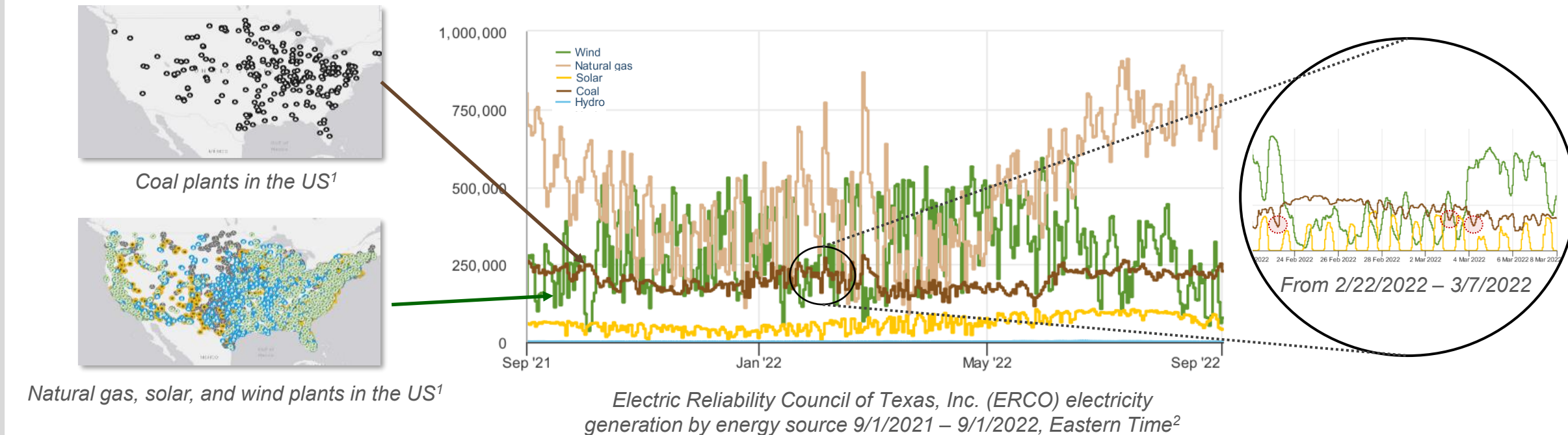


Multiperiod Generalized Disjunctive Programming Optimization in IDAES: Simultaneous Operation and Design of an Integrated Energy System

Edna Soraya Rawlings^a, Jaffer Ghouse^b, Naresh Susarla^b, John Siirola^a, David Miller^b
^a Sandia National Laboratories, ^b National Energy Technology Laboratory

