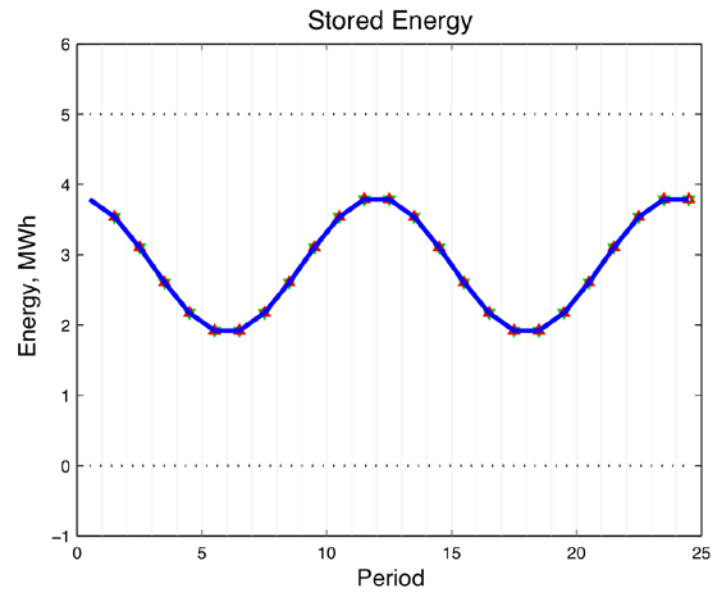
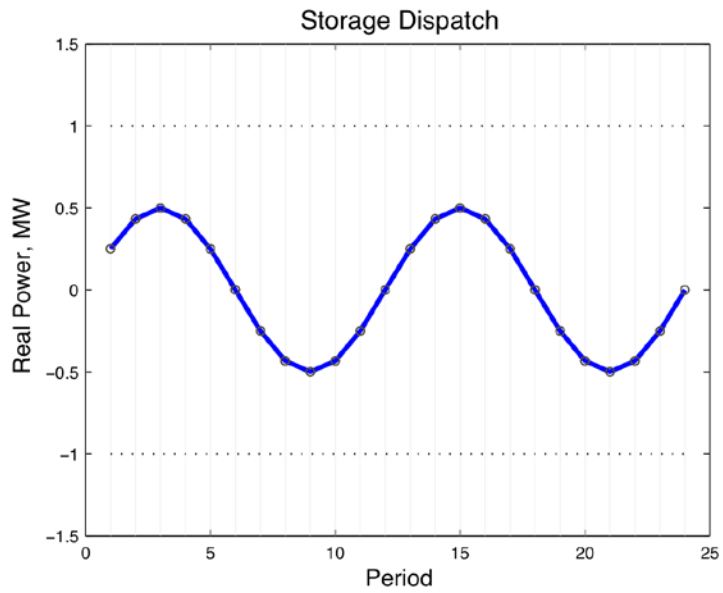
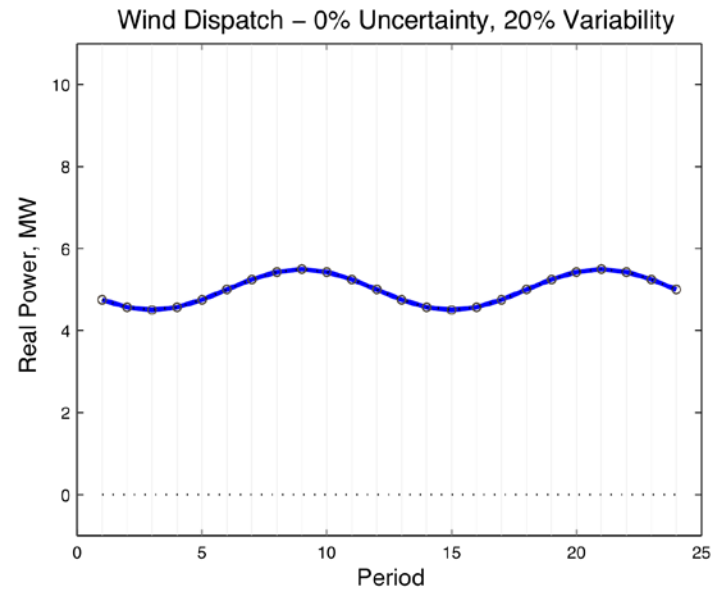
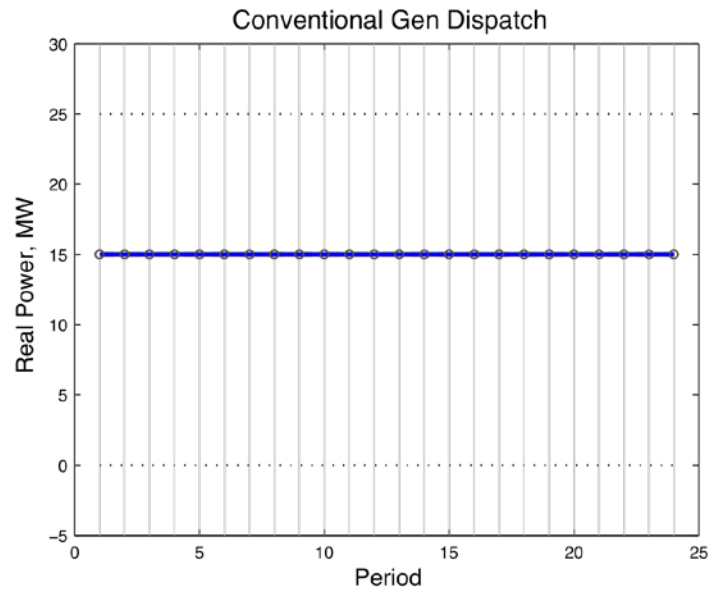


Storage Dispatch

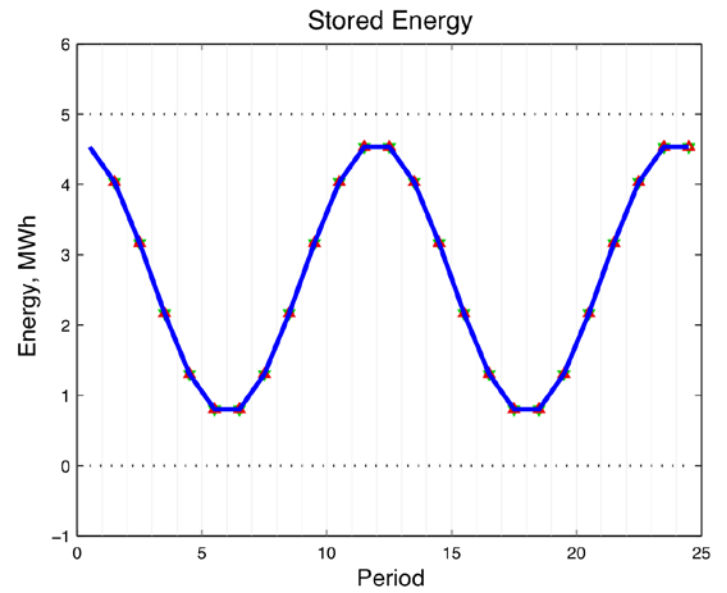
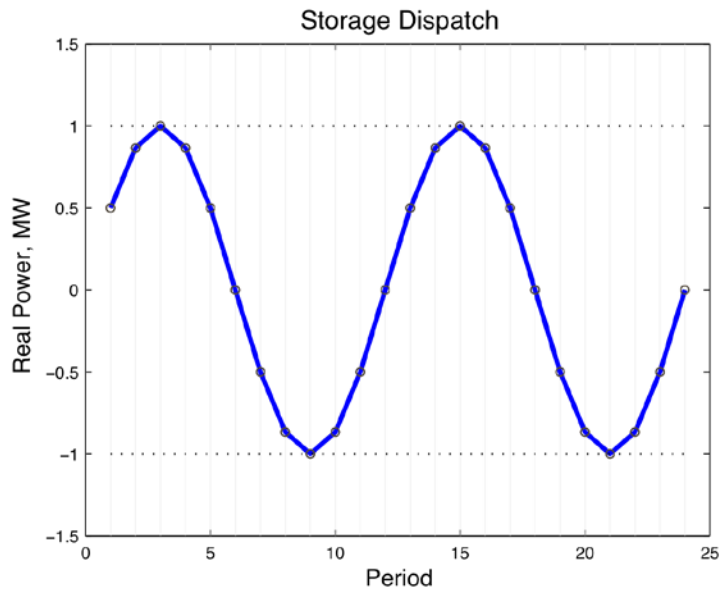
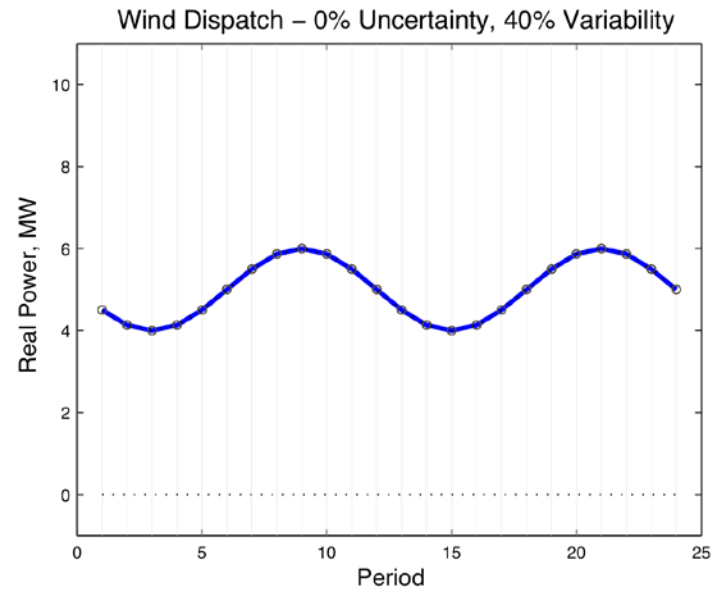
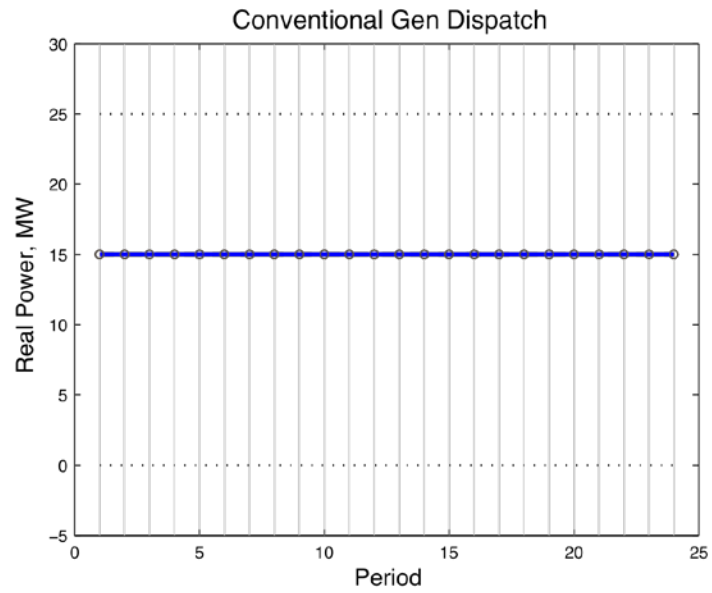


Stored Energy

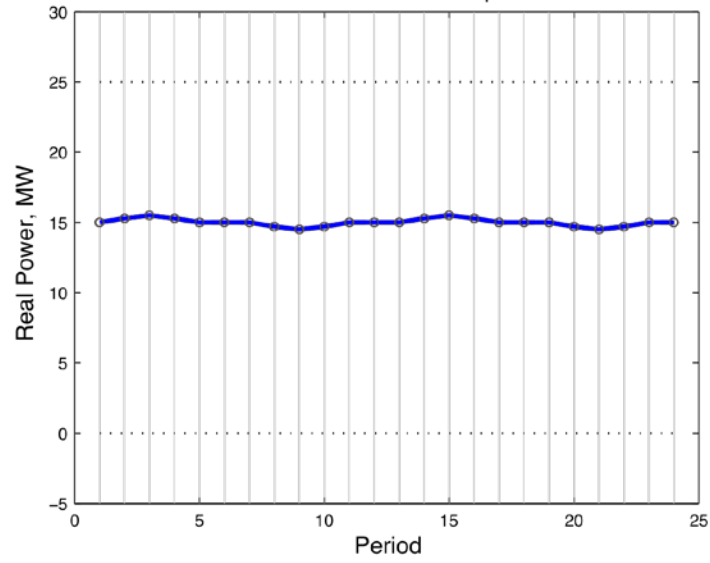




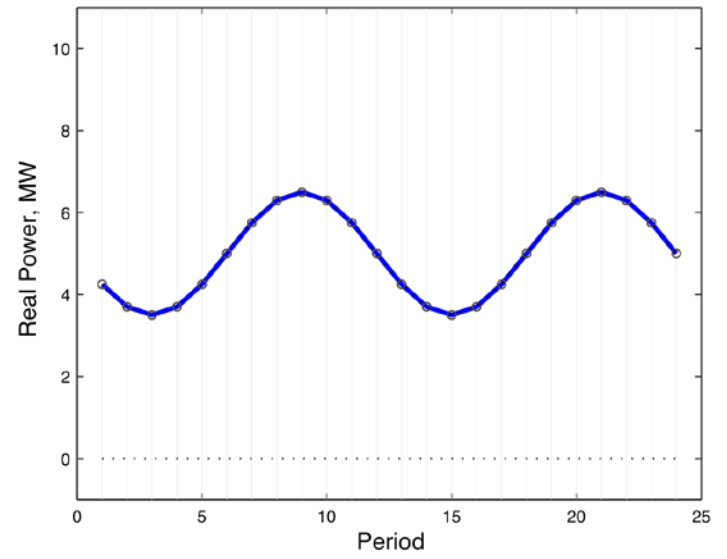




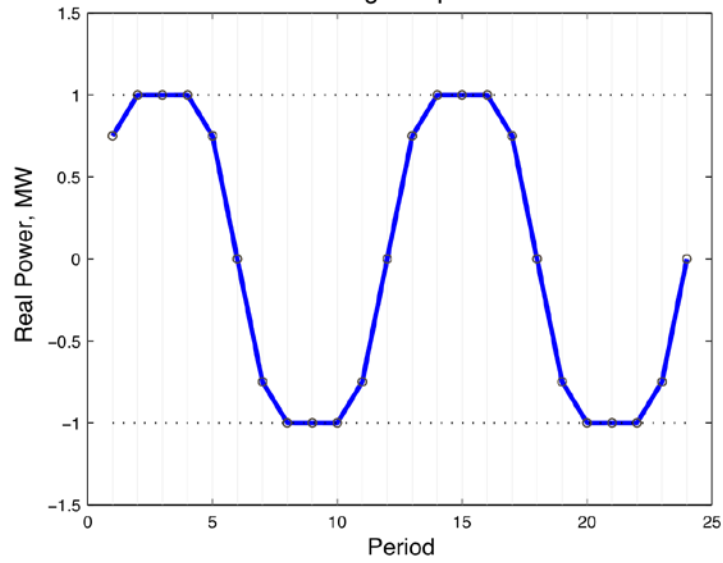
Conventional Gen Dispatch



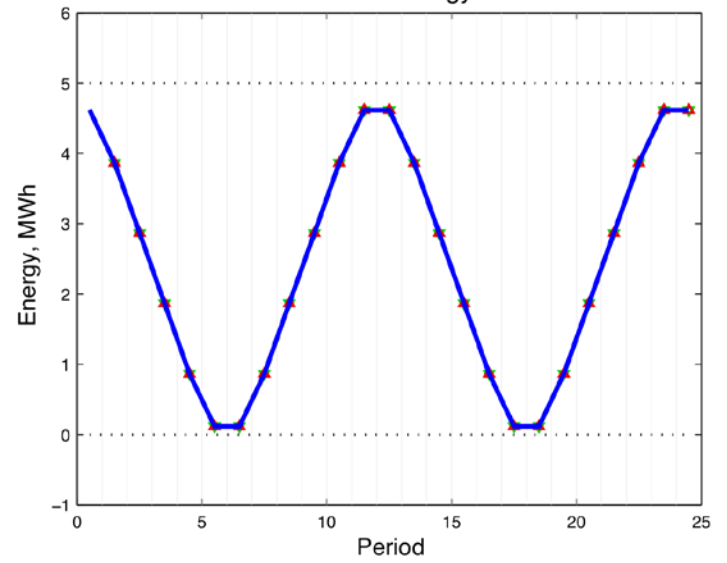
Wind Dispatch – 0% Uncertainty, 60% Variability

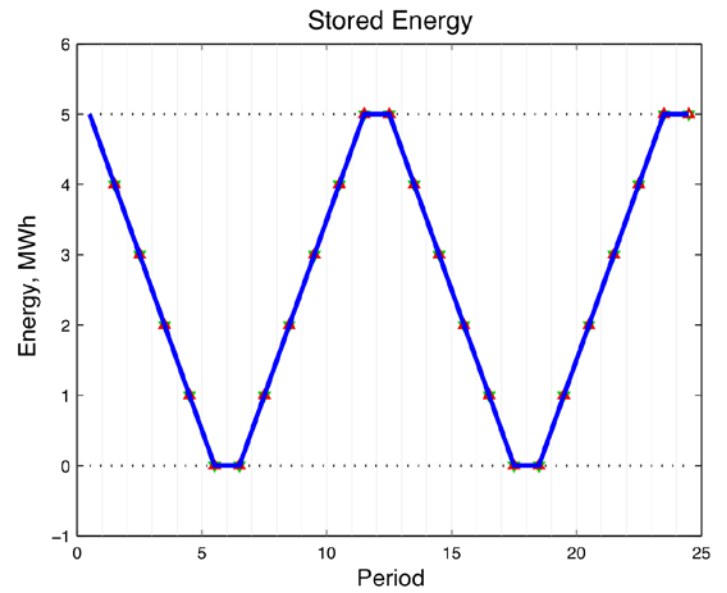
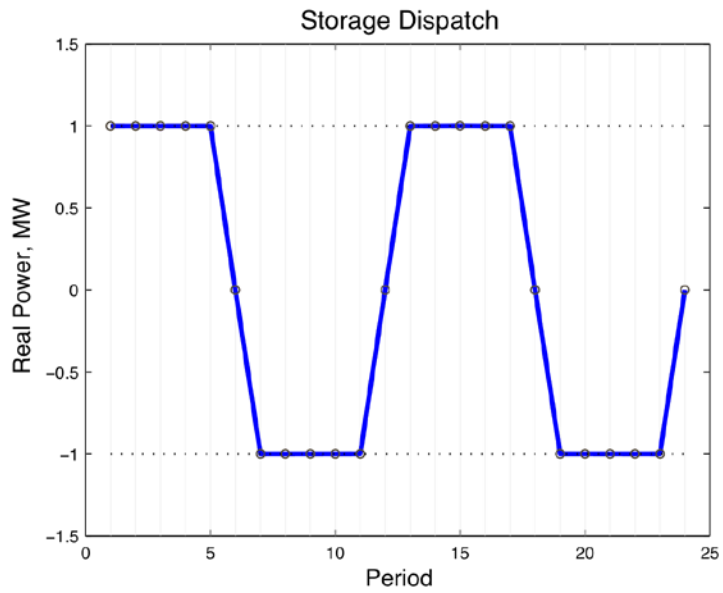
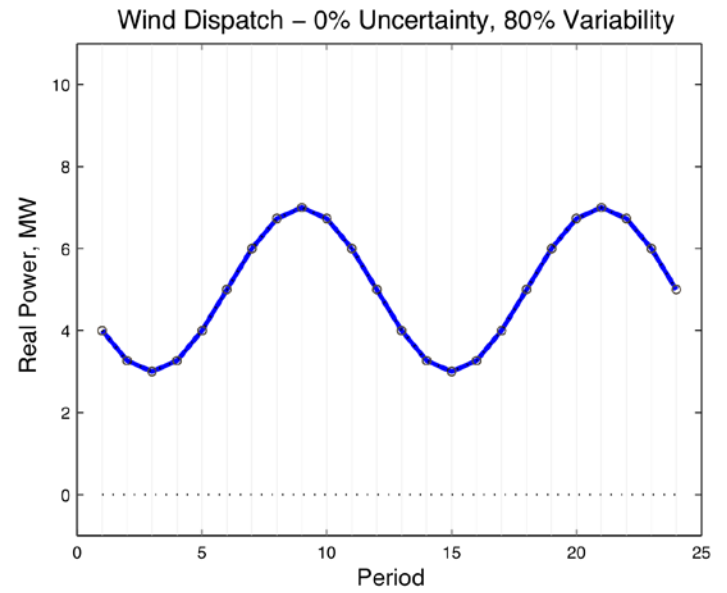
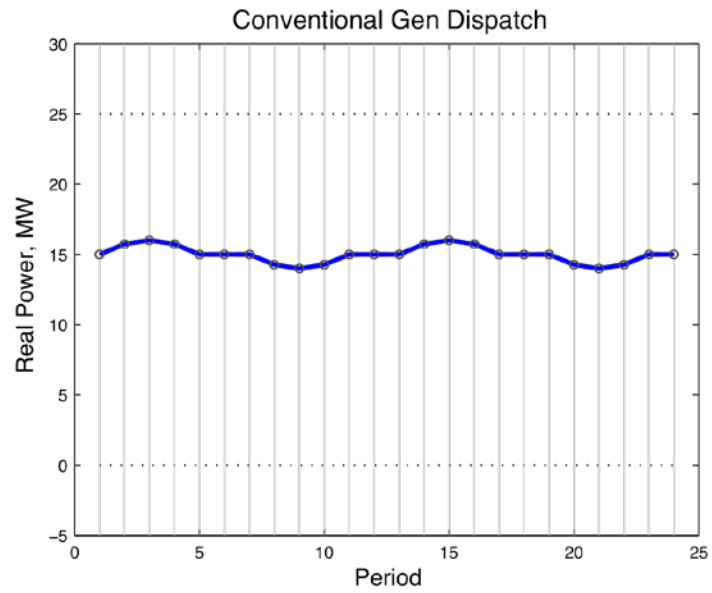


Storage Dispatch

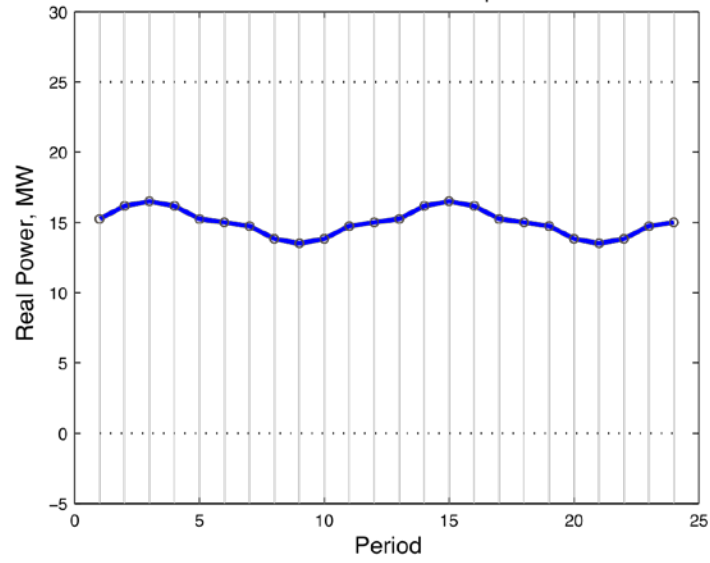


Stored Energy

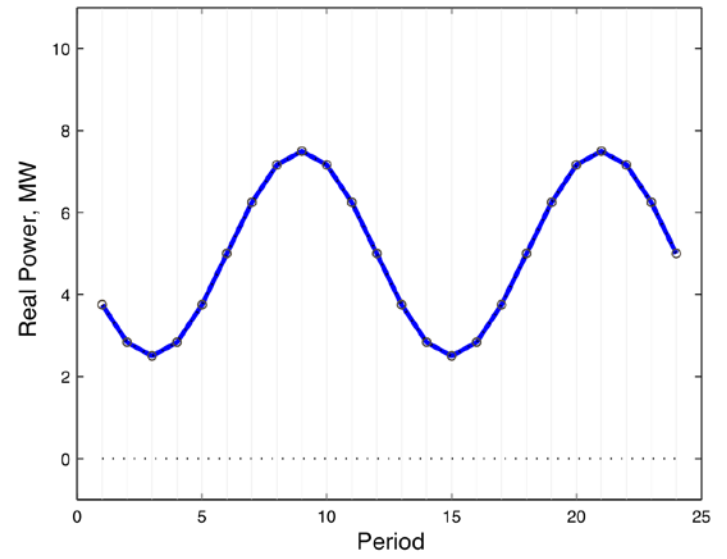




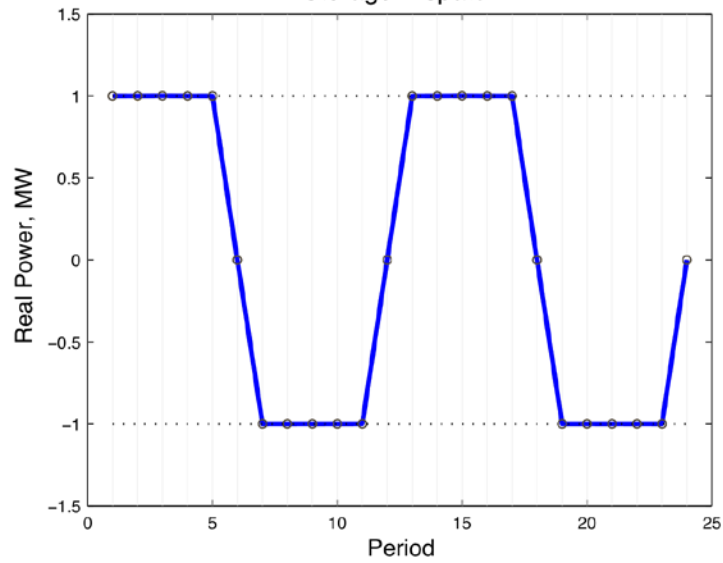
Conventional Gen Dispatch



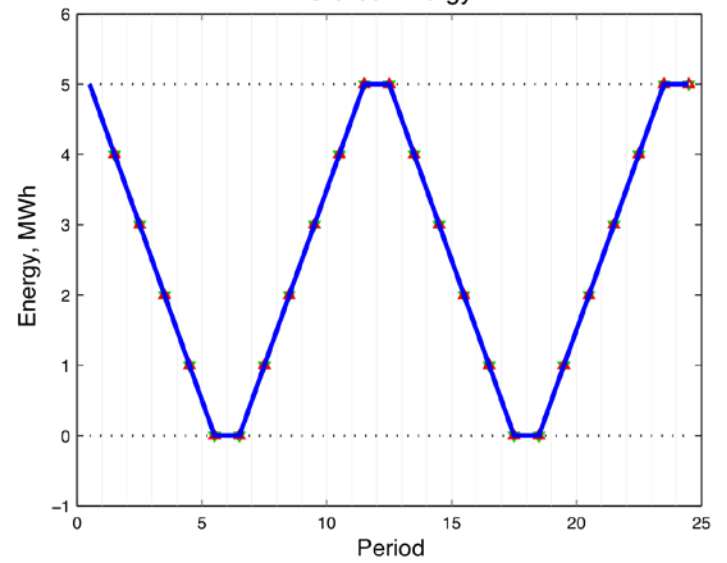
Wind Dispatch – 0% Uncertainty, 100% Variability

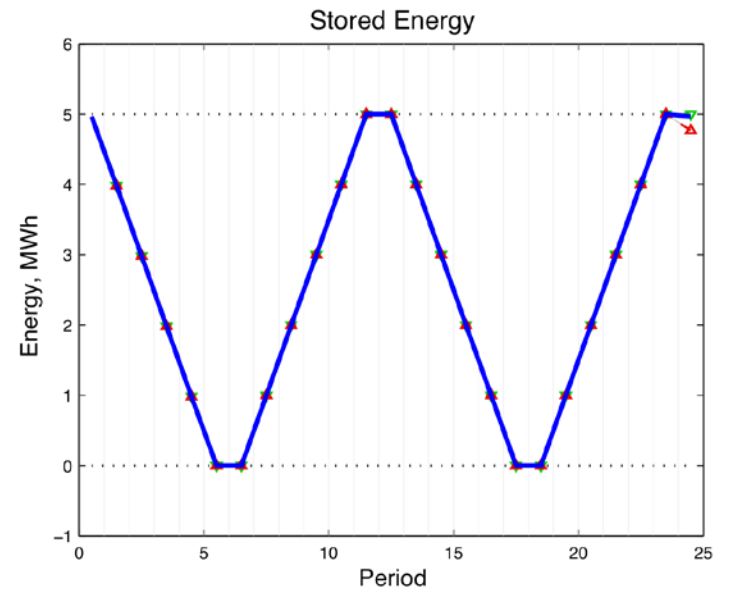
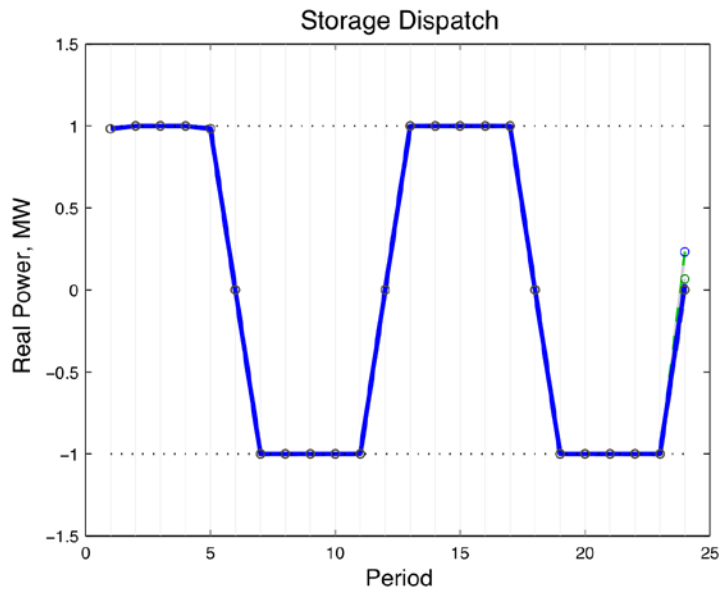
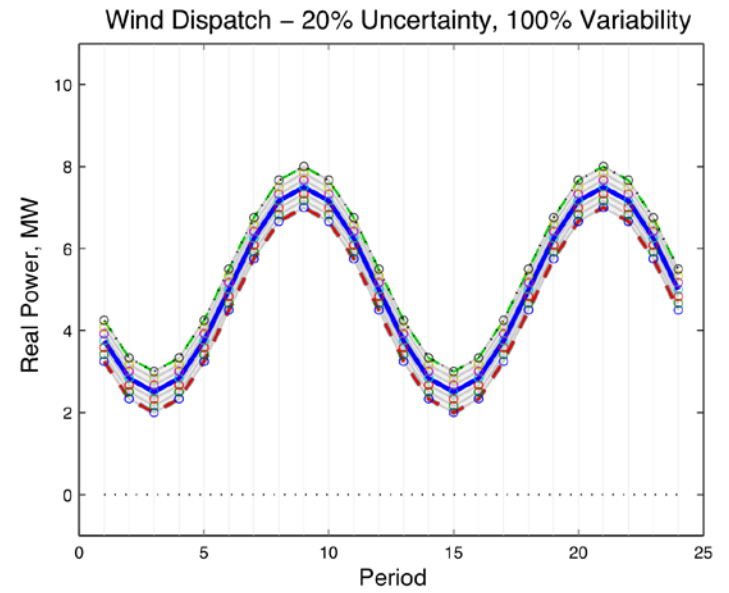
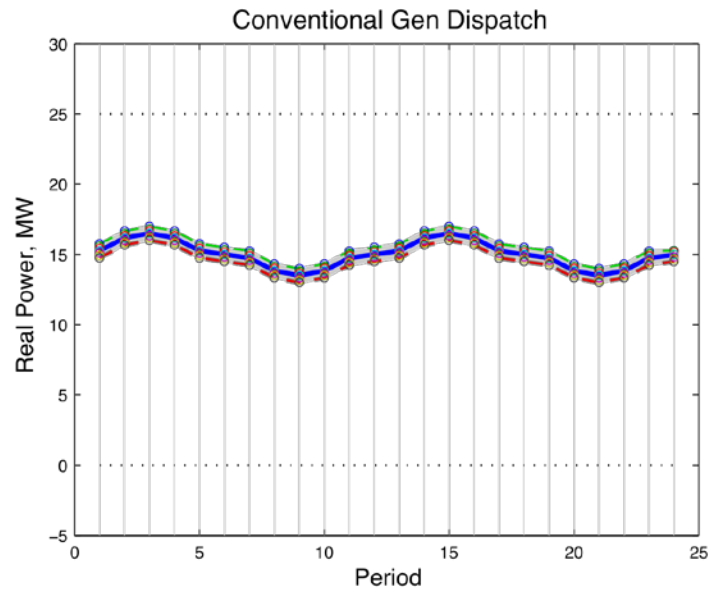


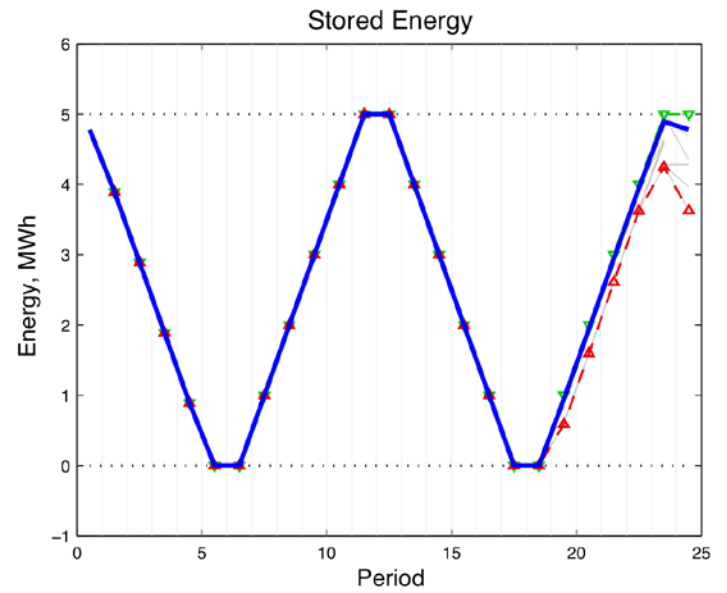
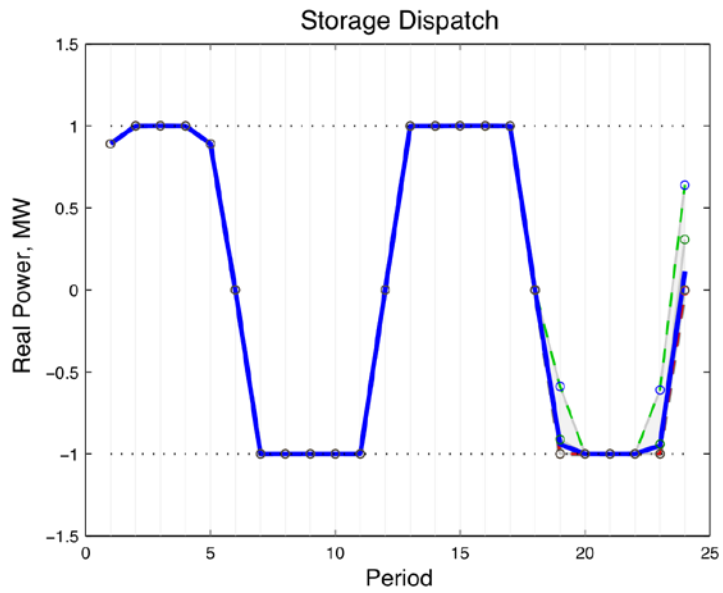
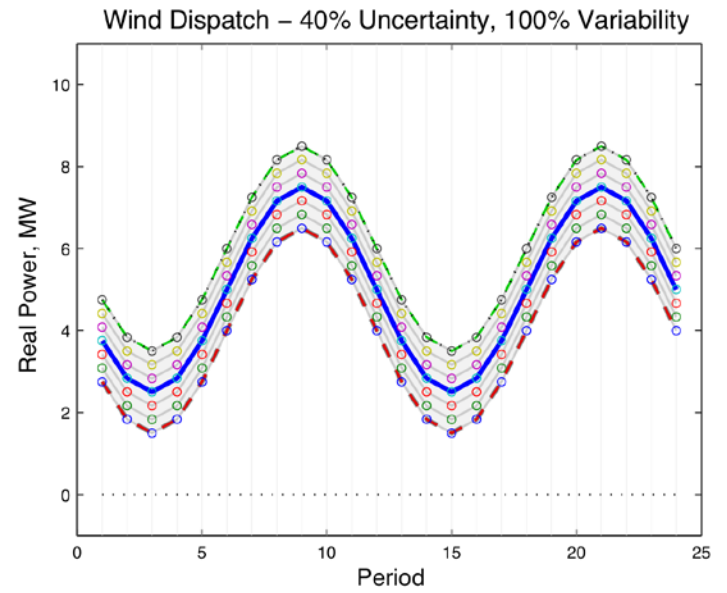
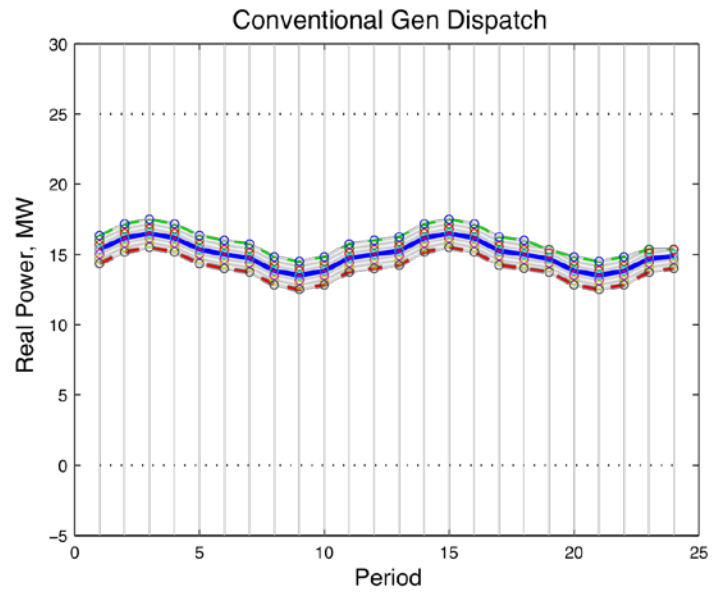
Storage Dispatch

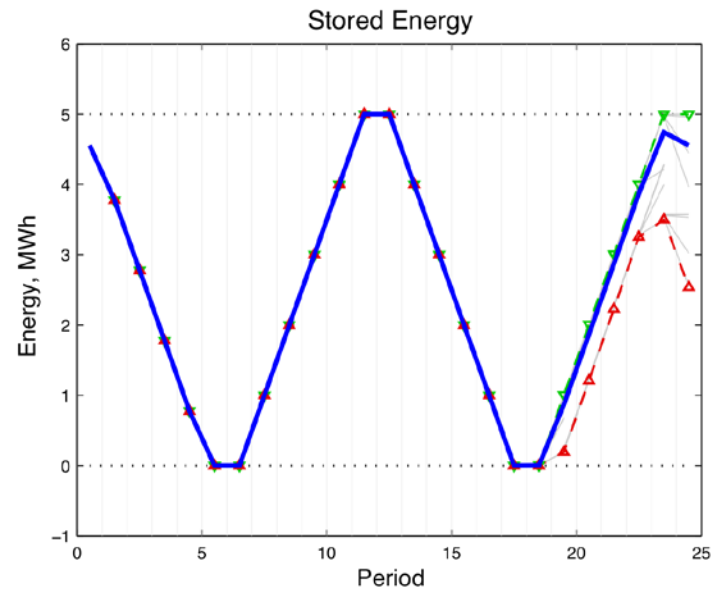
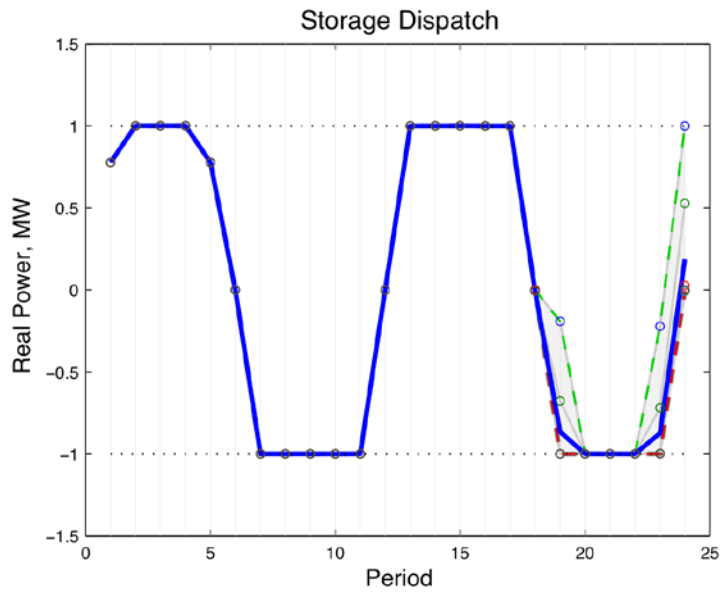
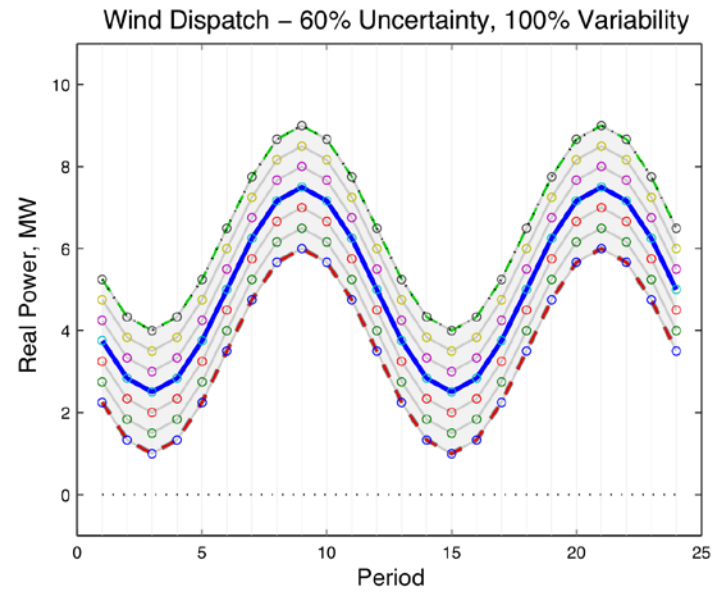
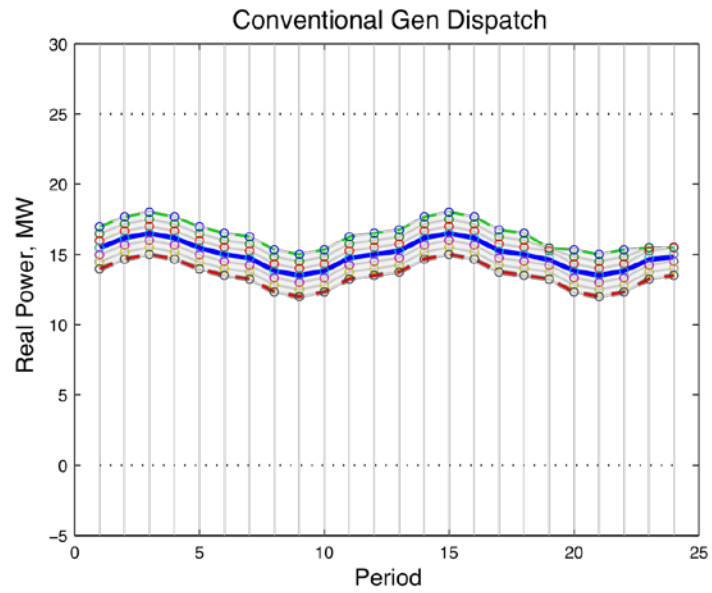


