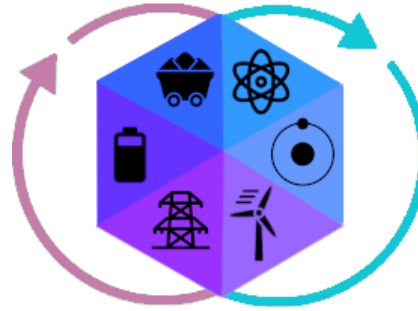


IDAES
Institute for the Design of
Advanced Energy Systems



DISPATCHES
Design Integration and Synthesis
Platform to Advance Tightly
Coupled Hybrid Energy Systems



Toward Future Energy Generation Systems: Multi-Scale Optimization with Market Interactions

AIChE Annual Meeting, November 10th, 2021, Boston, MA

Jordan Jalving¹, Jaffer Ghouse², Ben Knueven³, Shawn Martin¹, Nicole Cortes⁴, Xian Gao⁴,
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³ National Renewable Energy Laboratory, Golden, CO

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Carnegie Mellon

West Virginia University

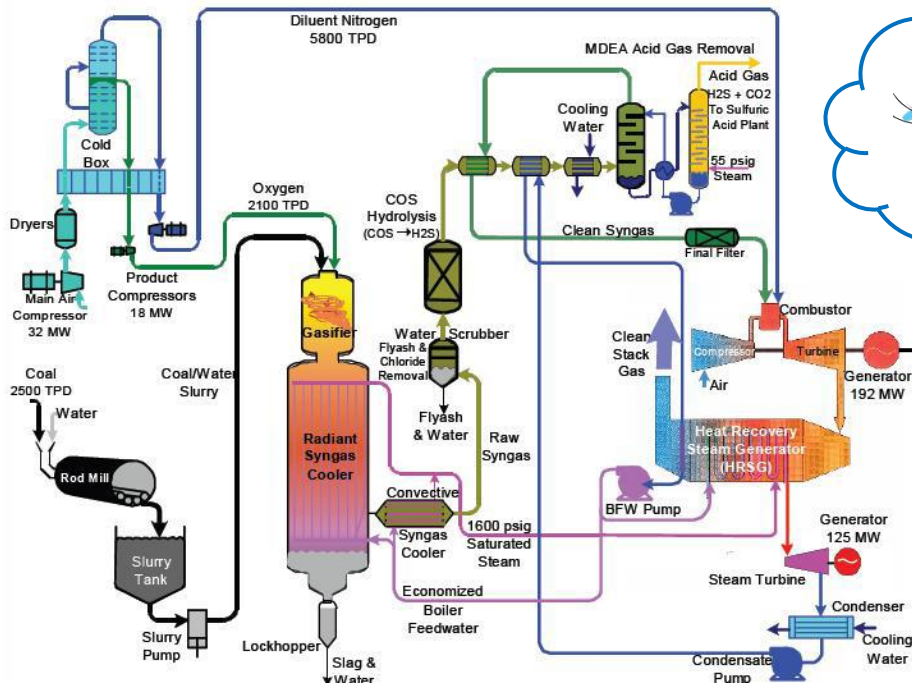


**U.S. DEPARTMENT OF
ENERGY**

Energy system analysis capabilities are applied in isolation

Process-centric Modeling

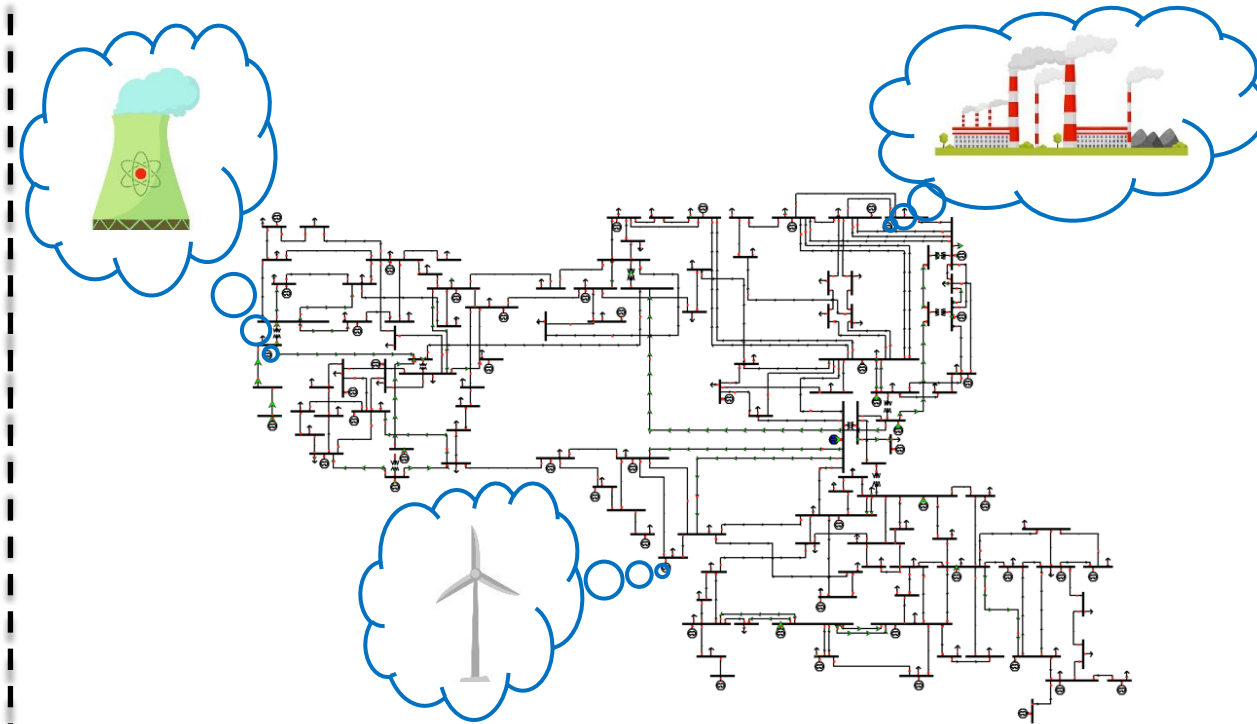
Detailed steady state or dynamic process models,
with the grid modeled as an infinite capacity bus



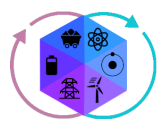
<https://www.netl.doe.gov/research/coal/energy-systems/gasification/gasifedia/igcc-config>

Grid-centric Modeling

Detailed power flow models,
with individual generators modeled as either
dispatchable point sources or stochastic "negative loads"