Motivation

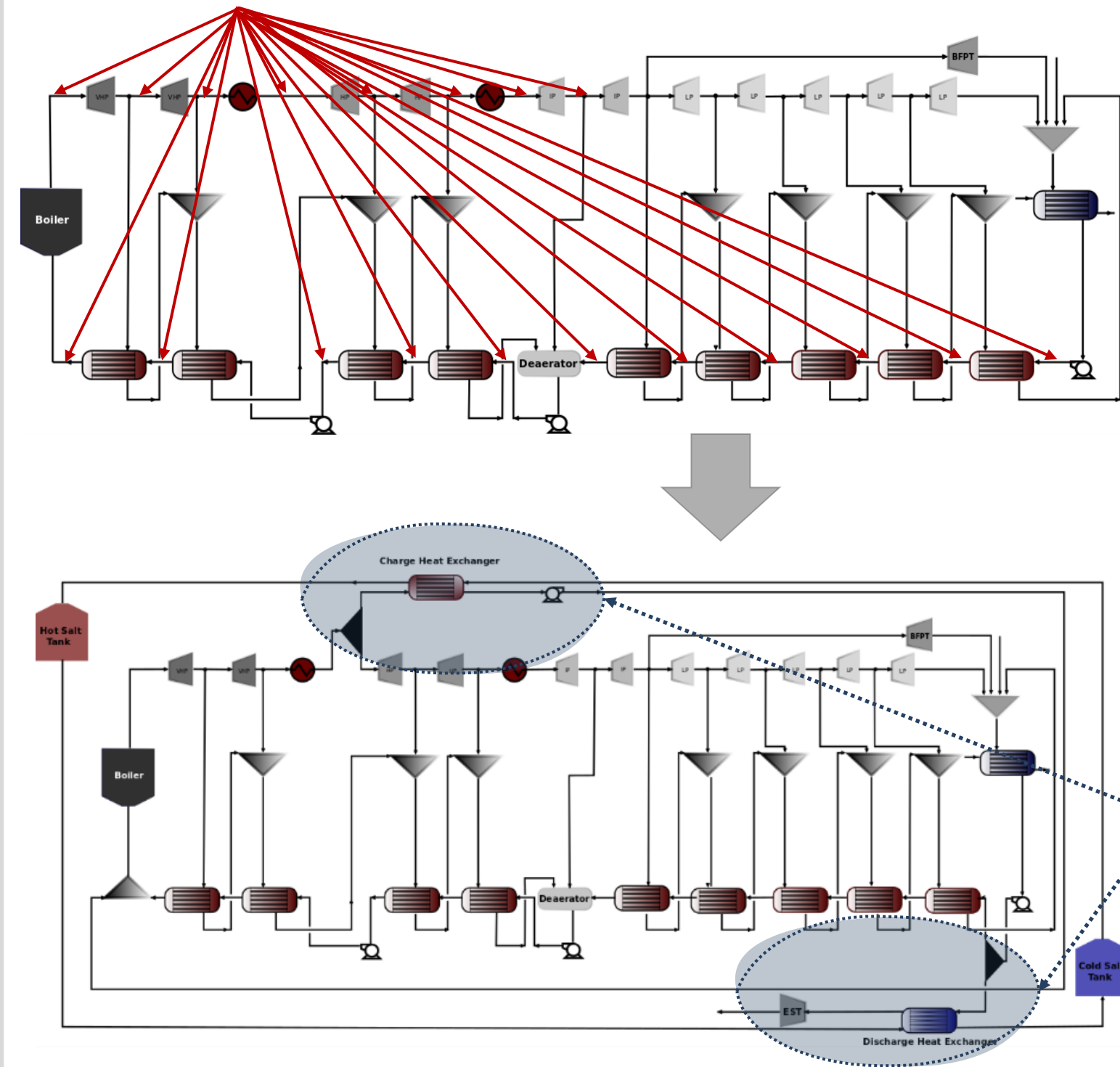
- Integration of variable renewable energy requires more flexible fossil generators
- Thermal energy storage (TES) systems allow generators to reduce frequent cycling and operate at higher efficiencies



¹ Source: U.S. Energy Information Administration, U.S. Energy Mapping System, <https://www.eia.gov/state/maps.php>, Accessed July 2, 2021. ² Source: U.S. Energy Information Administration, Hourly Electric Grid Monitor, <https://www.eia.gov/electricity/gridmonitor/expanded-view/custom/pending/GenerationByEnergySource-4/edit>, Accessed September, 2022

Problem Statement

Where to integrate the storage system in a real plant?



To include the TES, decide:

What **storage material** is the best and cheapest?

What is the best **design** for the storage system?

Where is the **optimal integration point** of the TES?

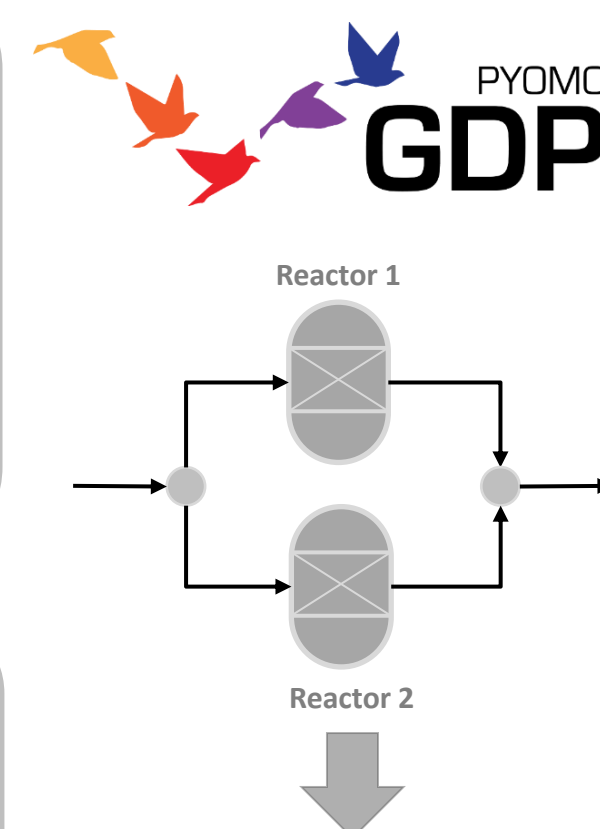
Once TES is integrated:

When and how to deploy storage to maximize profit?

Methodology

Conceptual Design using IDAES

- Use IDAES unit models and interconnections to create rigorous superstructures
- Use **Generalized Disjunctive Programming** (GDP) to incorporate discrete design decisions as disjunctions
- Use of advanced solver **GDPopt**, to explore entire design space within IDAES



Multi-period Approach

- Pricetaker assumption to determine optimal schedule and design of storage operating cycles
- Two formulations for determining optimal schedule: Nonlinear programming (NLP) and GDP

$$\begin{bmatrix} z_1 \\ h_1(x) = 0 \\ g_1(x) \leq 0 \end{bmatrix} \vee \begin{bmatrix} z_2 \\ h_2(x) = 0 \\ g_2(x) \leq 0 \end{bmatrix}$$

It can be applied for designing storage systems for **fossil, nuclear and geothermal energy systems**

Conceptual Design of Energy Systems

- Build two superstructures to identify optimal points of integration of storage system with thermal generator
- Rigorous flowsheet using first-principle models for steam cycle and TES with accurate properties for steam/water using **IAPWS95**

Charge Cycle

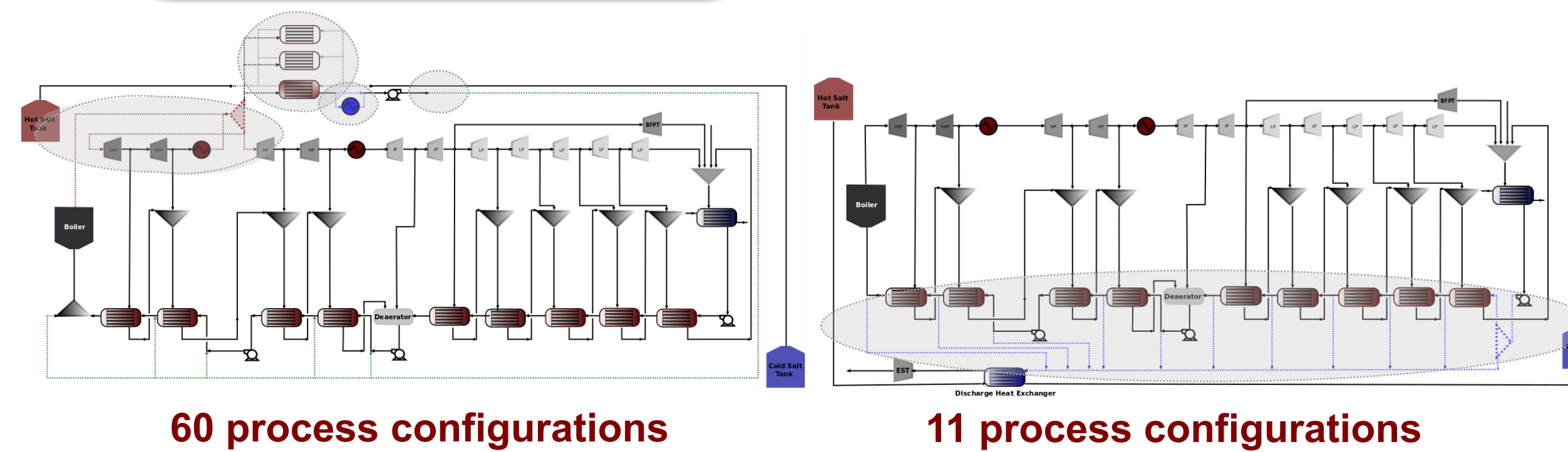
Discrete decisions:

- Steam source** to heat up storage material
- Condensed steam return**
- Storage material:** Solar, Hitec, Thermal oil
- Use of a **cooler** after TES

Discharge Cycle

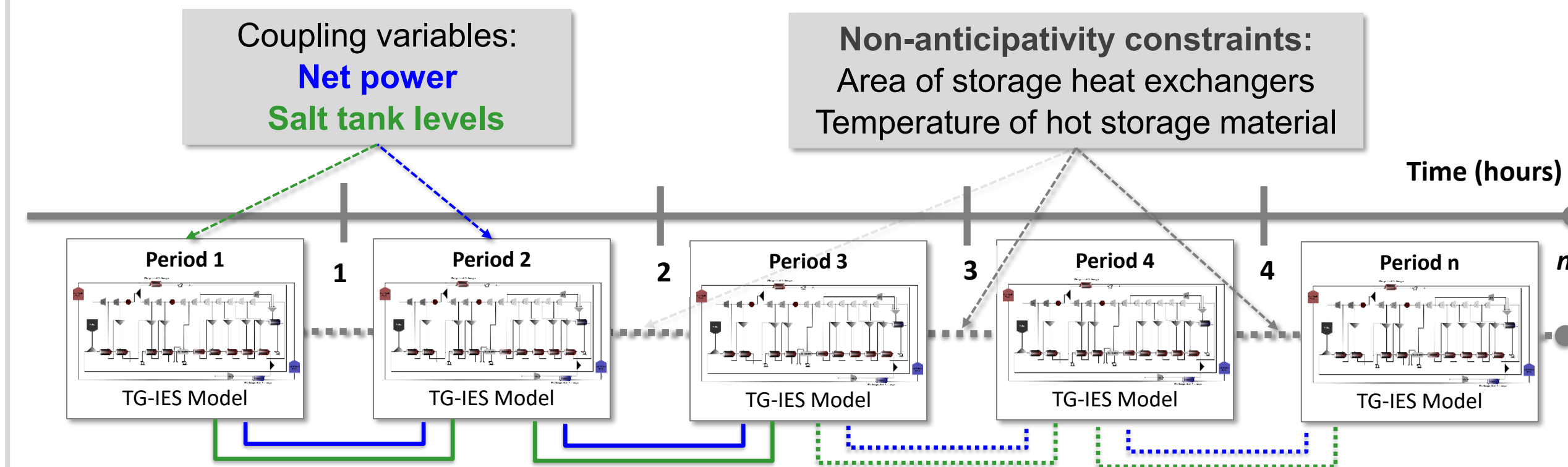
Discrete decision:

- Condensed steam source** using stored energy in storage material
 - Dedicated turbine connected to the discharge system



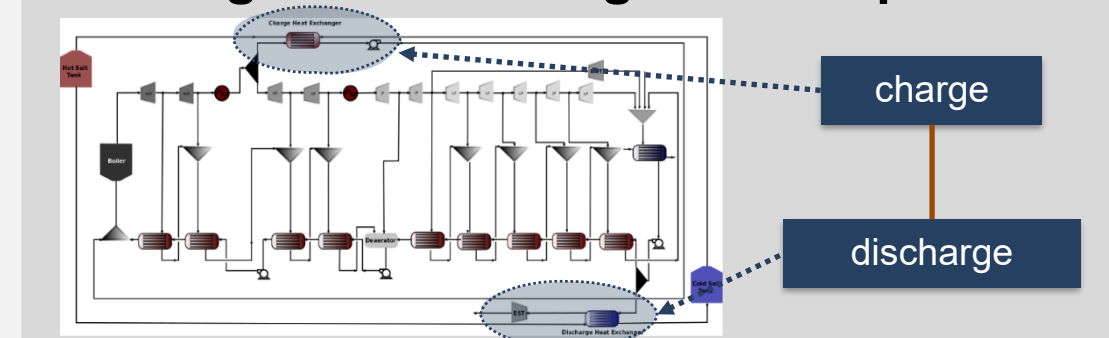
Multi-period Storage Schedule and Design