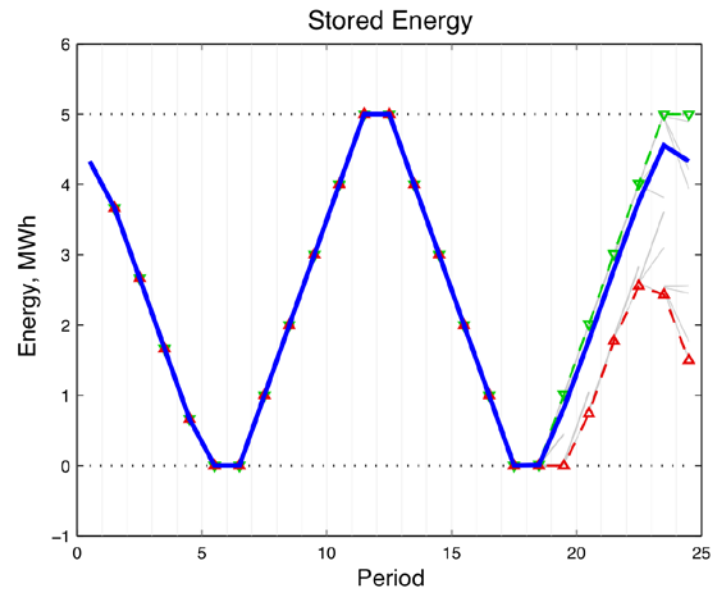
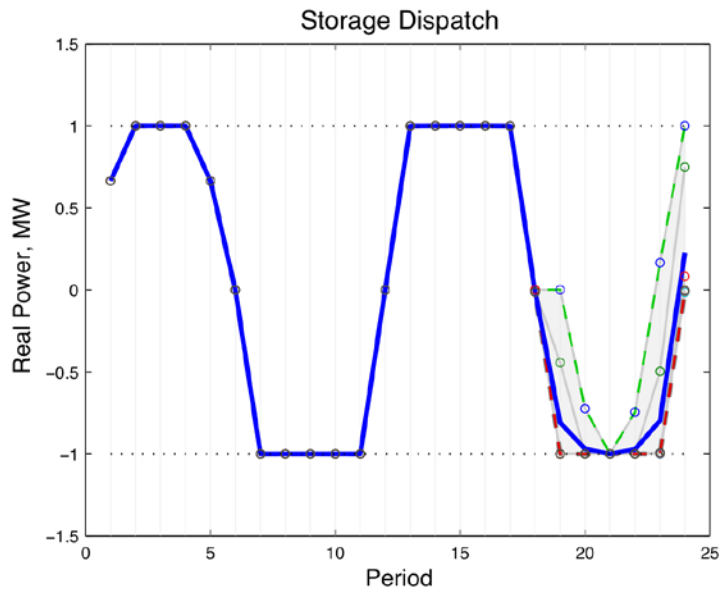
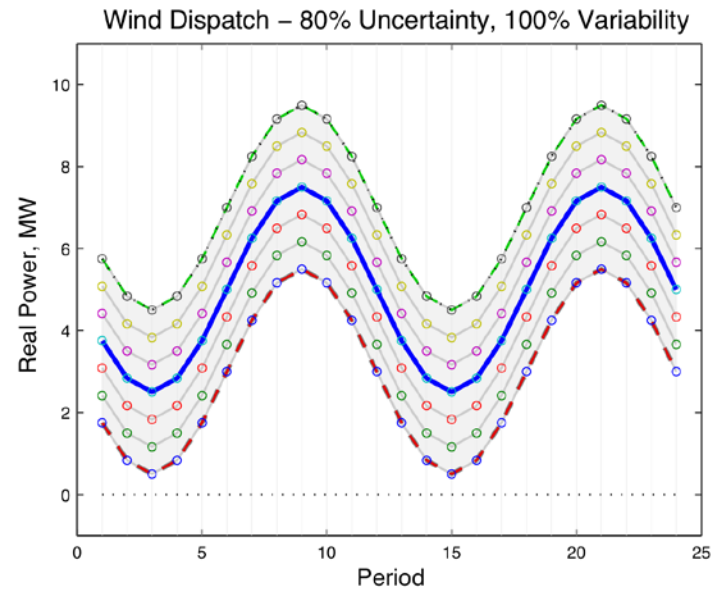
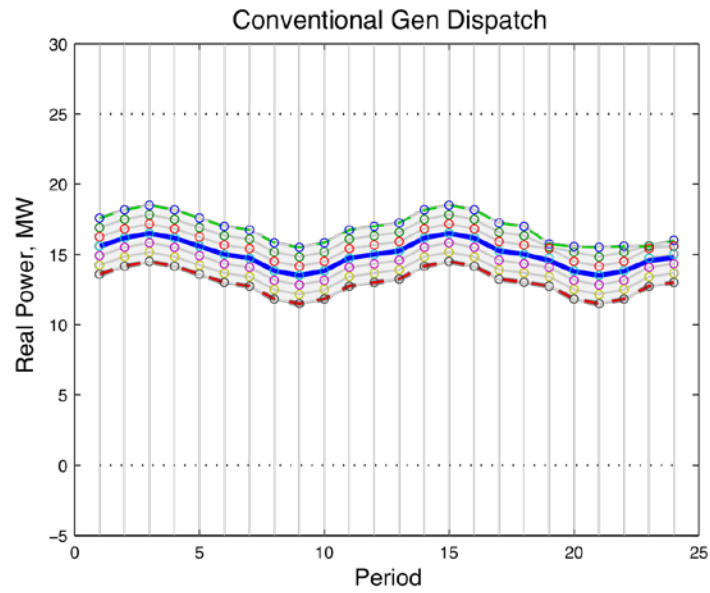
Stored Energy



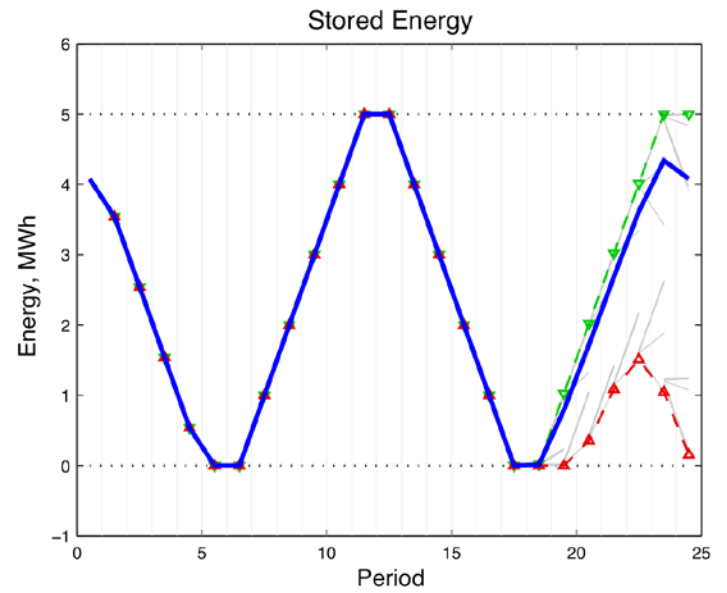
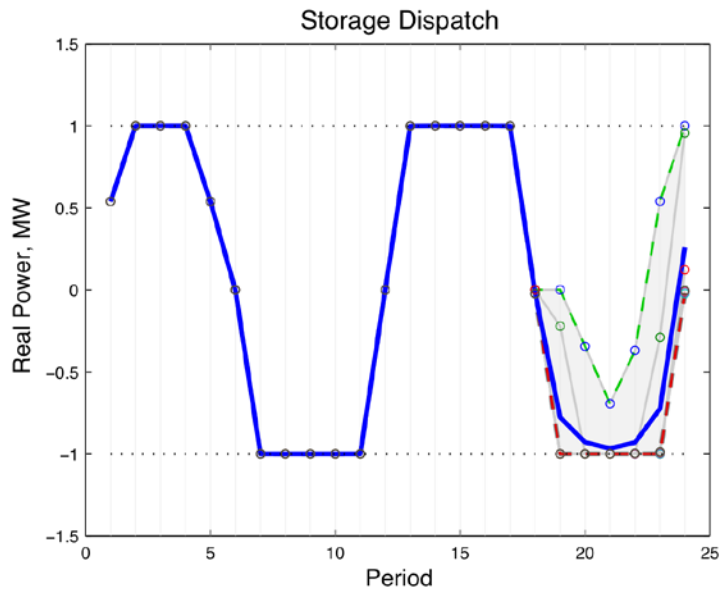
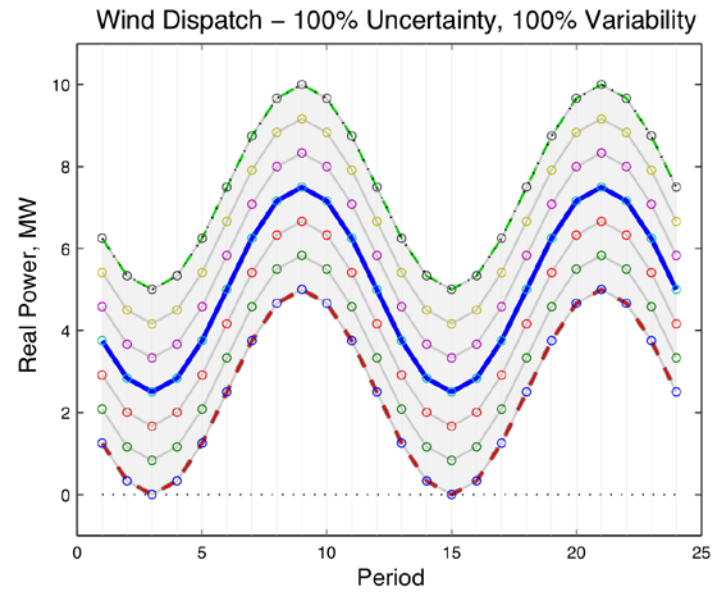
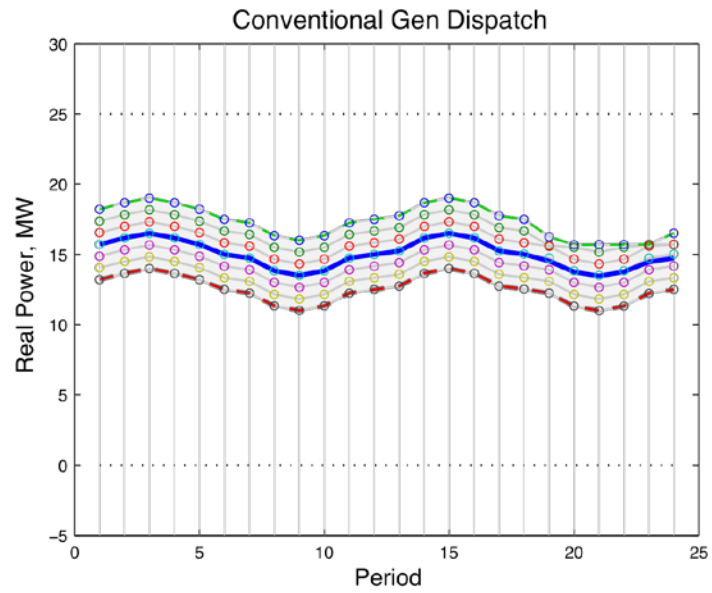


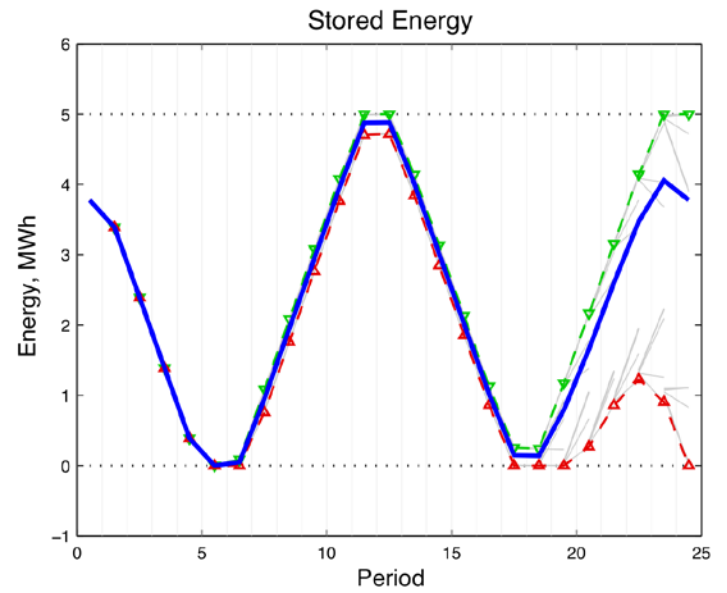
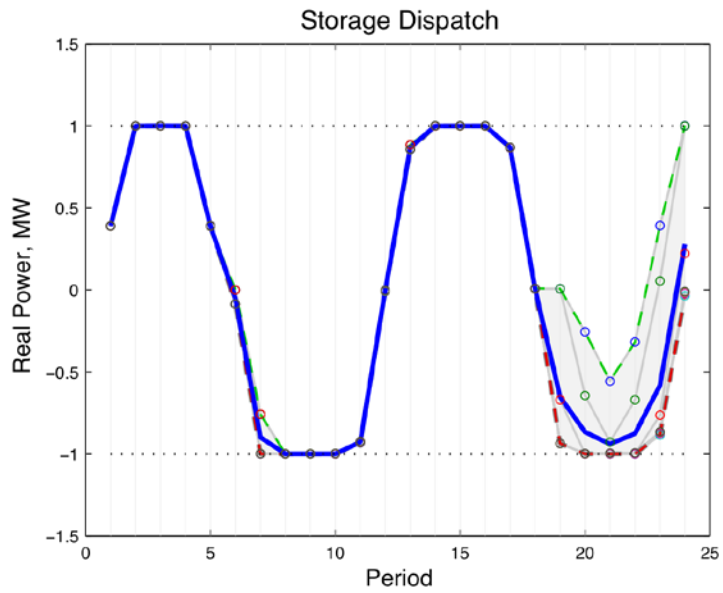
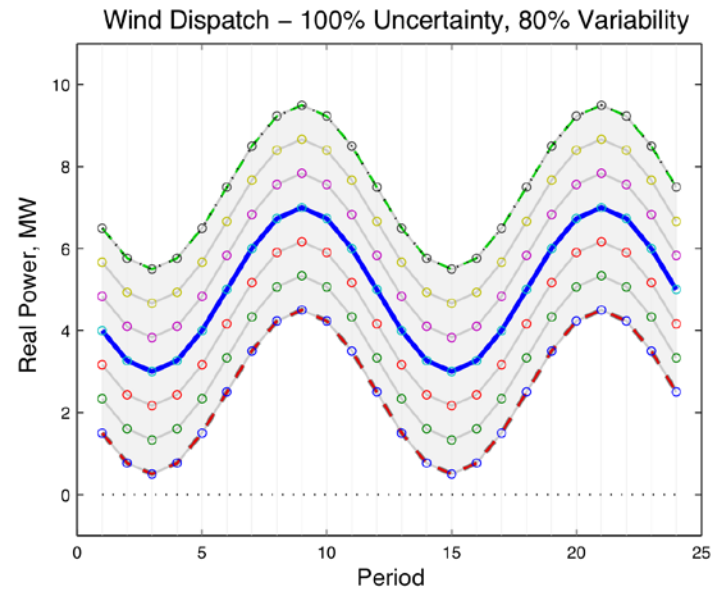
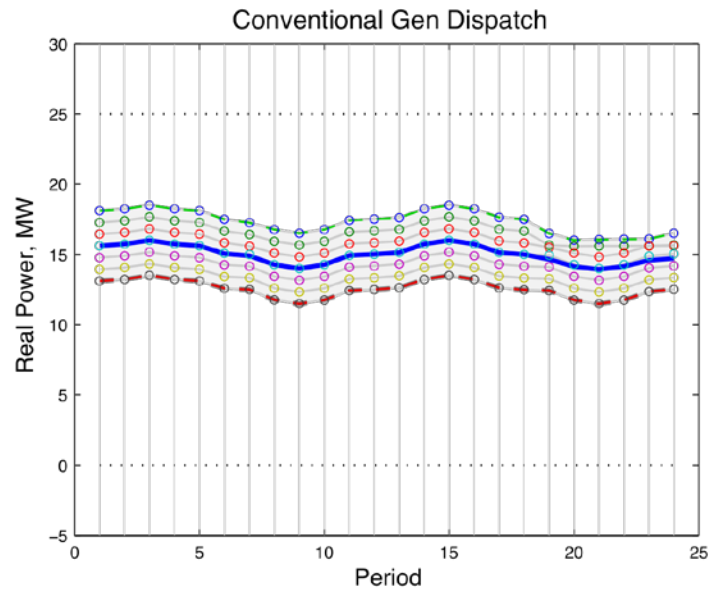


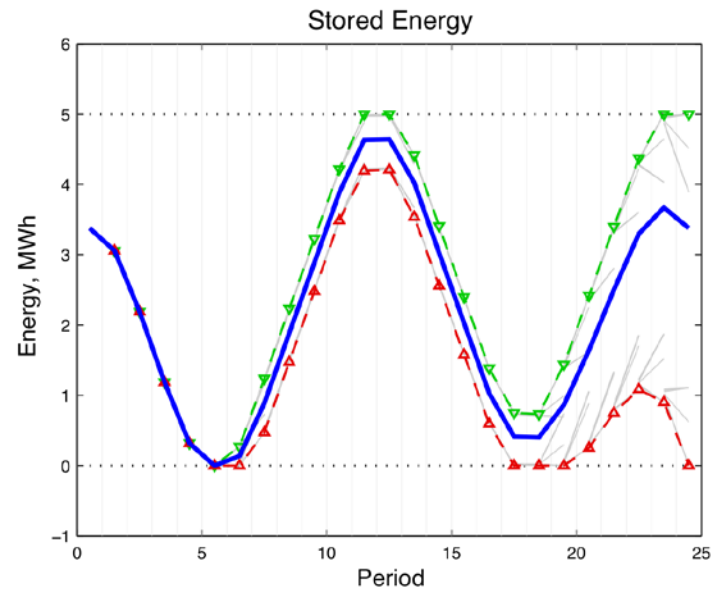
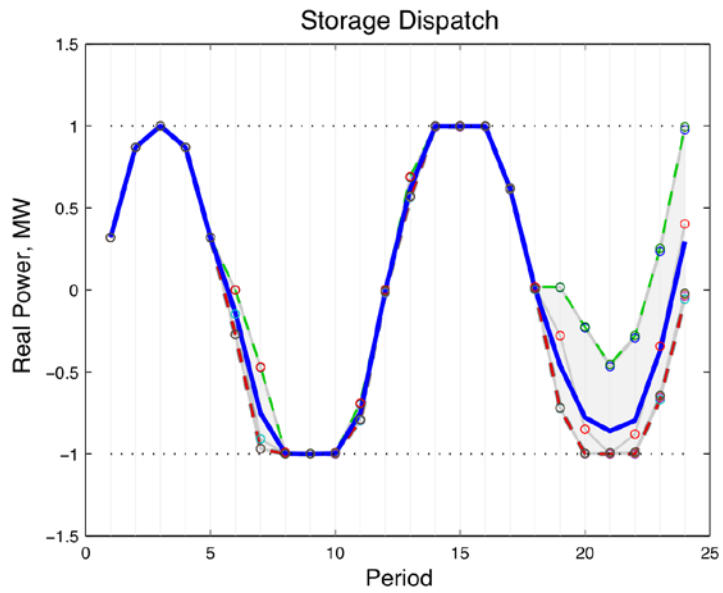
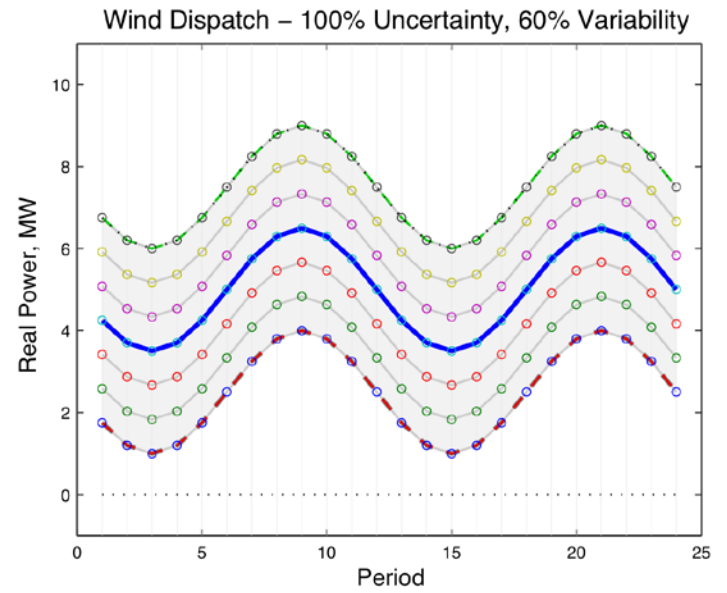
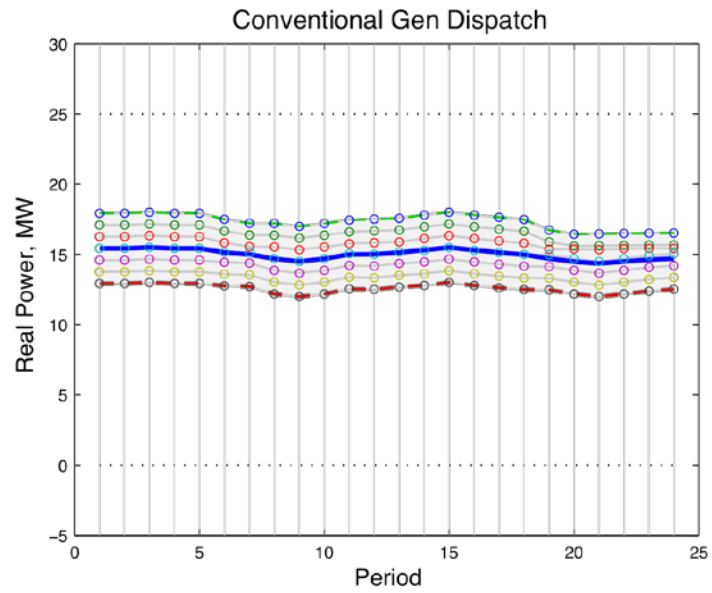


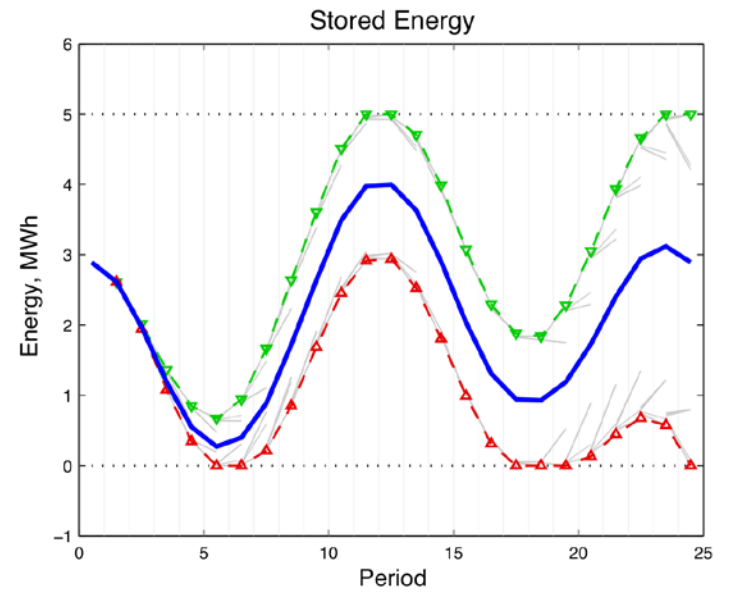
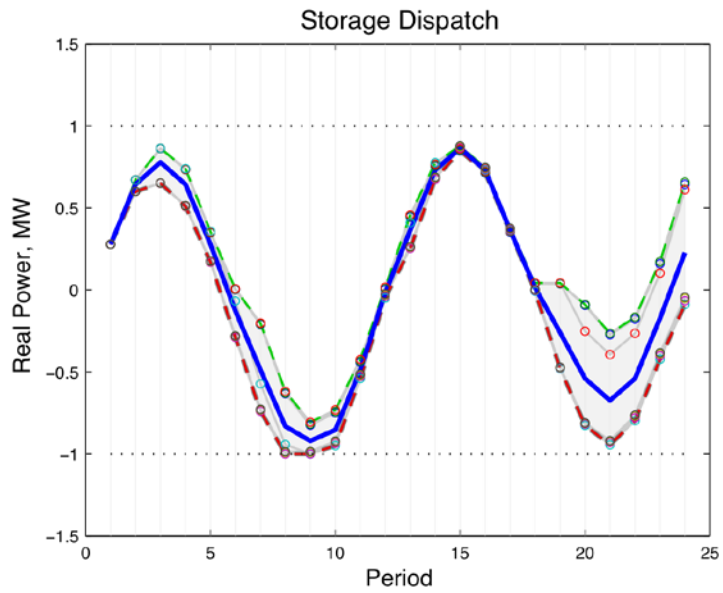
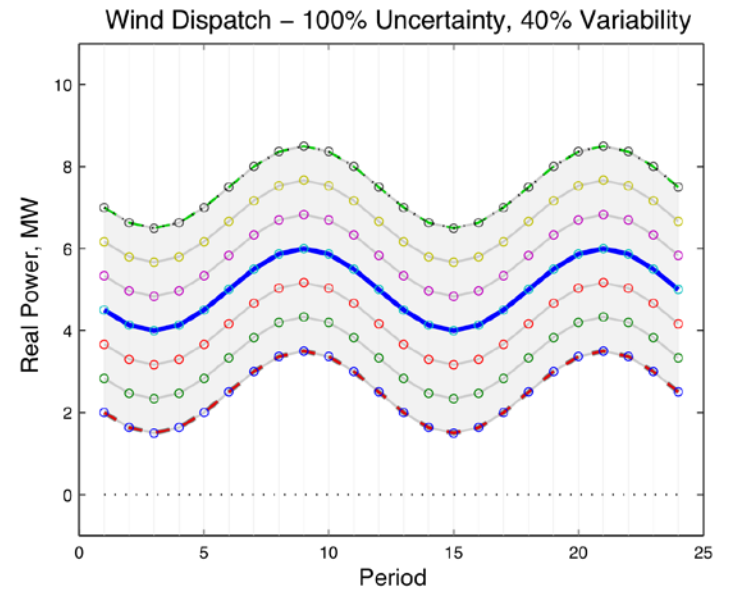
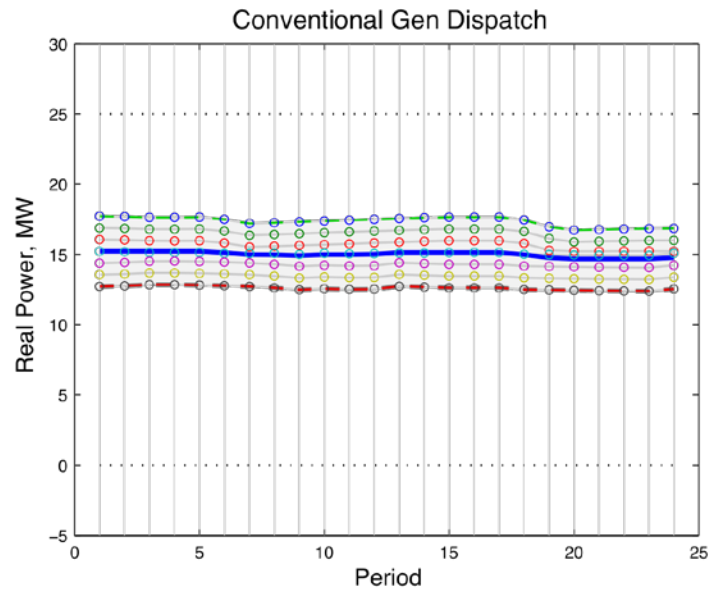




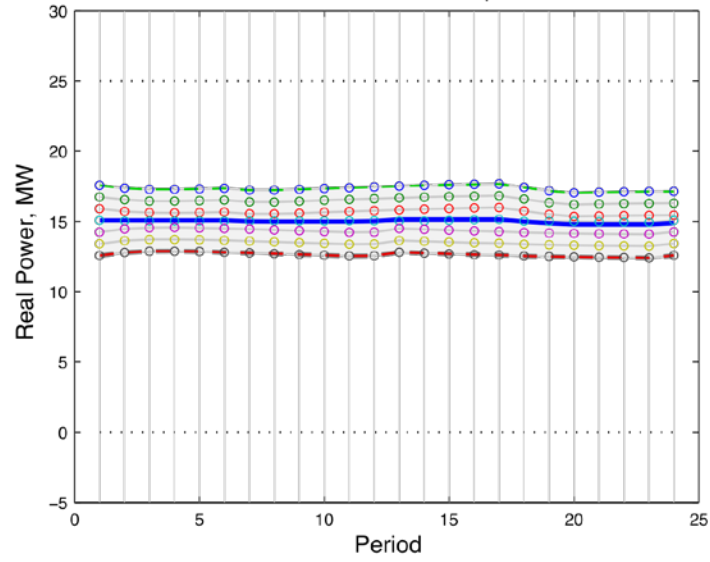




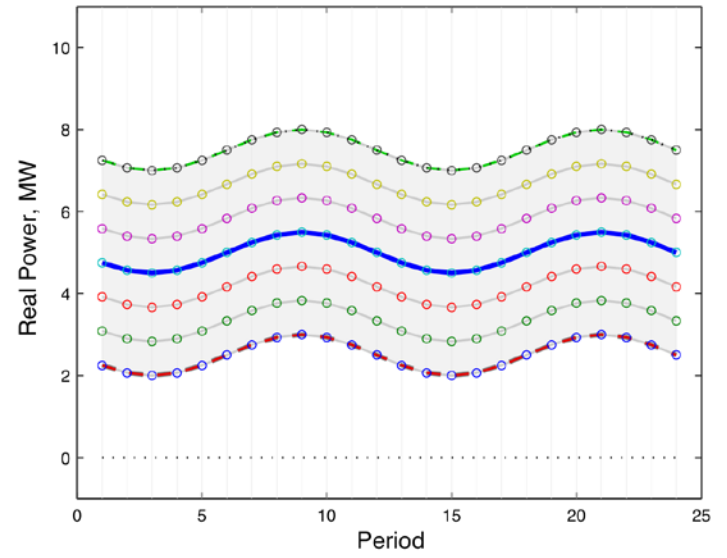




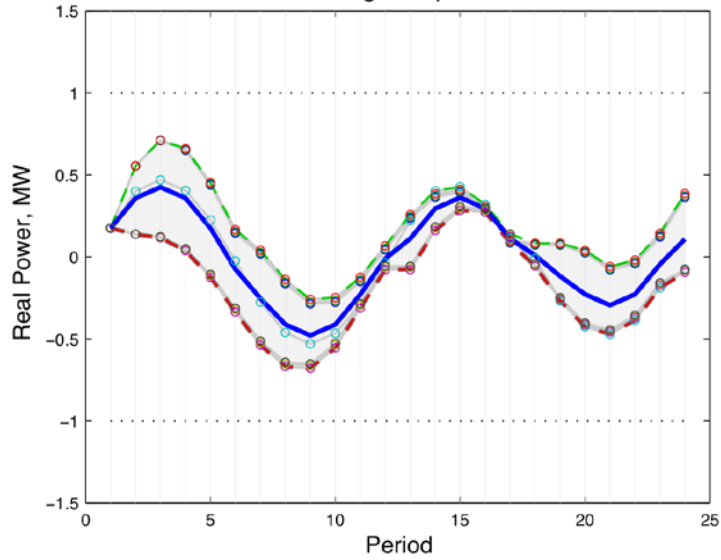
Conventional Gen Dispatch



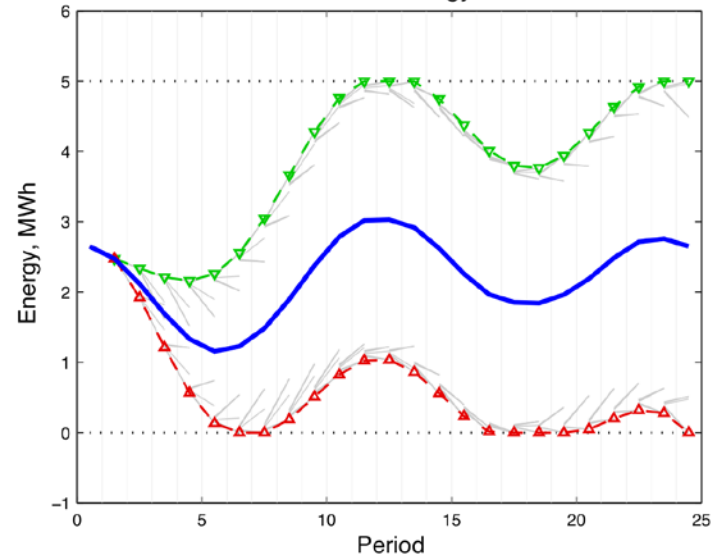
Wind Dispatch – 100% Uncertainty, 20% Variability



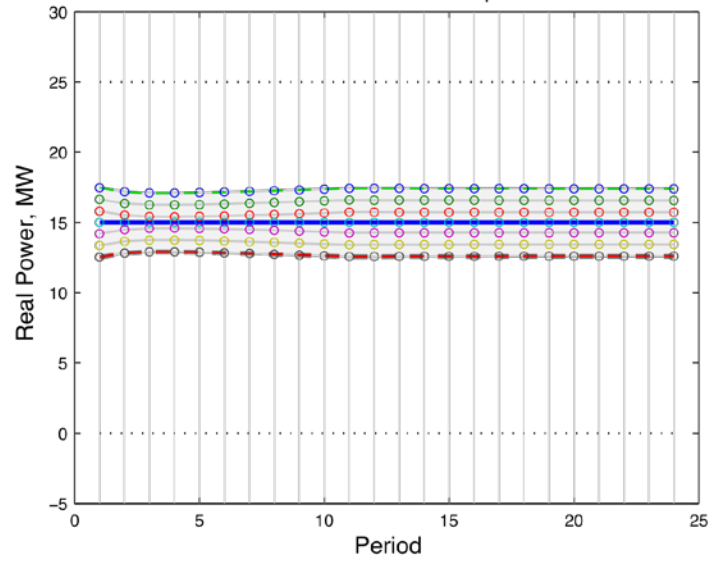
Storage Dispatch



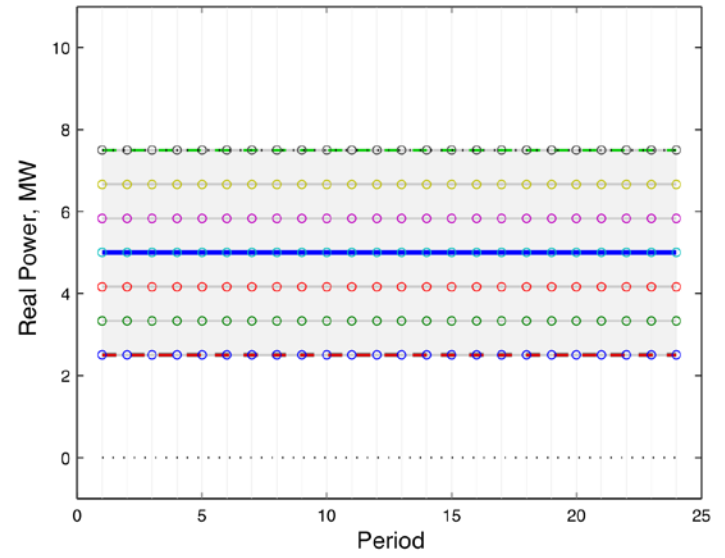
Stored Energy



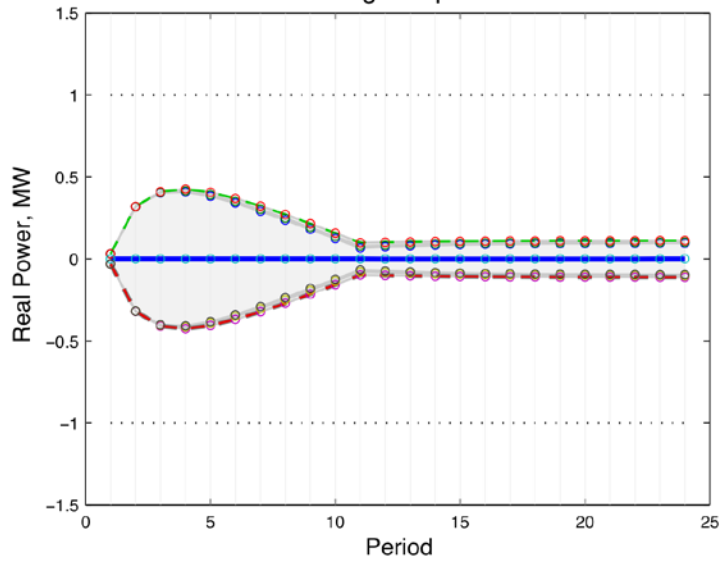
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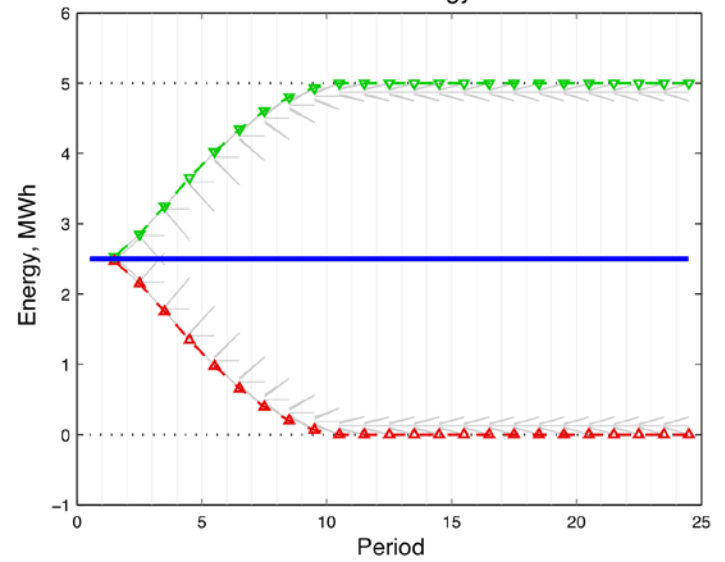
Wind Dispatch – 100% Uncertainty, 0% Variability

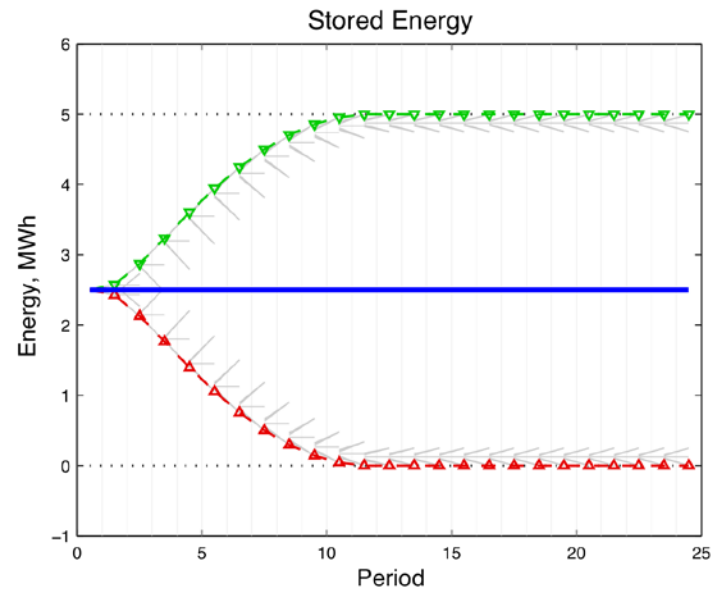
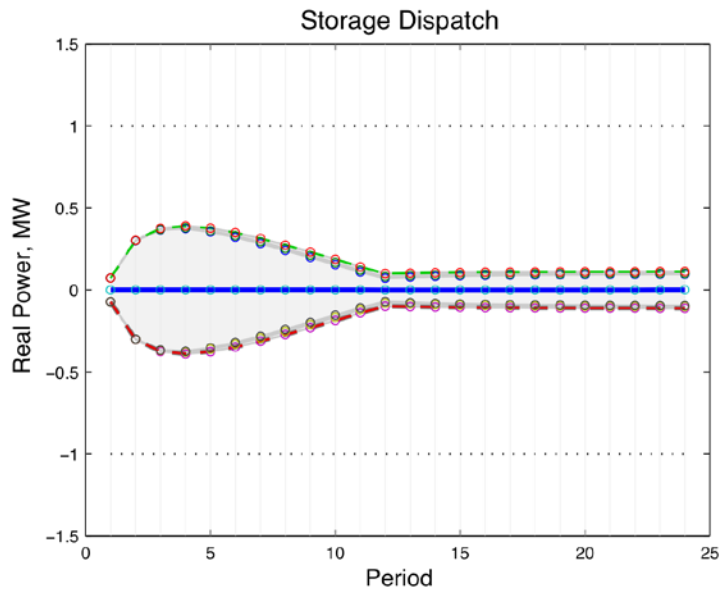
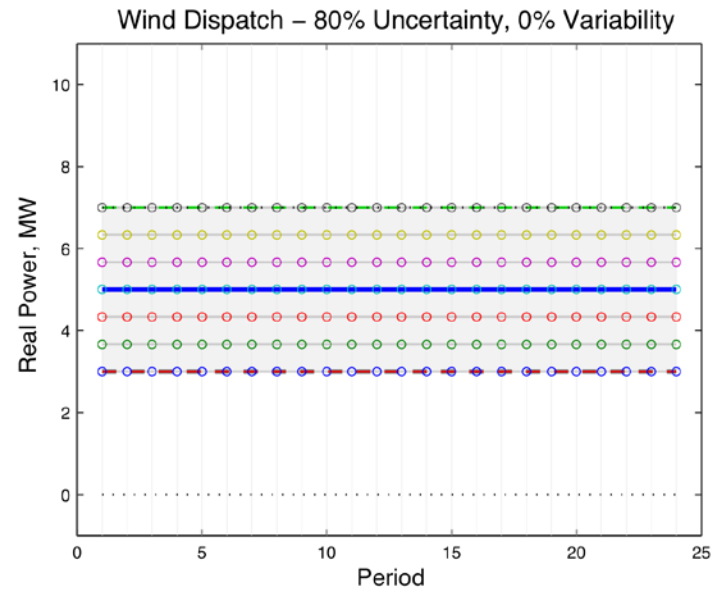
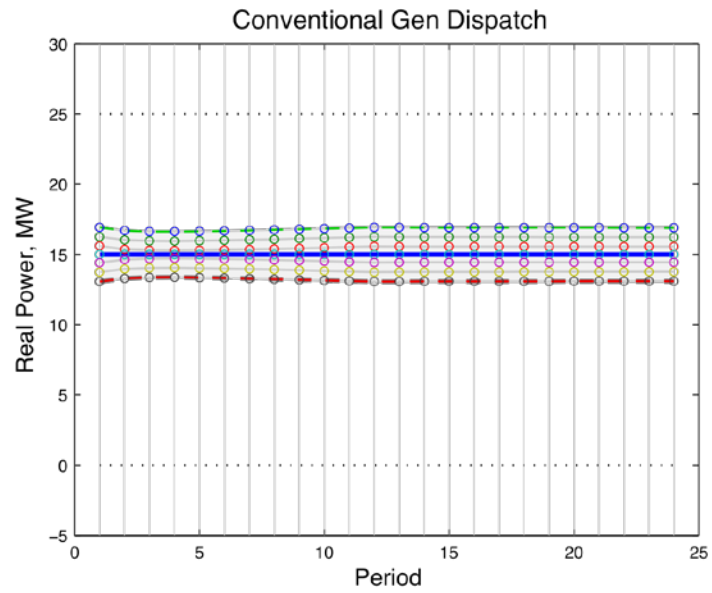