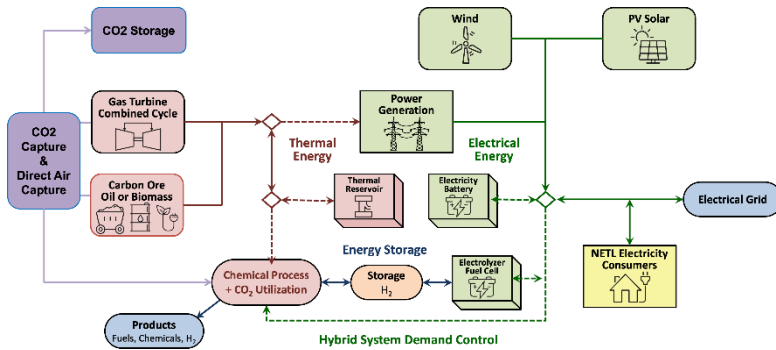


<https://icseg.iti.illinois.edu/files/2013/10/IEEE118.png>



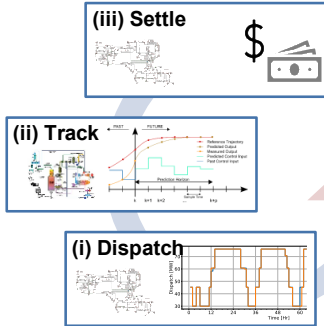
Challenge of increasingly integrated & dynamic grid/generation

Generation & Process Modeling

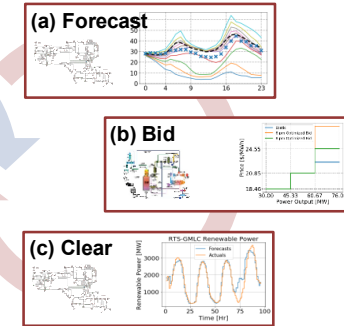


Integrated Resource-Grid Model

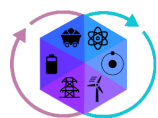
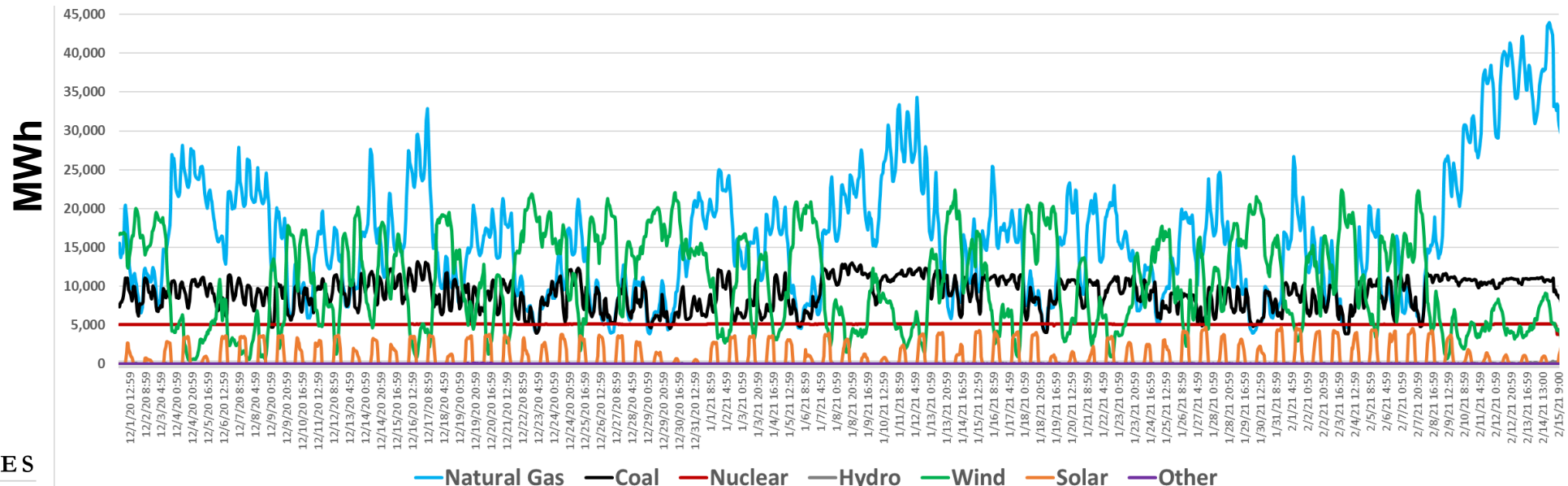
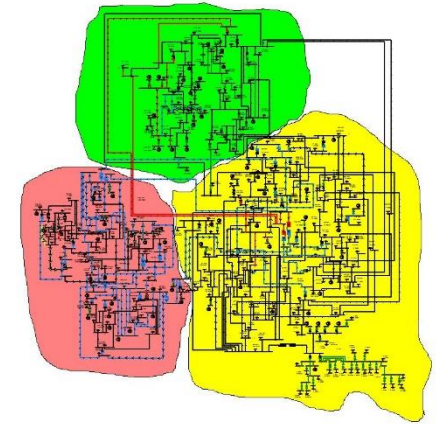
Real-Time Market Loop (1 cycle = 1 hour)



Day-Ahead Market Loop (1 cycle = 1 day)



Grid Modeling

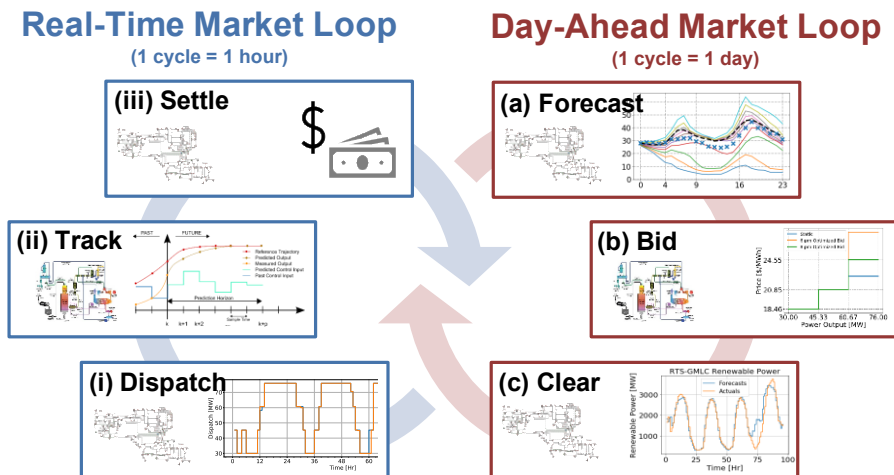


DISPATCHES

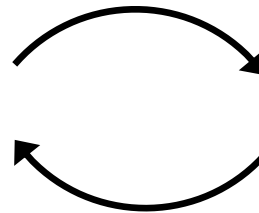
Design Integration and Synthesis
Platform to Advance Tightly
Coupled Hybrid Energy Systems