- Optimal design and operation** of the thermal generator and storage system in a given time horizon
- Optimal points of integration of TES with thermal generator obtained during GDP conceptual design
- NLP and GDP** models to determine the optimal hourly operating design and schedule of the thermal energy storage system



NLP Model

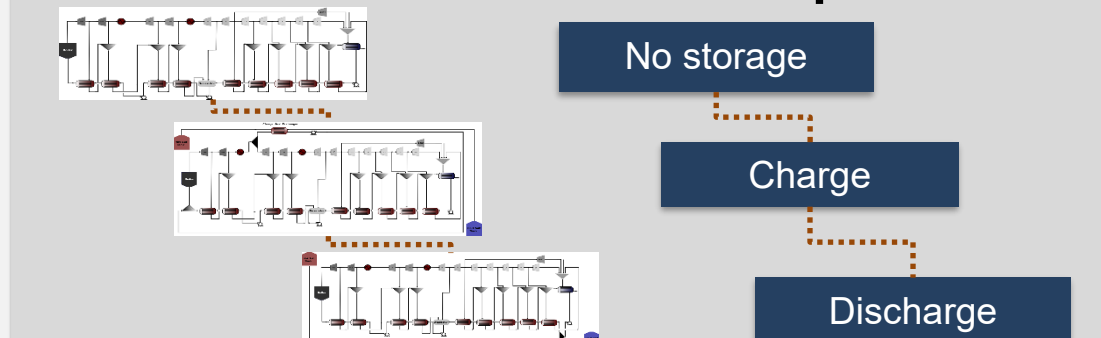
Charge and discharge at each period:



- Minimum flow requirement for the storage operating cycles
- Simplified scheduling analysis to determine when to use storage

GDP Model

Three alternatives at each period:



- Disjunction for the selection of operation mode
- Problem size from **millions** to **quadrillions: 3ⁿ alternatives**