Storage Dispatch

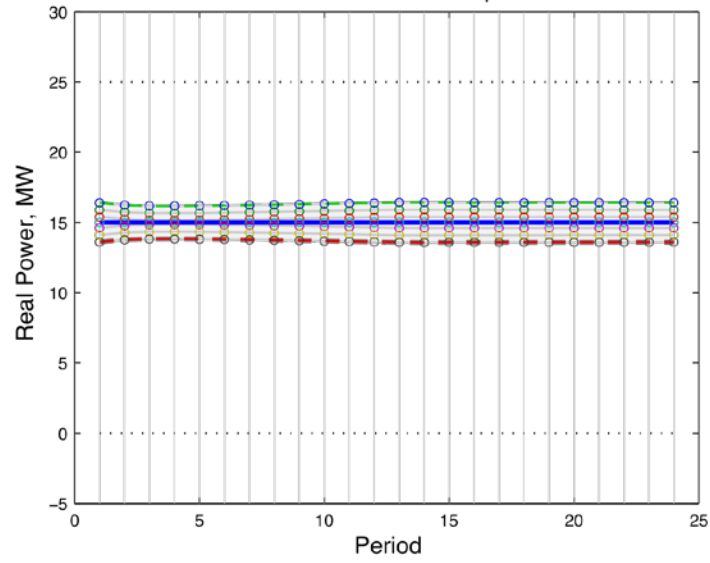


Stored Energy

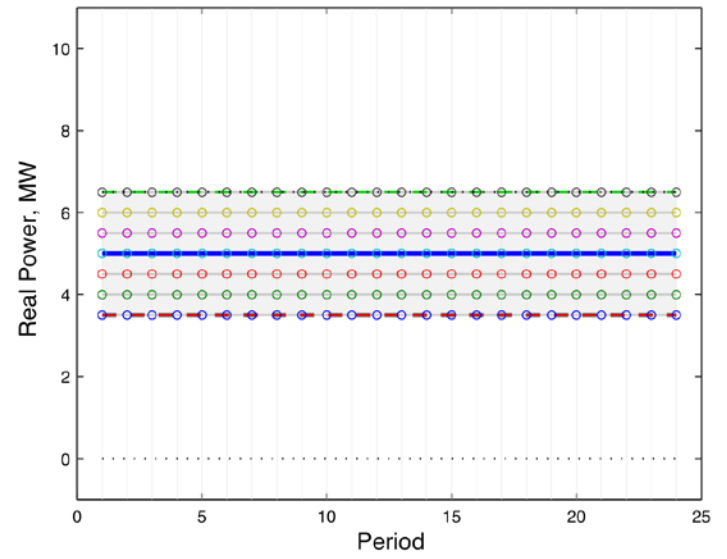




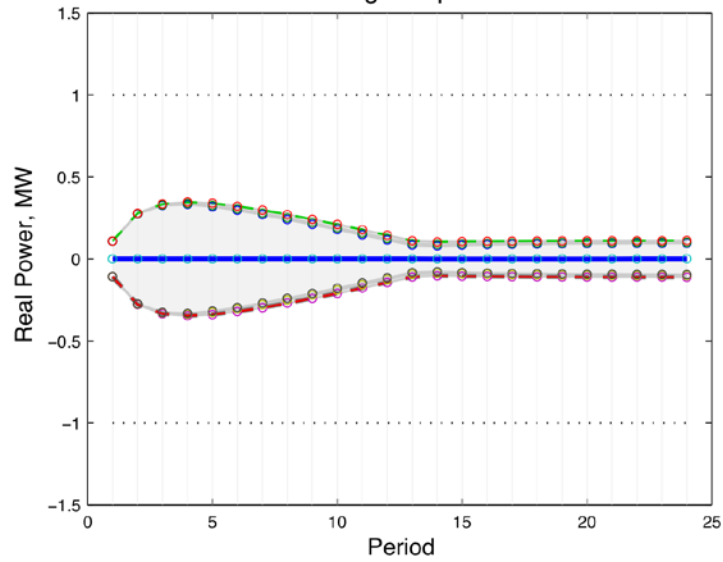
Conventional Gen Dispatch



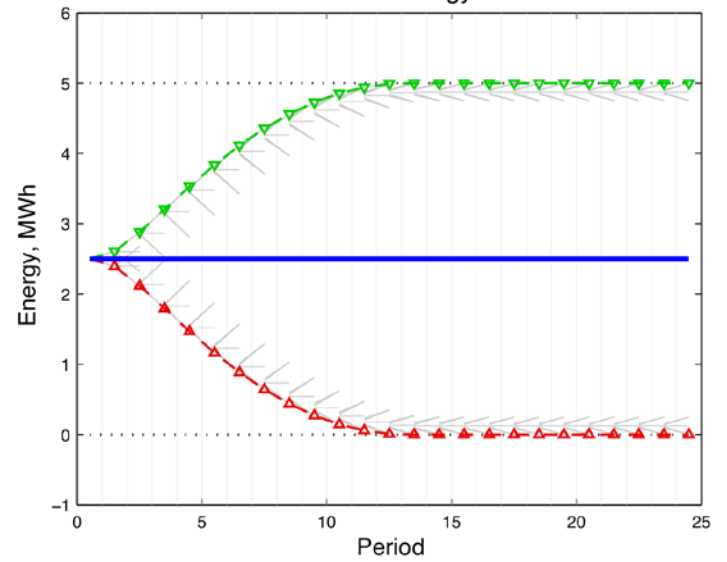
Wind Dispatch – 60% Uncertainty, 0% Variability



Storage Dispatch

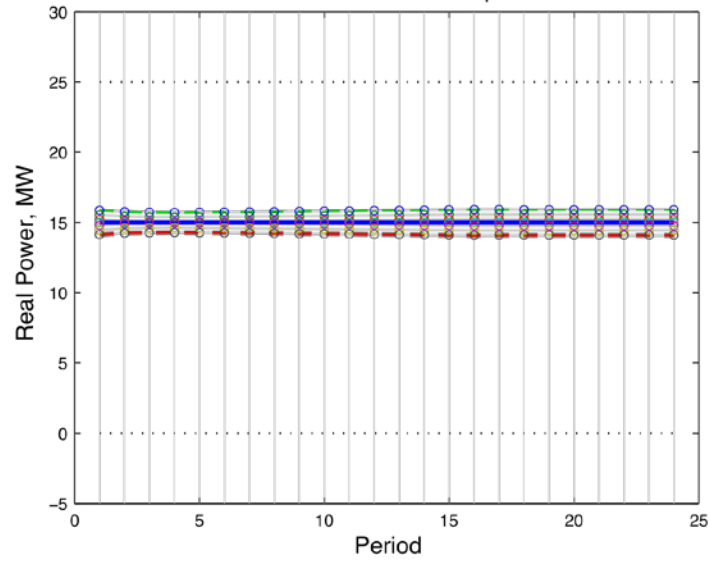


Stored Energy

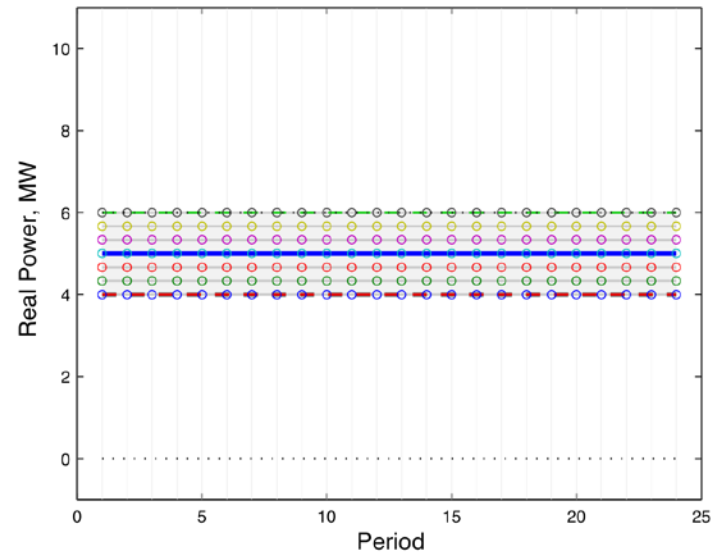




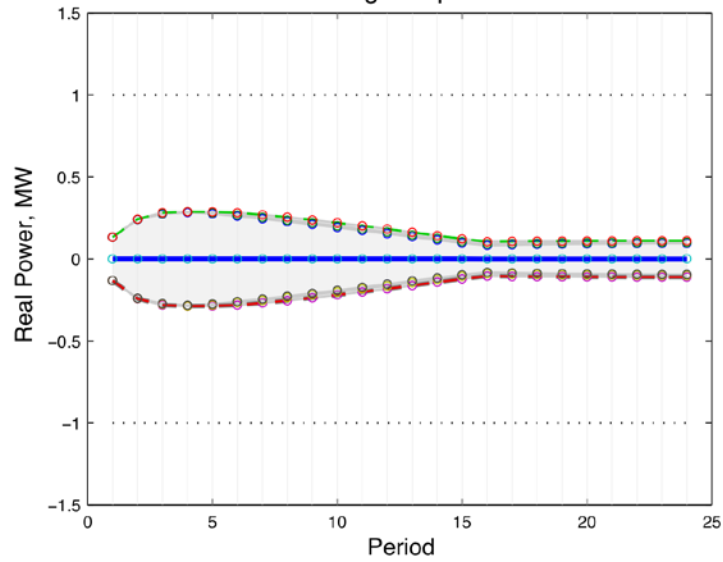
Conventional Gen Dispatch



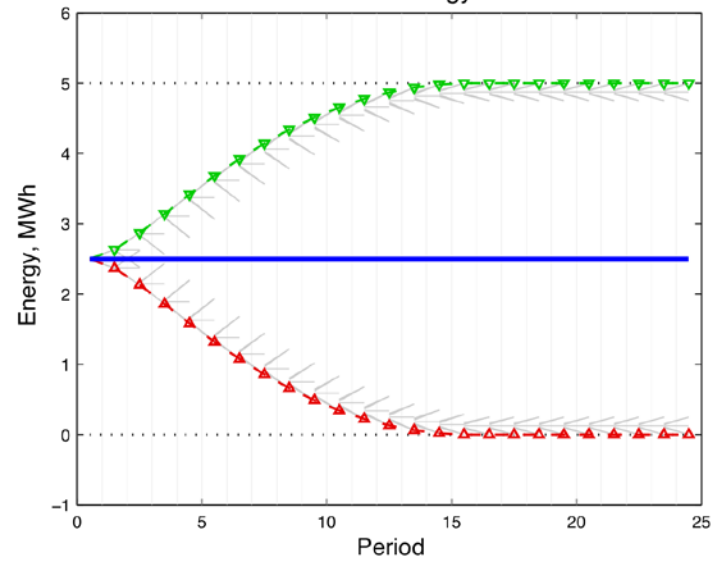
Wind Dispatch – 40% Uncertainty, 0% Variability



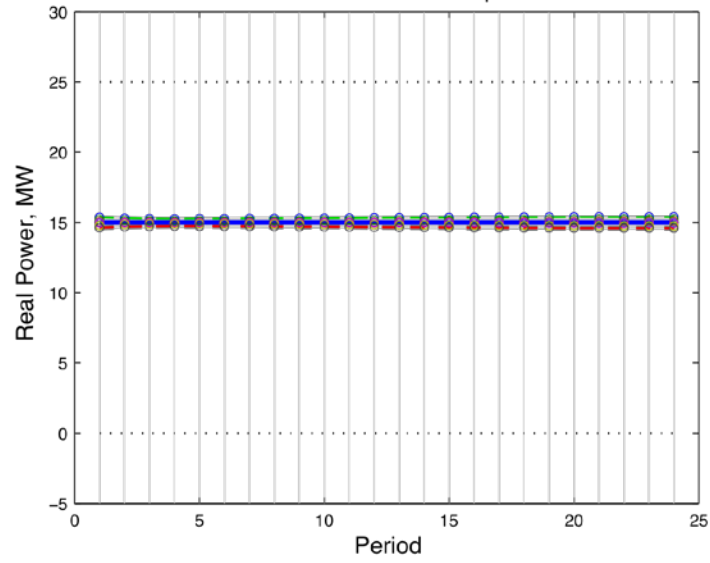
Storage Dispatch



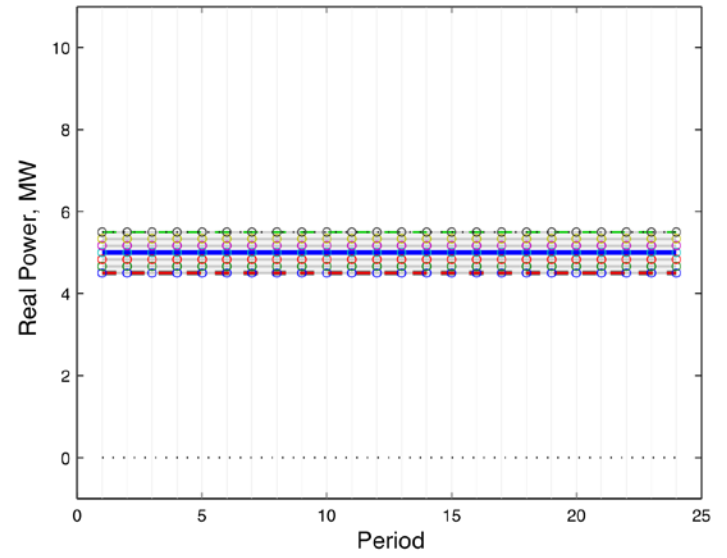
Stored Energy



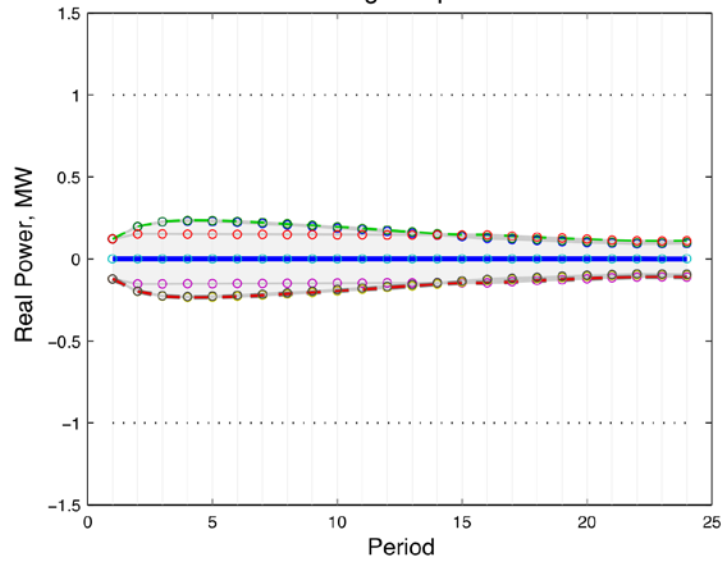
Conventional Gen Dispatch



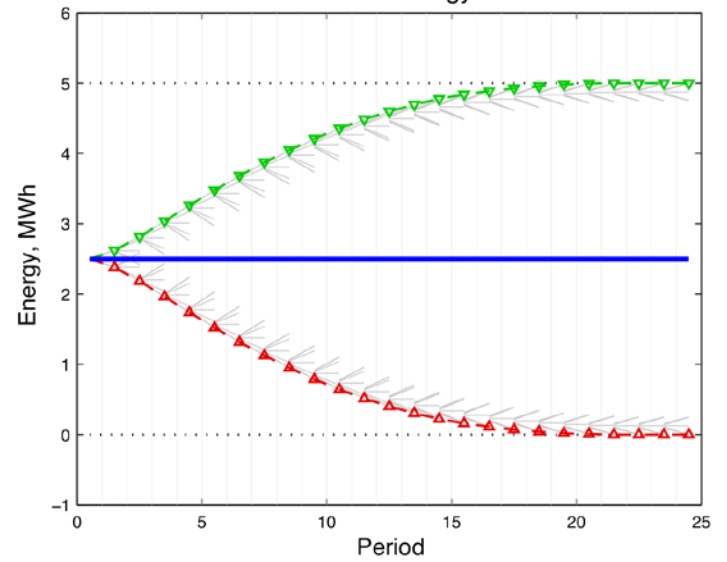
Wind Dispatch – 20% Uncertainty, 0% Variability

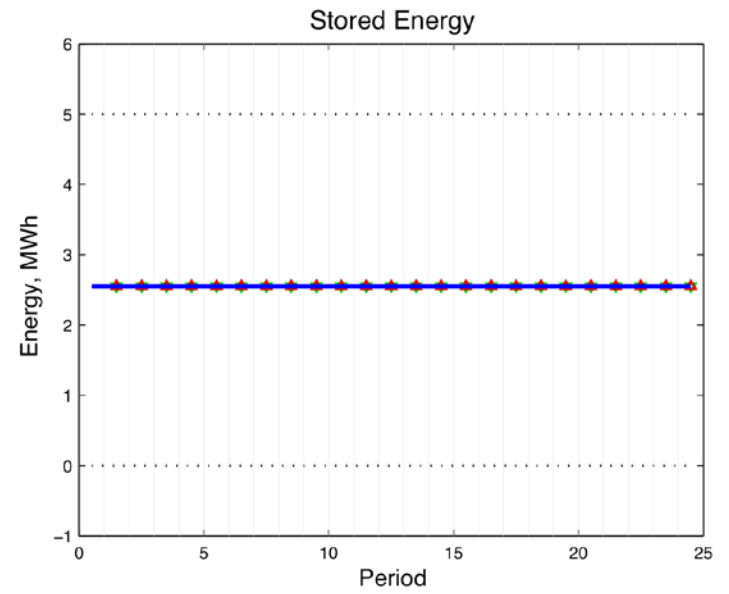
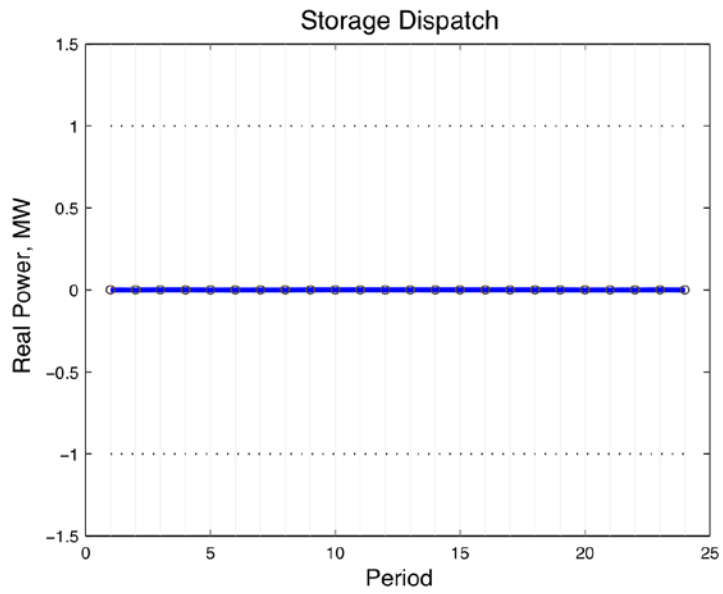
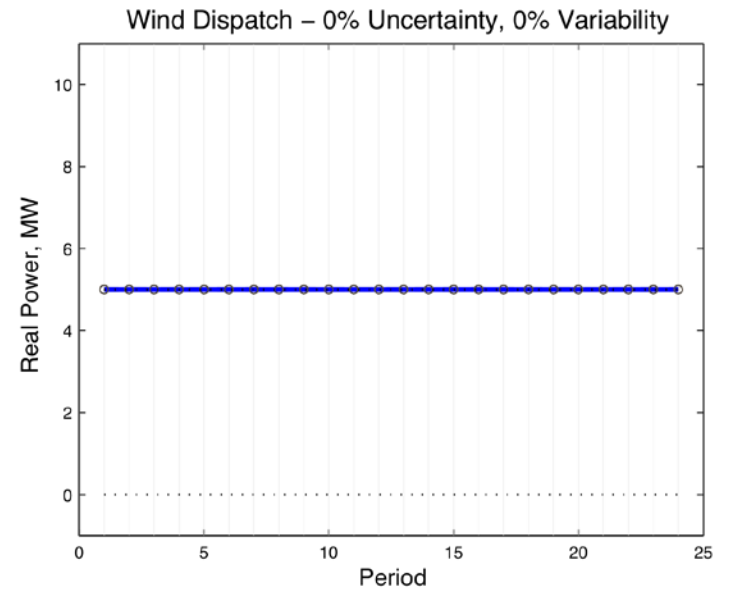
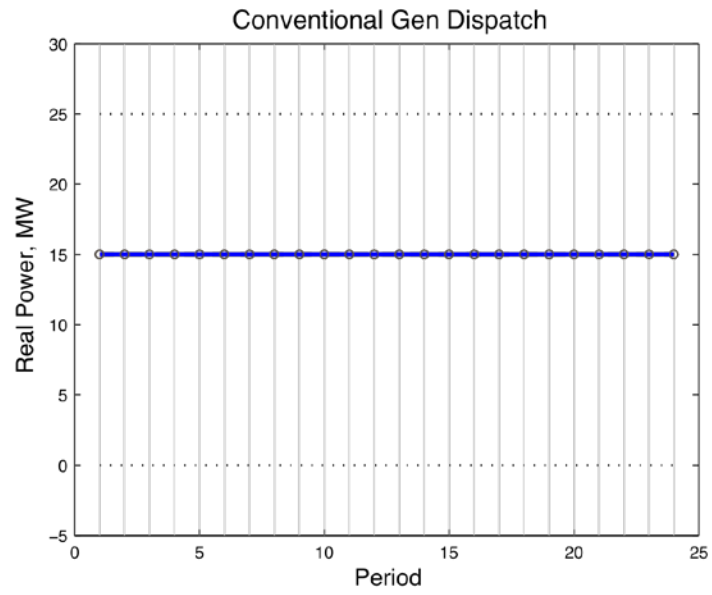


Storage Dispatch

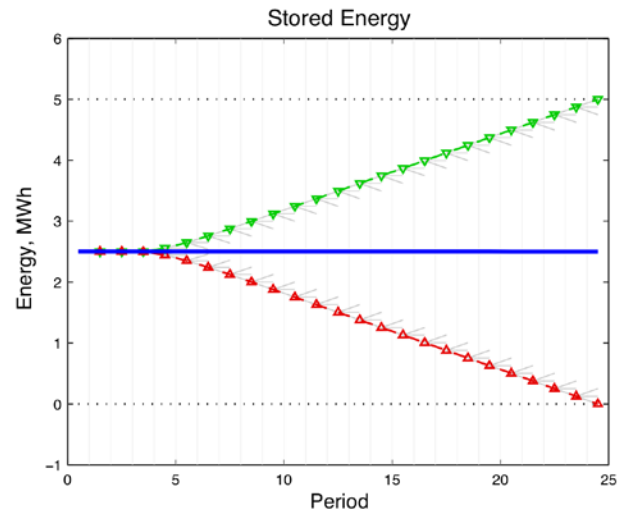
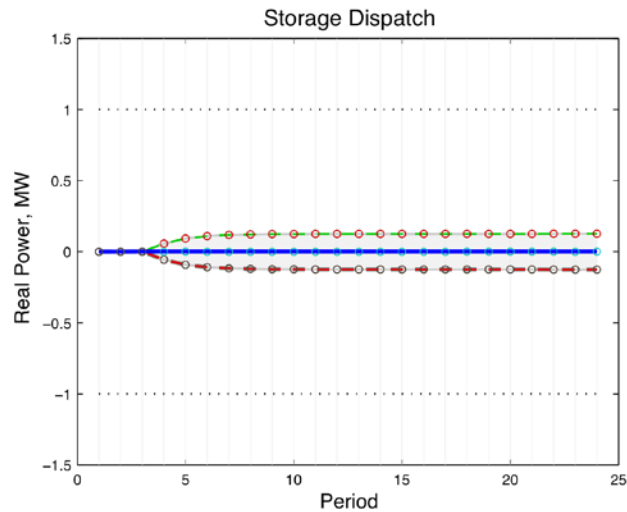
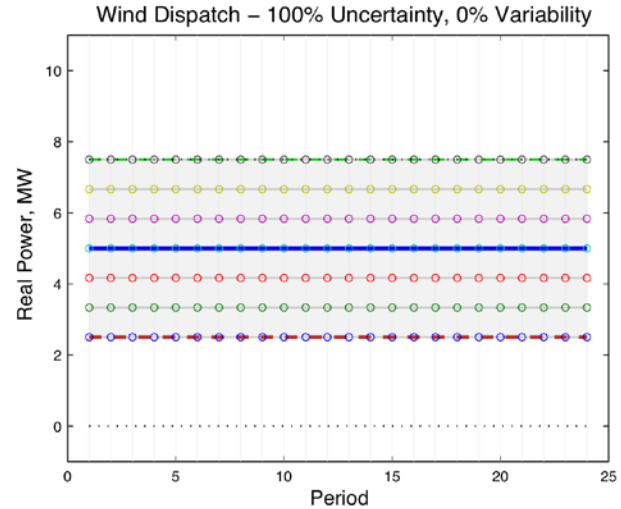
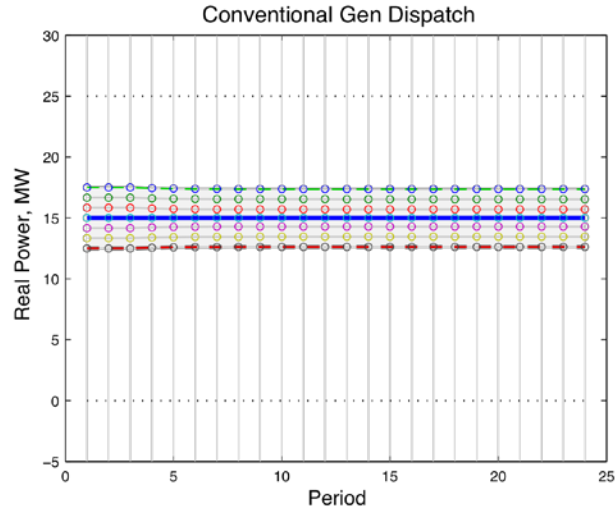


Stored Energy

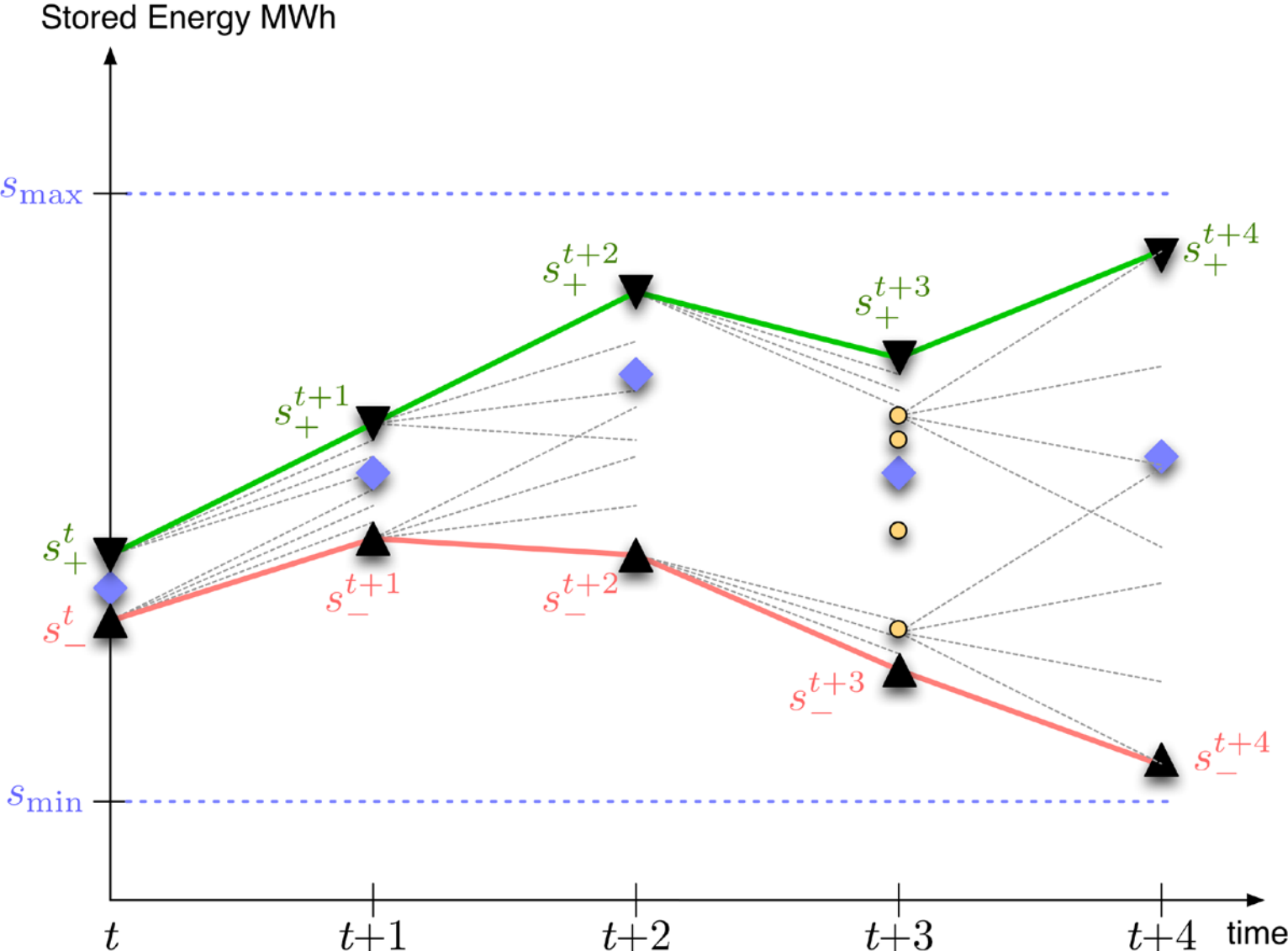


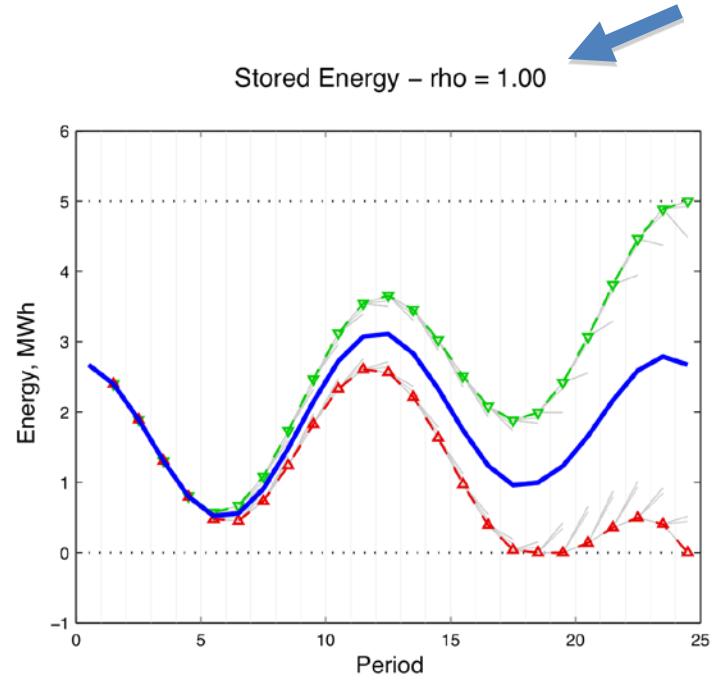
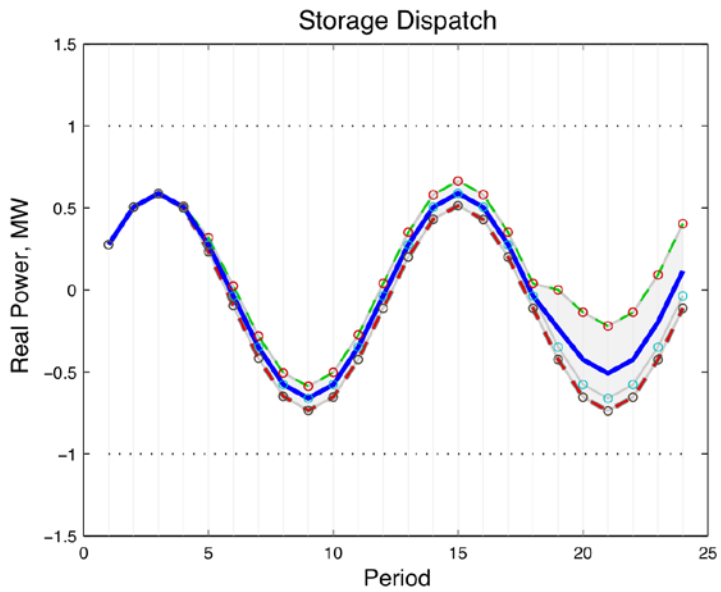
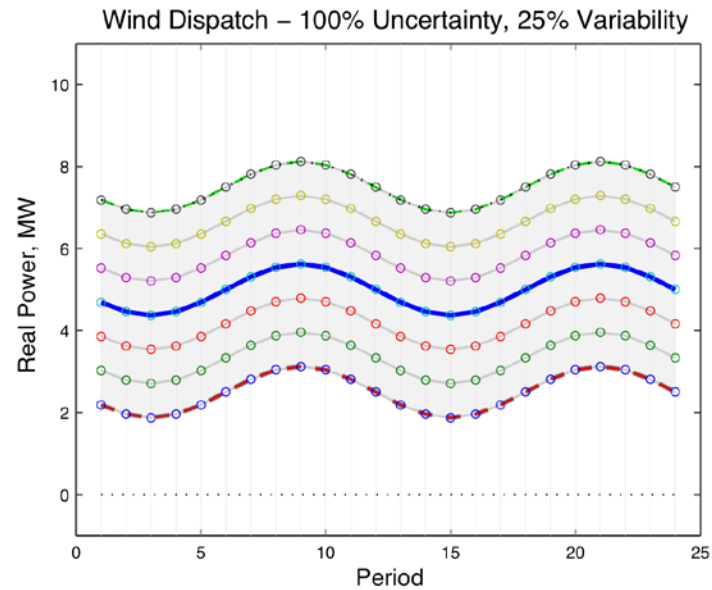
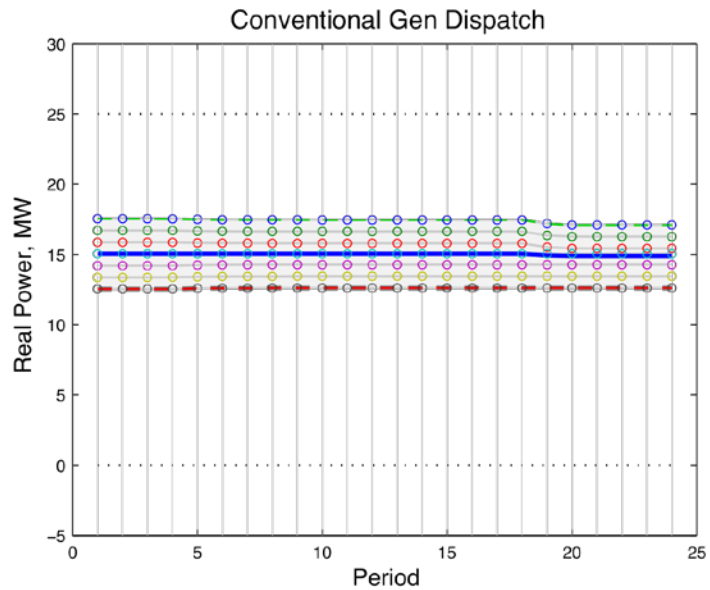


# Worst Case Storage Constraint Problem

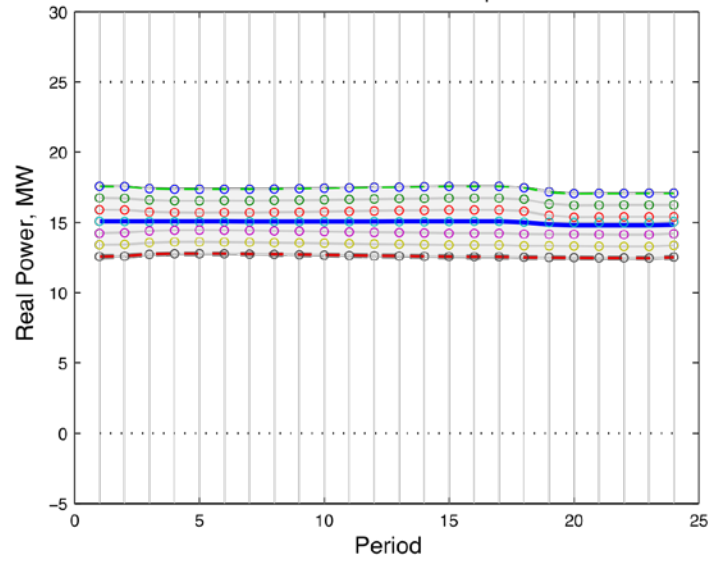


# Relaxing Worst Case Storage Constraints

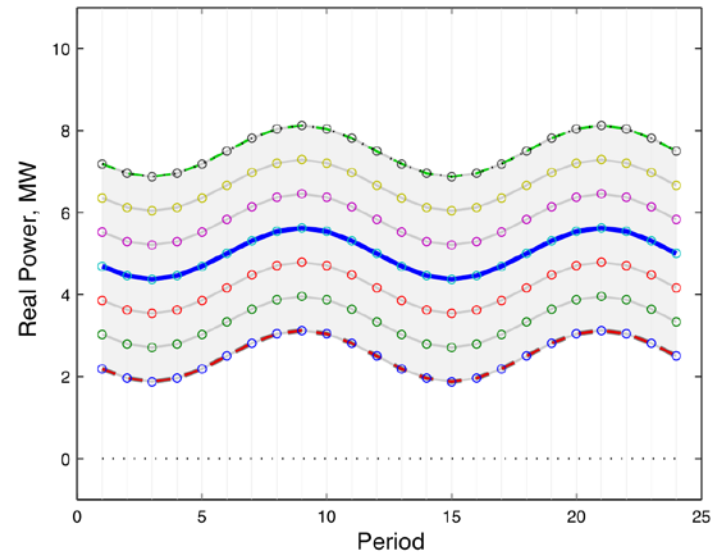




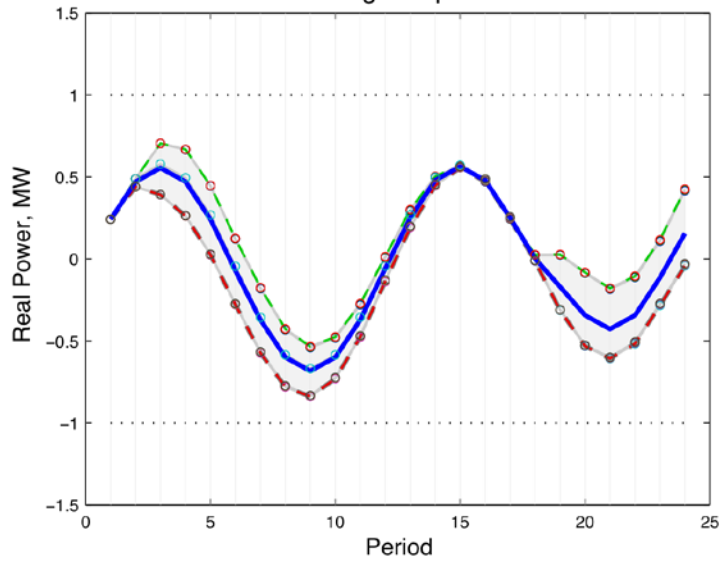
Conventional Gen Dispatch



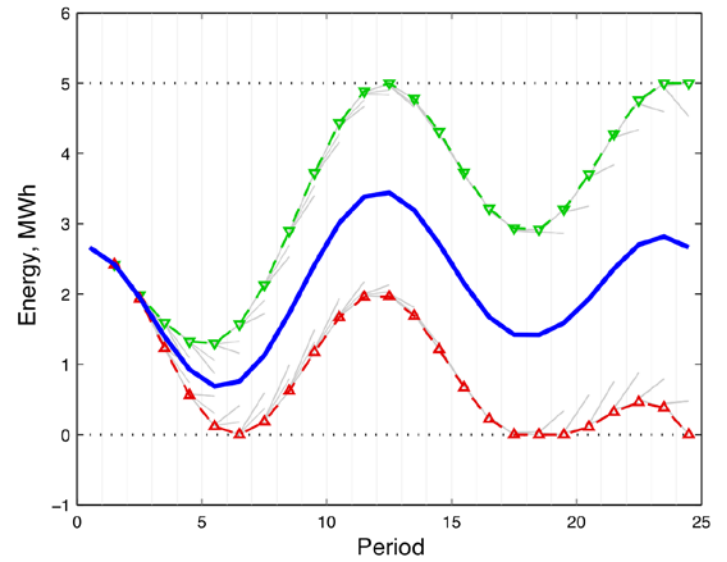
Wind Dispatch – 100% Uncertainty, 25% Variability