Multiscale Market-Based Optimization of IES

Multiscale Simulation to Quantify Grid/IES Interactions



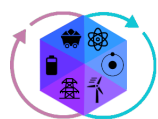
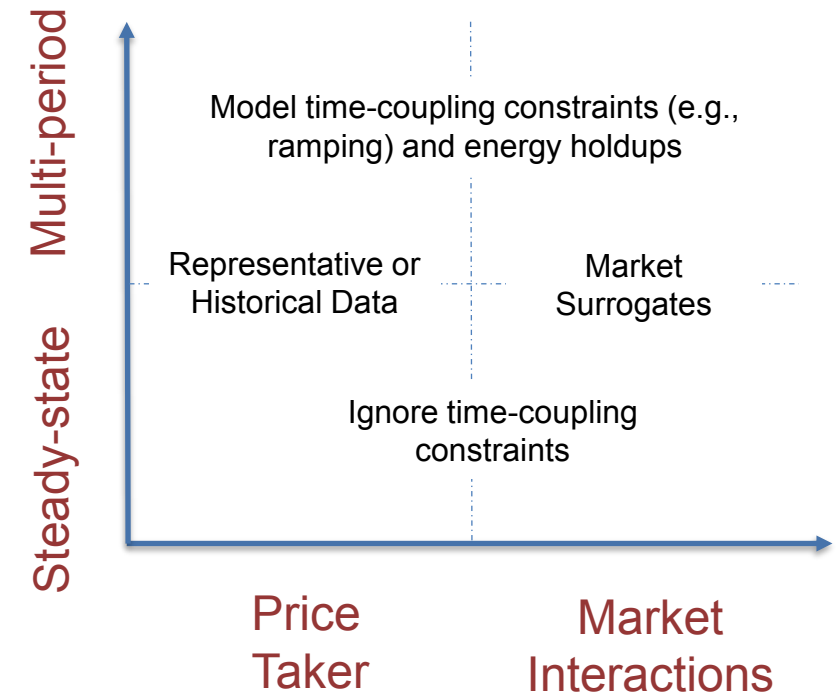
Data to refine market representation (e.g., prices, surrogates)



Candidate IES designs (i.e., IDAES models) to evaluate

Superstructure Optimization

Maximize Net Present Value



Abstract Integrated Energy System

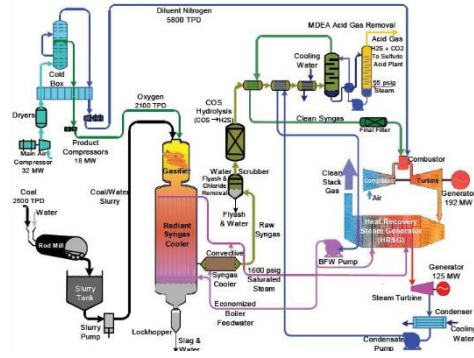
Integrated Energy System

Design Decisions: d

Operating Decisions: u

Net Power Output: δ

Steady-State Model (Constraints): $g(d, u, \delta) = 0$



Power δ

Energy Market



Price π

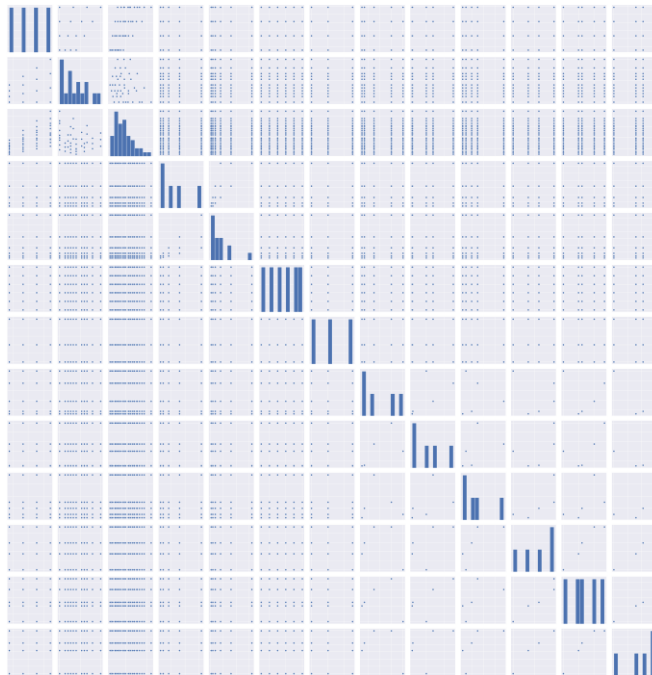
This abstraction is easy to extend to...

- multiple products (electricity, heat, H_2 , chemicals)
- multiple market timescales (day ahead, real time, ancillary services, capacity)

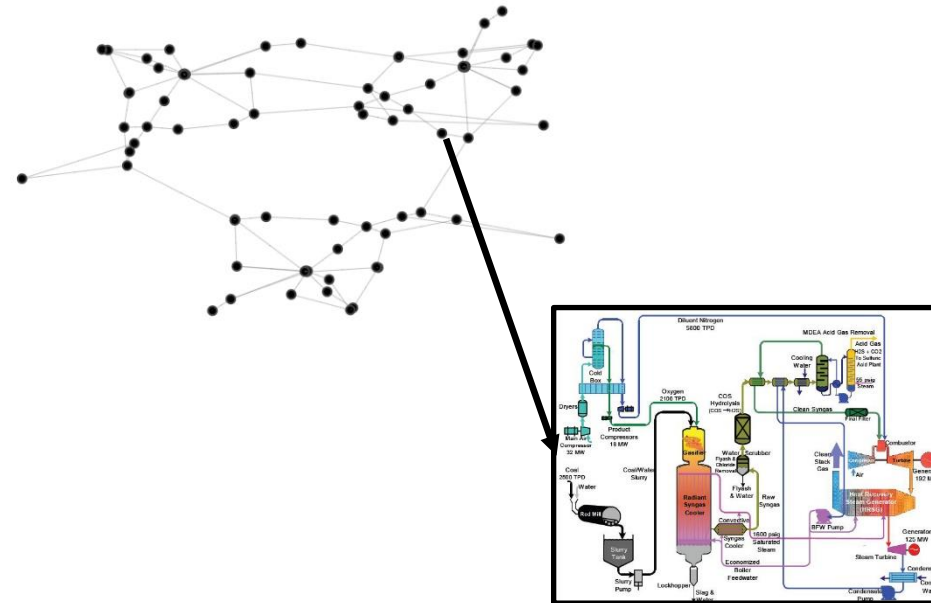
How would a new generator change market outcomes?

