## Techno-Economic Optimization of Advanced Energy Plants with Integrated Thermal, Mechanical, and Electro-Chemical Storage

Award#:DE-FE0031771 PI: Debangsu Bhattacharyya<sup>a</sup> Co-PI: M. M. Faruque Hassan<sup>b</sup> <sup>a</sup> Department of Chemical and Biomedical Engineering, West Virginia University <sup>b</sup> Department of Chemical Engineering, Texas A&M

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#### • Motivation

- Opportunities of Proposed Decentralized Storage
- Challenges of Decentralized Storage
- Energy Storage Alternatives
- Optimal Synthesis and Suggested Case Studies
- Preliminary Results
- Project Schedule: Milestones



## **Motivation**



- Increasing renewable energy penetration into the grid is resulting in increasing cycling operation of conventional generating units
- Increasing cycling leads to efficiency loss, higher emissions, and adverse effect on plant health
- Incorporating energy storage can lower power plant ramp rates while providing necessary power
- Information available in the current open literature shows that storage integration at grid level (centralized storage) is being mainly considered: large capacity and costs
- When power plant level storage(decentralized storage) is considered, synergy with the power plant operation is not exploited
- Here we propose storage integration at power plant level (decentralized storage) exploiting the synergy with the host plant



#### CAISO Duck Curve<sup>[1]</sup>



Source: www.caiso.com.





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## **Opportunities of Proposed Decentralized Storage**

- Greatly improves the number of potential storage options
- **Reduce costs** and **utilize** existing equipment items in host power plants
  - Since all storage options should finally produce electricity, existing equipment items from power plants can be utilized rather than purchasing new equipment items for centralized options
  - Decentralized storage is also likely to reduce the O&M cost including personnel cost
- Decentralized thermal storage can be easily achieved at various exergy levels thus providing high flexibility and efficiency
- Exploit synergy with the host power plant
  - For example, the hot air from battery cooling can be added to the combustion air of a power plant
- Lower investment
  - Much lower capital investment than centralized storage customized on the desired load-following capability of the host plant
  - Future power plants with integrated storage option may have lower nominal rating thus reducing capital investment





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## WestVirginiaUniversity. Challenges of Decentralized Storage



- Maximum ramp rate, achievable lowest load, and penalty (efficiency, emission, health) of various types of power plants are vastly different.
  - For optimal synthesis of the energy storage technologies and their ratings, dynamics of the host plant need to be considered along with the dynamics of the storage technology.
  - Impact of the cycling operation of the auxiliary equipment items provided for storage on their efficiency and health needs to be considered in the cost function.

Therefore the optimization problem for storage synthesis is dynamic optimization and can be considerably large and expensive.

- Decentralized storage option should align with the business strategy of the host plant.
  - For example, storing energy as a chemical to be sold may not align well with the business model of the power plants
- Should have admissible impact on power plant operation and configuration
- Storage technologies vary widely based on the maturity of technologies, their costs, capacity, life, availability, efficiency, footprint, dynamics, safety and environmental hazards. These aspects need to be considered in optimal selection of technology/technologies for a host power plant.





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### **Energy Storage Alternatives**



• Note: Not necessarily each type of storage will be considered for each type of power plant

#### Synergistic Thermal Storage in an SCPC Plant Using Molten Salt

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#### Synergistic Thermal Storage in an NGCC Plant Using Phase Change Materials (PCMs)





West Virginia University.

## West Virginia University. Synergistic Mechanical Storage in an NGCC Plant Using Compressed Air





#### Synergistic Chemical Storage in an IGCC Plant Using Hydrogen

West Virginia University.