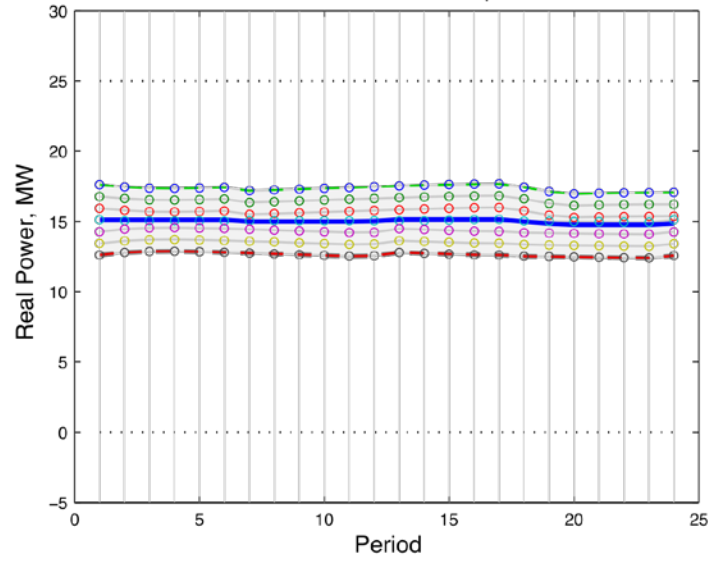
Storage Dispatch



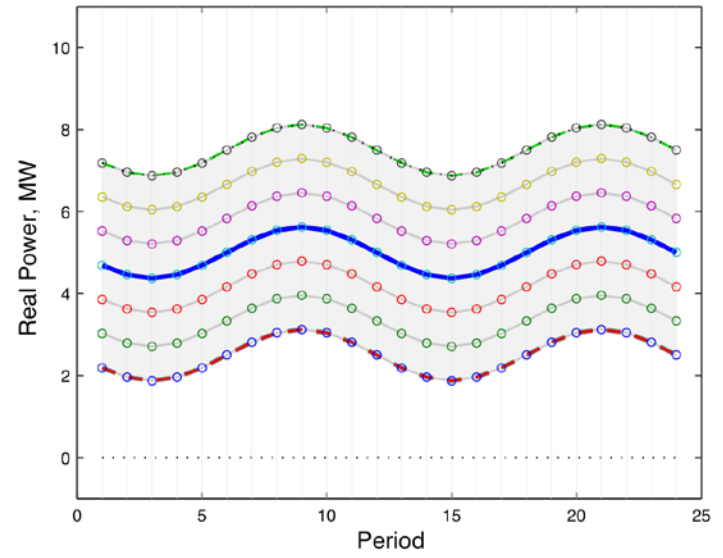
Stored Energy – rho = 0.98



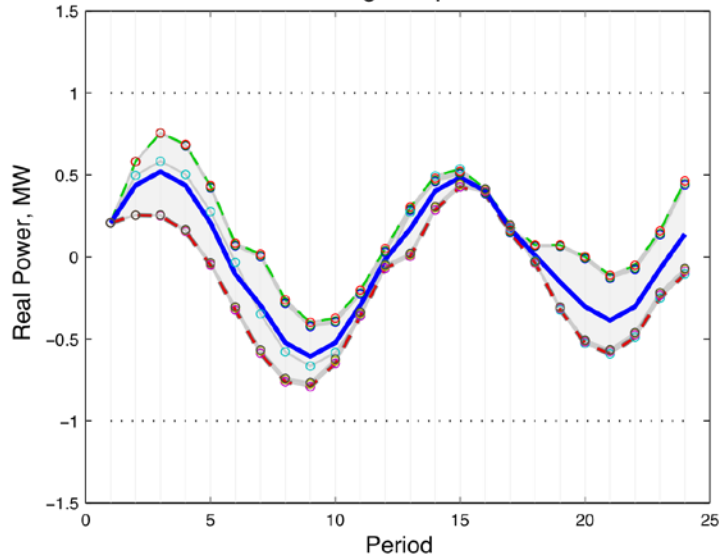
Conventional Gen Dispatch



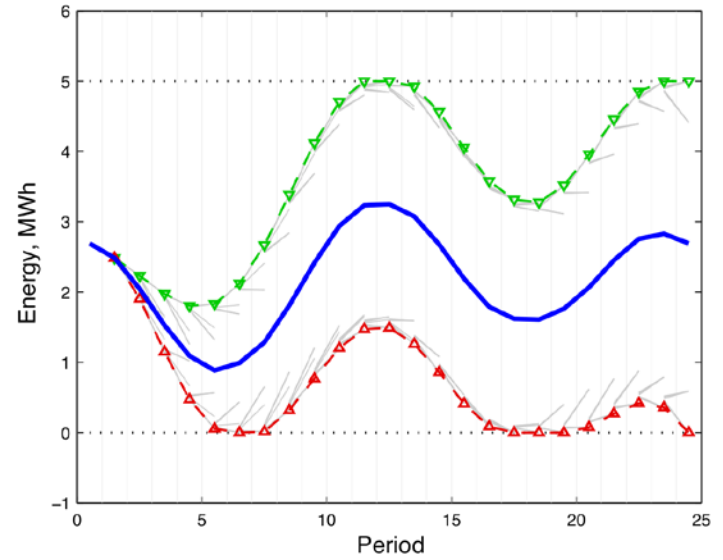
Wind Dispatch – 100% Uncertainty, 25% Variability



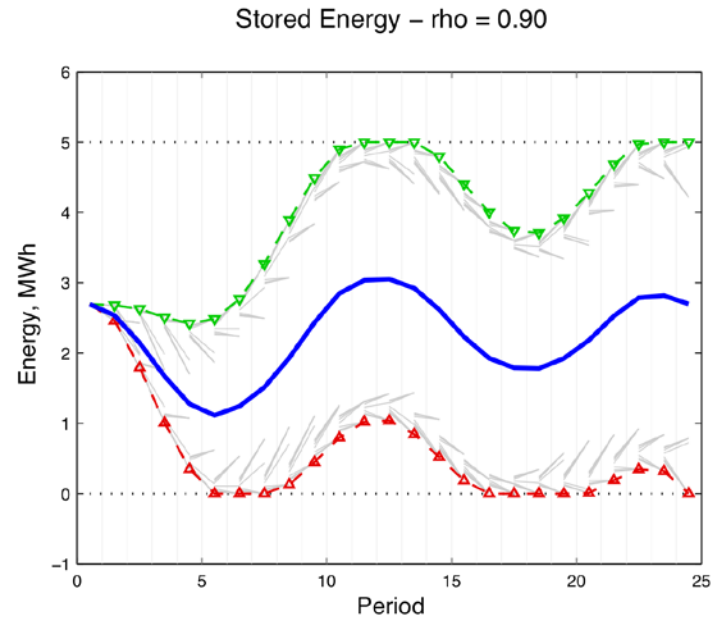
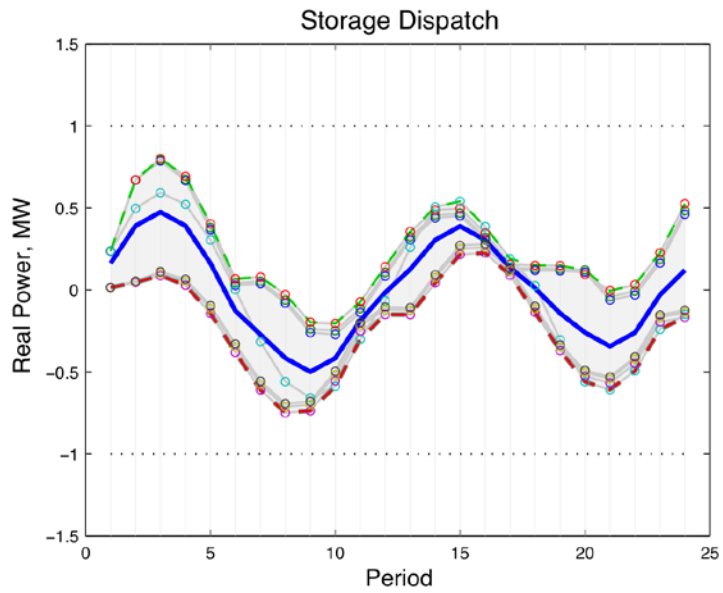
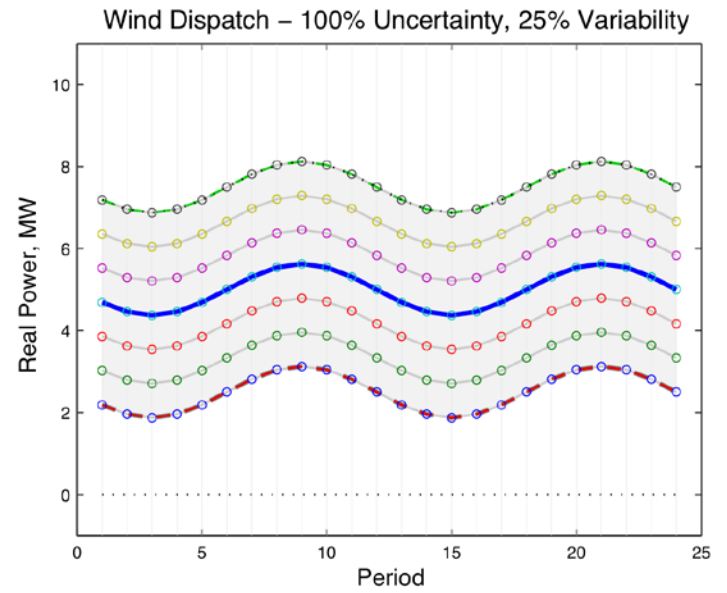
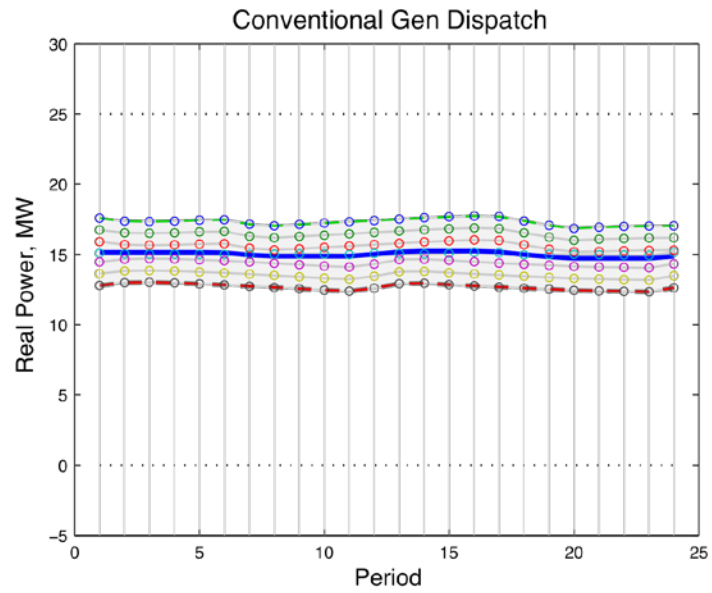
Storage Dispatch

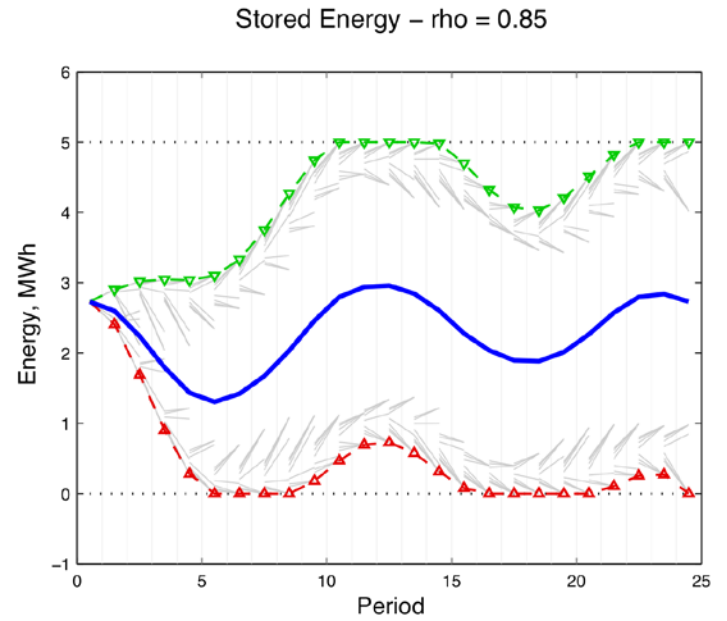
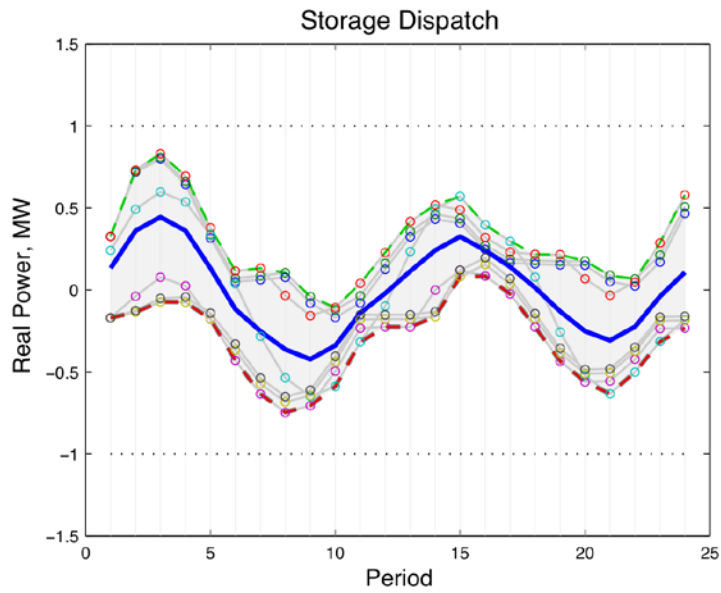
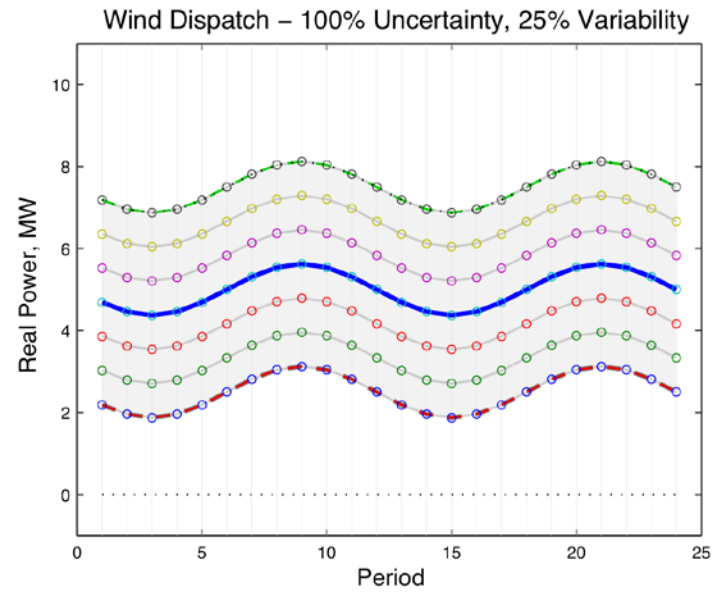
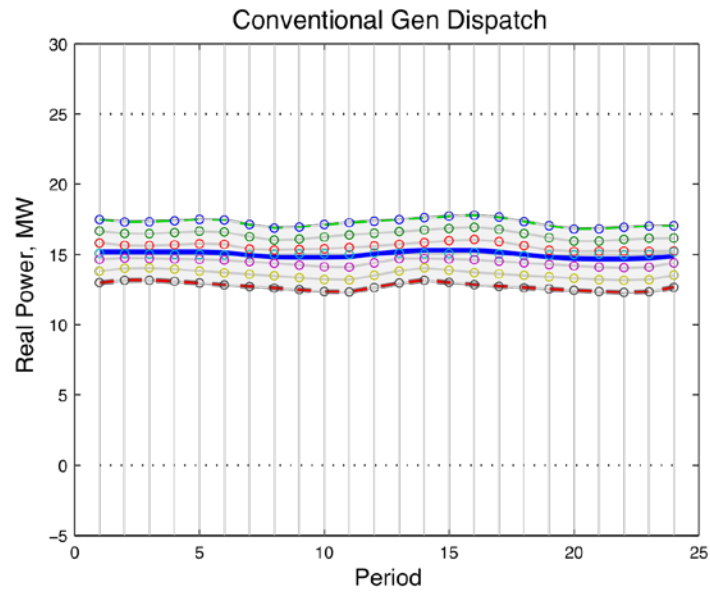


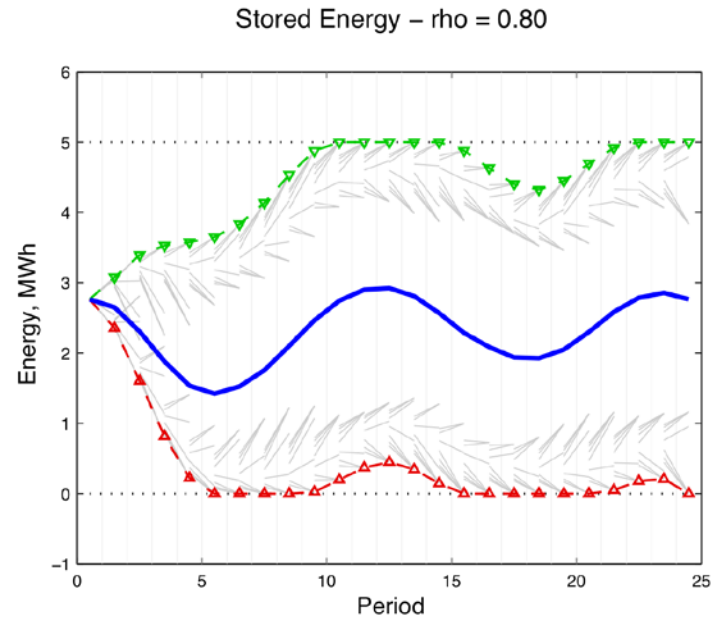
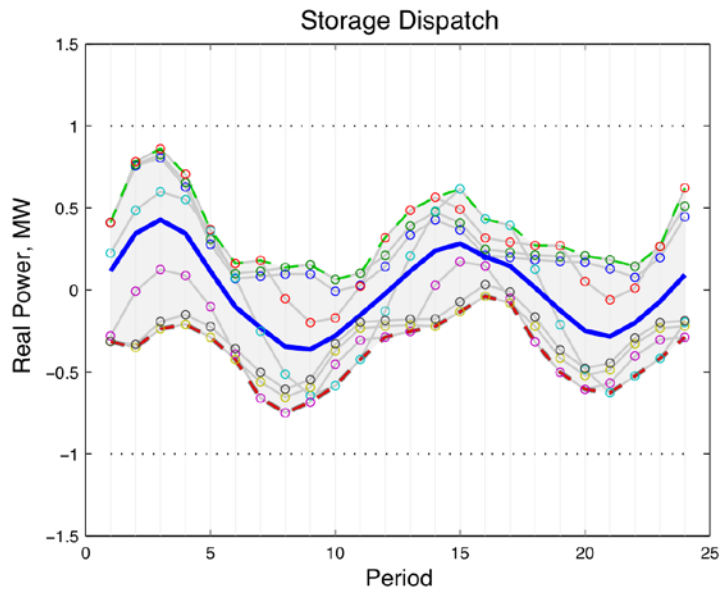
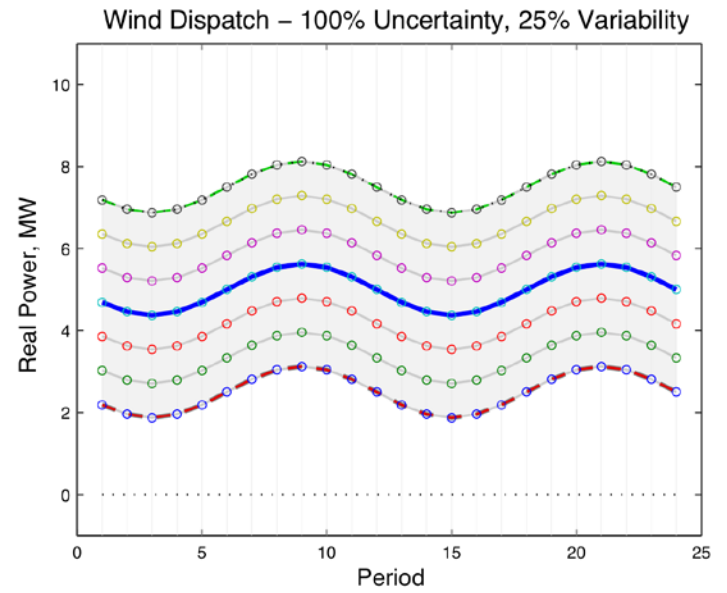
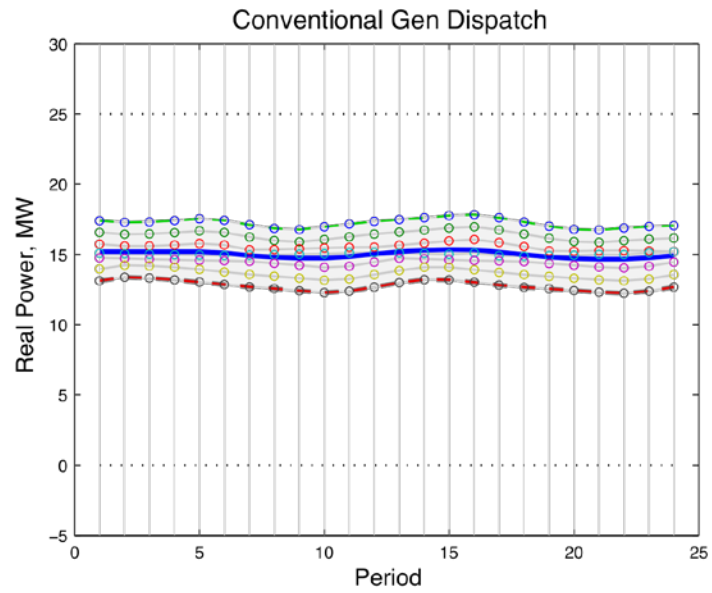
Stored Energy – rho = 0.95

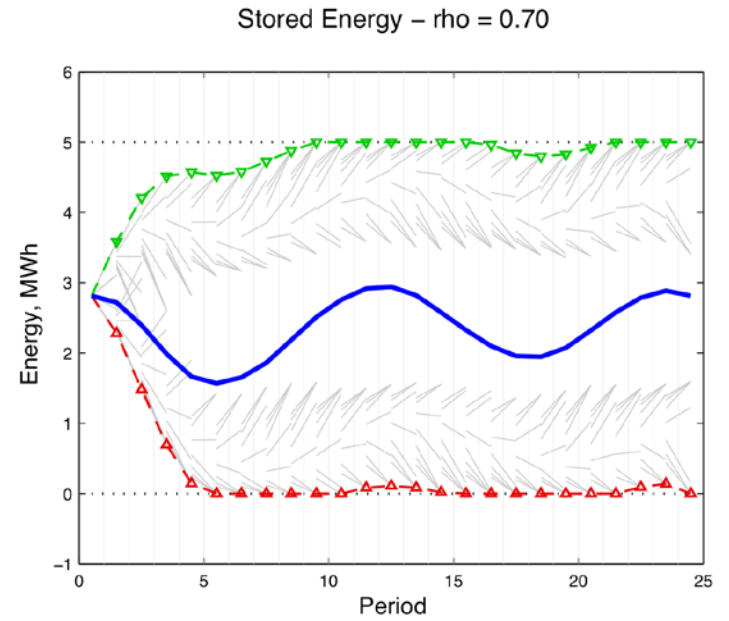
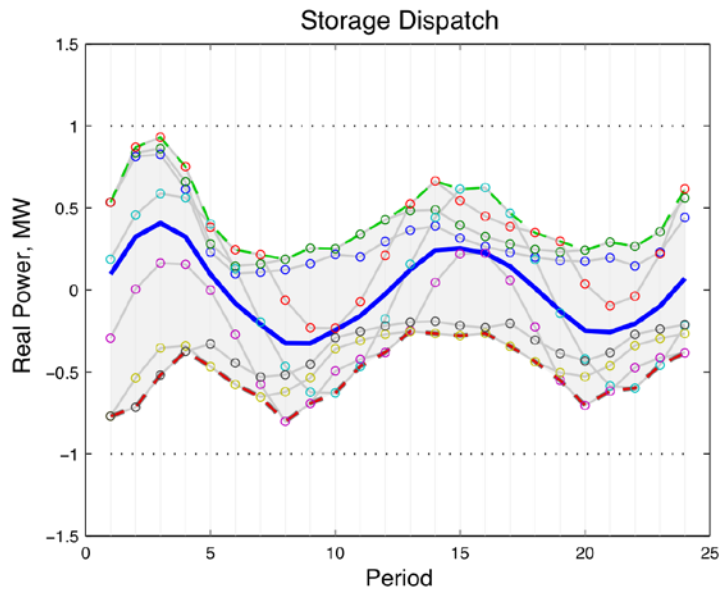
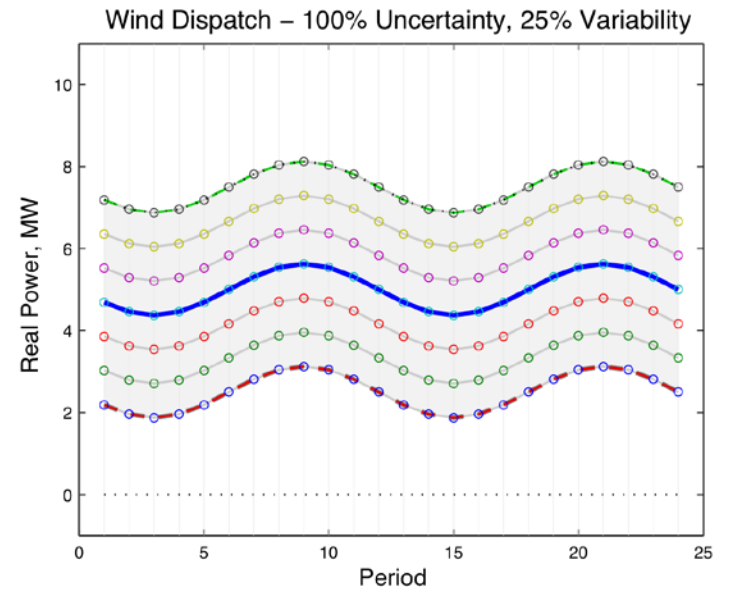
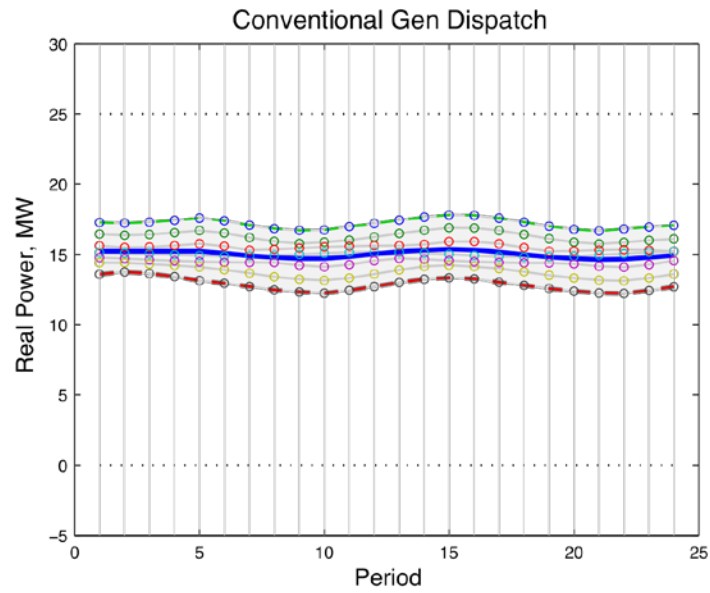


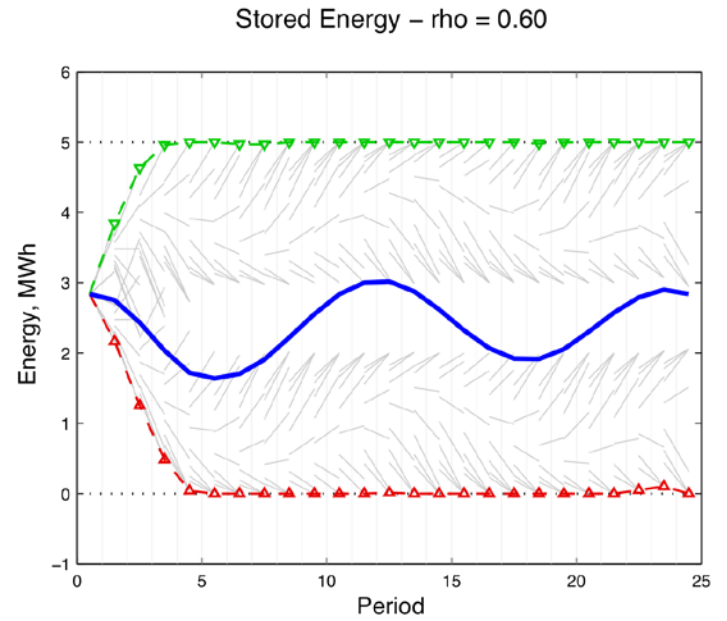
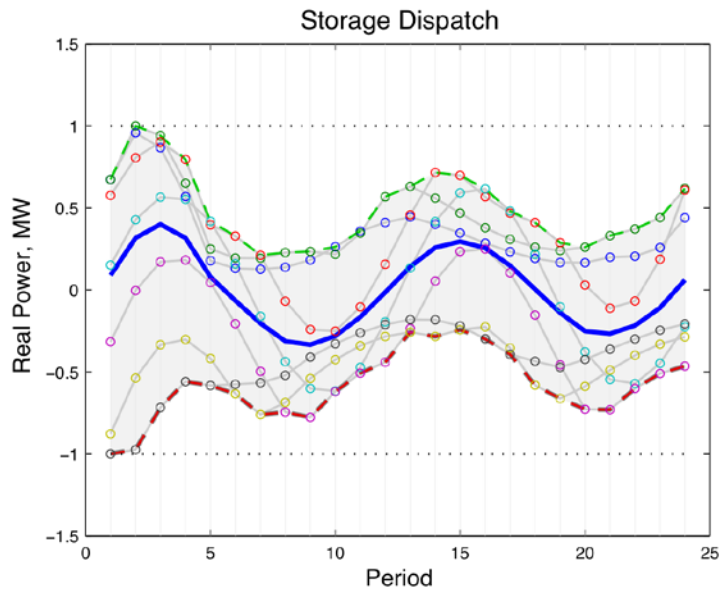
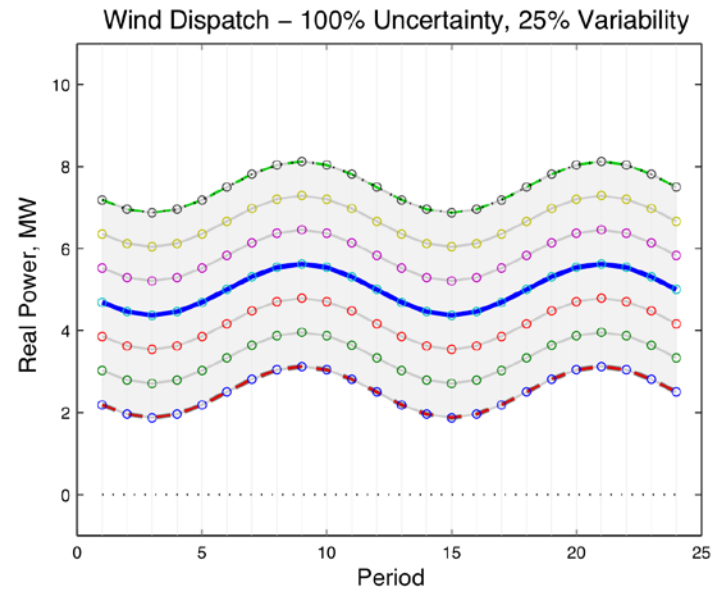
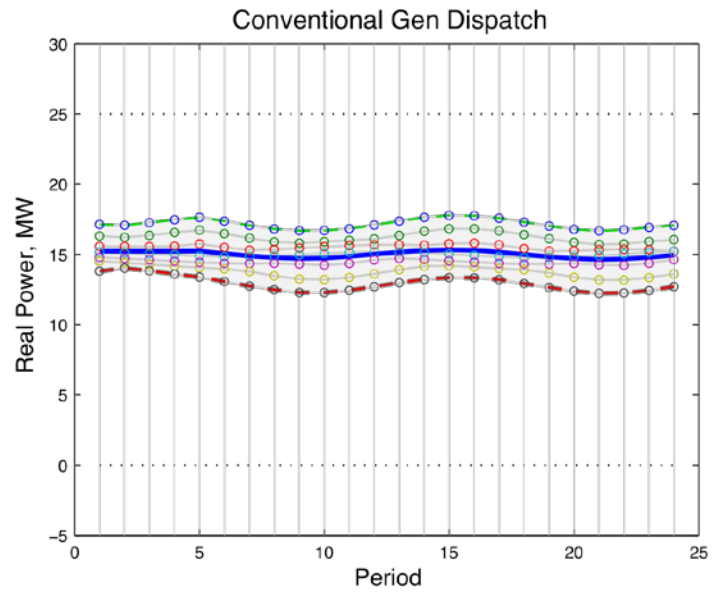


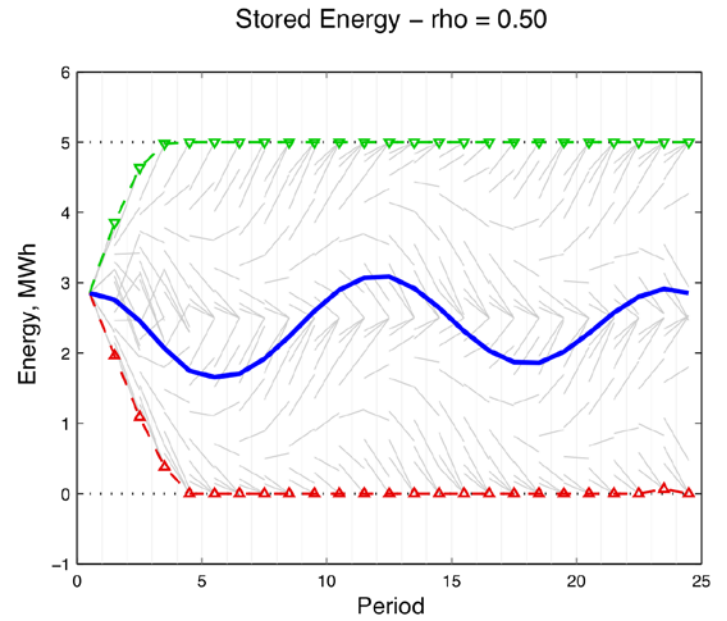
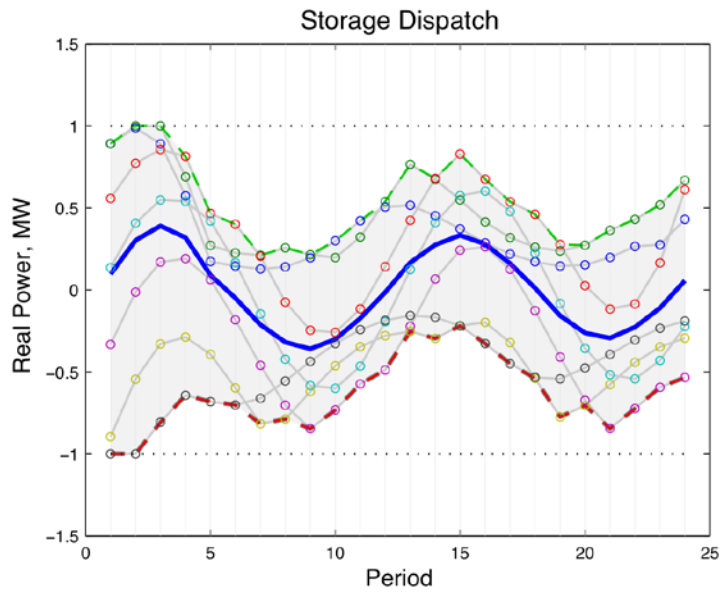
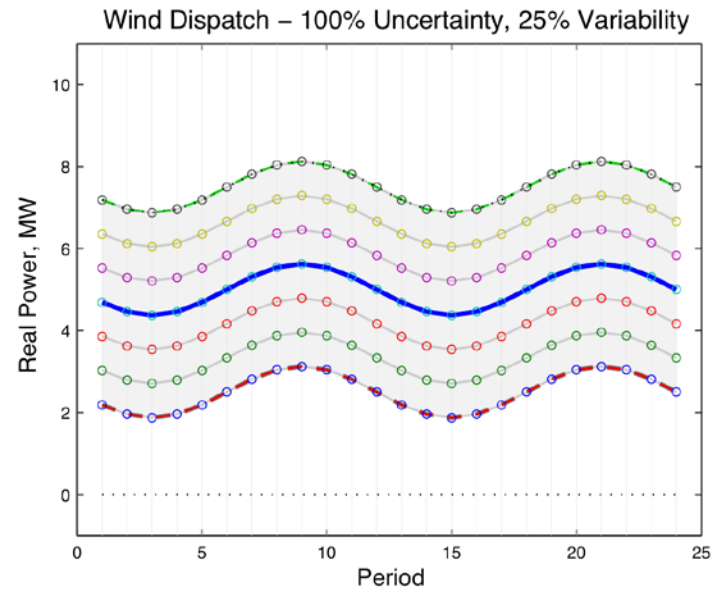
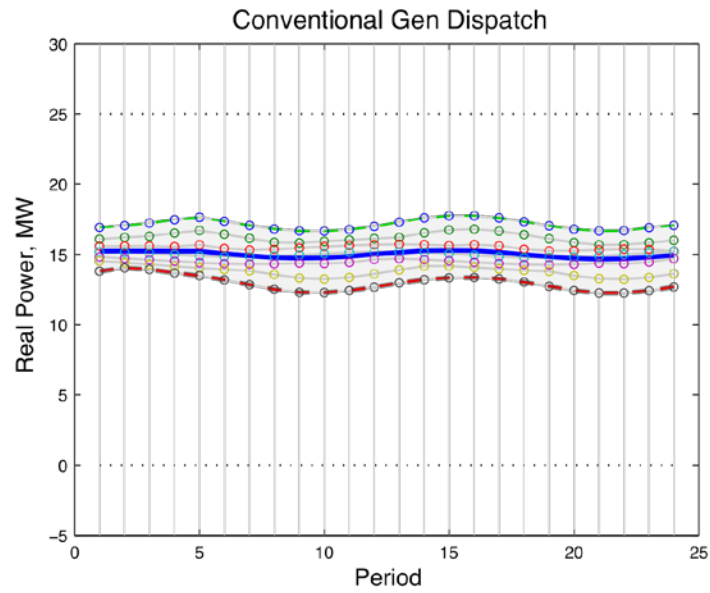


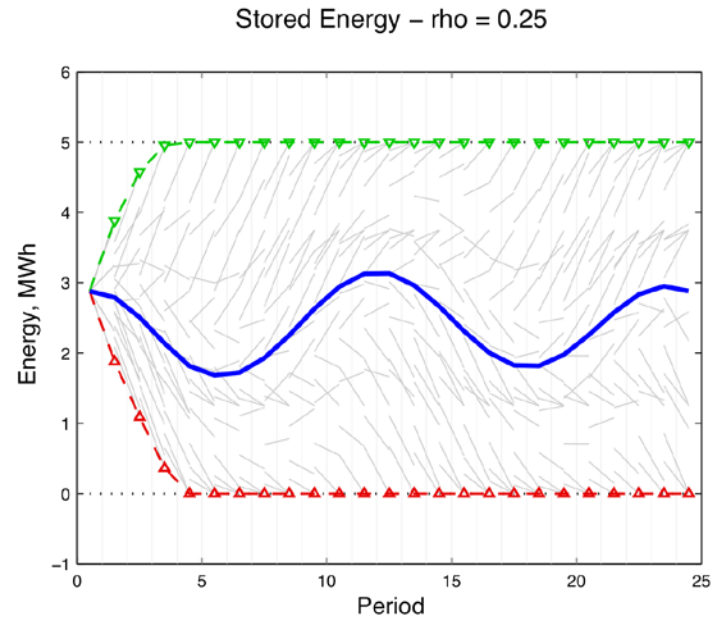
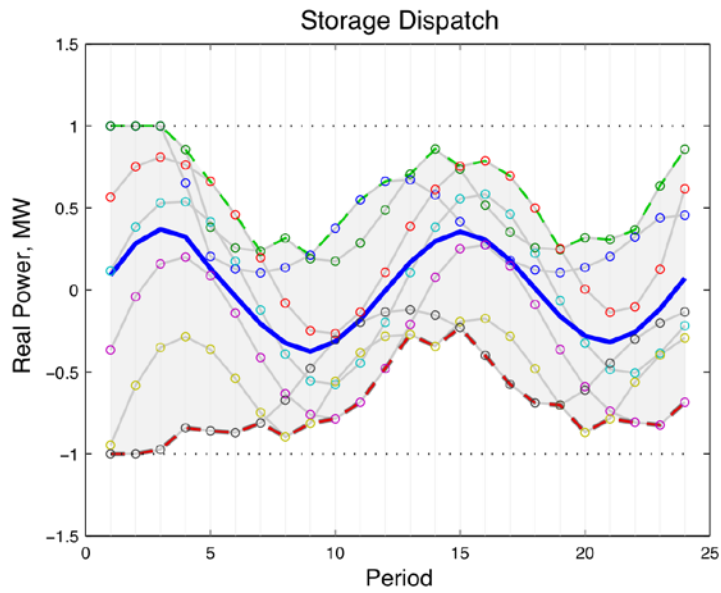
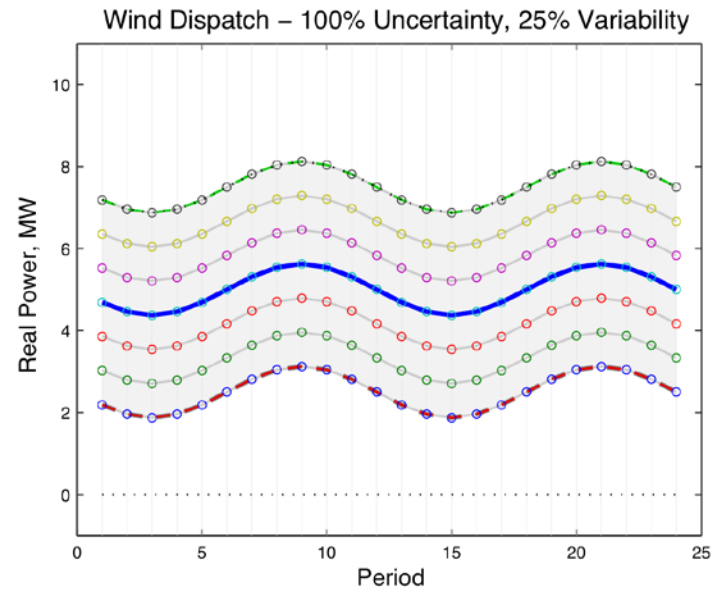
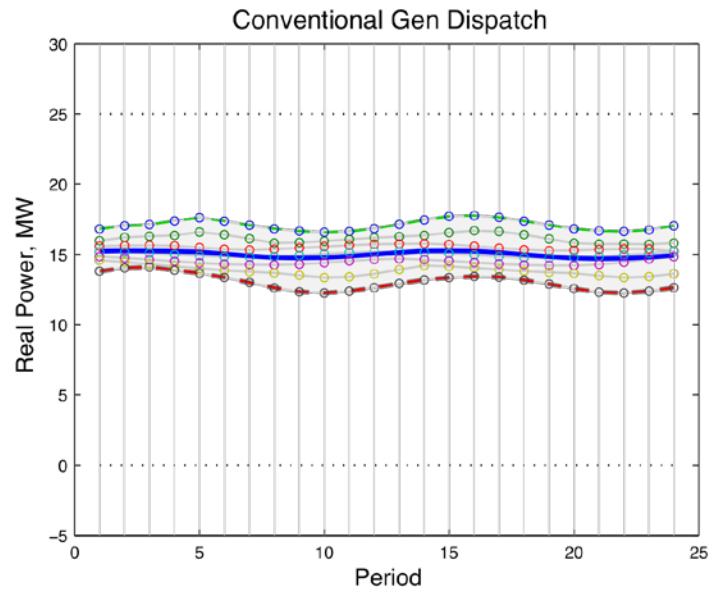