Simulation Design

	Market Inputs		
1	PMax [MW]	7	No Load Cost [\$ /hr]
2	PMin [MW]	8	Start Time Hot [Hr]
3	Ramp Rate [MW/hr]	9	Start Time Warm [Hr]
4	Min Up Time [Hr]	10	Start Time Cold [Hr]
5	Min Down Time [Hr]	11	Start Cost Hot [\$]
6	Marginal Cost [\$ /MWh]	12	Start Cost Warm [\$]
		13	Start Cost Cold [\$]

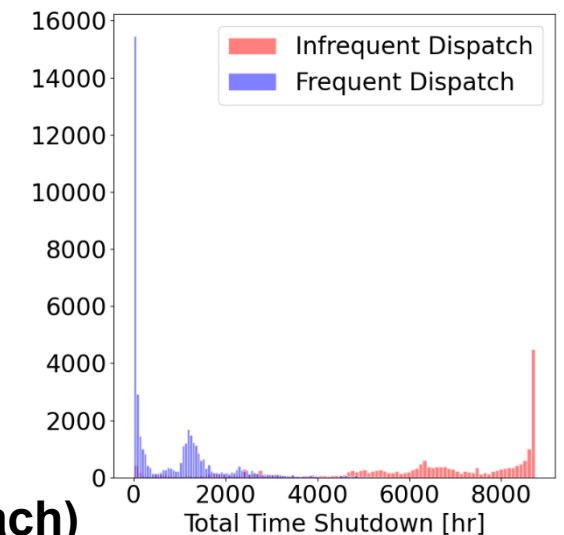
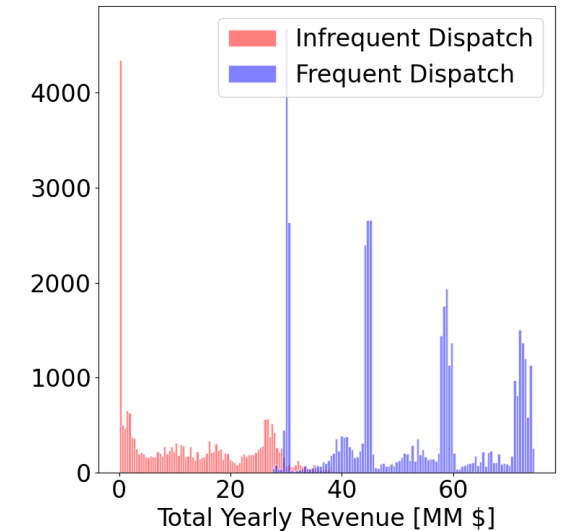


Production Cost Modeling: RTS-GMLC Test System

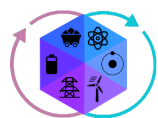


PRESCIENT

Revenue & Dispatch Results



64,800 1-year Prescient simulations (@ 3 hr each)



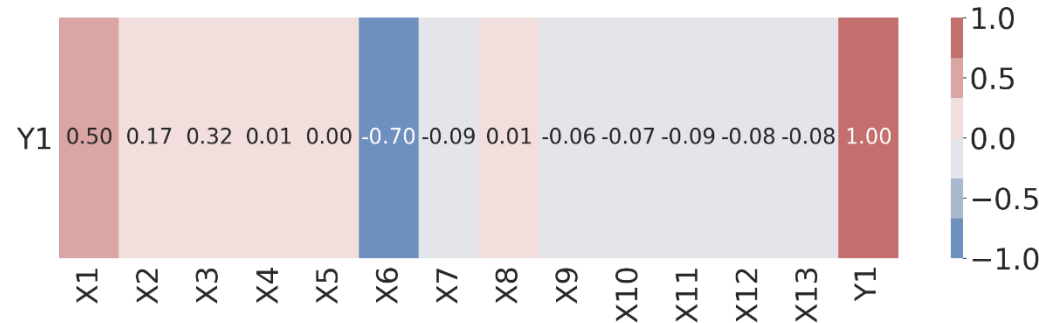
DISPATCHES

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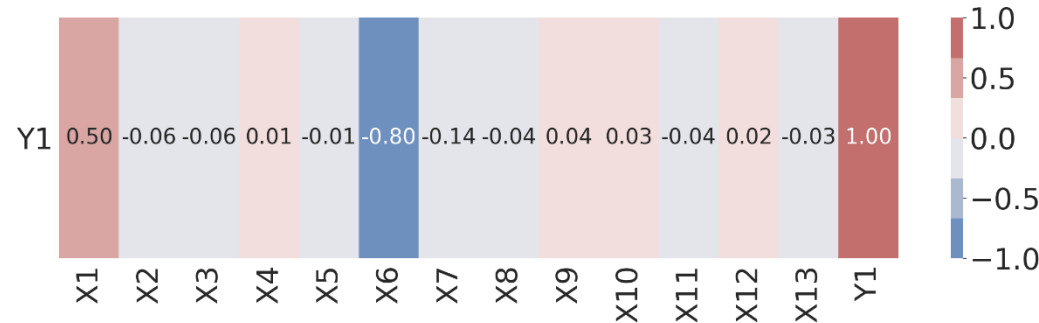
Global Sensitivity Analysis

	Market Parameters
X1	P_{\max} [MW]
X2	P_{\min} [MW]
X3	Ramp Rate [MW/hr]
X4	Min Up Time [Hr]
X5	Min Down Time [Hr]
X6	Marginal Cost [\$/MWh]
X7	No Load Cost [\$/hr]
X8	Start Time Hot [Hr]
X9	Start Time Warm [Hr]
X10	Start Time Cold [Hr]
X11	Start Cost Hot [\$]
X12	Start Cost Warm [\$]
X13	Start Cost Cold [\$]
Y1	Revenue [MM\$]

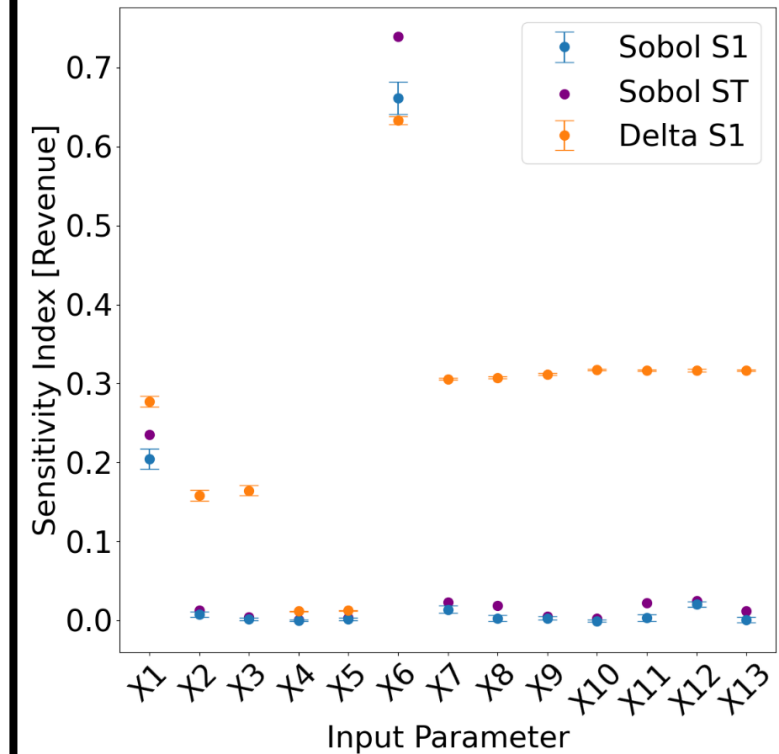
Spearman Correlations



PRCC (Ranked Partial Correlations)