#### Synergistic Electro-Chemical Storage in an NGCC Plant Using Sodium Sulfur Battery





Blower

West Virginia University,





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## **Optimal Synthesis of Storage Technologies**



• Trade-offs between capital costs and efficiency for technologies

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- Discrete decisions for selection of storage technologies: Mixed Integer Nonlinear Programming (MINLP) problem
- Integrated system must be able to follow a time-varying grid demand
- Need for dynamic models of storage and power plants
- Including system dynamics in the optimization framework results in Mixed-Integer, Nonlinear and Ordinary Differential Equation (MINODE) models







- Data-Driven Black-Box Optimization used
- Consists of Feasibility and Optimization phases
- Both phases based on trust-region framework
- Utilize input-output simulation data for constructing surrogate models
- Can handle infeasible initial points
- General form:

 $\min_{x} f(x) \qquad \mathbf{P1}$ s. t.  $g_k(x) \le 0$ ,  $\forall k = 1, ..., K$  Functions with hidden expressions in a black-box simulator  $g_u(x) \le 0$ ,  $\forall u = 1, ..., U$  Functions with known expressions  $x_i \in [x_i^L, x_i^U] \quad \forall i = 1, ..., n$ 

Provide initial guess,  $x_0$  and number of samples to be maintained No Feasibility  $\theta(x_0) \le \epsilon$ Phase Yes Optimization Phase

Bajaj et al. "A trust region-based two phase algorithm for constrained black-box and grey-box optimization with infeasible initial point." Computers & Chemical Engineering 116 (2018): 306-321.



## **Suggested Case Studies**



#### One Summer Day Imbalance from California ISO Considered as 'Today'

#### Net demand (demand minus solar and wind) AS OF 17.05

This graph illustrates how the ISO meets demand while managing the quickly changing ramp rates of variable energy resources, such as solar and wind. Learn how the ISO maintains reliability while maximizing clean energy sources.



#### One Winter Day Imbalance from California ISO Considered as 'Extreme'

#### Net demand (demand minus solar and wind) AS OF 17.05

This graph illustrates how the ISO meets demand while managing the quickly changing ramp rates of variable energy resources, such as solar and wind. Learn how the ISO maintains reliability while maximizing clean energy sources.







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## West Virginia University. Preliminary Results: NGCC Plant with Integrated Sodium Sulfur Battery



Development of reduced order model (ROM) for the NGCC plant to be used for optimal downselection:

- Linearization of the nonlinear NGCC plant model at different load. (e.g. full load, 80% load, 60% load, etc.)
- Large-scale state-space model includes about 600 state variables

APD

FOM

ROM

12

15

9

Time(hr)

660

550

440

330

220

0

3

6

Total Power Output(MWe)

 ROM is generated using the balanced truncation method, based on Hankel singular value (HSV) decomposition



**APD:** High-fidelity model in Aspen plus dynamics; FOM: Full-order model linearized at full load, Size(FOM)= **582** ROM: Reduced-order model, Size(ROM)=**35** 

Time(hr)

450

375

300

225

150

0

GT Power Output(MWe)

## West Virginia University. Preliminary Results: NGCC Plant with Integrated Sodium Sulfur Battery



Development of reduced order model (ROM) for the Sodium-Sulfur Battery to be used for optimal downselection:

- NAARX (Non-linear Additive Autoregressive with Exogenous Input) model
- Model selection using Akaike Information Criterion

#### **Model Validation: Discharging in 2-Phase Region**









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| Milestone | Task/Sub | Milestone Title and Descriptions   | Planned                | Verification Method   |
|-----------|----------|--|------------------------|---|
| No.       | task     |  | <b>Completion Date</b> |   |
| 1         | 4.1      | Complete the simplified input-output<br>model of the proposed mechanical,<br>chemical and electrochemical energy<br>storage technologies | 10/31/20               | Root mean squared error (RMSE) is within 10% of rigorous dynamic model and time to simulate the model takes less than 1/100 <sup>th</sup> of real time so that they can be used in optimization |
| 2         | 4.2      | Complete the simplified input-output<br>model of the proposed thermal<br>energy storage technologies                                     | 10/31/20               | Root mean squared error (RMSE) is within 10% of rigorous dynamic model and time to simulate the model takes less than 1/100 <sup>th</sup> of real time so that they can be used in optimization |
| 3         | 5.0      | Complete the energy storage technology downselection   | 12/31/20               | MINLP algorithm successfully satisfied the convergence criteria   |
| 4         | 8.0      | Complete the TEA of six selected system concepts   | 12/31/21               | Energy storage technologies can satisfy the<br>imbalance with at least 10% excess storage, which<br>is the design margin  |





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# Thank you for your attention

# Questions?