Results

Conceptual design of storage superstructures

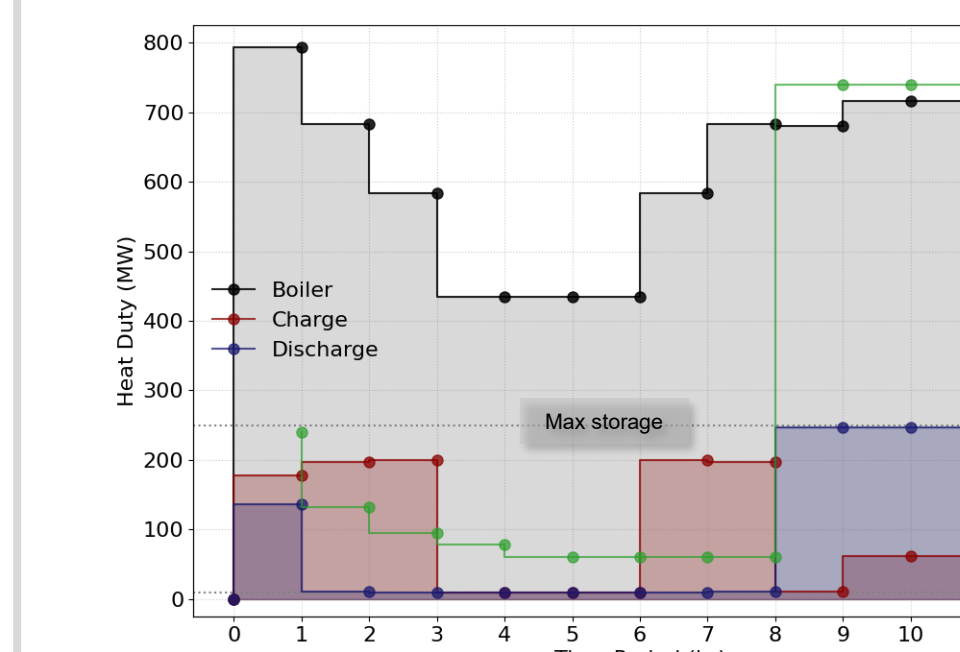
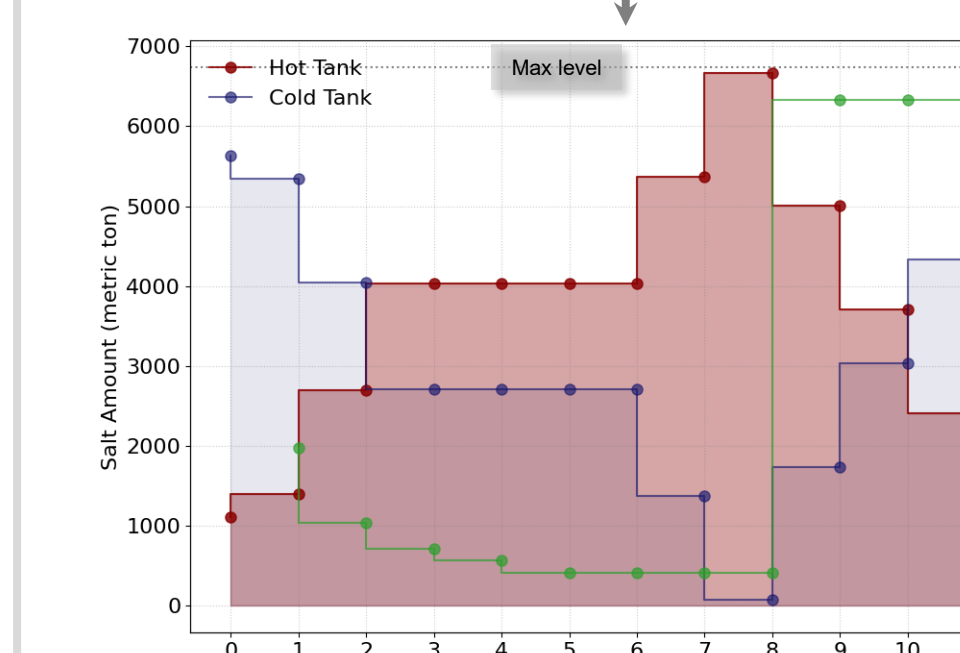
- Optimal design for **charge superstructure**: Solar salt as storage material, a high pressure steam, condensed steam to return to the boiler, and no cooler
- Optimal design for **discharge superstructure**: Condensed steam source after condenser

Multi-period Analysis for Storage Schedule and Design

- Determine heat exchanger areas and operating temperatures
- Optimal schedule varies based on electricity prices
- GDP multi-period allows for the search when **3¹² alternatives** are available