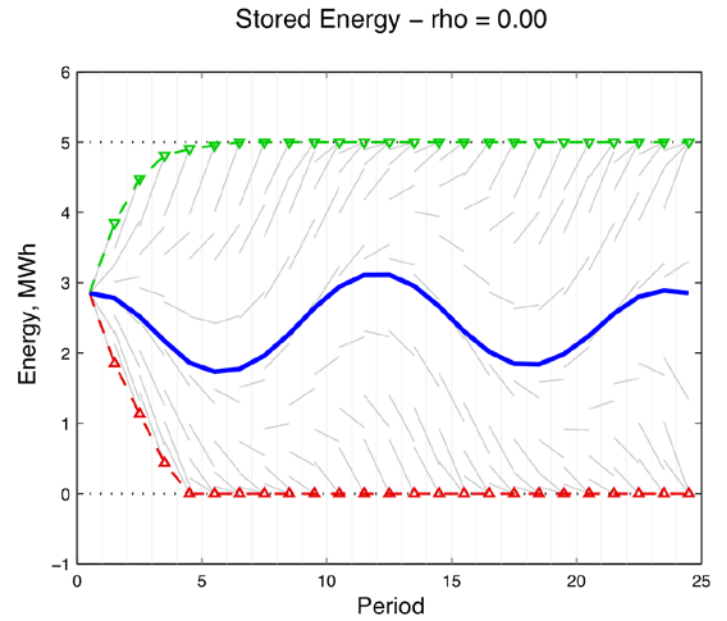
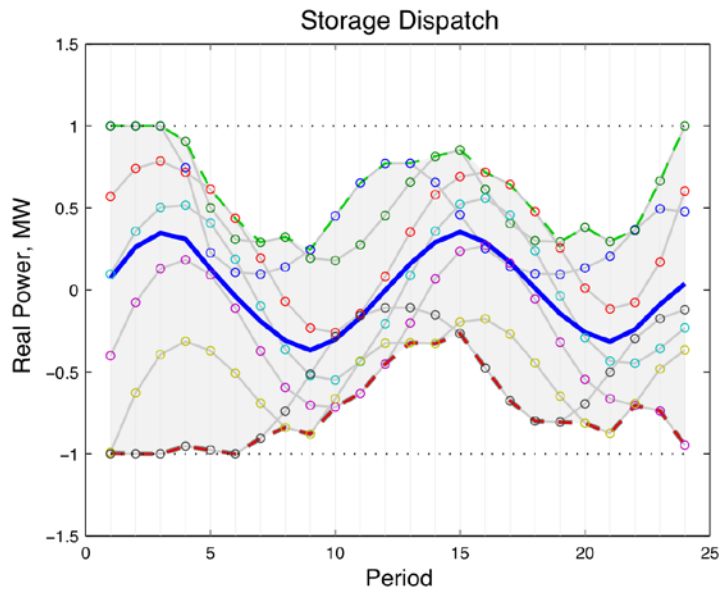
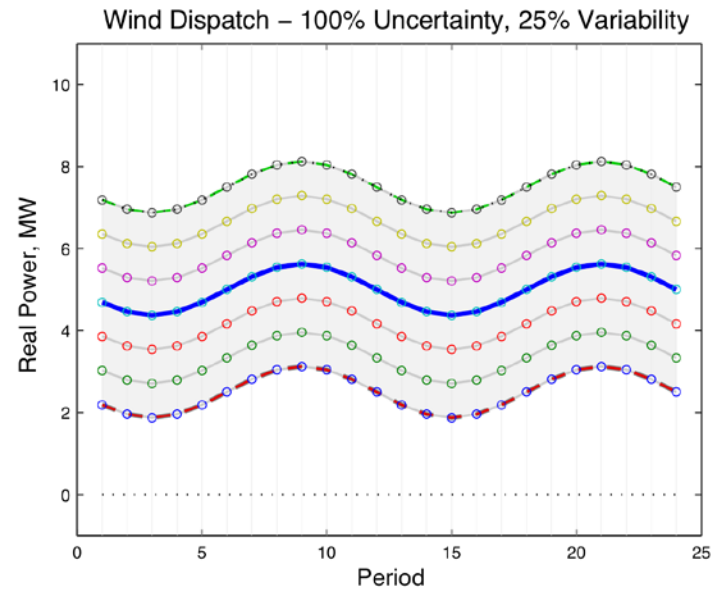
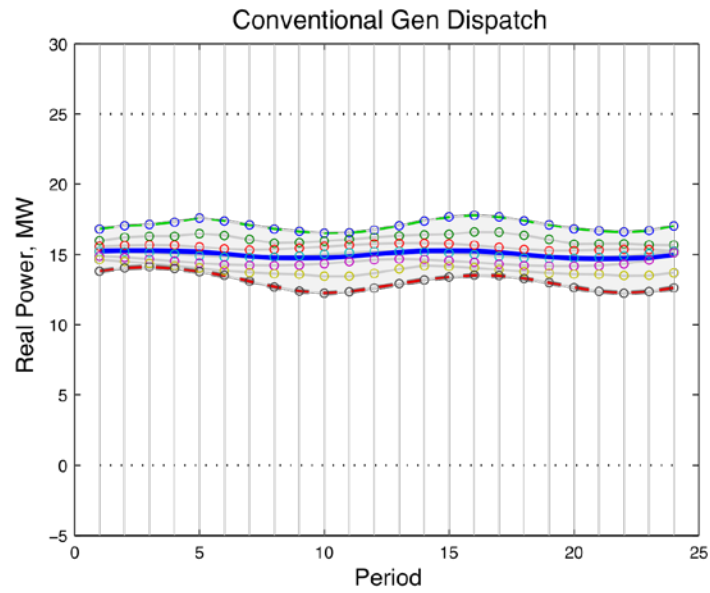








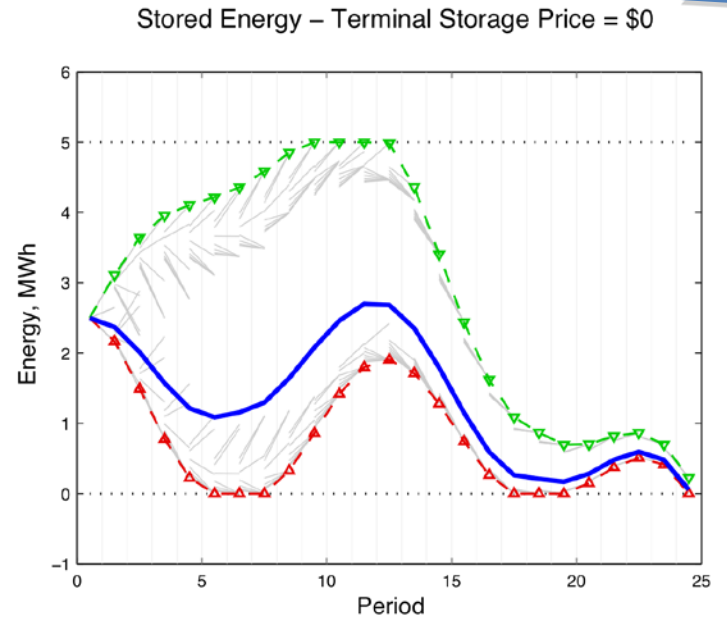
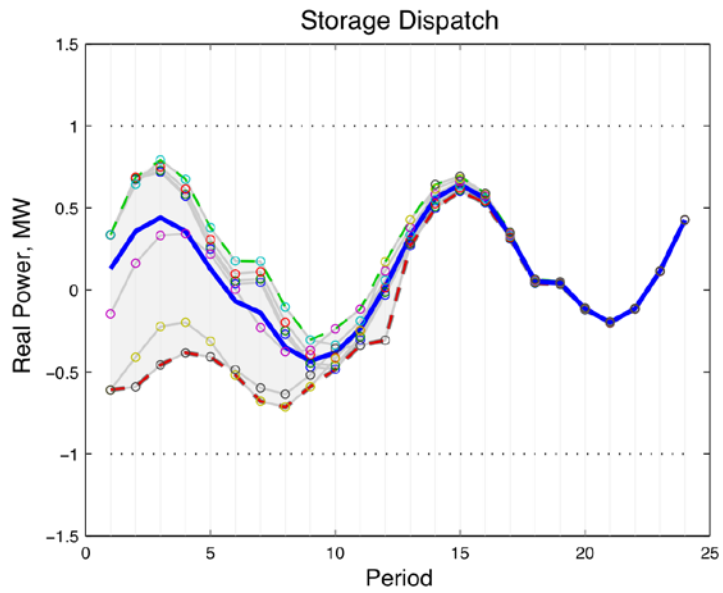
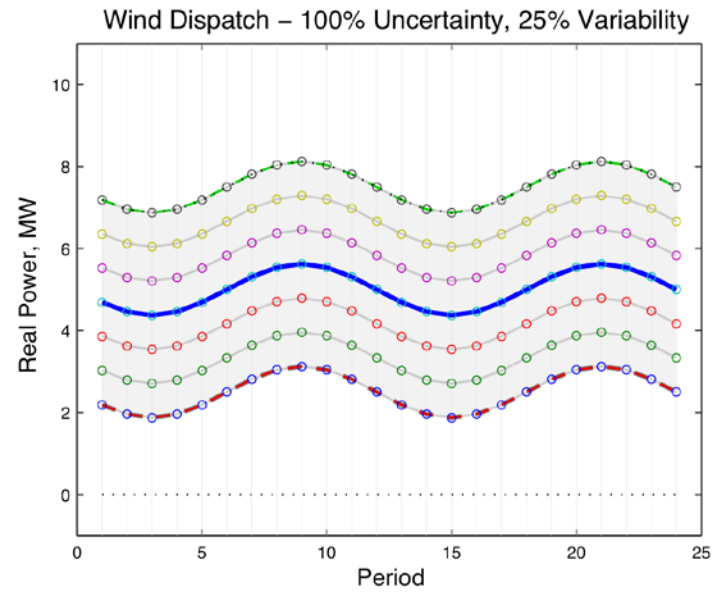
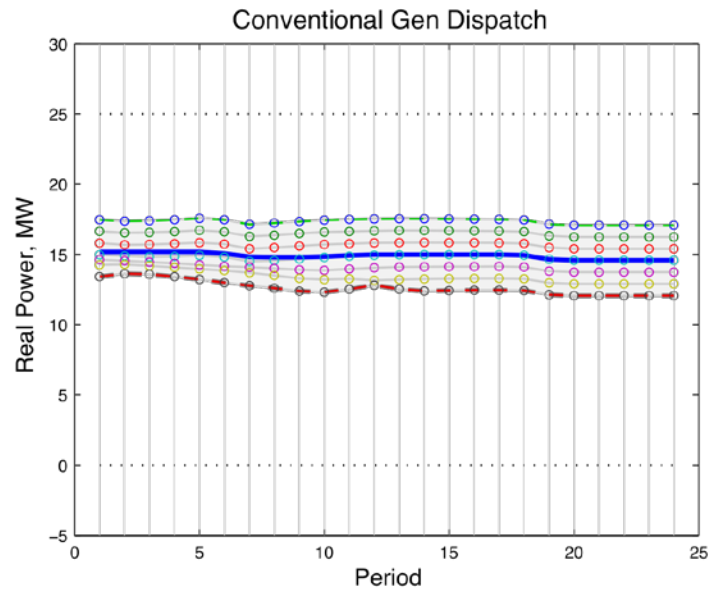




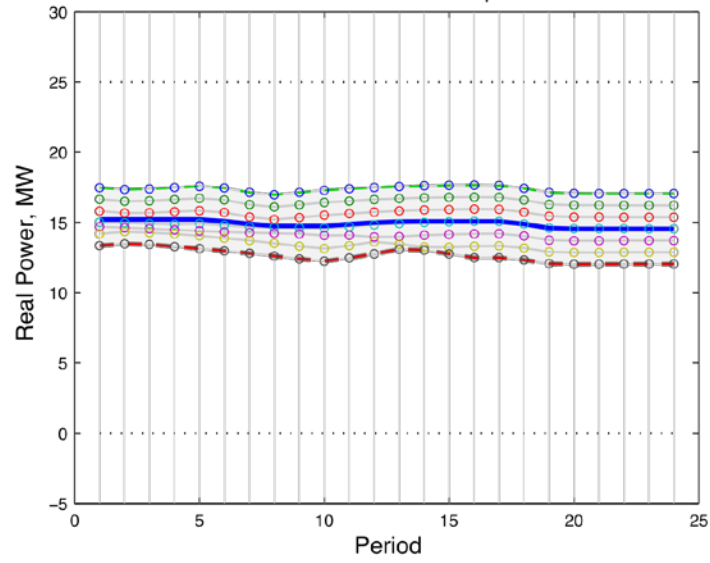


# Value of Leftover Storage in Terminal States

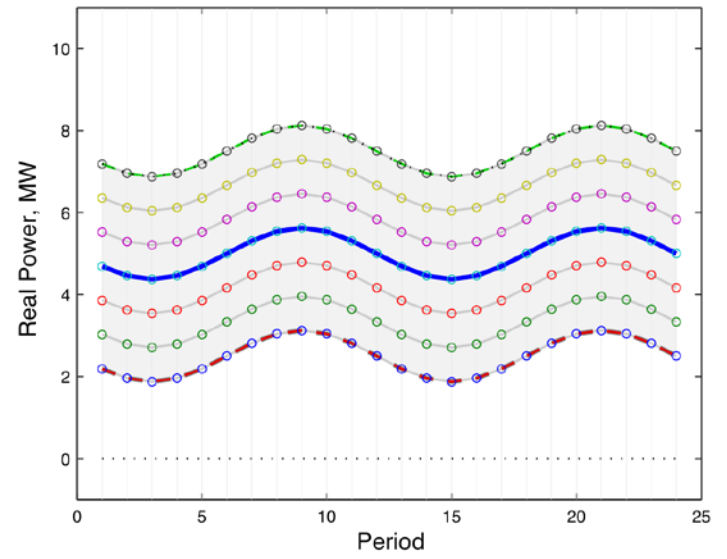
- 5 price model, allowing to specify a unique value for each of the following 5 types of contributions to the total expected value of leftover stored energy:
  - charging or discharging in non-terminal states
  - charging in terminal end-of-horizon base states
  - discharging in terminal end-of-horizon base states
  - charging in terminal contingency states
  - discharging in terminal contingency states
- optional cyclic storage constraint
  - initial stored energy is a variable, constrained to equal expected final stored energy
- optional target constraint for expected stored energy



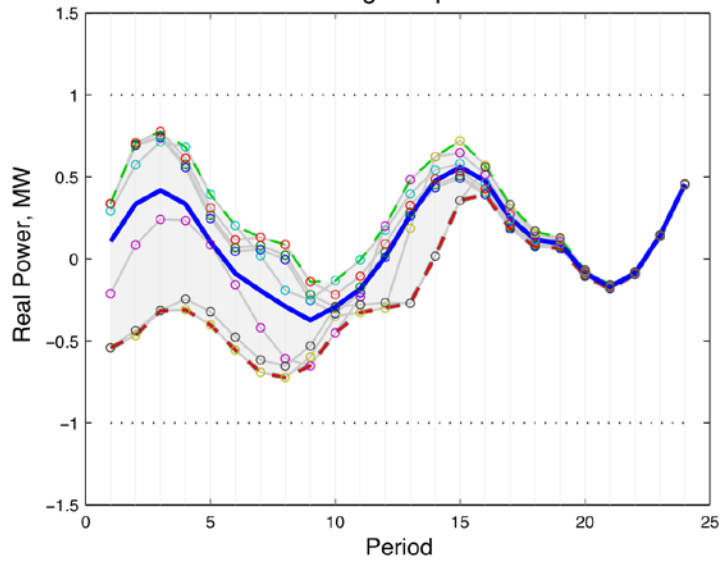
Conventional Gen Dispatch



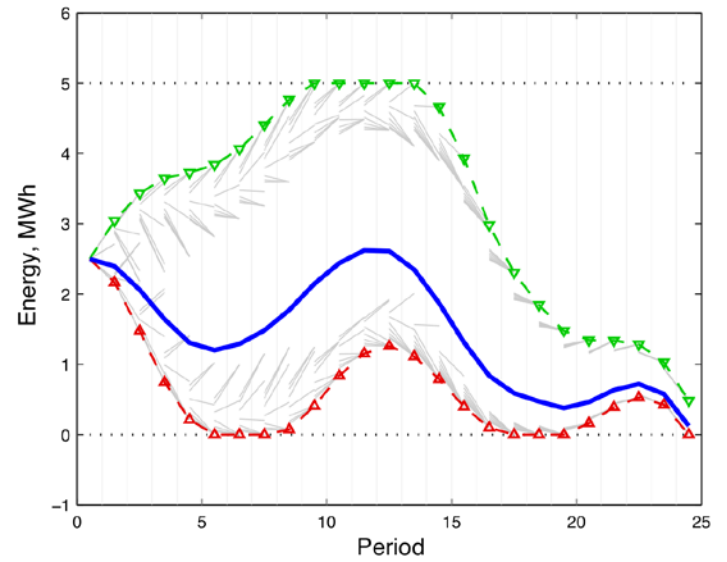
Wind Dispatch – 100% Uncertainty, 25% Variability

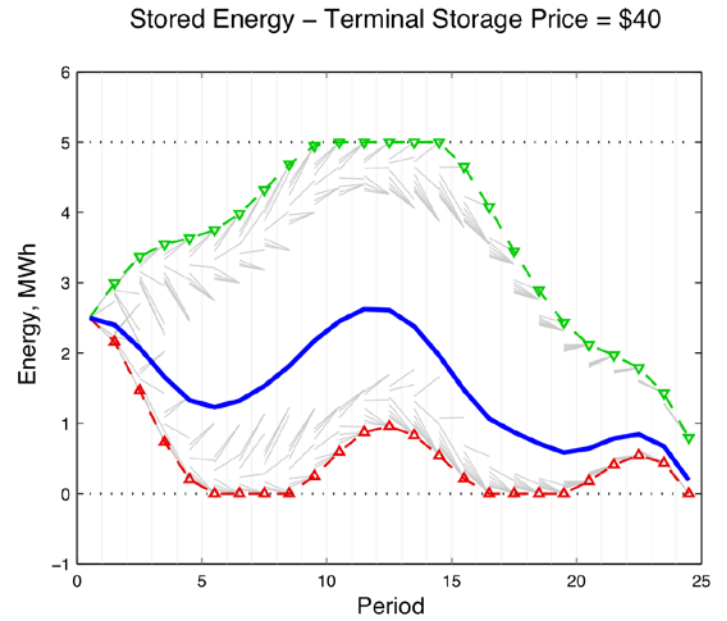
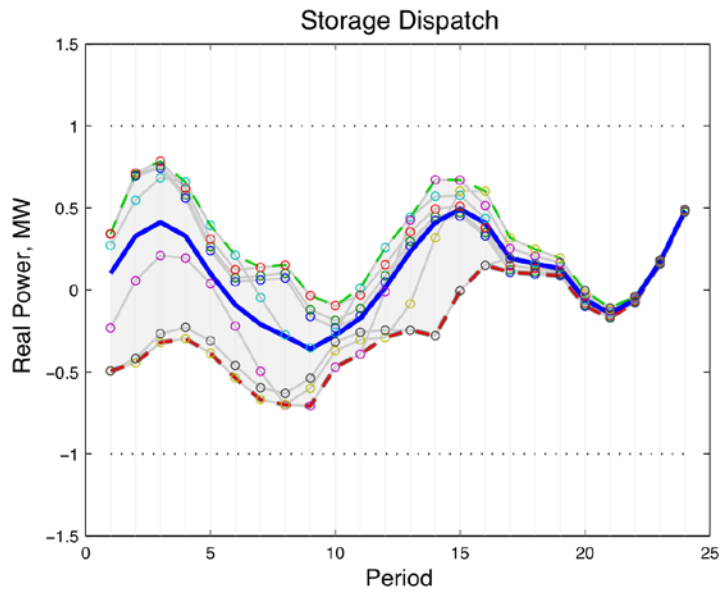
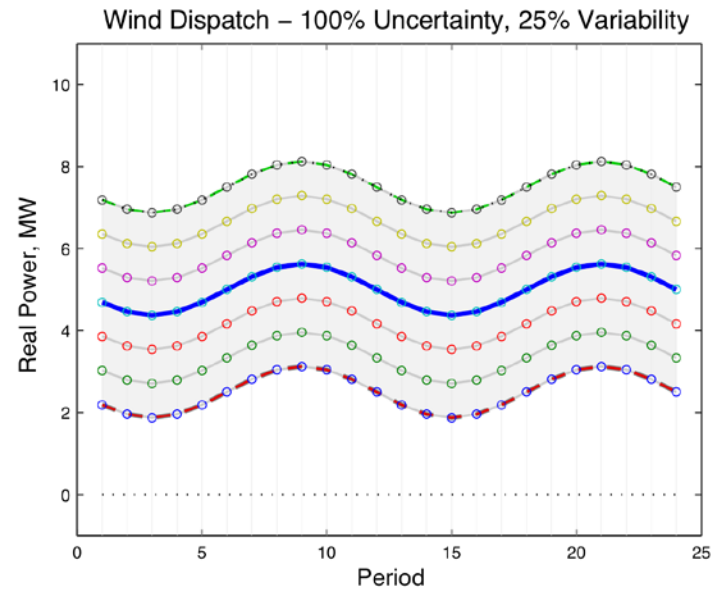
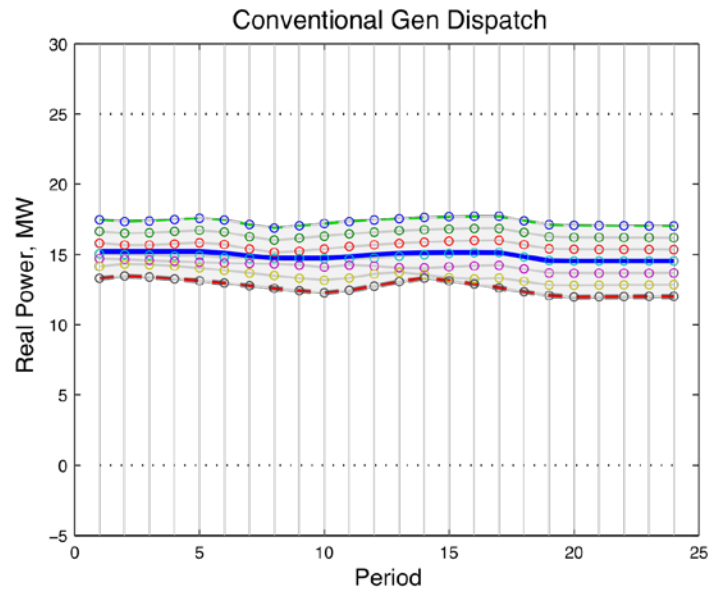


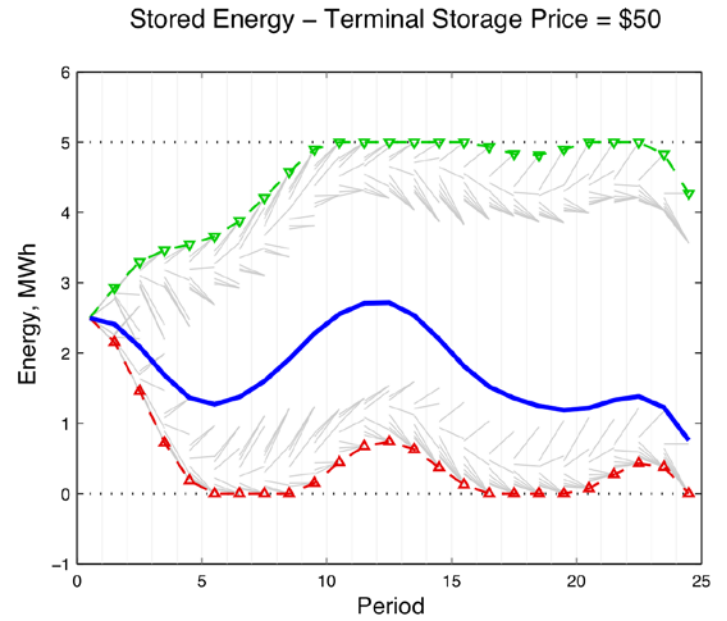
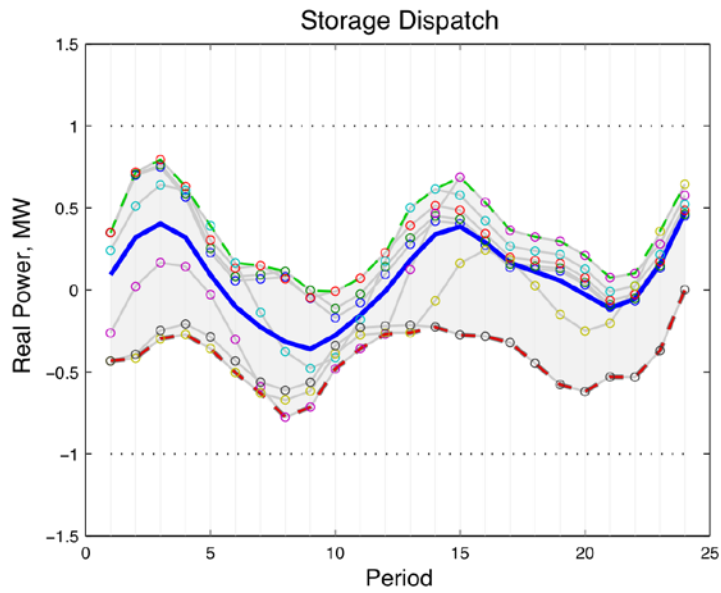
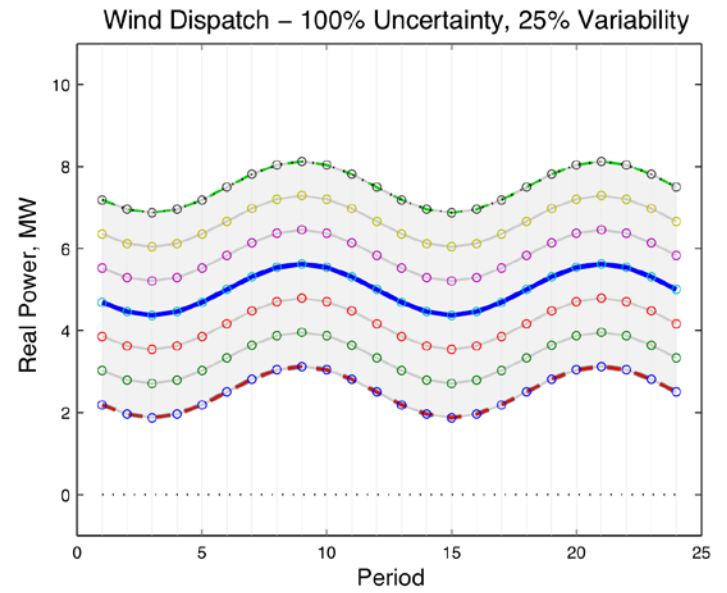
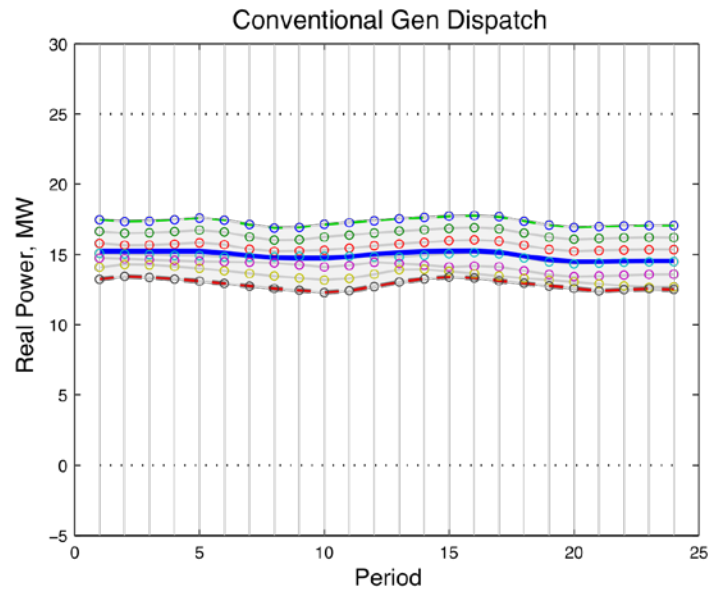
Storage Dispatch



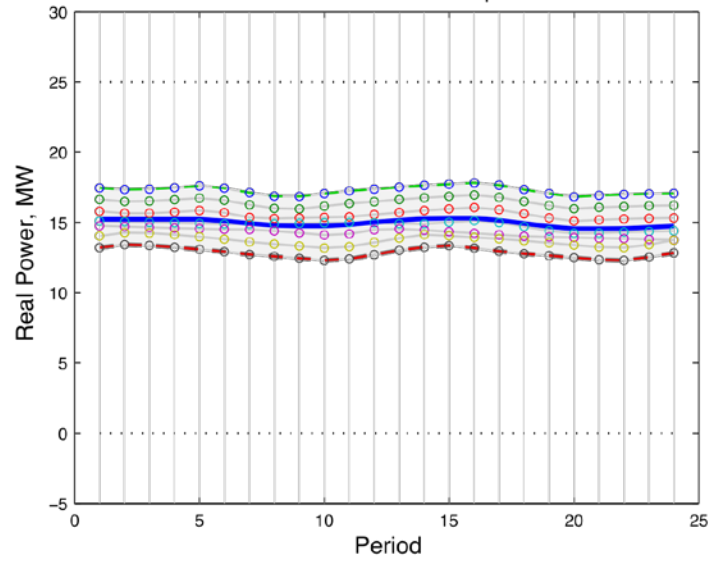
Stored Energy – Terminal Storage Price = \$30



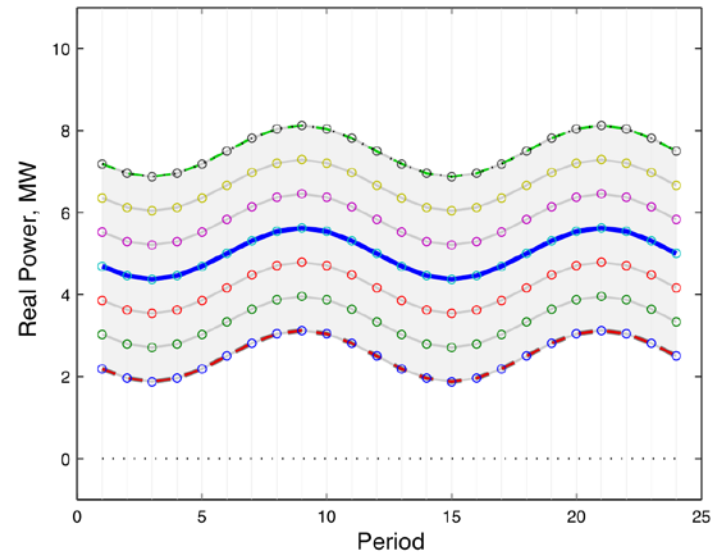




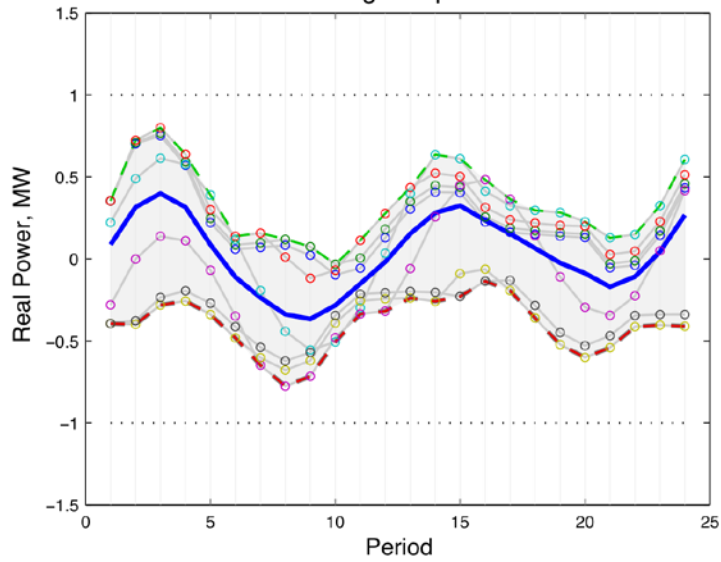
Conventional Gen Dispatch



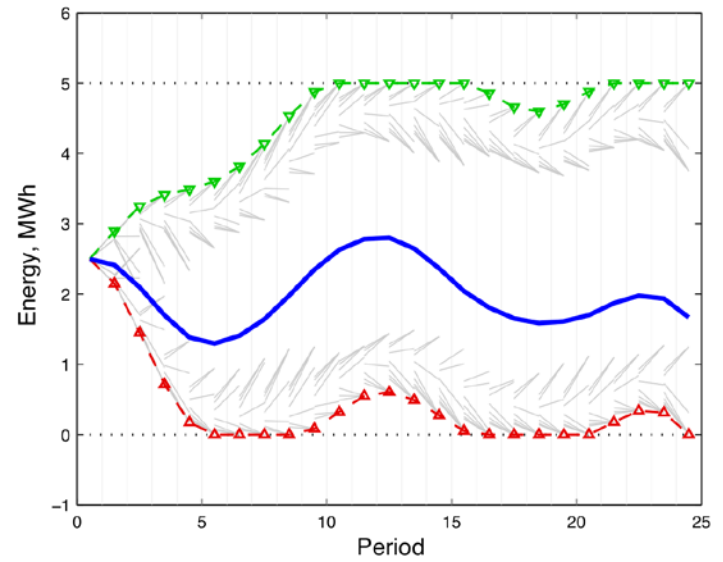
Wind Dispatch – 100% Uncertainty, 25% Variability

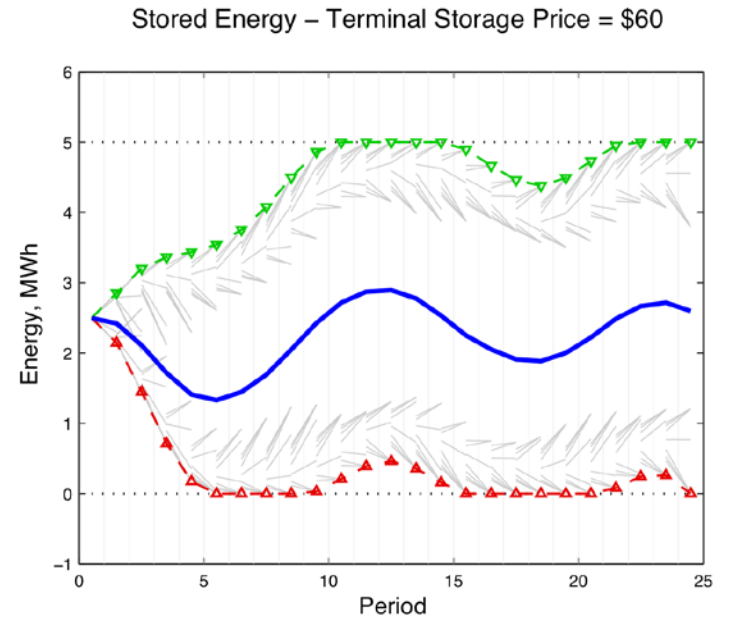
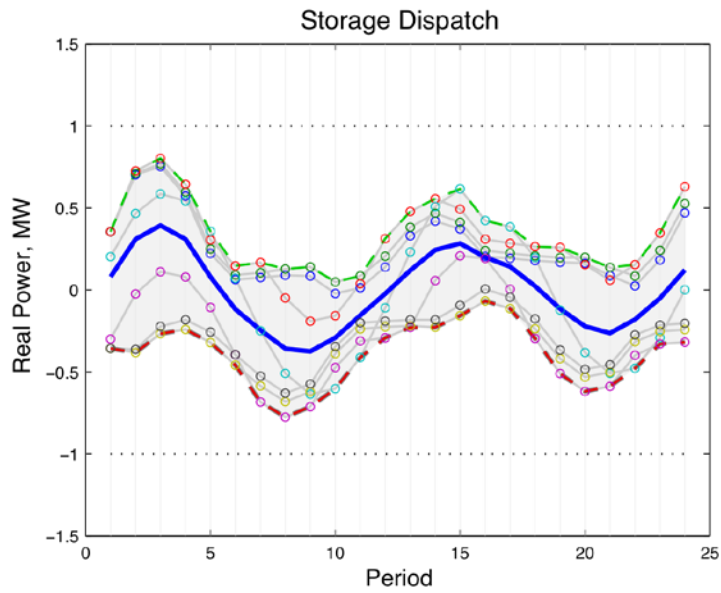
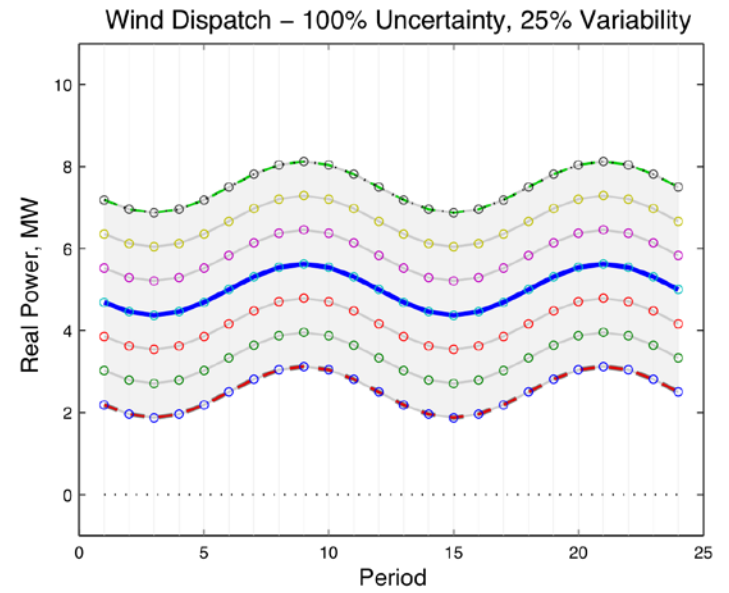
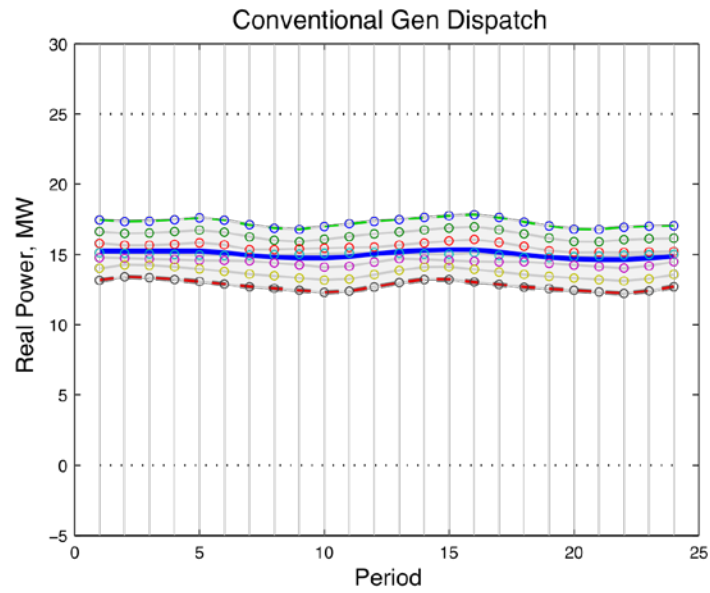