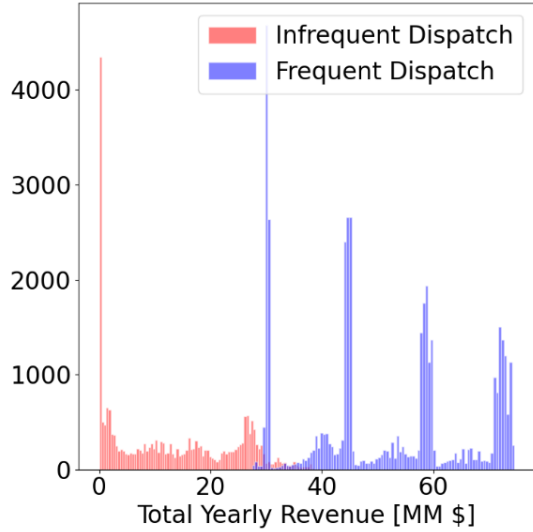


Sensitivity Indices

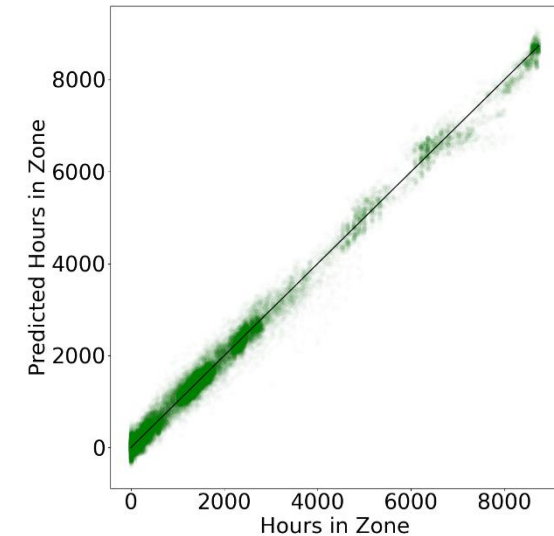
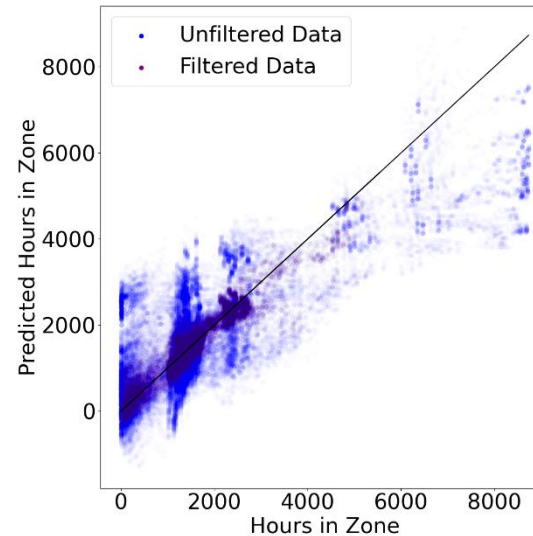
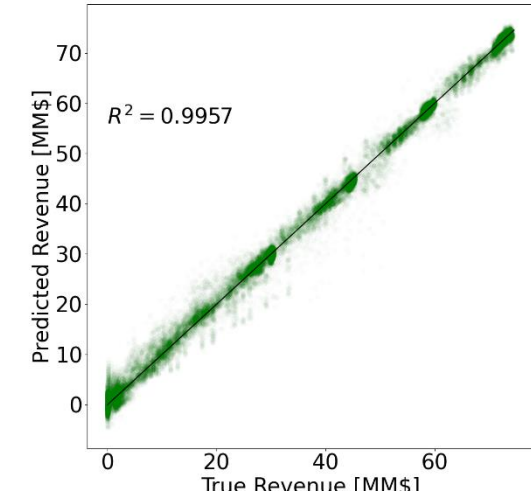
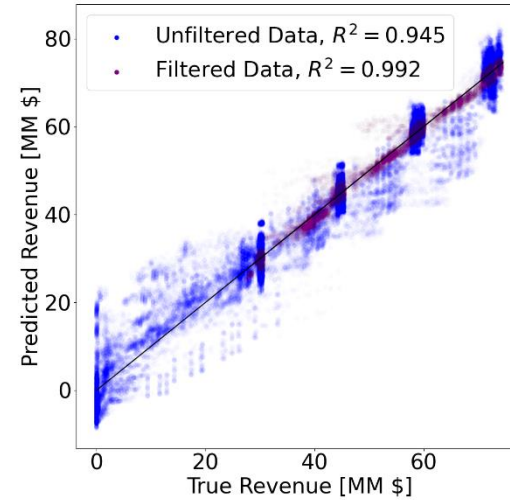
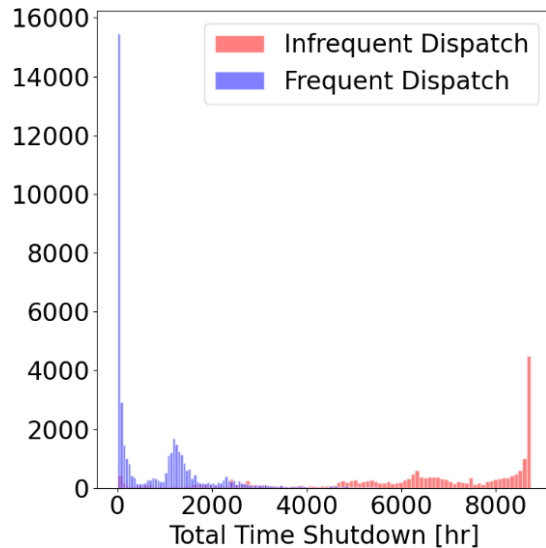
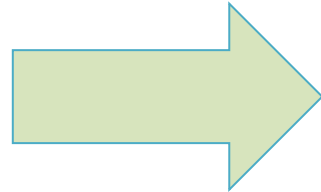


- Sensitivity analysis tends to confirm intuition
 - P_{\max} and marginal cost describe most variation in output
 - However, all inputs improve overall surrogate fit