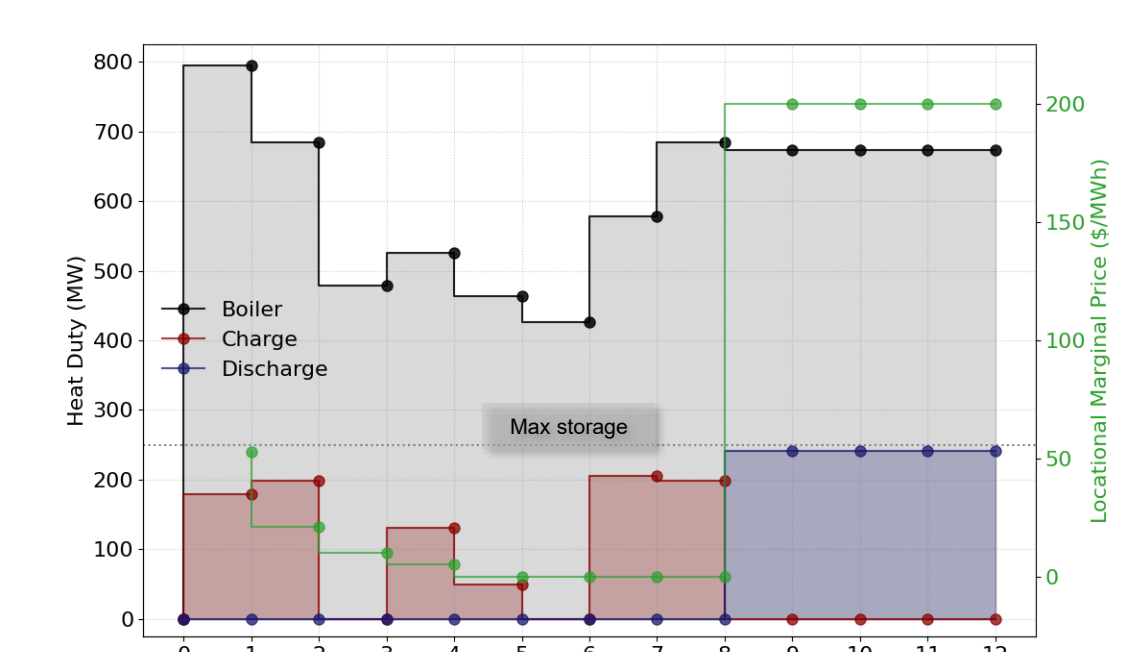
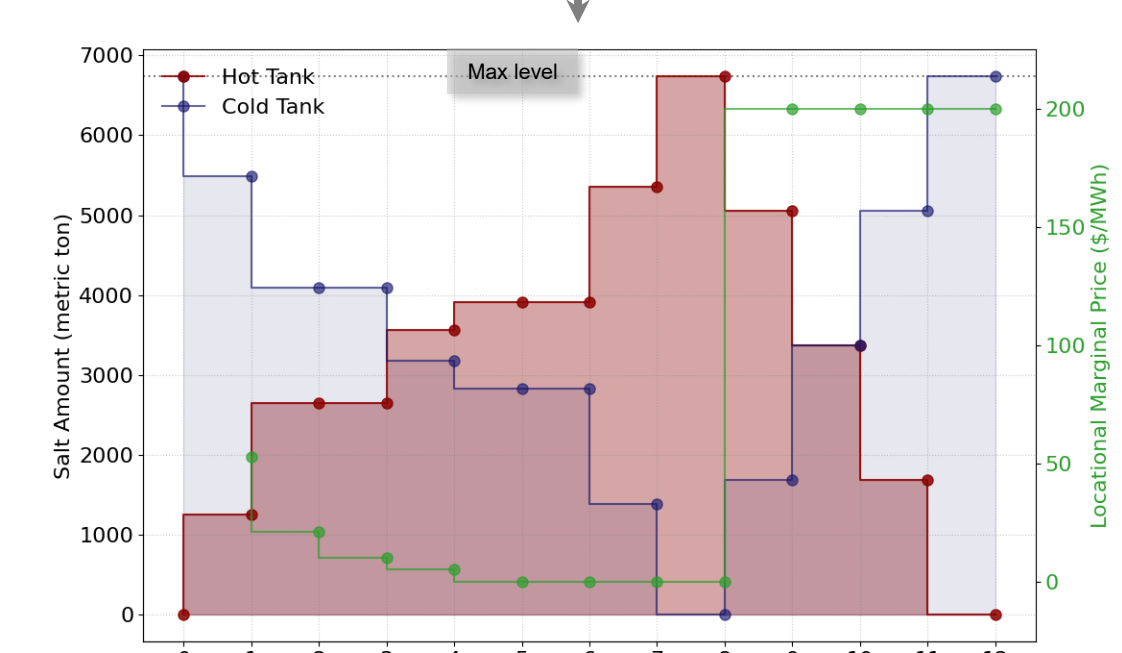
NLP multi-period approach

Area heat exchangers
Charge: 2987 m²
Discharge: 3000 m²
Profit: \$311,953



GDP multi-period approach

Area heat exchangers
Charge: 3000 m²
Discharge: 2486 m²
Profit: \$344,947



GDP approach improve profit by 10% when comparing with NLP approach with minimum flow constraints

Future Work

- Include using discharge steam in CCS system as a potential design option
- Extend analysis of NLP and GDP multi-period models to consider electricity market surrogates, i.e. remove pricetaker assumption

Contact: Edna Soraya Rawlings, esrawil@sandia.gov

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