Storage Dispatch

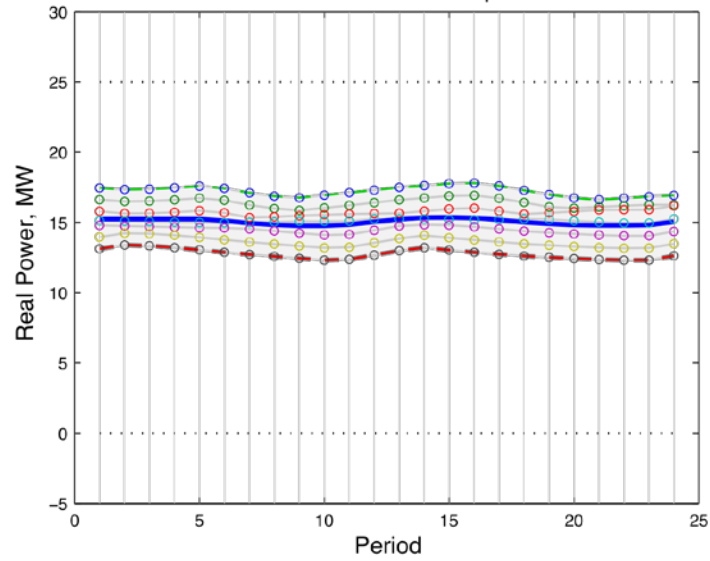


Stored Energy – Terminal Storage Price = \$55

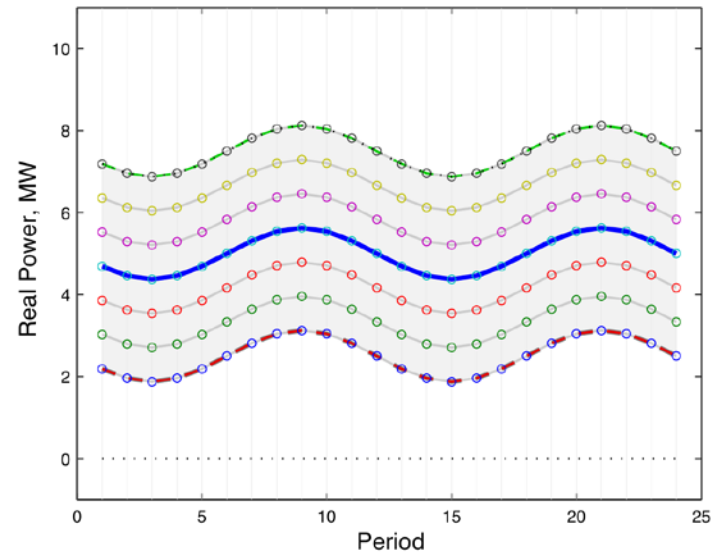




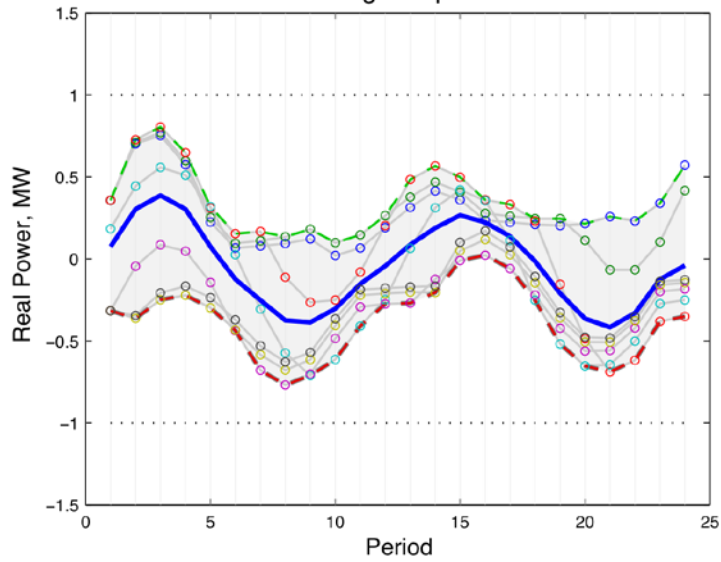
Conventional Gen Dispatch



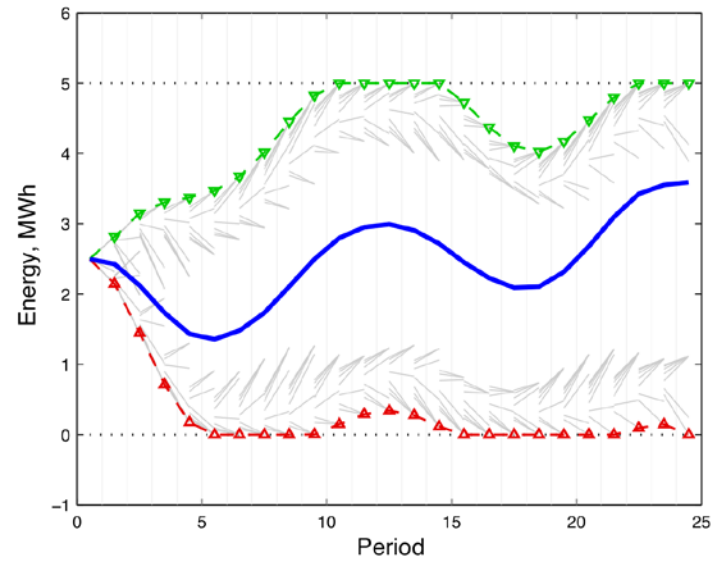
Wind Dispatch – 100% Uncertainty, 25% Variability



Storage Dispatch

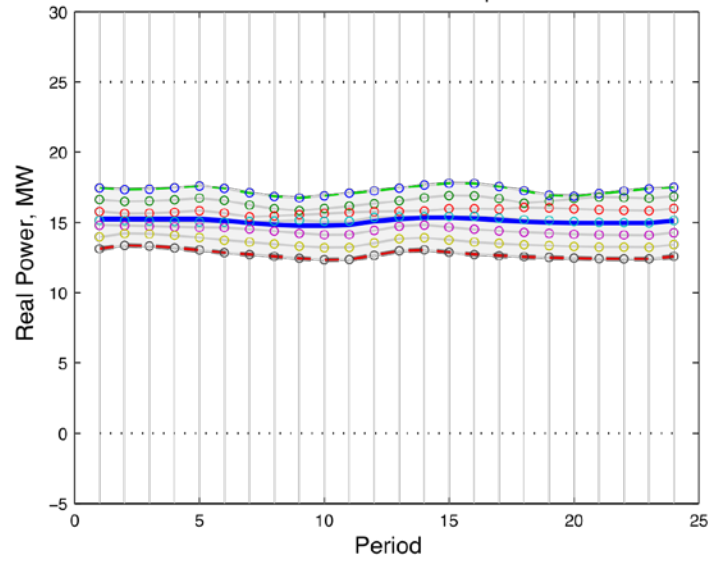


Stored Energy – Terminal Storage Price = \$65

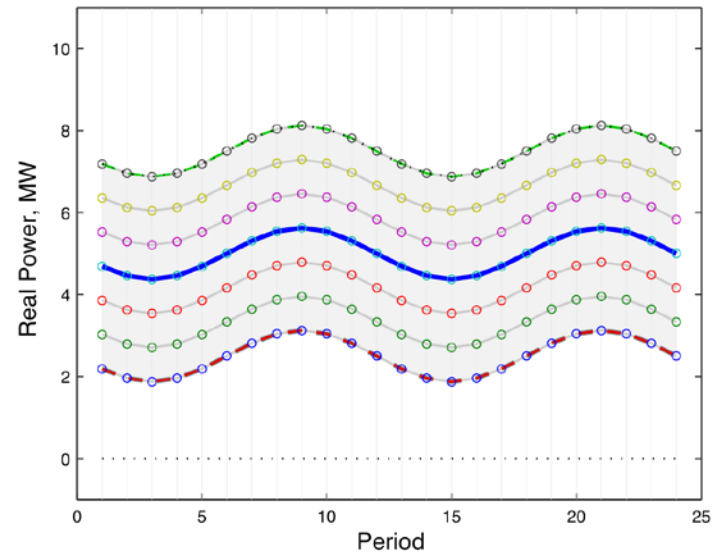




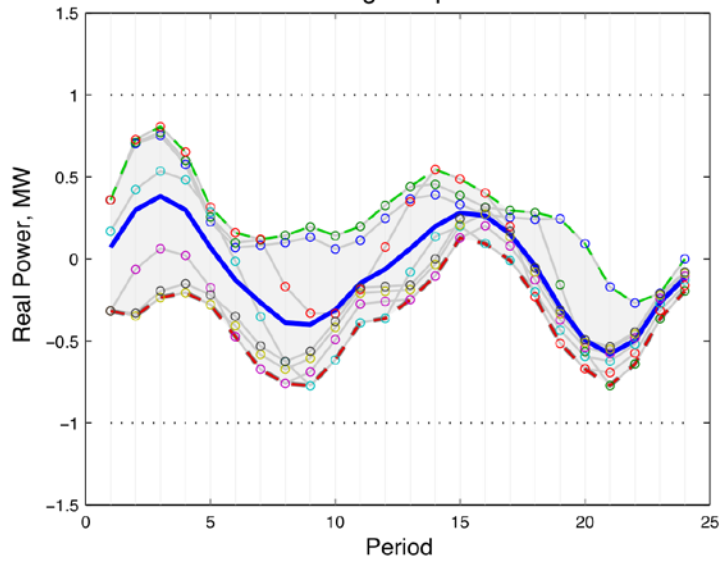
Conventional Gen Dispatch



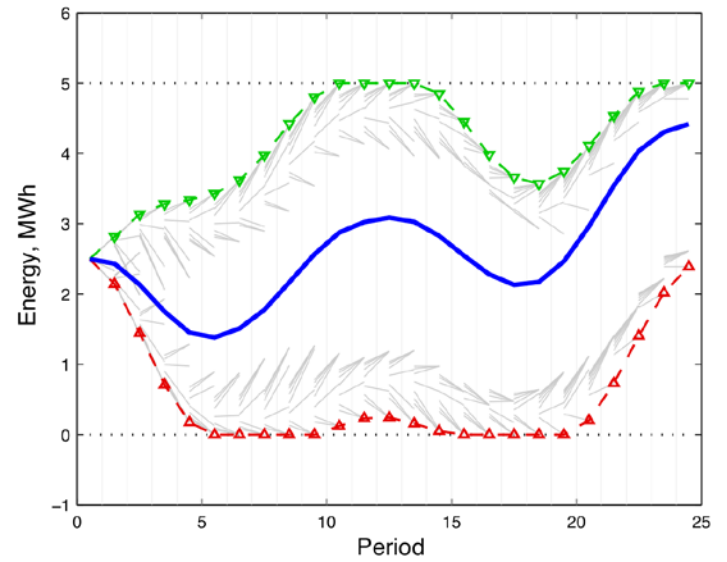
Wind Dispatch – 100% Uncertainty, 25% Variability



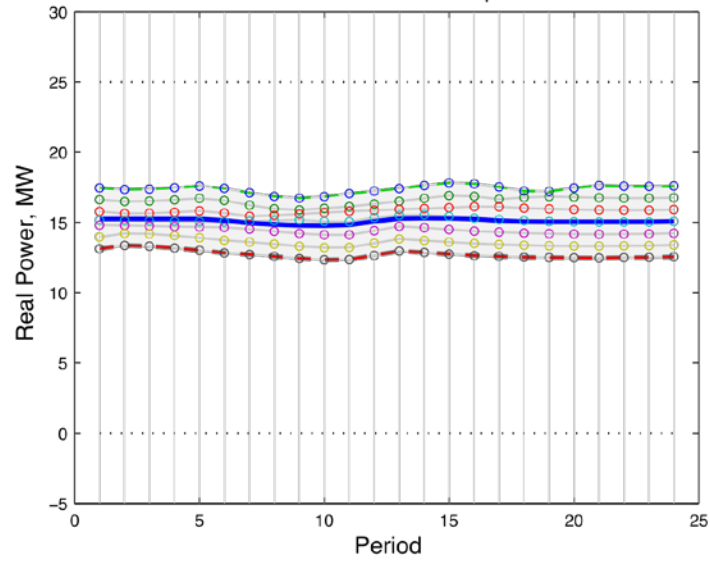
Storage Dispatch



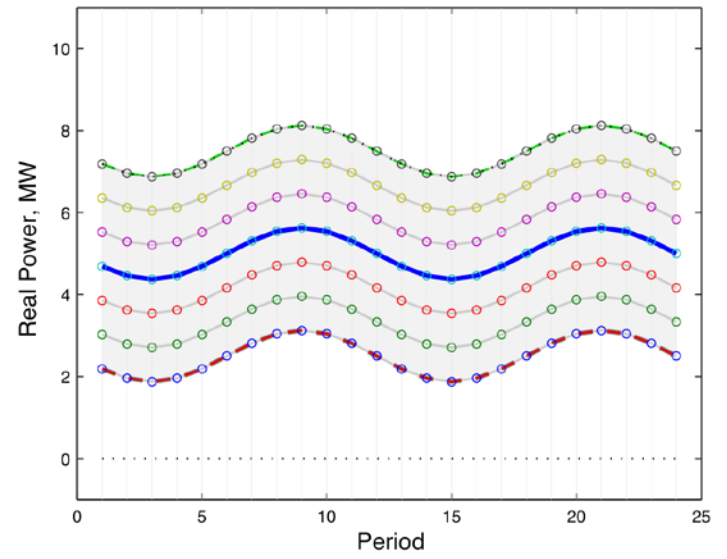
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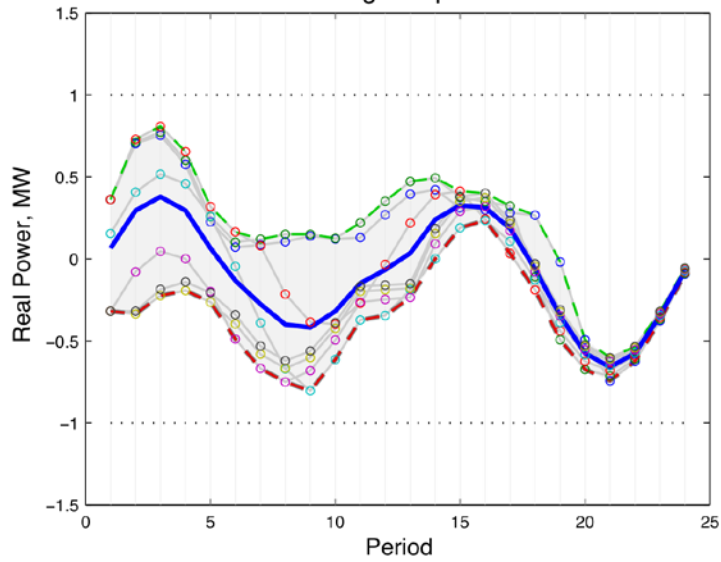
Conventional Gen Dispatch



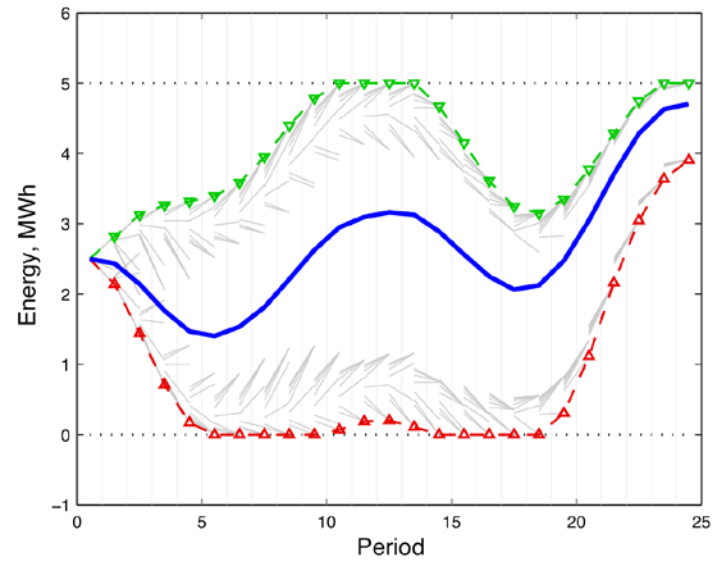
Wind Dispatch – 100% Uncertainty, 25% Variability



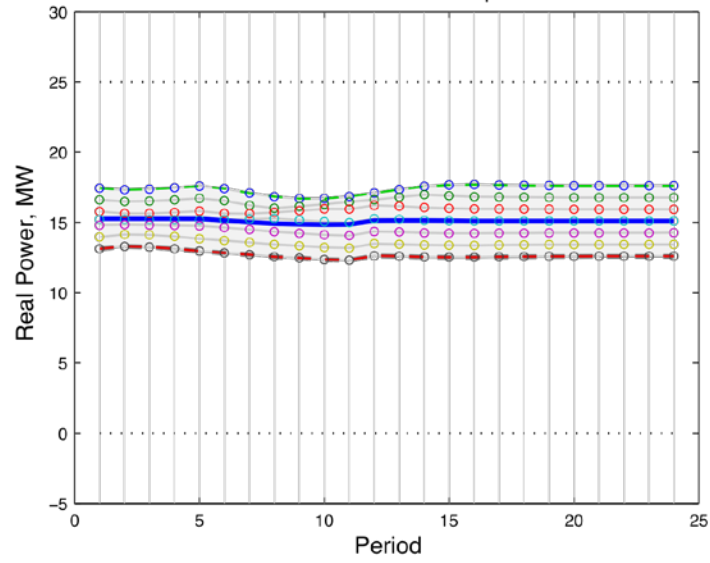
Storage Dispatch



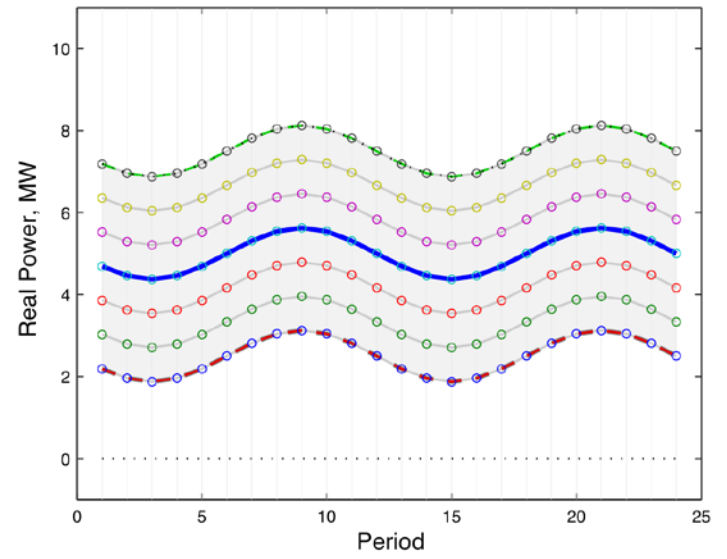
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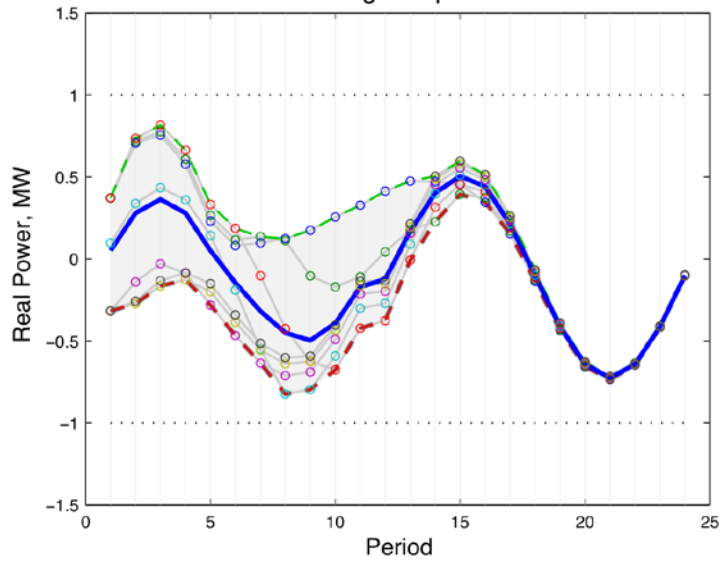
Conventional Gen Dispatch



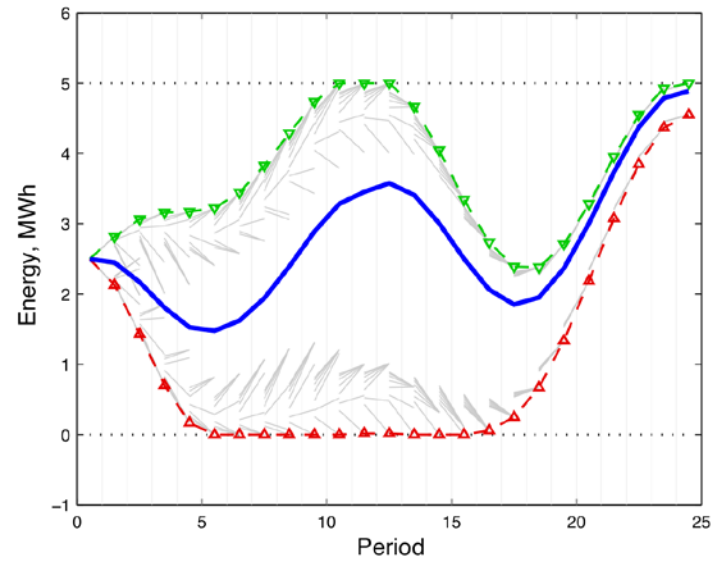
Wind Dispatch – 100% Uncertainty, 25% Variability



Storage Dispatch

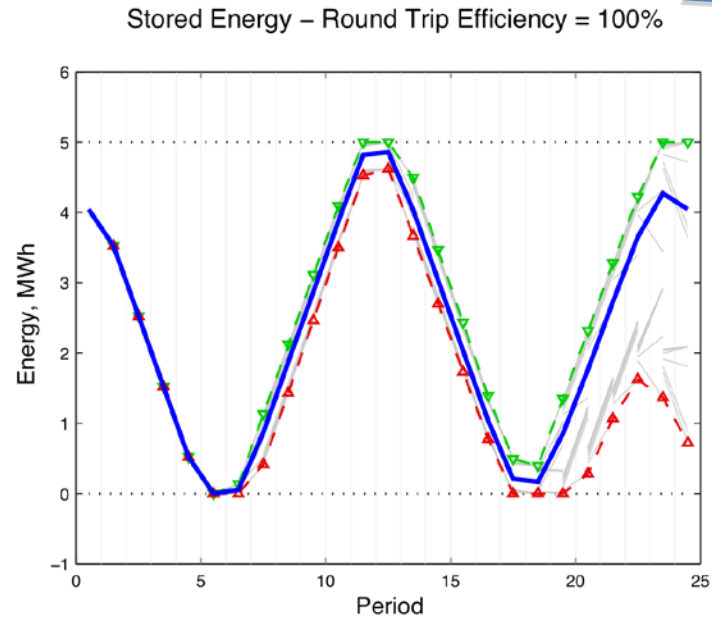
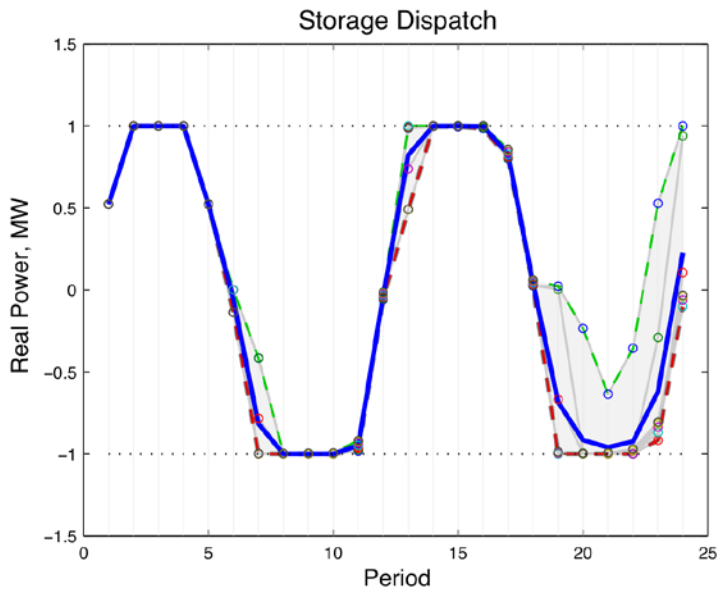
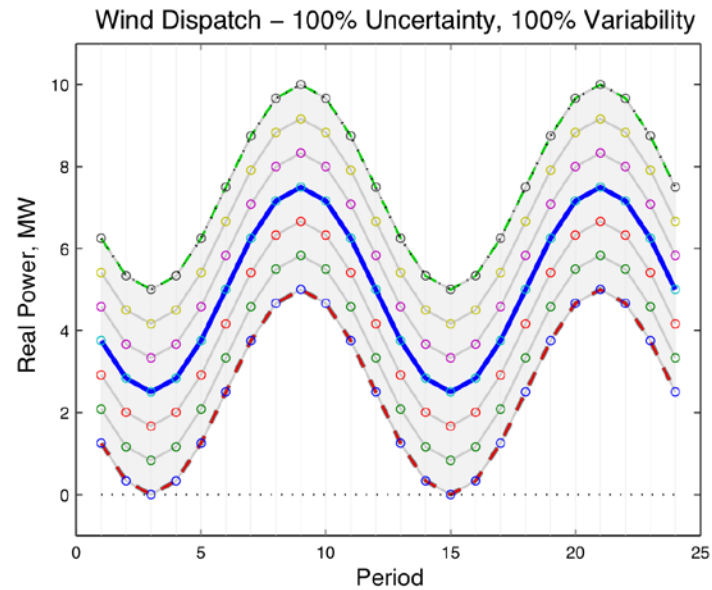
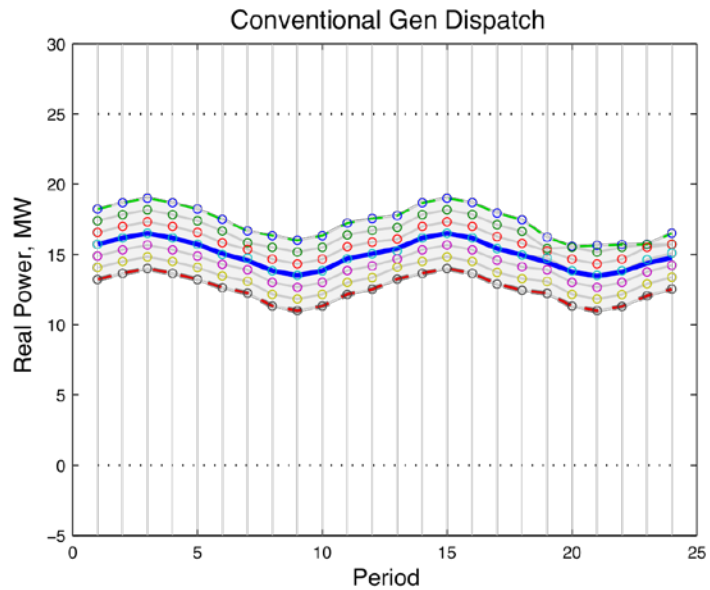


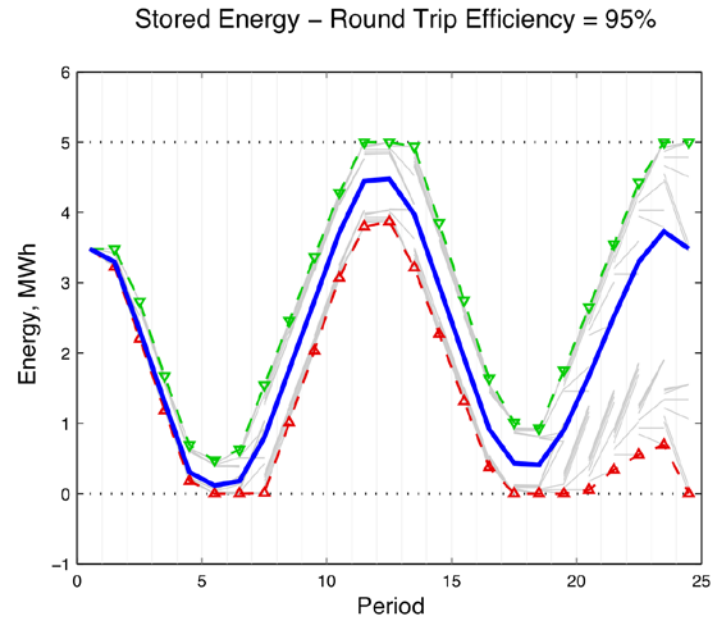
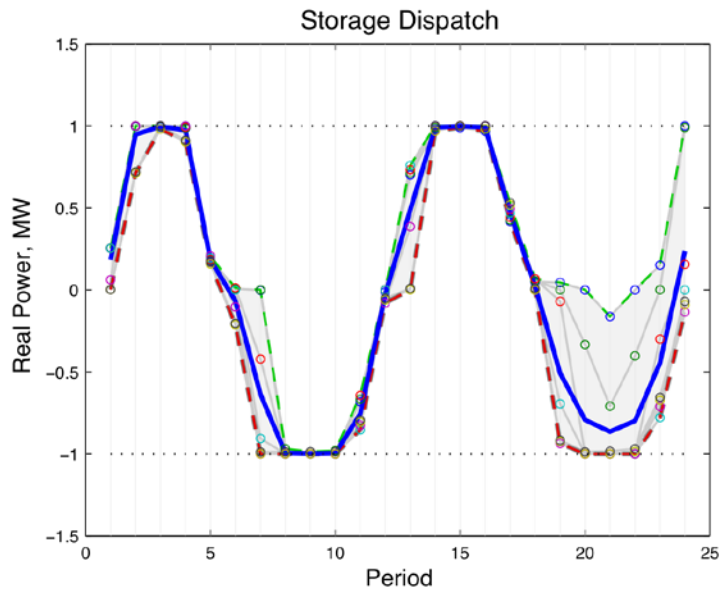
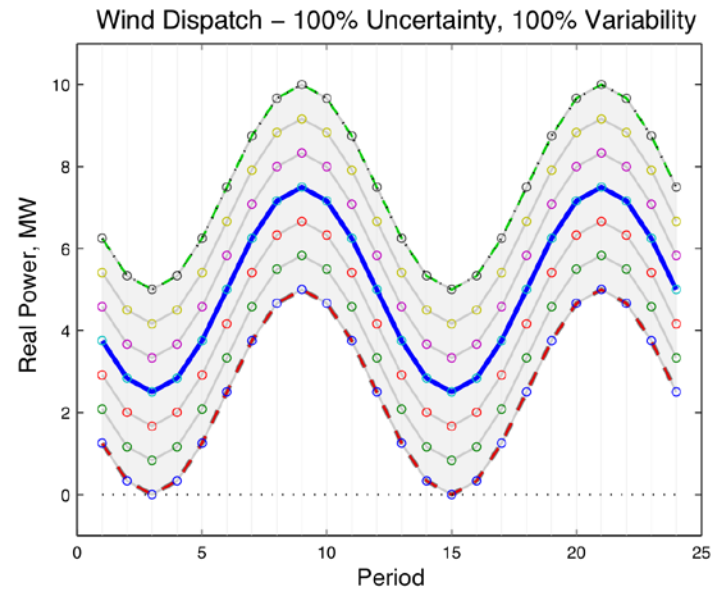
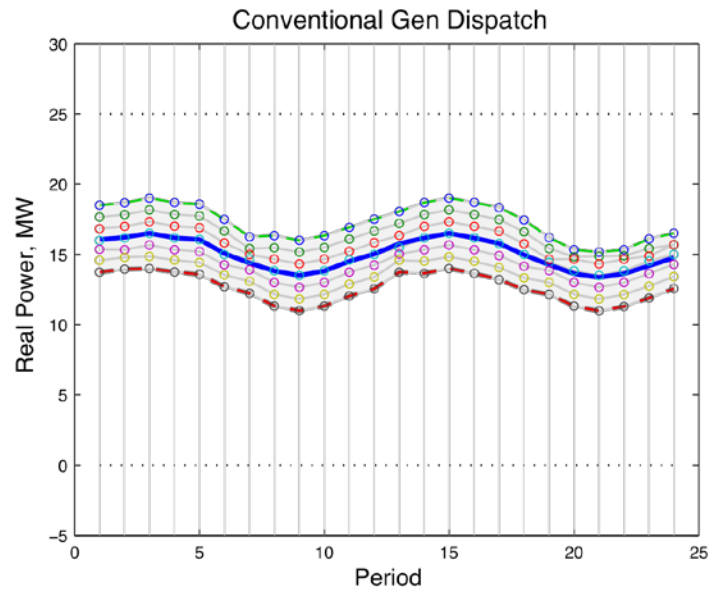
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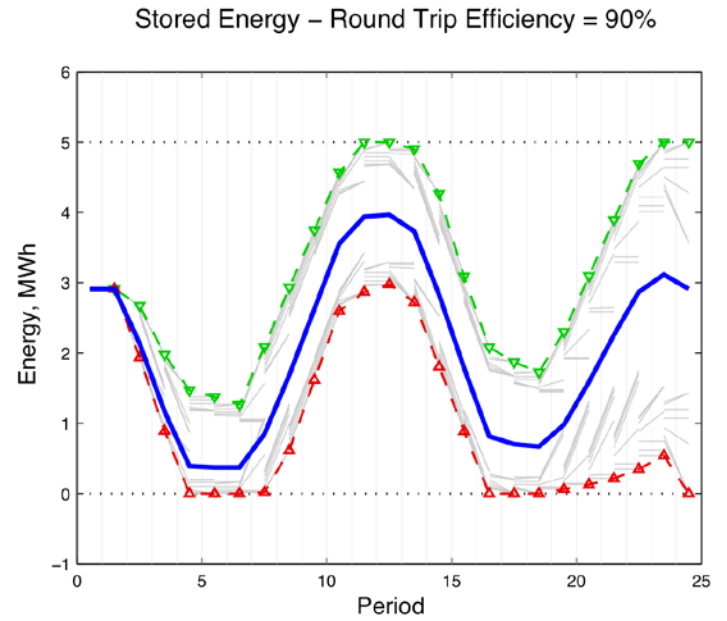
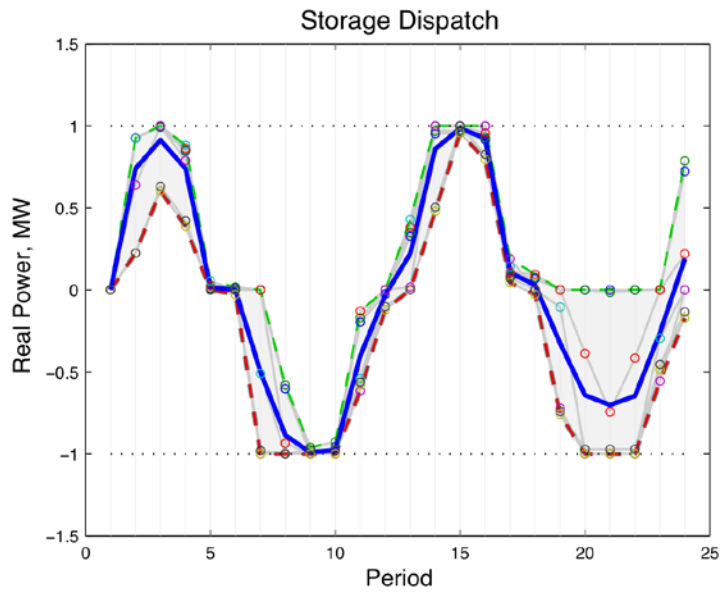
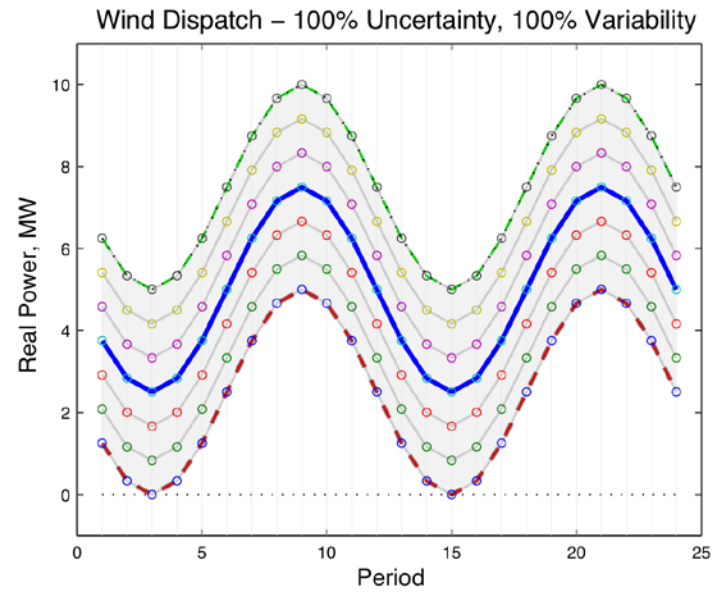
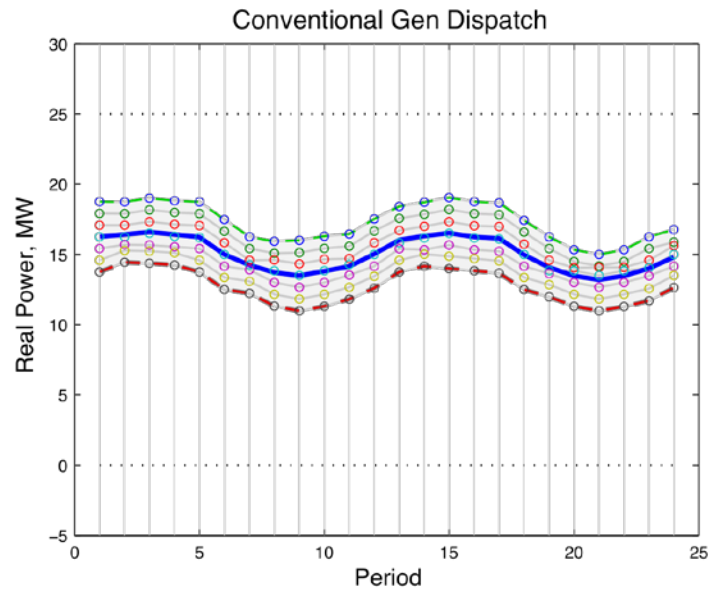


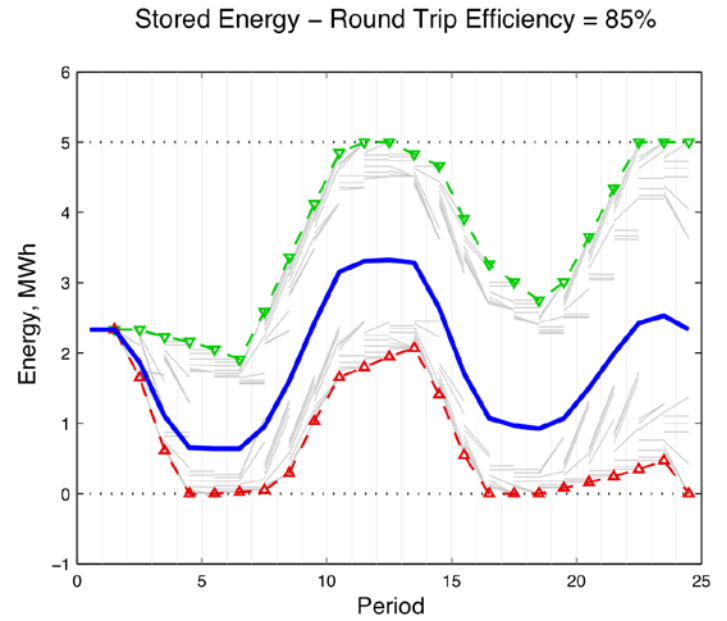
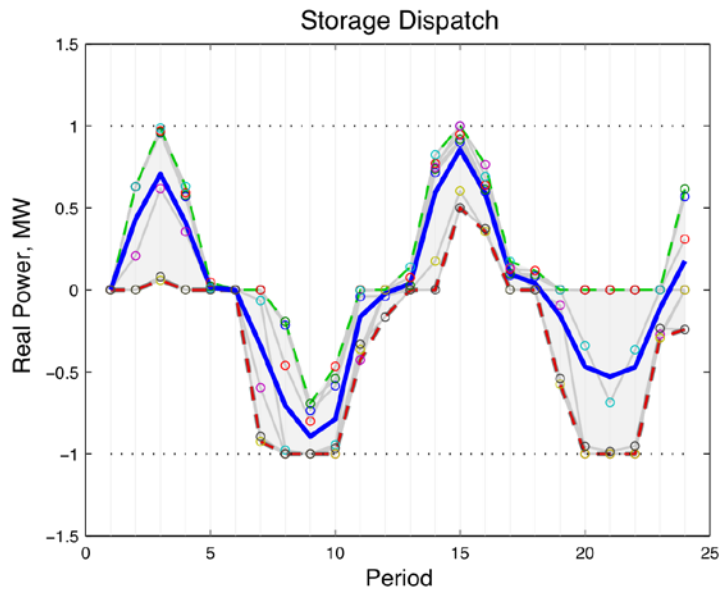
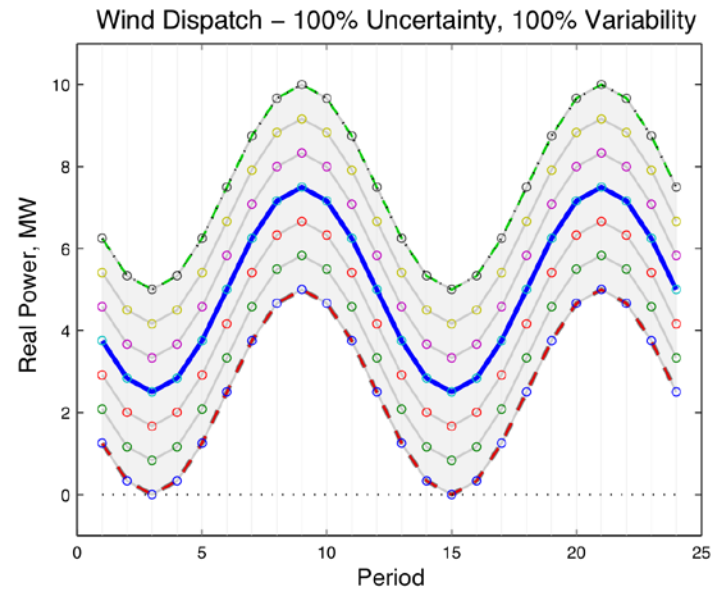
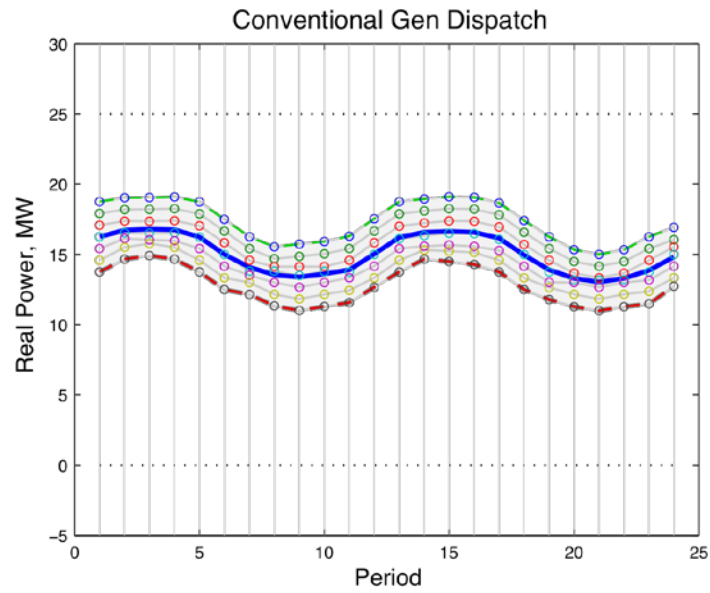
# Storage Efficiency