Surrogates Accurately Predict Market Outcomes



Fit Market
Surrogates



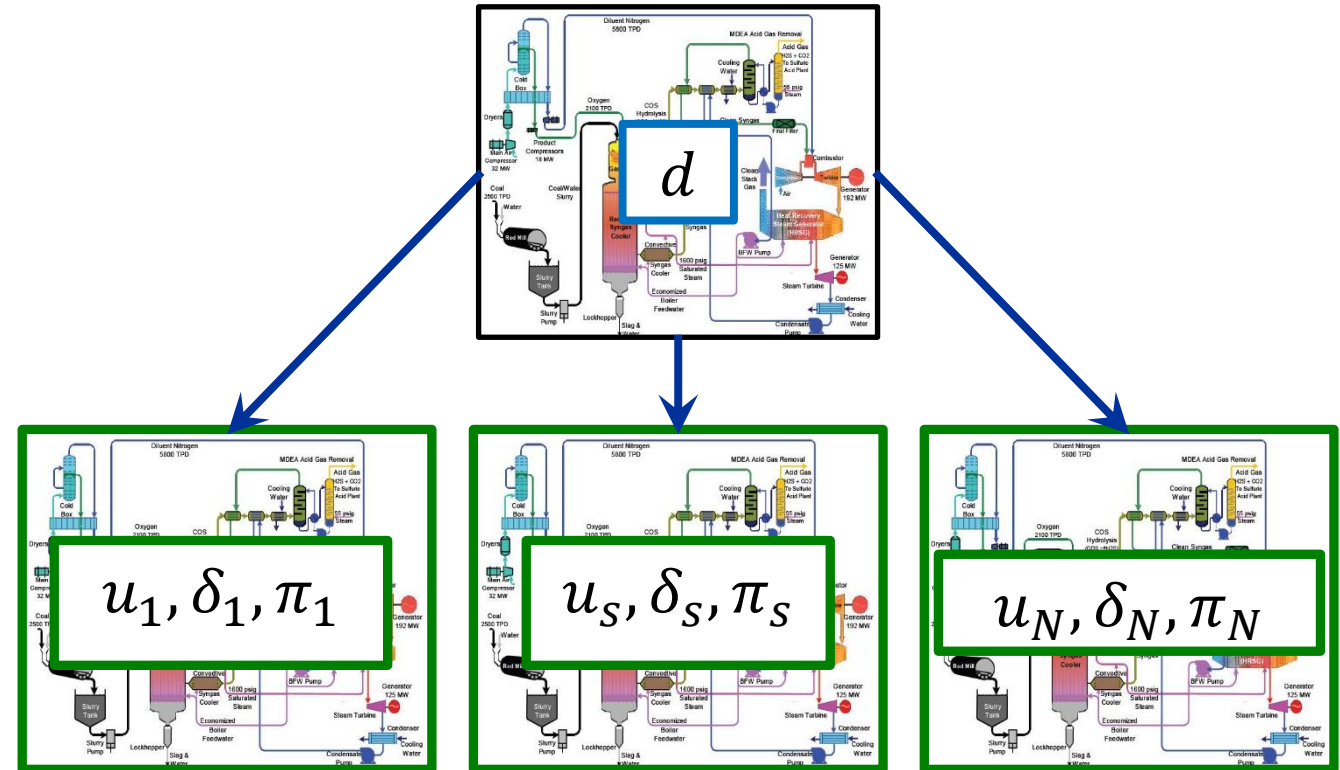
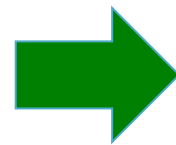
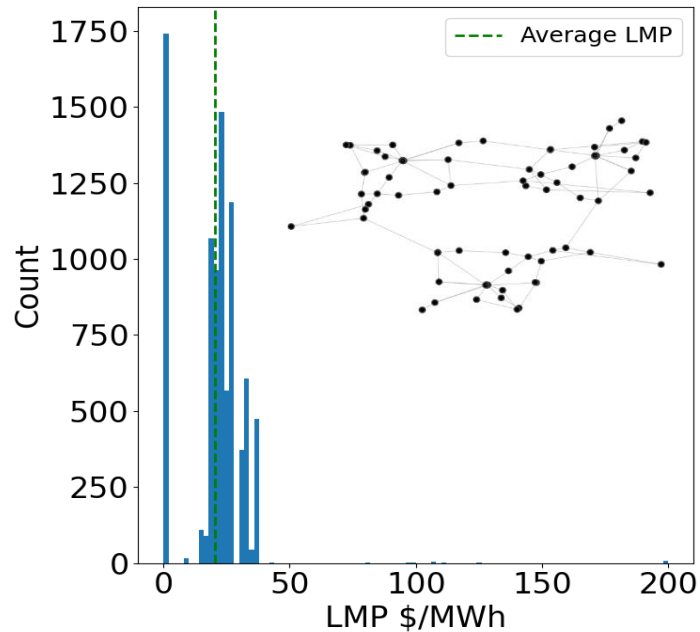
Steady-State Price Taker (Self-Schedule)

$$\max_{d, u, \delta} \sum_{s \in S} \overbrace{w_s}^{\text{Weight / Frequency}} \overbrace{[R(d, u_s, \delta_s, \pi_s) - C(d, u_s, \delta_s)]}^{\text{Revenue (Operating + Capital)}}$$

$$g(d, u_s, \delta_s) = 0, \quad \forall s \in S$$

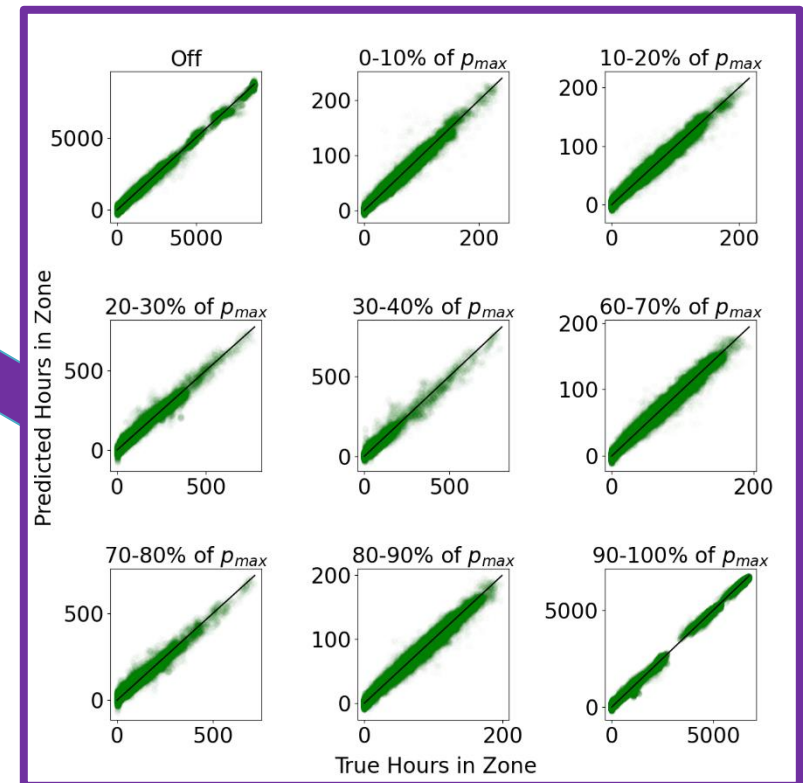
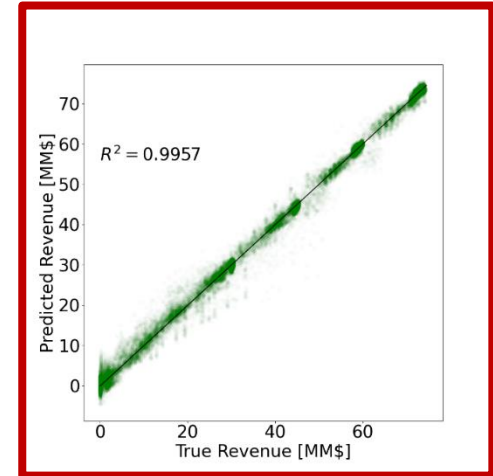
d	Design decisions
δ_s	Power output decision for scenario
u_s	Operating decisions for scenario
π_s	Scenario price (data)
$R()$	Revenue: function of decisions and prices
$C()$	Cost: function of decisions