- input efficiency
- output efficiency
- losses

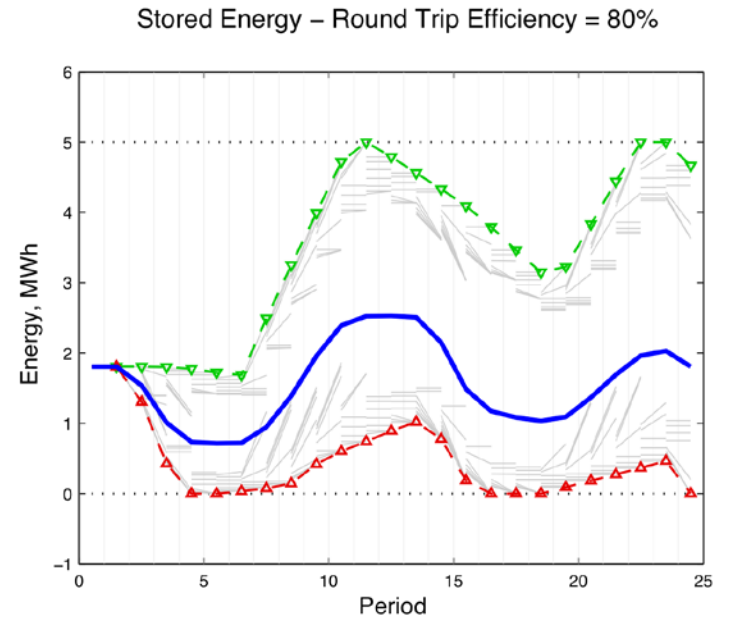
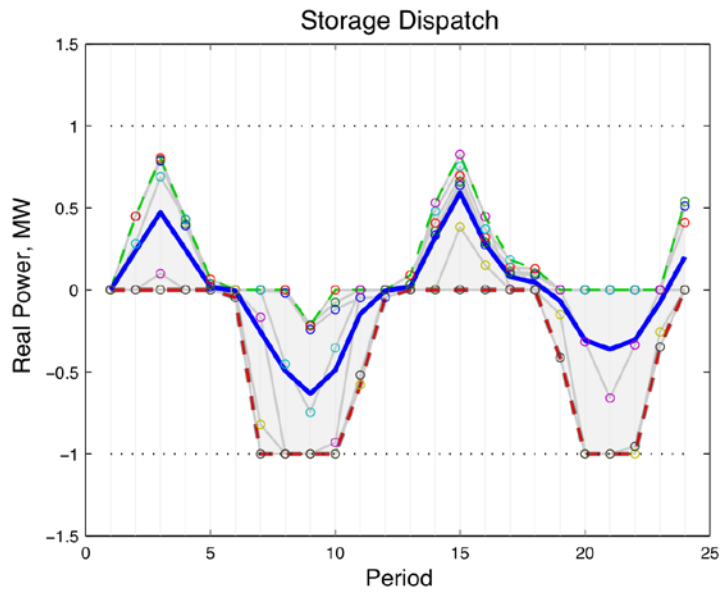
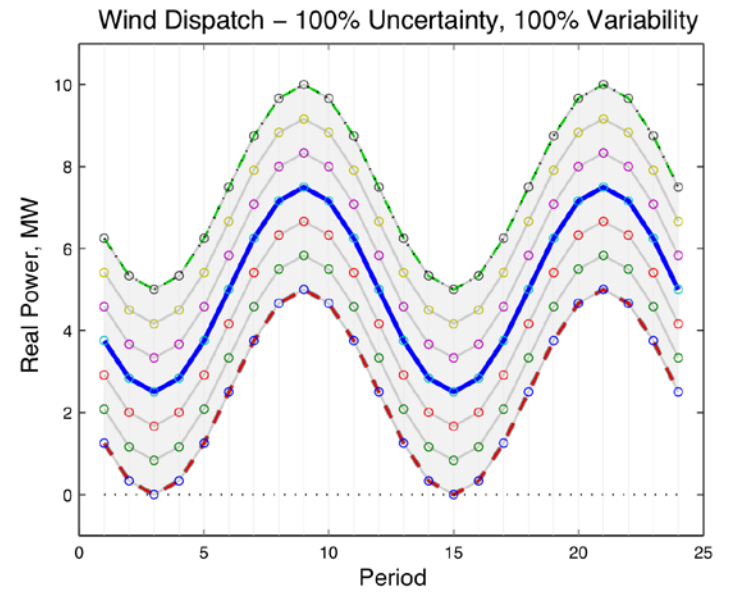
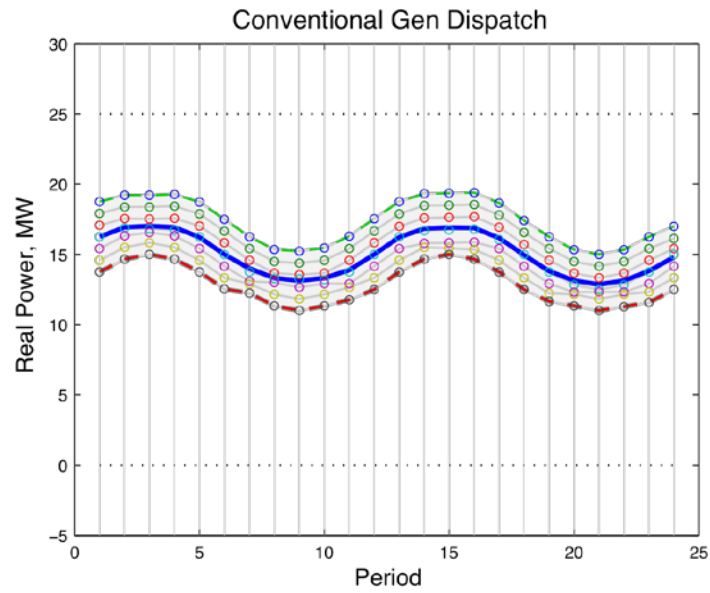


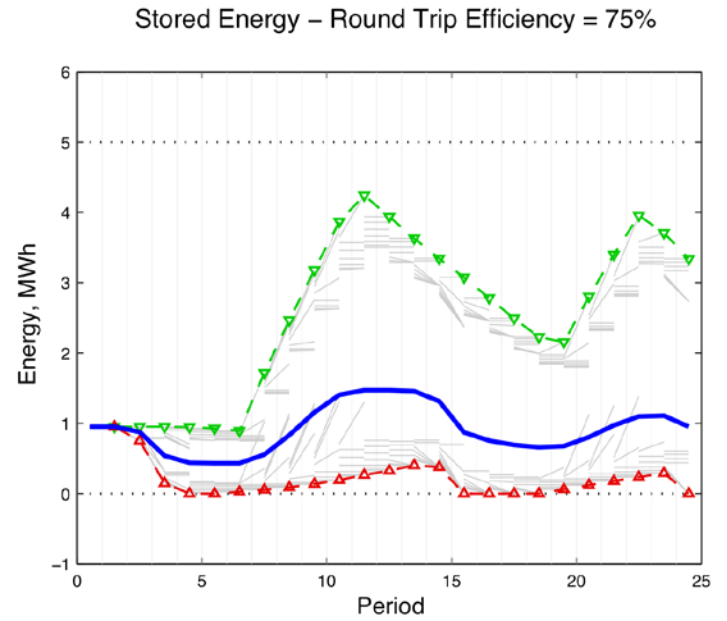
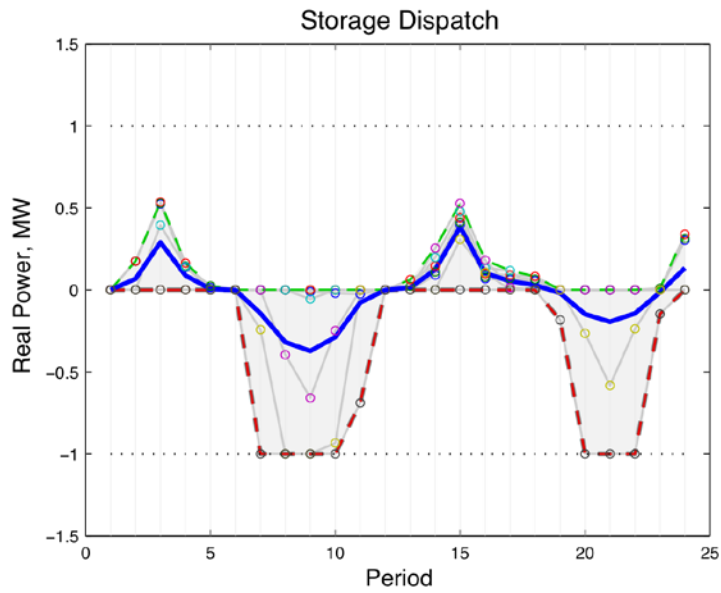
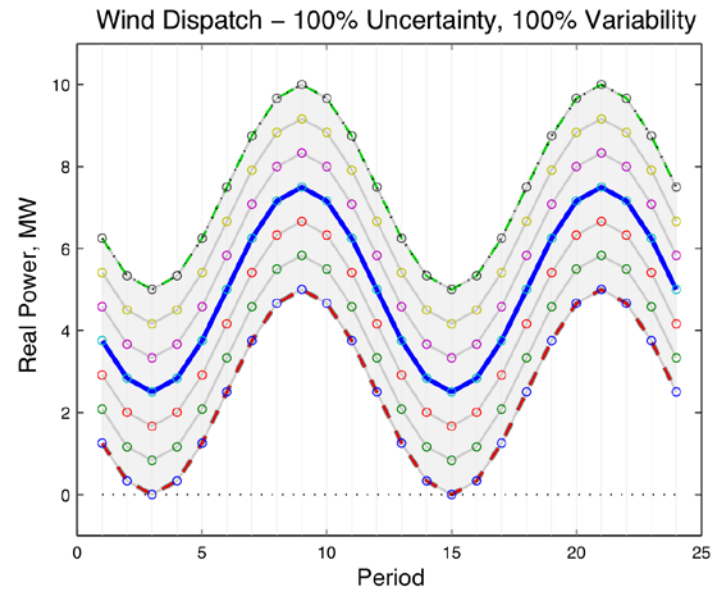
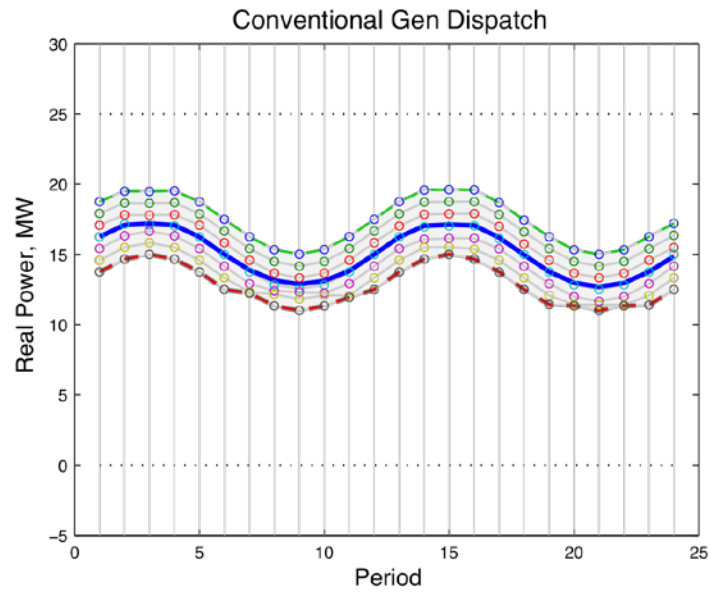


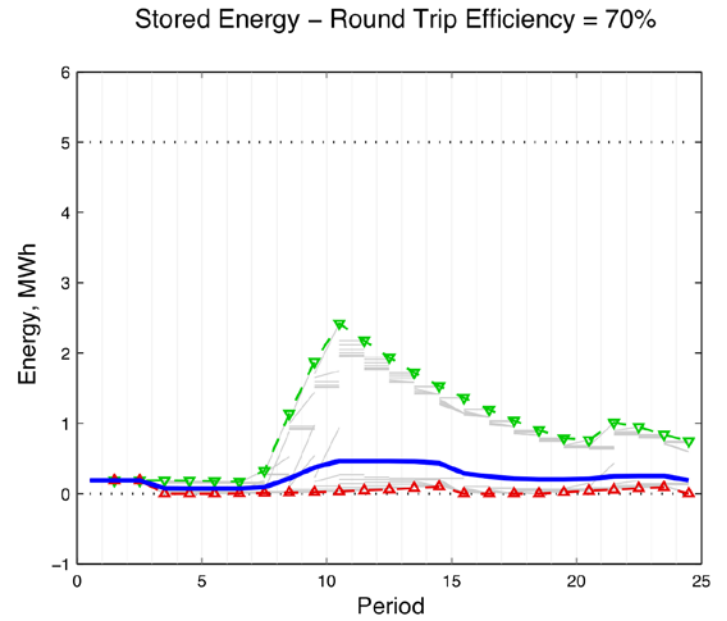
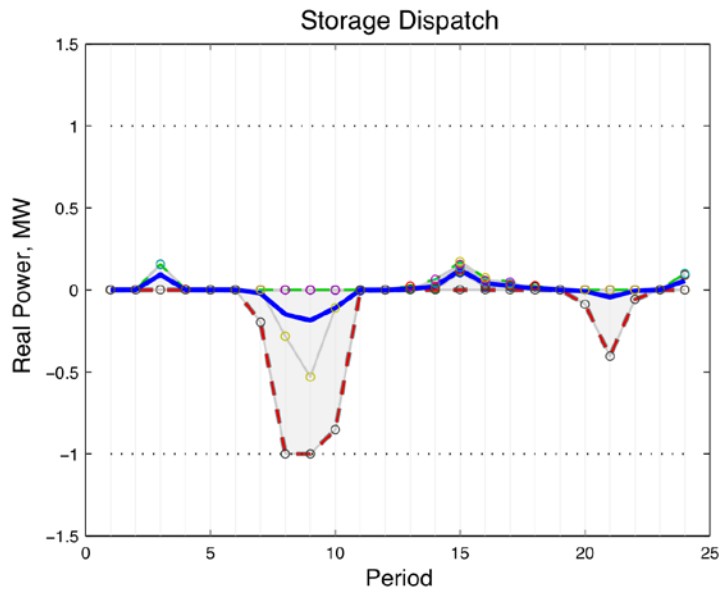
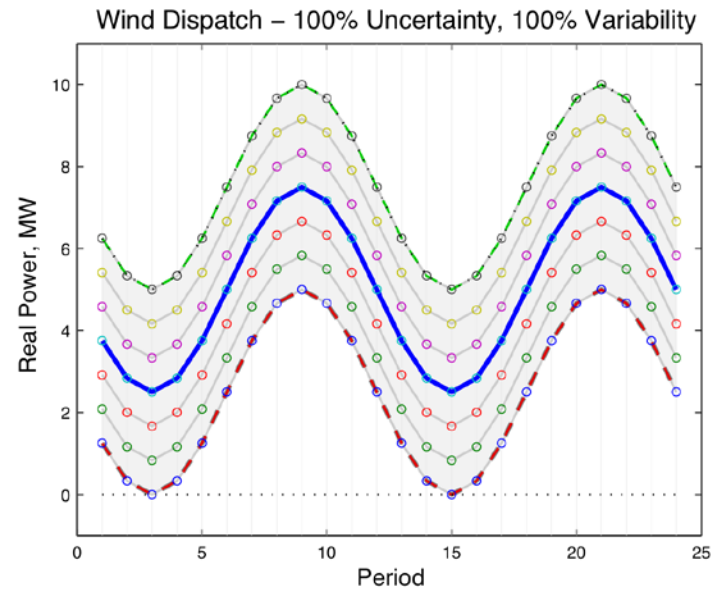
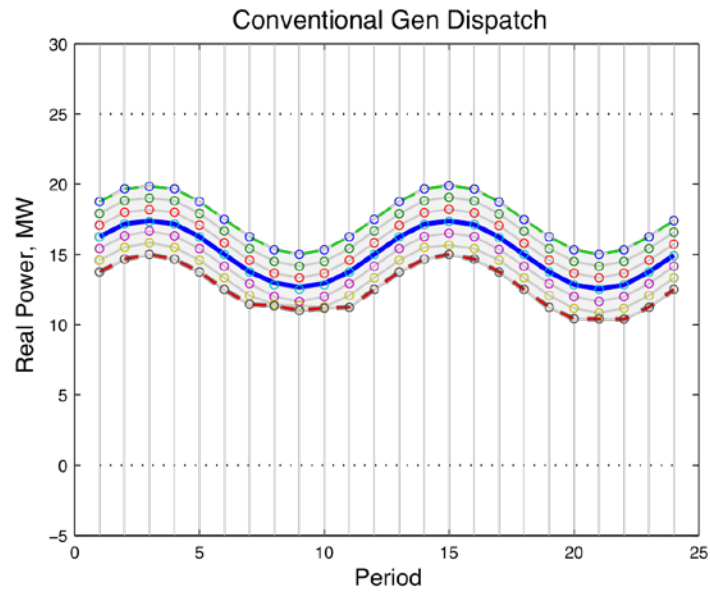


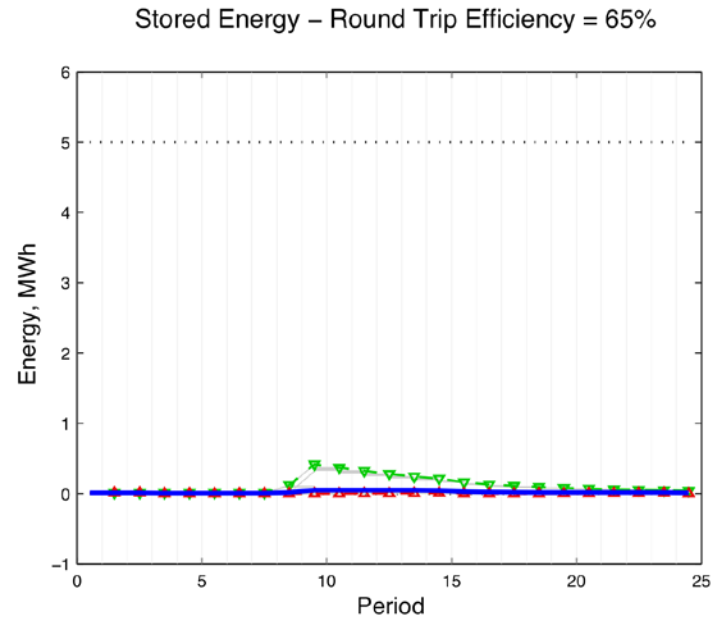
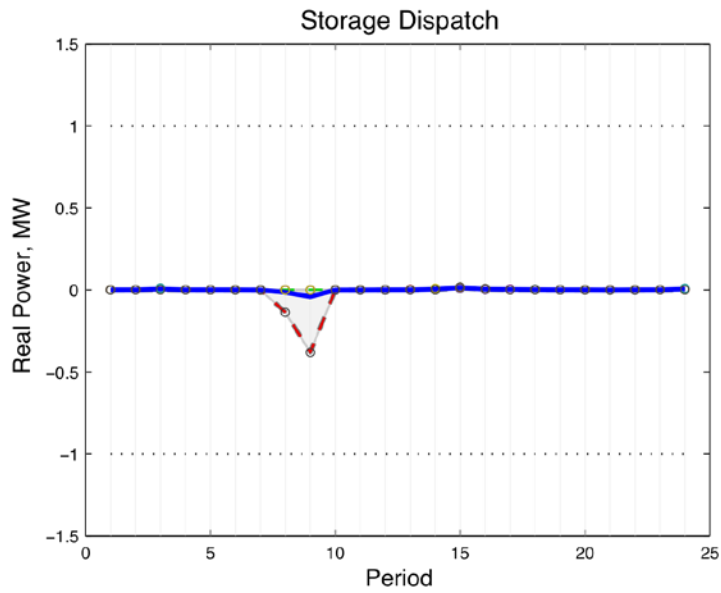
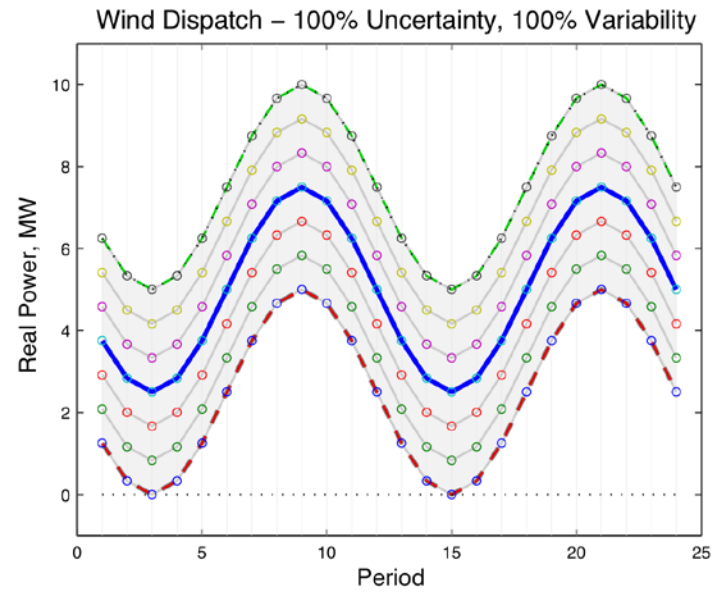
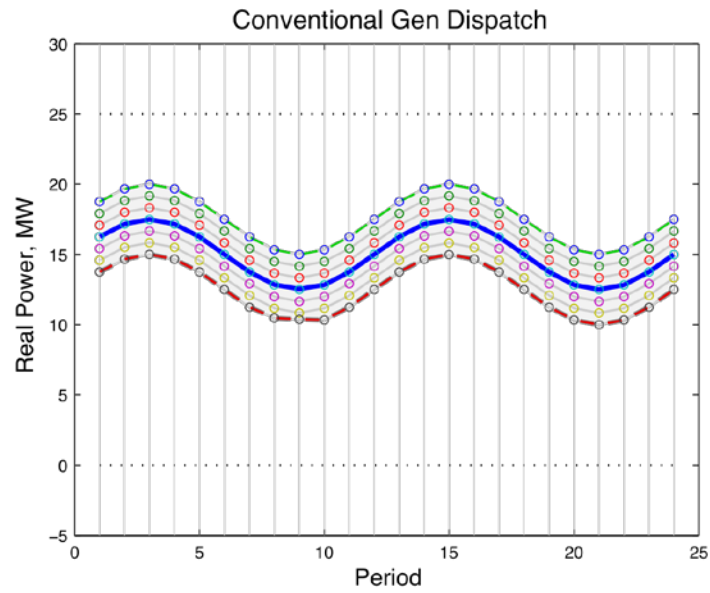


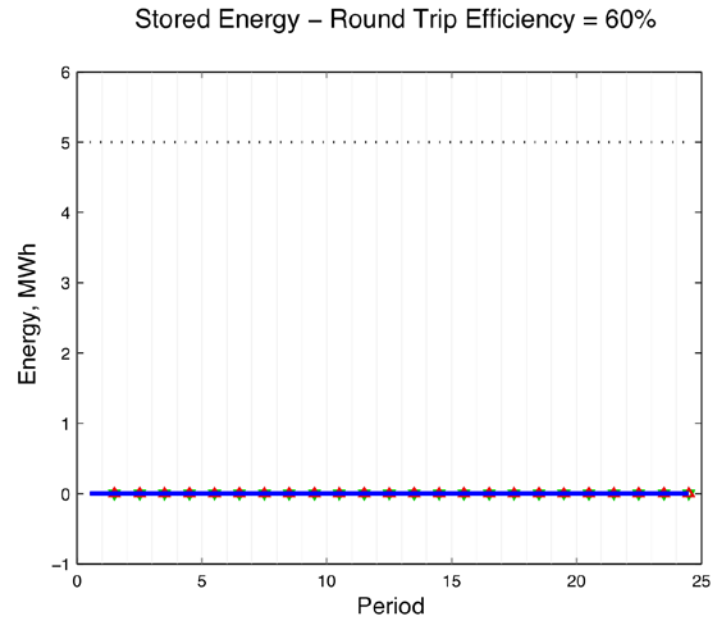
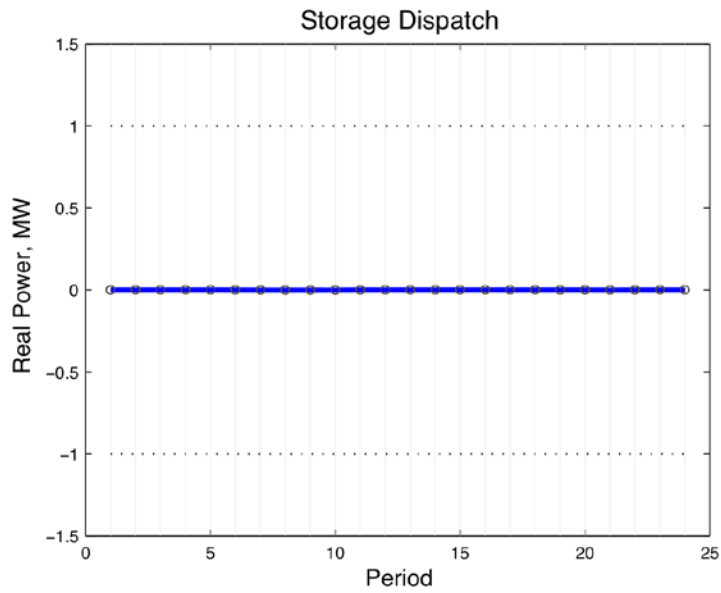
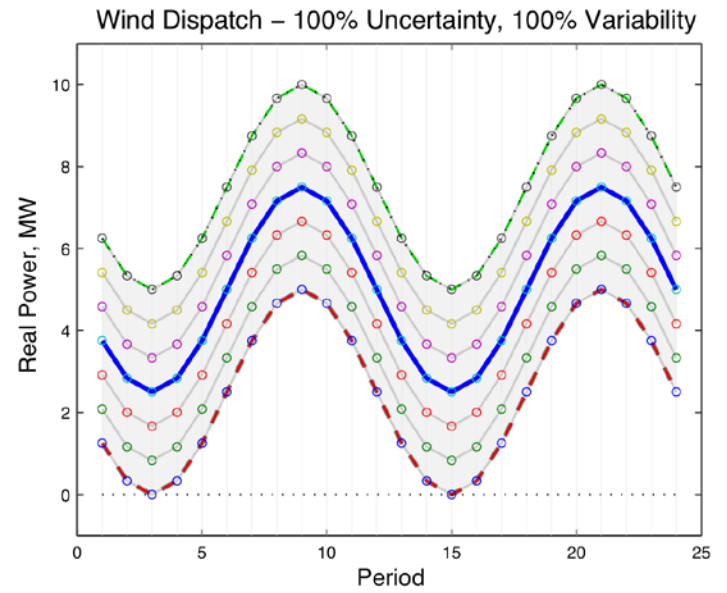
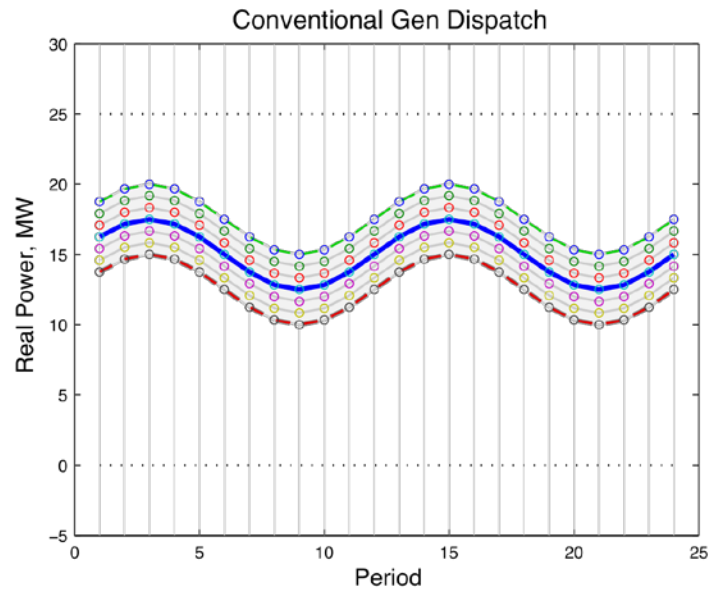








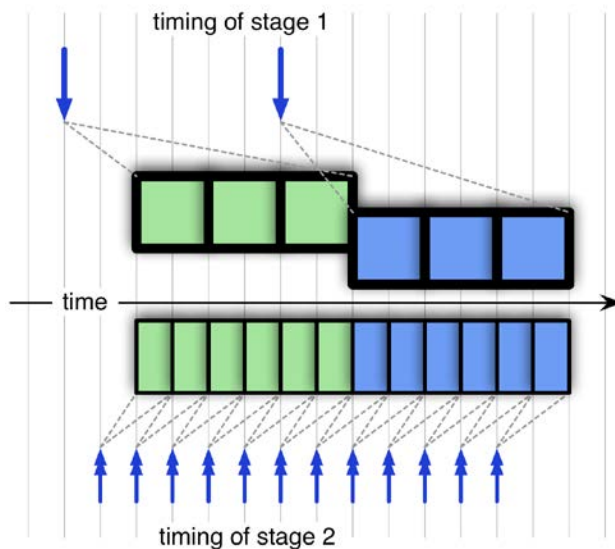




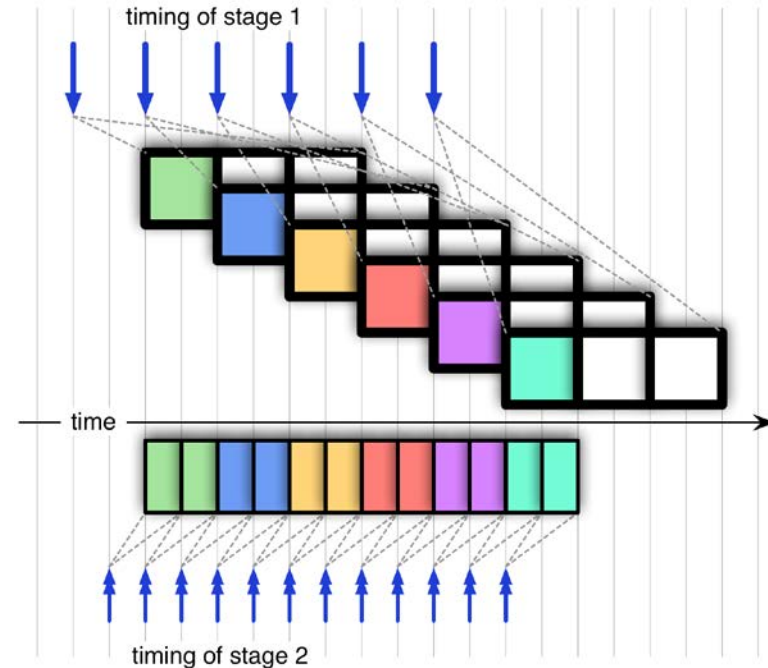
# Simulation Environment

- multi-period SuperOPF (2<sup>nd</sup>/3<sup>rd</sup> gen)
- two stage structure
  - stage 1 – day-ahead / hour-ahead
    - multiperiod – determines contracts for energy, reserve/ramping capacity and unit commitment
    - computes 24-hour plan
  - stage 2 – real-time / balancing
    - determines balancing energy, real-time prices
    - executes based on the plan and resolved uncertainties

# Day-at-a-Time vs. Receding Horizon



Traditional Approach – Stage 1 runs *once-per-day*, finds hourly solution for *full day*; stage 2 runs intra-hour, finds single period solution subject to *day-ahead* contracts.



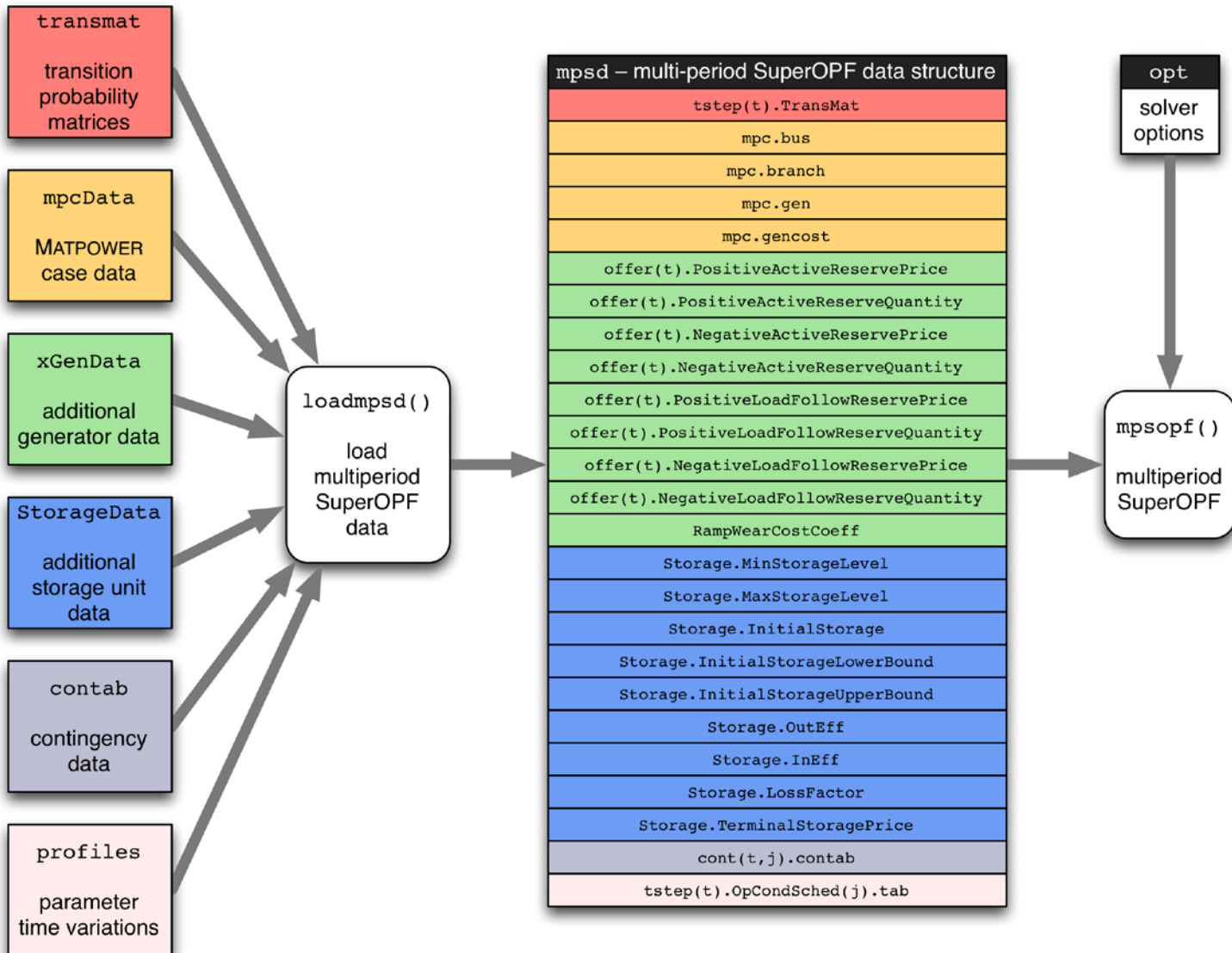
Receding Horizon Approach – Stage 1 runs *hourly*, finds solution for *first hour* with hourly full-day look-ahead; stage 2 runs intra-hour, finds single period solution subject to *hour-ahead* contracts.

# Stage 1 Requires Input Scenarios

- Uncertainty characteristics must reflect increased accuracy of shorter term forecasts
- Previous approach
  - generate simulated forecast for entire wind/load data sets
  - select forecasts “similar” to planning day
  - cluster to generate scenarios for each hour and transition probabilities
  - Issues
    - not enough “similar” days
    - even “similar” days don't have shared starting point (current operating state)
    - resulting scenarios do not adequately cover tails
- New approach
  - estimate models for temperature, load and wind
  - use model to generate many potential realizations of planning day based on common history
  - cluster to generate scenarios for each hour and transition probabilities
    - with new techniques for incorporating outliers and scenarios with specific behaviors



# Input Data File Standards



# Unit Commitment SuperOPF (3<sup>rd</sup> gen)

Same as 2<sup>nd</sup> gen Multi-period SuperOPF, with addition of ...

- integer unit commitment decisions
- startup/shutdown costs
- minimum up/down times

# Implementation is Flexible

Can be used to solve DC versions of:

- standard deterministic OPF
- single period secure, stochastic OPF (*1<sup>st</sup> gen SuperOPF*)
- multiperiod deterministic OPF (with ramping, storage)
- multiperiod secure, stochastic OPF (*2<sup>nd</sup> gen SuperOPF*)
- deterministic UC w/economic dispatch
- deterministic UC w/OPF constraints
- secure, stochastic UC with individual trajectories
- secure, stochastic UC with full transition probabilities

Plan to integrate into upcoming version of MATPOWER for wide distribution.

# References

- Two papers
  - Carlos E. Murillo-Sánchez, Ray D. Zimmerman, C. Lindsay Anderson and Robert J. Thomas, “A Stochastic, Contingency-Based Security-Constrained Optimal Power Flow for the Procurement of Energy and Distributed Reserve”, *Decision Support Systems*, 2013, <http://dx.doi.org/10.1016/j.dss.2013.04.006>.
  - Carlos E. Murillo-Sánchez, Ray D. Zimmerman, C. Lindsay Anderson and Robert J. Thomas, “Secure Planning and Operations of Systems with Stochastic Sources, Energy Storage and Active Demand”, *accepted for the Special Issue of IEEE Transactions on Smart Grid on “Optimization Methods and Algorithms Applied to Smart Grid”*
- Presentation at FERC Software Conference “Increasing Real-time and Day-ahead Market Efficiency Through Improved Software”
  - <http://www.ferc.gov/industries/electric/indus-act/market-planning/2013-conference.asp>

# Discussion of UC SuperOPF Implementation and Preliminary Results

Carlos E. Murillo-Sánchez

# Traditional MIP UC formulation

$$u^{ti} P_{\min}^{tijk} \leq p^{tijk} \leq u^{ti} P_{\max}^{tijk}$$

$$u^{ti} Q_{\min}^{tijk} \leq q^{tijk} \leq u^{ti} Q_{\max}^{tijk}$$

$$u^{ti} - u^{(t-1)i} = v^{ti} - w^{ti}$$

$$\sum_{y=t-\tau_i^++1}^t v^{yi} \leq u^{ti}, \quad \sum_{y=t-\tau_i^-+1}^t w^{yi} \leq 1 - u^{ti}$$

- In practice, only  $u$  variables need to be defined as binary.
- A  $u$  variable shuts down all injections related to a given generator in a given time slice (all scenarios, all contingencies and base cases).
- On input, a unit can have “available for commitment decision”, “forced on” or “forced off” status.
- However, a number of issues must be addressed for incorporation into the SuperOPF framework.

# The lighter issues

- Several changes introduced in the code so that all required injections are represented.
- Forced-off injections trigger superfluous contingencies filters and pruning occurs.



# The issue of structure

- The SuperOPF tree can be thought of as a probability tree with recombination of scenarios in the central path.
- A contingency might be defined in which one of the generators available for the commitment decision goes offline.
- If  $u=0$ , that contingency is superfluous and the base case in the corresponding scenario should assimilate that contingency's probability.
- Indeed,  $u=0$  or  $u=1$  changes the structure of the probability tree and the corresponding probability weights in the cost function!
- Can adopt a cost formulation that switches on or off certain cost segments (for linear or piecewise linear costs), but it is messy and introduces  $u$ 's in the cost in a complicated manner (otherwise each  $u$  simply triggers the fixed part of the corresponding operation cost into the objective) .

# A simplification

- Ignore the issue of structural change in probability tree; ok if probabilities of injection outages are small.
- Then, if  $u=0$ , the contingency for which the generator in question is ousted generates a power flow that is identical to that of the base case, so no reserve requirements will be set by the superfluous contingency flow.
- If  $u=1$  all is fine.
- The relative weighting of the base case can be a little bit off if  $u=0$  because of this simplification.

# Looking out to other formulations

- There are tighter UC MIP formulations, but the better ones are aimed at network-less power balance and fixed reserve formulations.
- The binary variables enter the simple power balance constraints: injections split in fixed  $u$ -weighted  $P_{\min}$  plus a variable ( $P_g - P_{\min}$ ) with zero lower bound.
- In a formulation with an explicit DC network flow (such as the single-QP version of the multi-period SuperOPF), the binary variables would enter into a staggering amount of network constraints, complicating the cuts.
- Not included yet: prescribed startup curves.

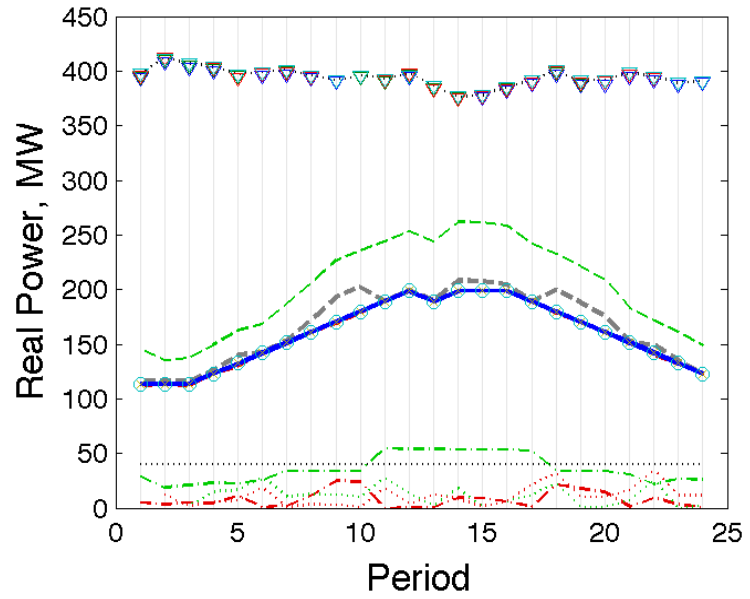
# Status

- Testing currently under way with 30 bus and 118 bus systems.
- A systematic exploration of parameter fine-tuning for CPLEX, Gurobi, is underway, and is necessary.
- Preliminary results in concordance with expectations given experience with continuous multi-period SuperOPF.
- Still not tested (but coded): Decomposition-based AC version. Each central problem much simpler than the corresponding MPSOPF with DC network.

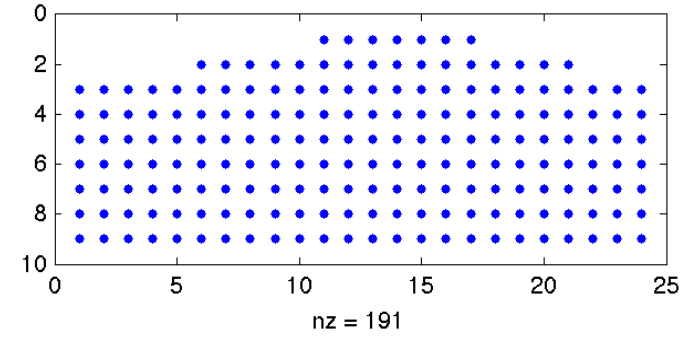
# Example: 30 bus

- Gens 1 & 2: large coal
- Gens 3 & 4: peakers
- Gens 5 & 6: medium coal
- Gen 7: storage (small)
- Gens 8 & 9: large wind
- Wind uncertainty increased from zero to actual forecast uncertainty; four wind scenarios (NREL)
- Contingencies: generator outages

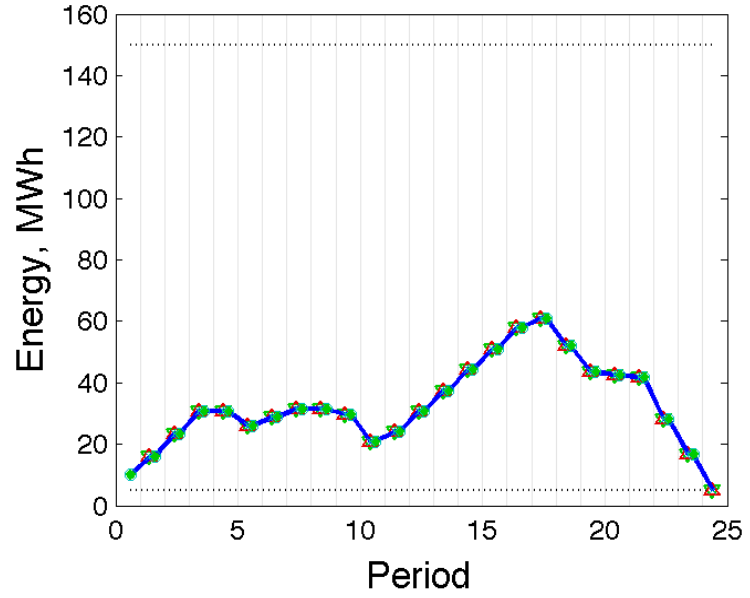
### Real Power Output for All Generators



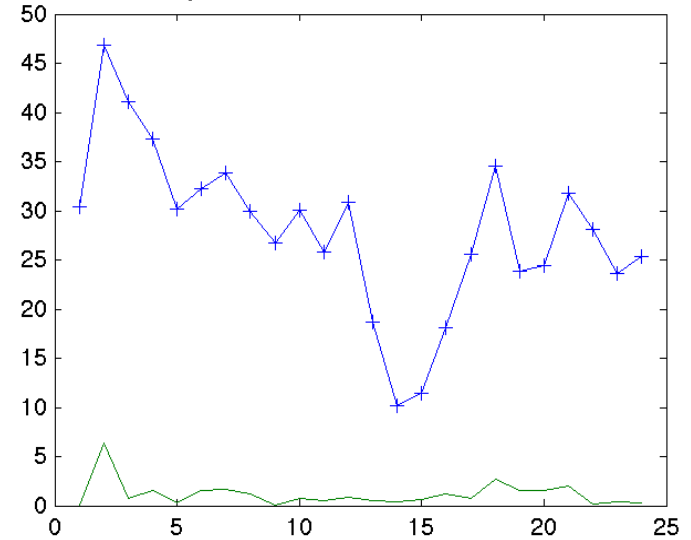
### Commitment Schedule



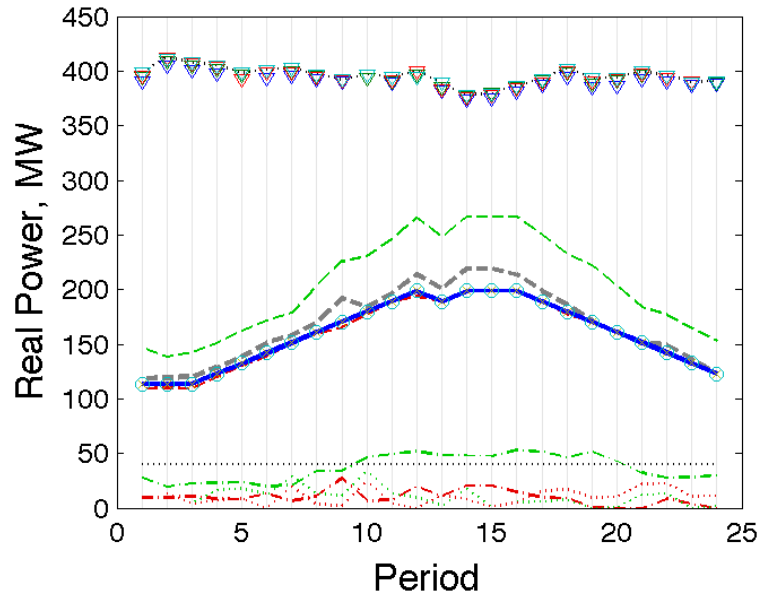
### Stored Energy for Storage Unit 1 (Gen 7) @ Bus 7



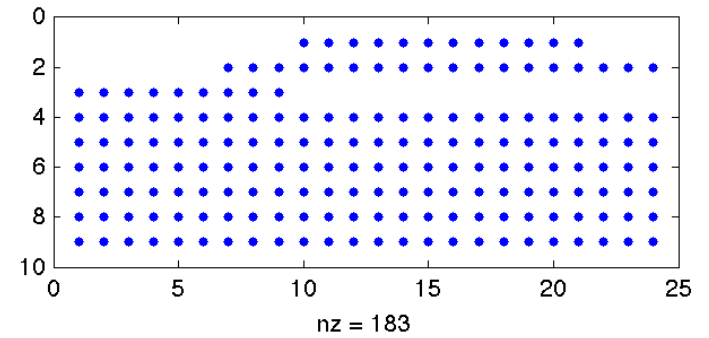
### Wind potential and curtailment



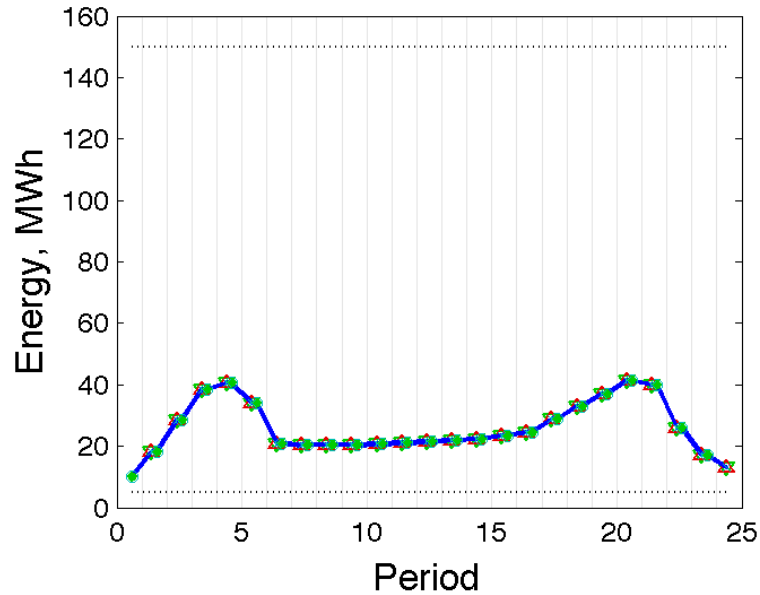
### Real Power Output for All Generators



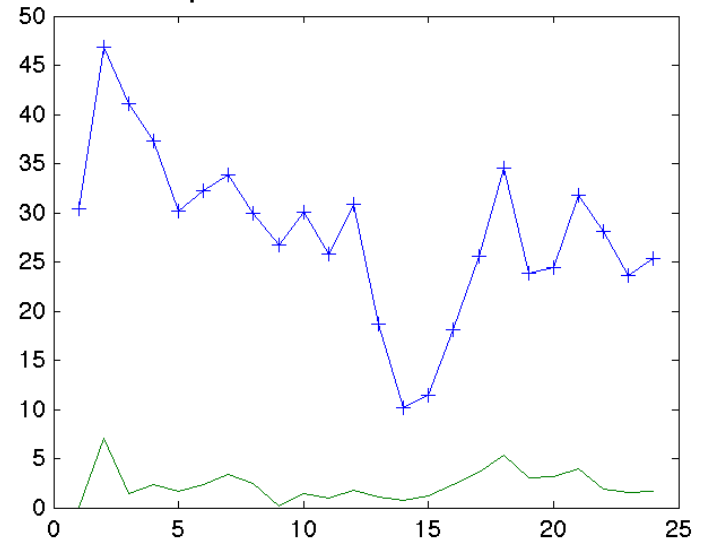
### Commitment Schedule



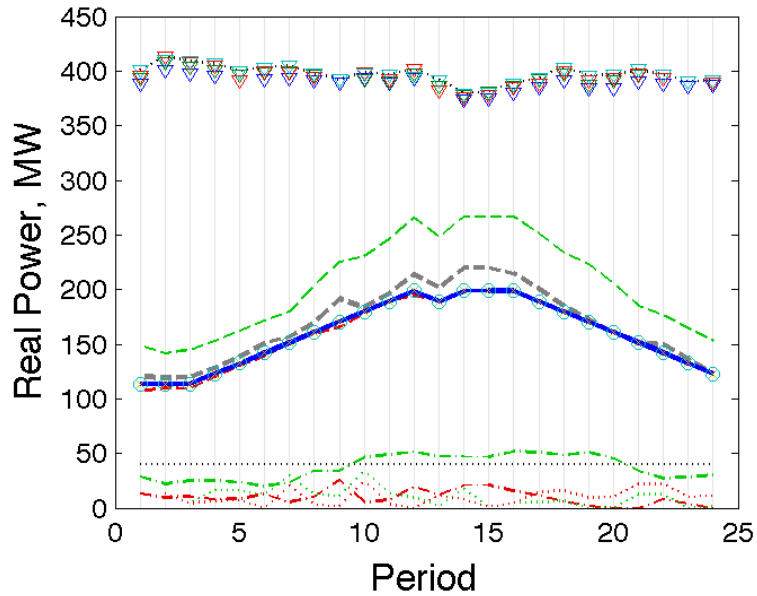
### Stored Energy for Storage Unit 1 (Gen 7) @ Bus 7



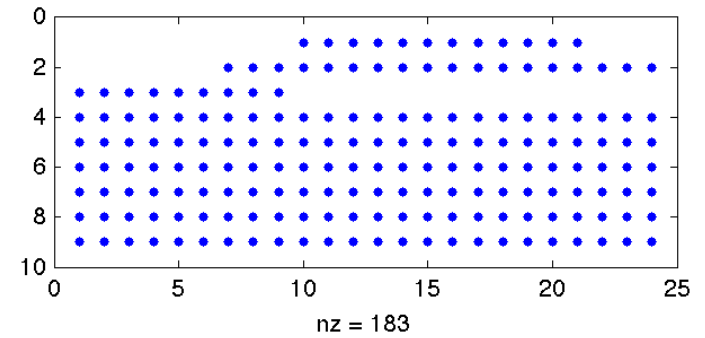
### Wind potential and curtailment



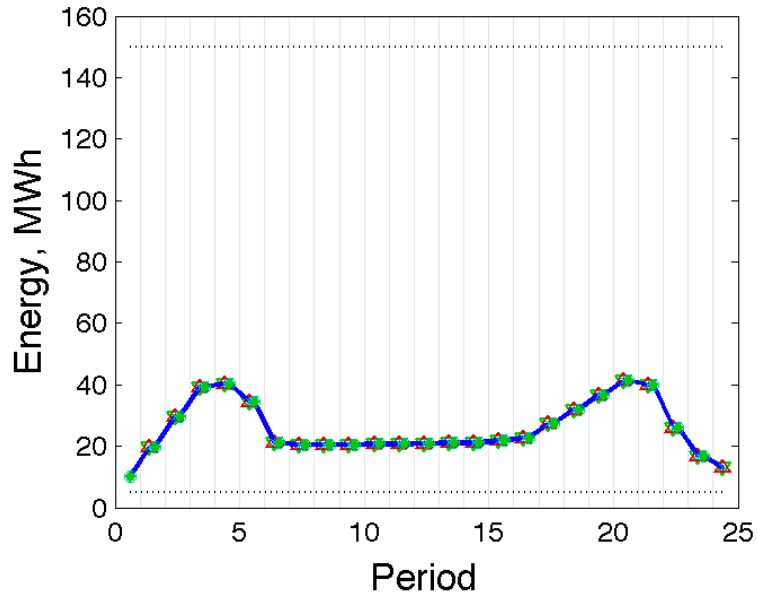
### Real Power Output for All Generators



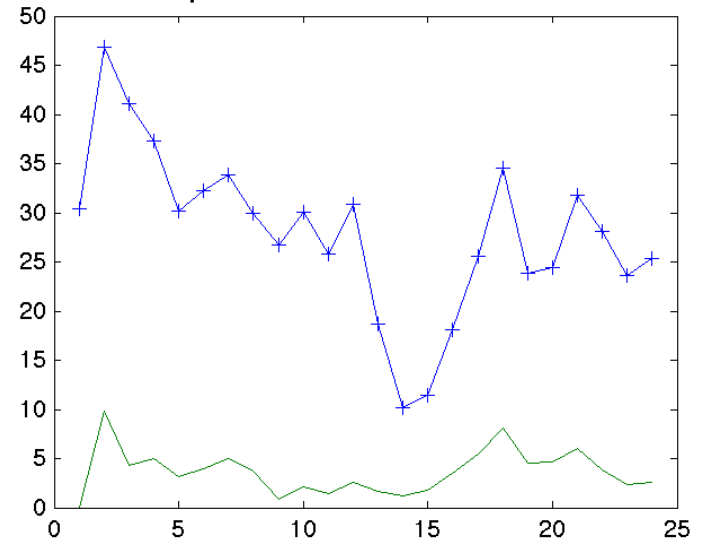
### Commitment Schedule



### Stored Energy for Storage Unit 1 (Gen 7) @ Bus 7

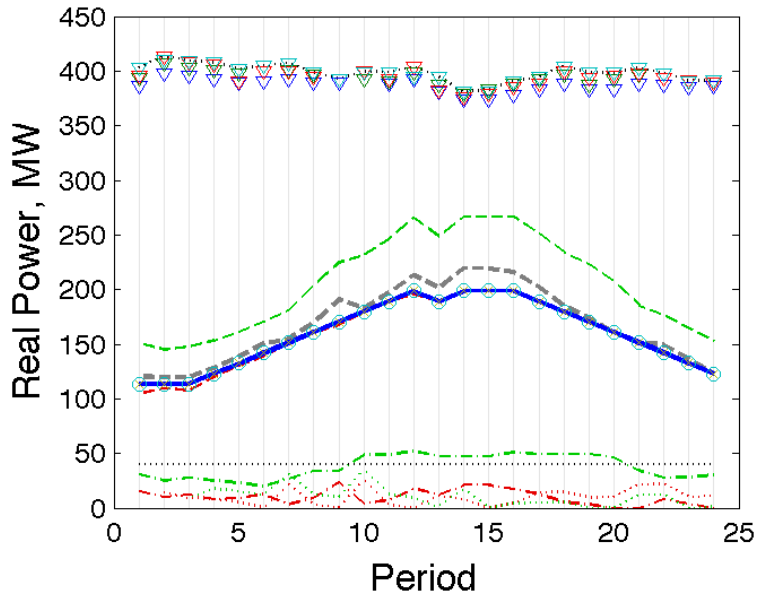


### Wind potential and curtailment

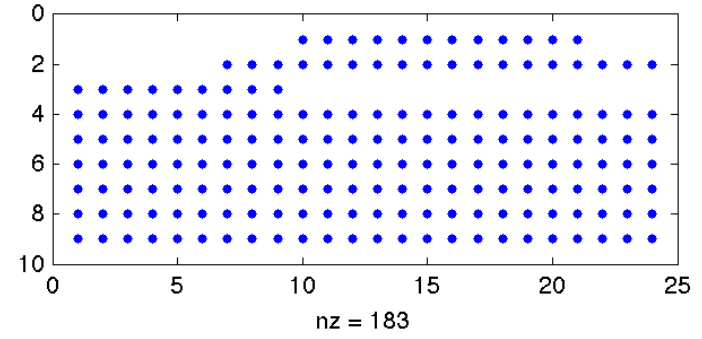




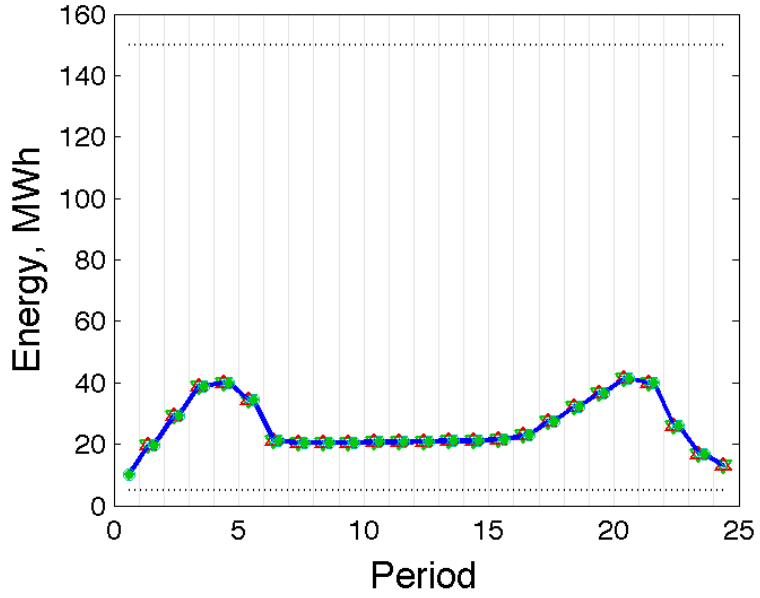
### Real Power Output for All Generators



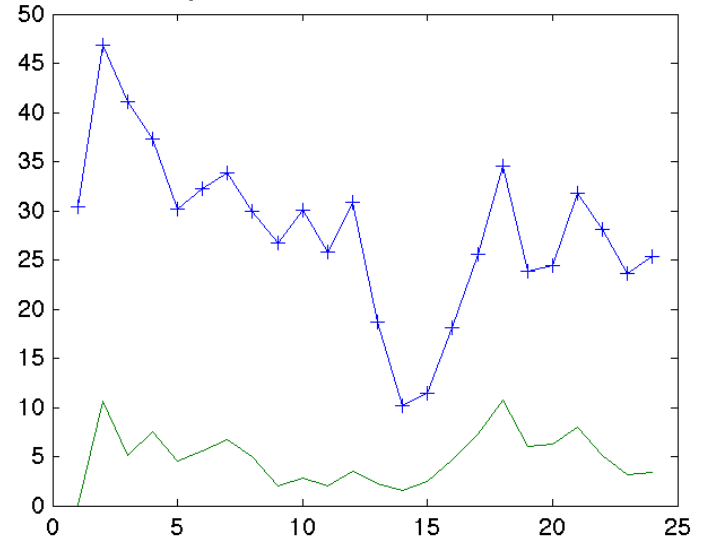
### Commitment Schedule



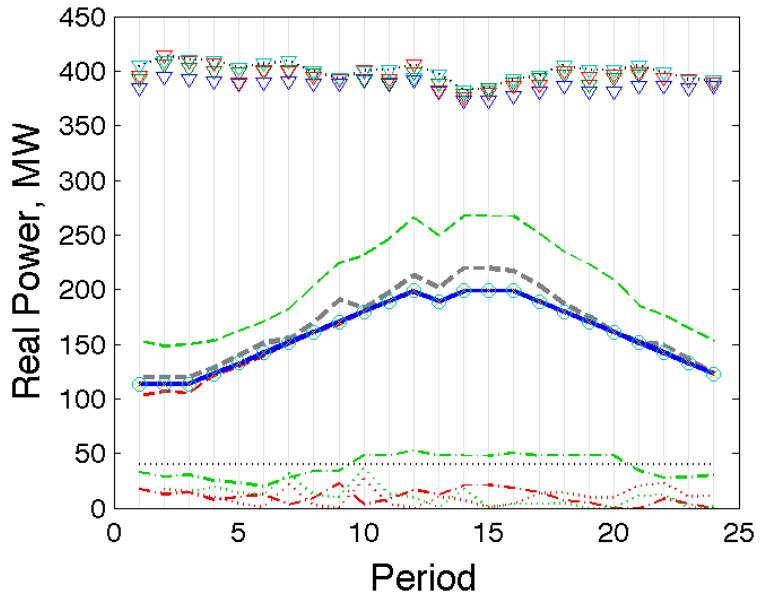
### Stored Energy for Storage Unit 1 (Gen 7) @ Bus 7



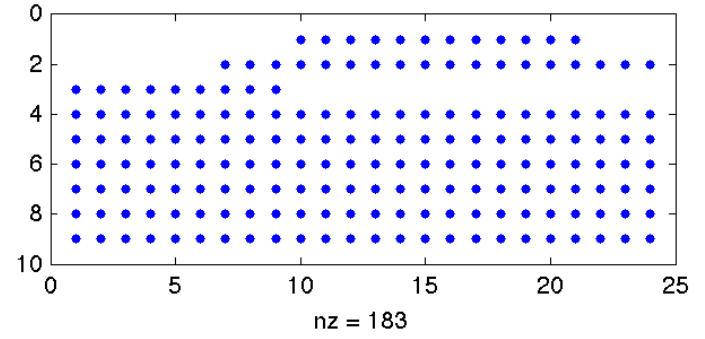
### Wind potential and curtailment



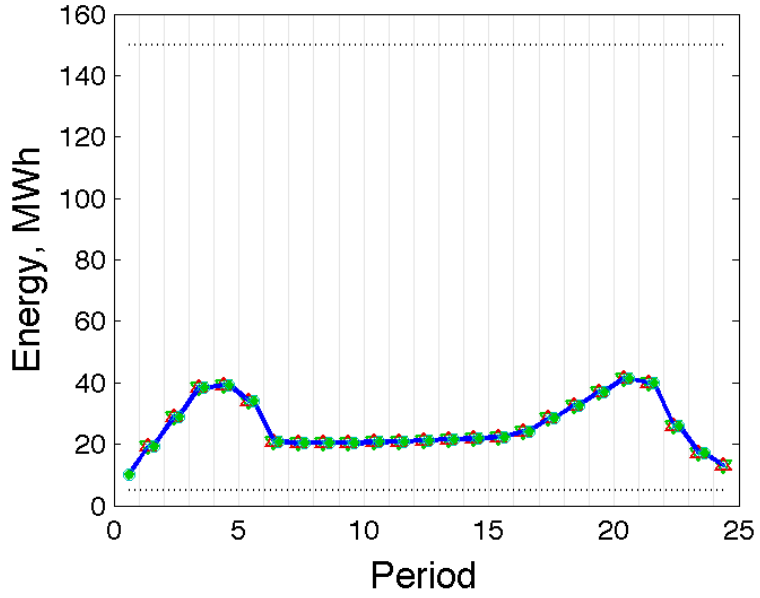
### Real Power Output for All Generators



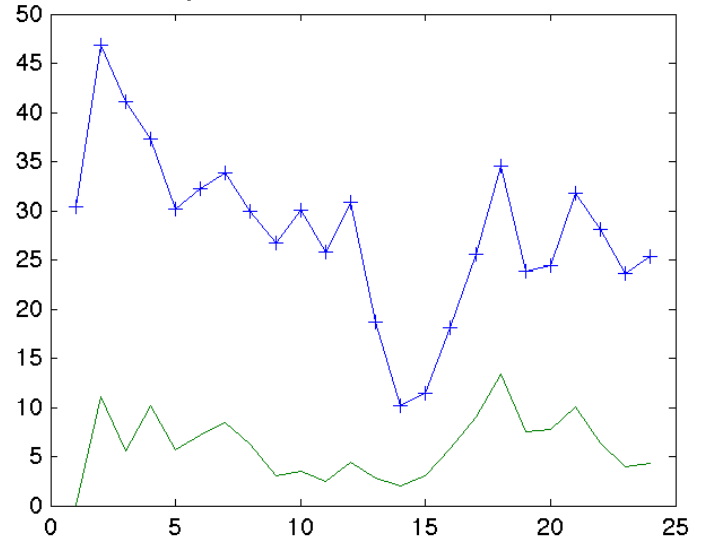
### Commitment Schedule



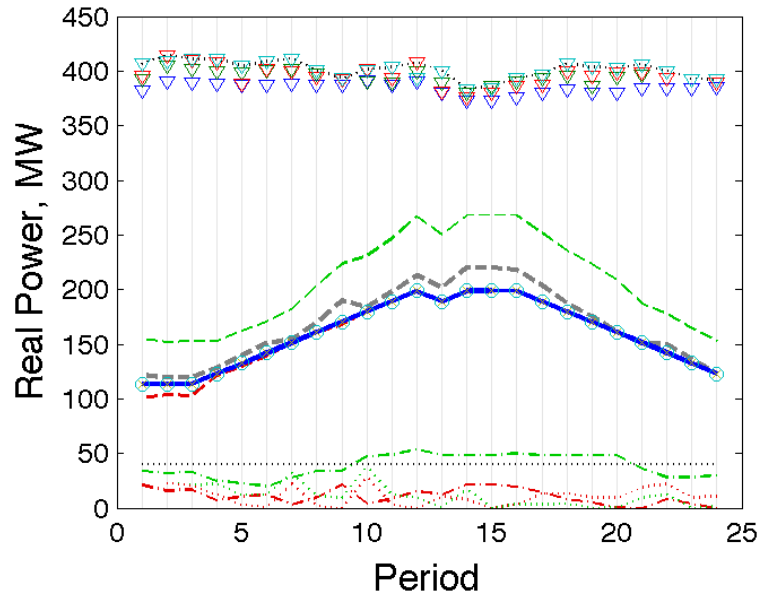
### Stored Energy for Storage Unit 1 (Gen 7) @ Bus 7



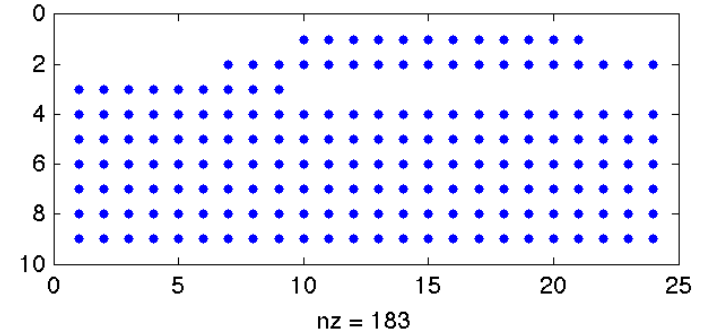
### Wind potential and curtailment



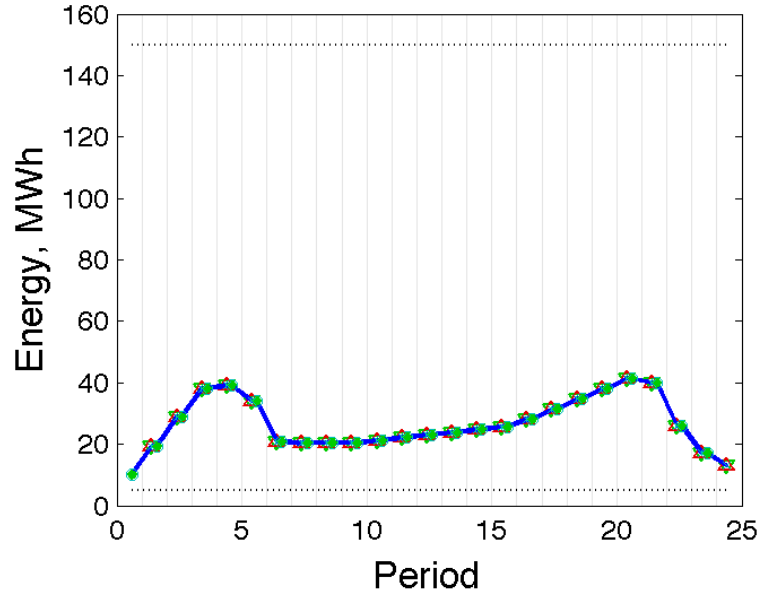
### Real Power Output for All Generators



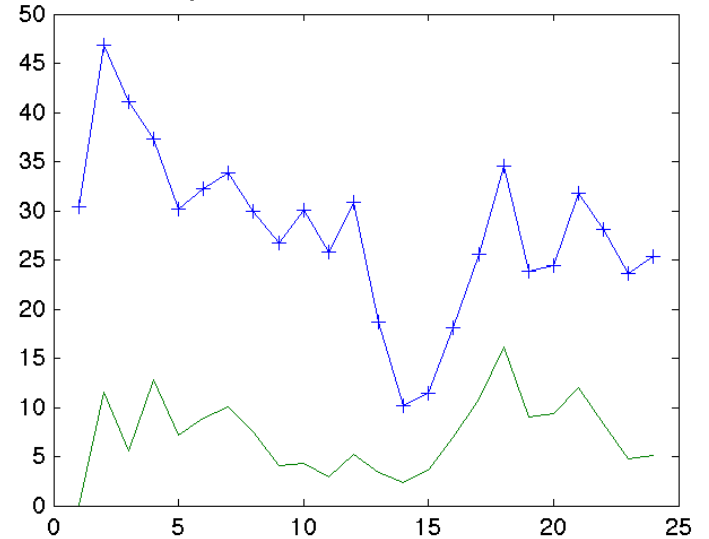
### Commitment Schedule



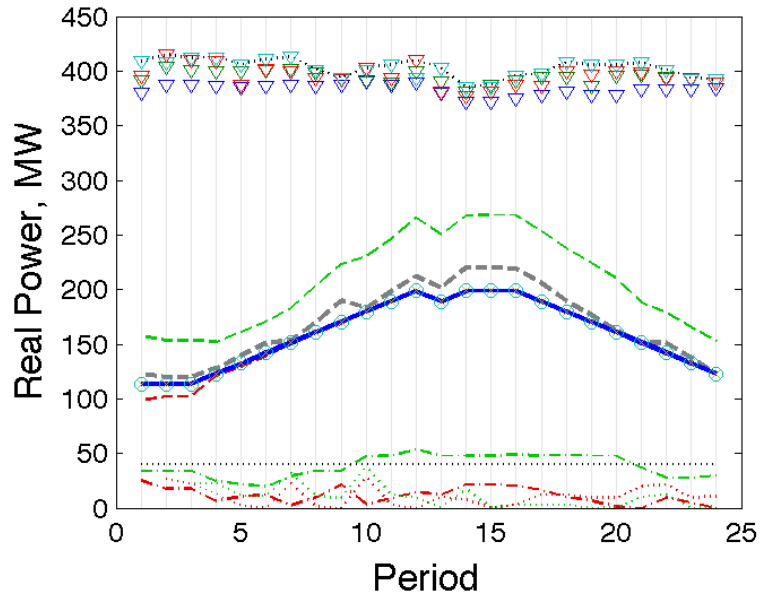
### Stored Energy for Storage Unit 1 (Gen 7) @ Bus 7



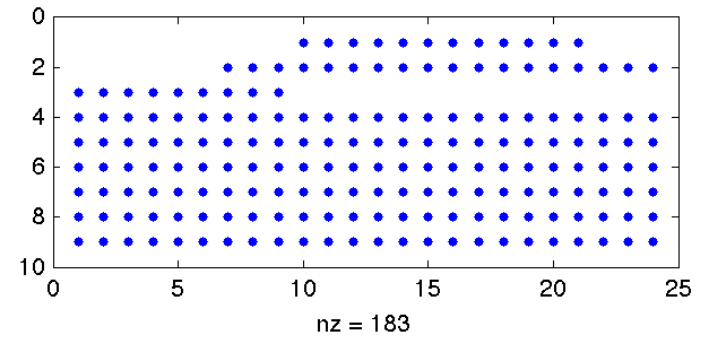
### Wind potential and curtailment



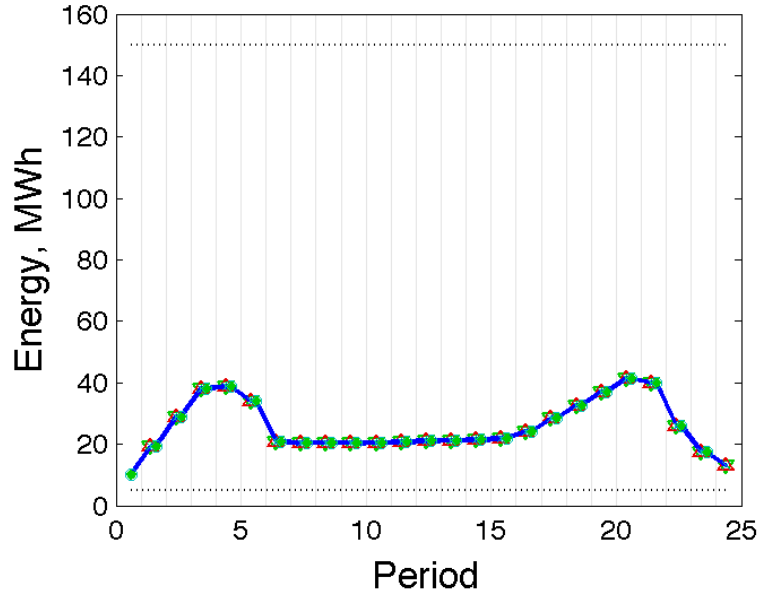
### Real Power Output for All Generators



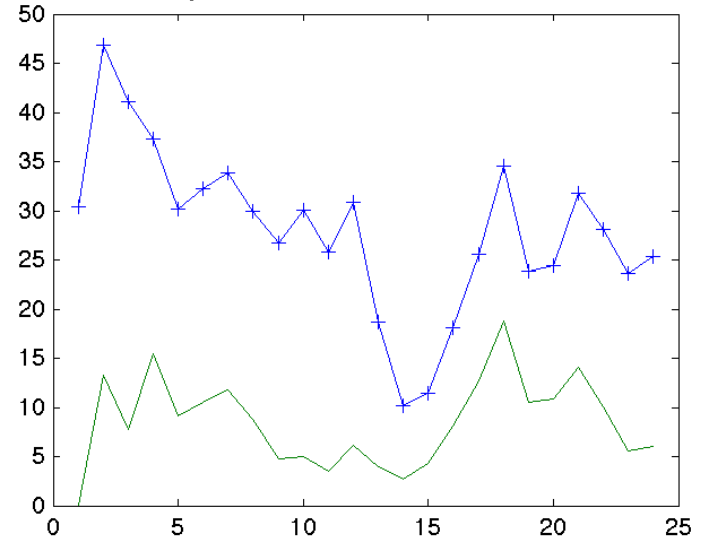
### Commitment Schedule



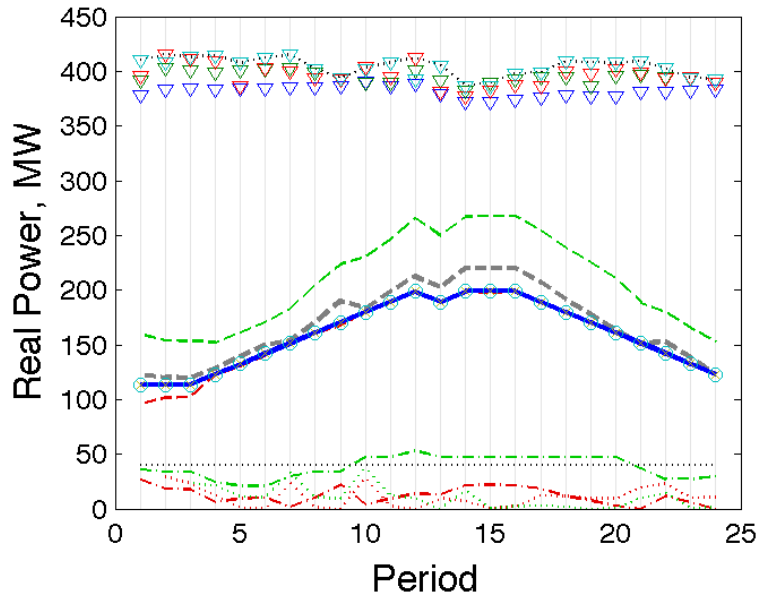
### Stored Energy for Storage Unit 1 (Gen 7) @ Bus 7



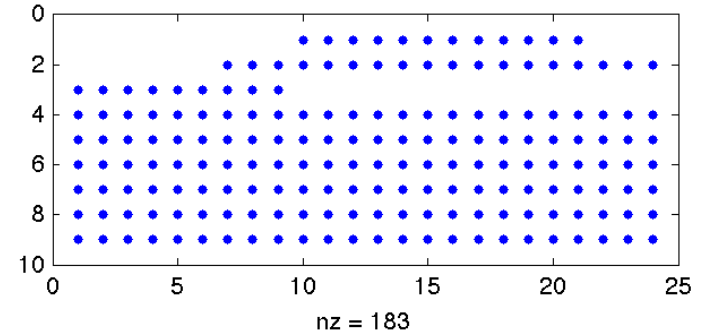
### Wind potential and curtailment



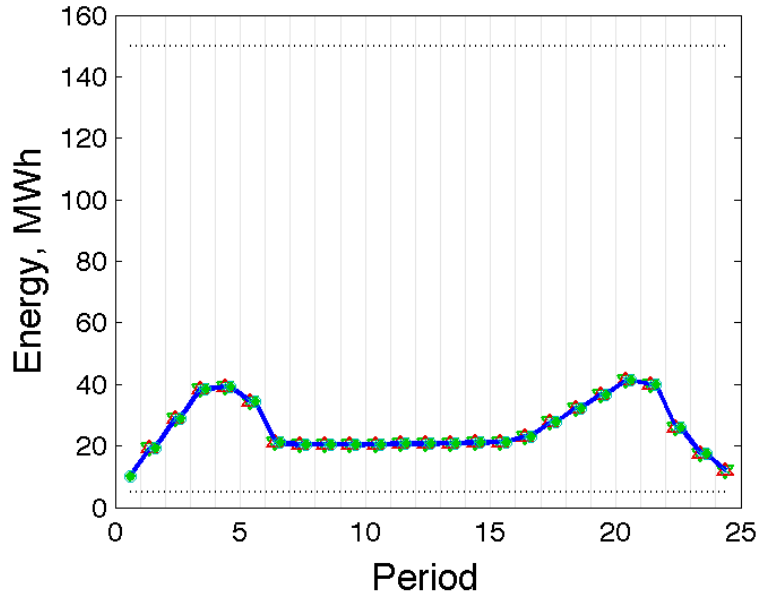
### Real Power Output for All Generators



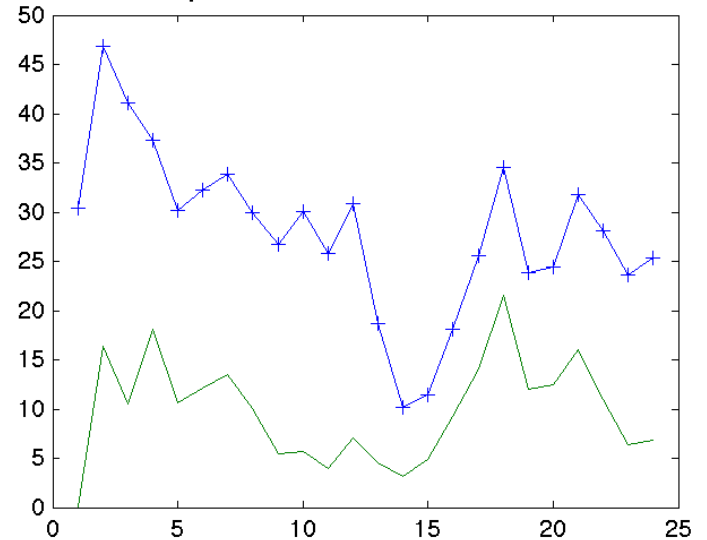
### Commitment Schedule



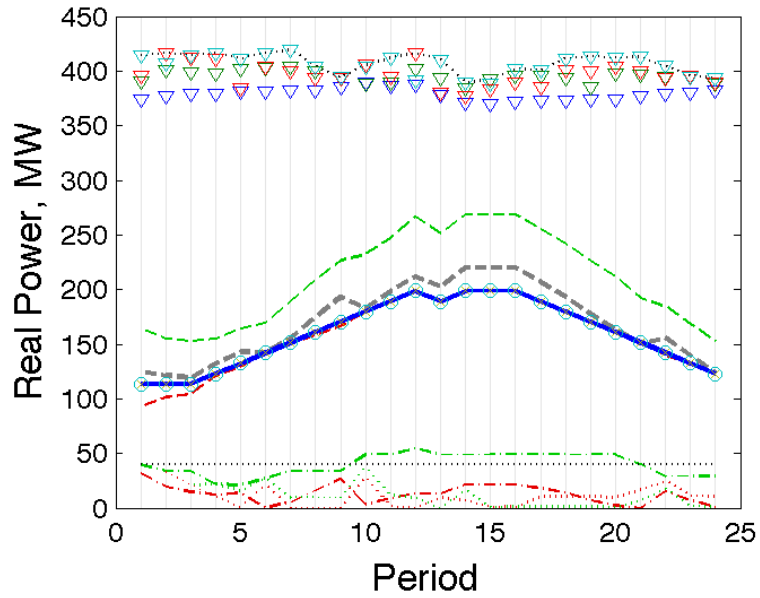
### Stored Energy for Storage Unit 1 (Gen 7) @ Bus 7



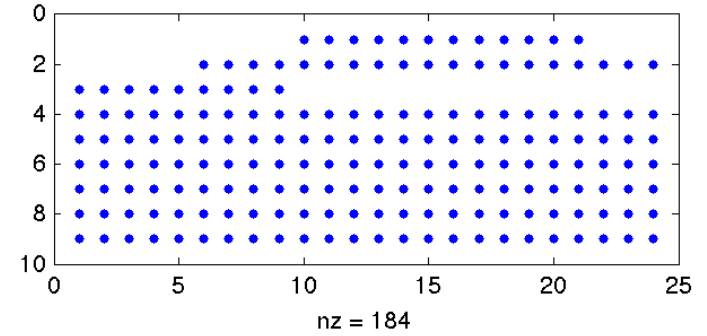
### Wind potential and curtailment



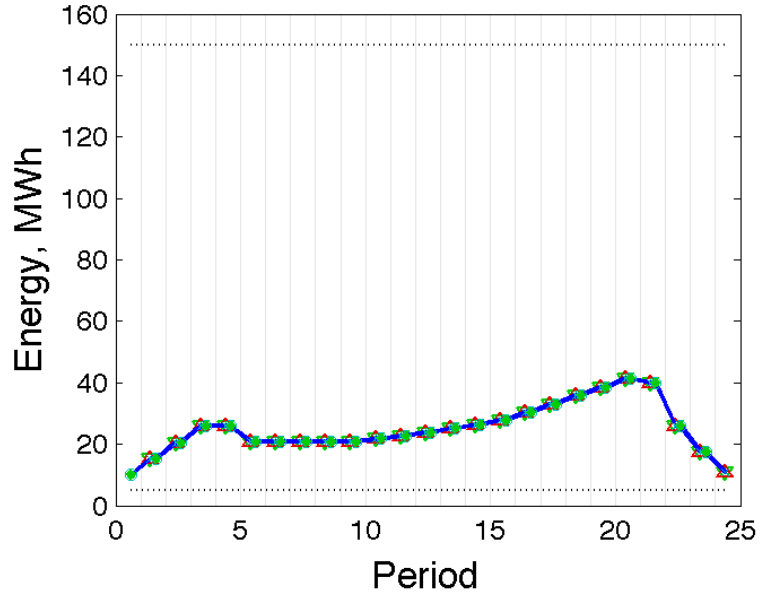
### Real Power Output for All Generators



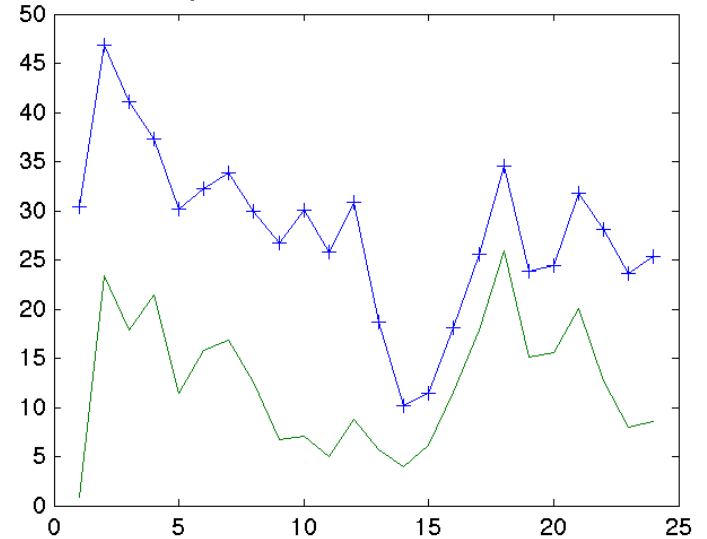
### Commitment Schedule



### Stored Energy for Storage Unit 1 (Gen 7) @ Bus 7



### Wind potential and curtailment



# Questions