Nominal RTS-GMLC Prices



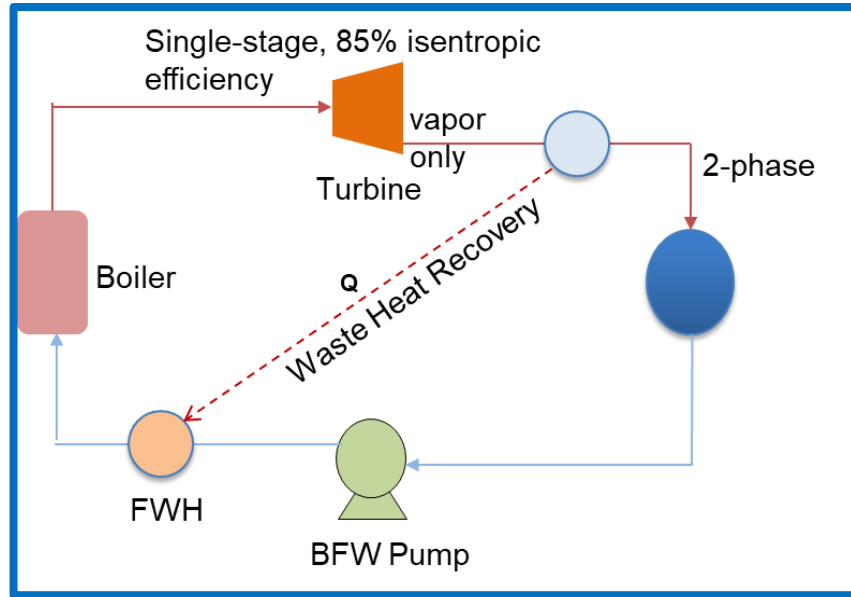
Steady-State with Market Surrogates (Bid)

$$\begin{aligned} & \text{Revenue} \\ & \max_{d, u, x} R(x) - \sum_{s \in S} w_s(x) [C(d, u_s, \delta_s)] \\ & \text{Cost (Operating + Capital)} \\ & g(d, u_s, \delta_s) = 0, \quad \forall s \in S \quad \text{Process Model} \\ & h(d, x) = 0 \quad \text{"Bid Rules"} \\ & R(x) = f_{rev}(x) \quad \text{Revenue Surrogate} \\ & w_s(x) = f_s(x), \quad \forall s \in S \quad \text{Dispatch Surrogates} \end{aligned}$$



d	Design decisions
δ_s	Power output data for scenario
u_s	Operating decisions for scenario
x	Market Inputs (bid decisions)
$R()$	Revenue: function of market inputs
$C()$	Cost: function of decisions
w_s	Scenario weight: function of market inputs

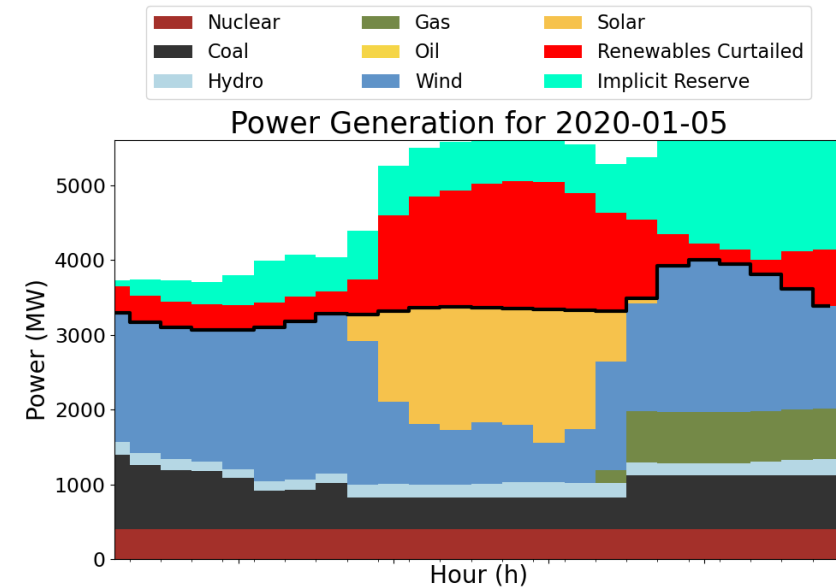
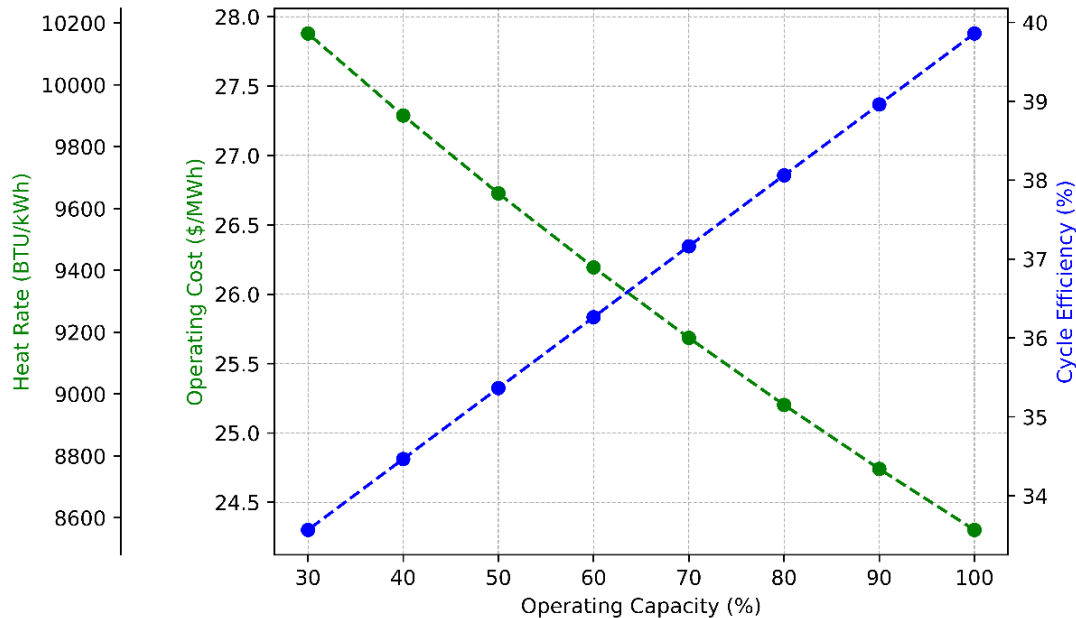
Example: Optimal Design of Rankine Cycle for RTS-GMLC



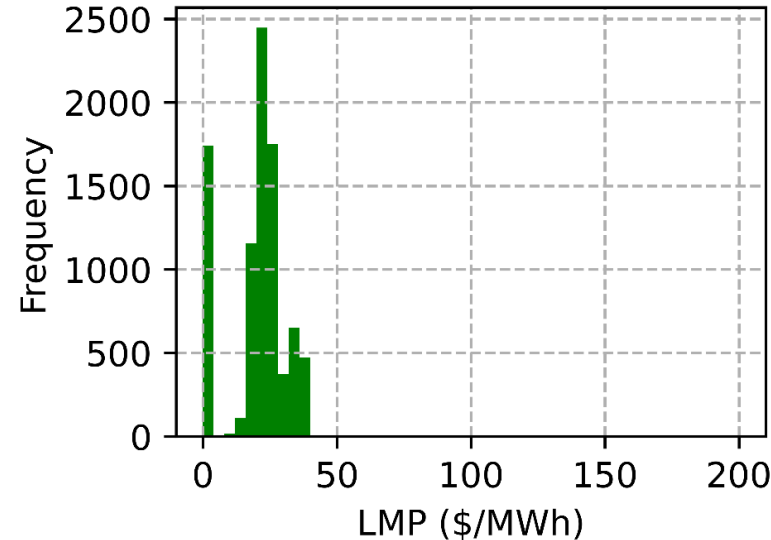
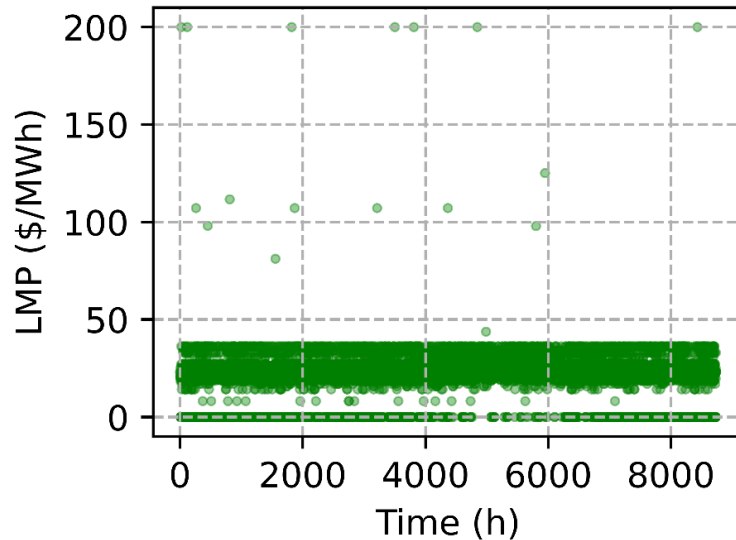
Power δ

Price π

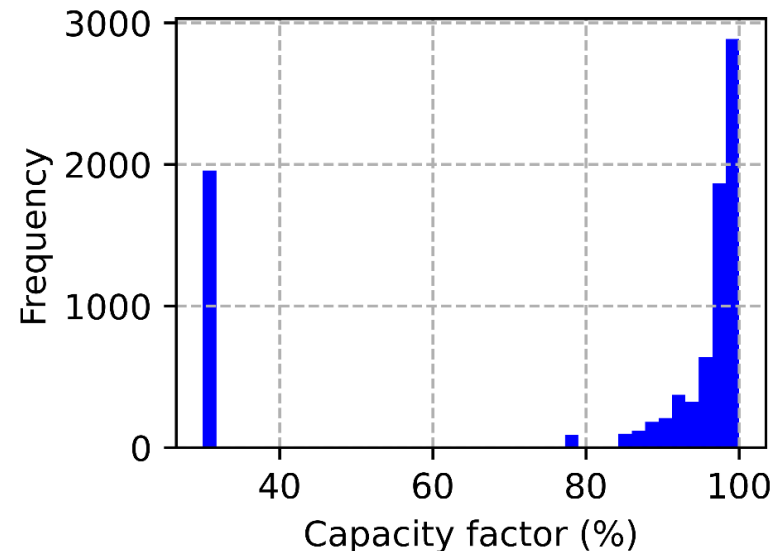
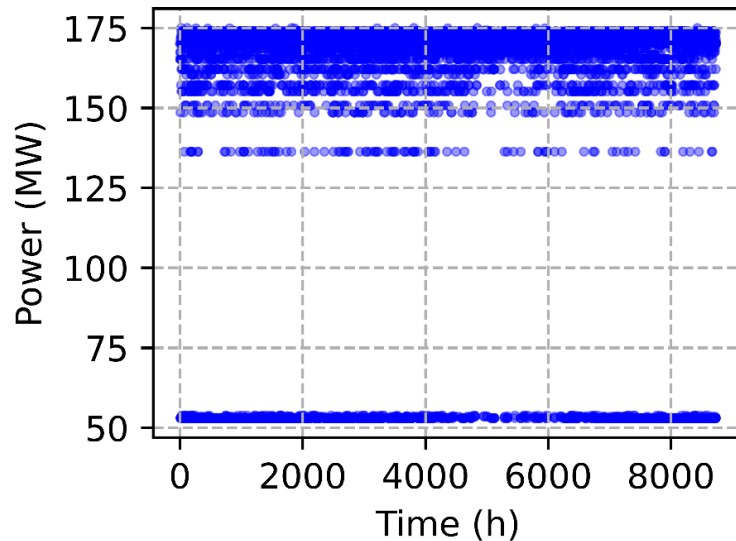
Energy Market



Steady-State Price Taker Results



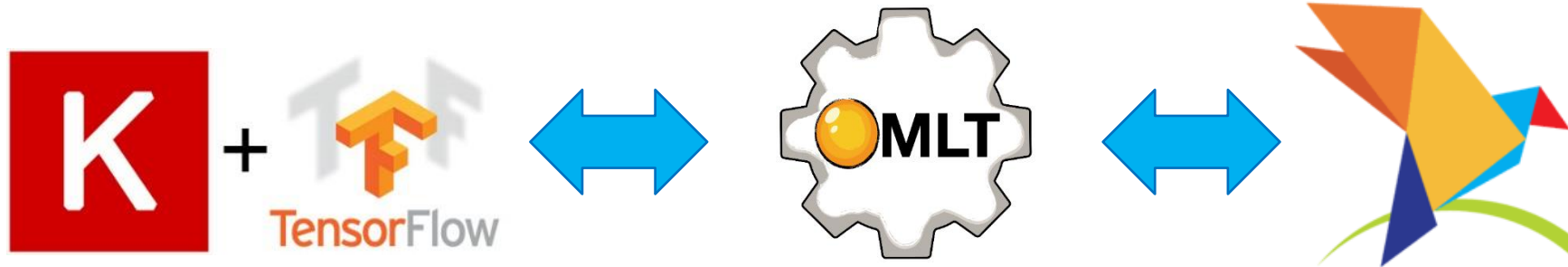
P_{\max}	175 MW
Marginal Cost	24 \$/MWh
Revenue	29.6 MM\$/yr
Opex	30.8 MM\$/yr
Capex	404 MM\$
20 year return	-428.8 MM\$



# Variables	419387
Solution Time	1846 sec (build) 49 sec (solve)

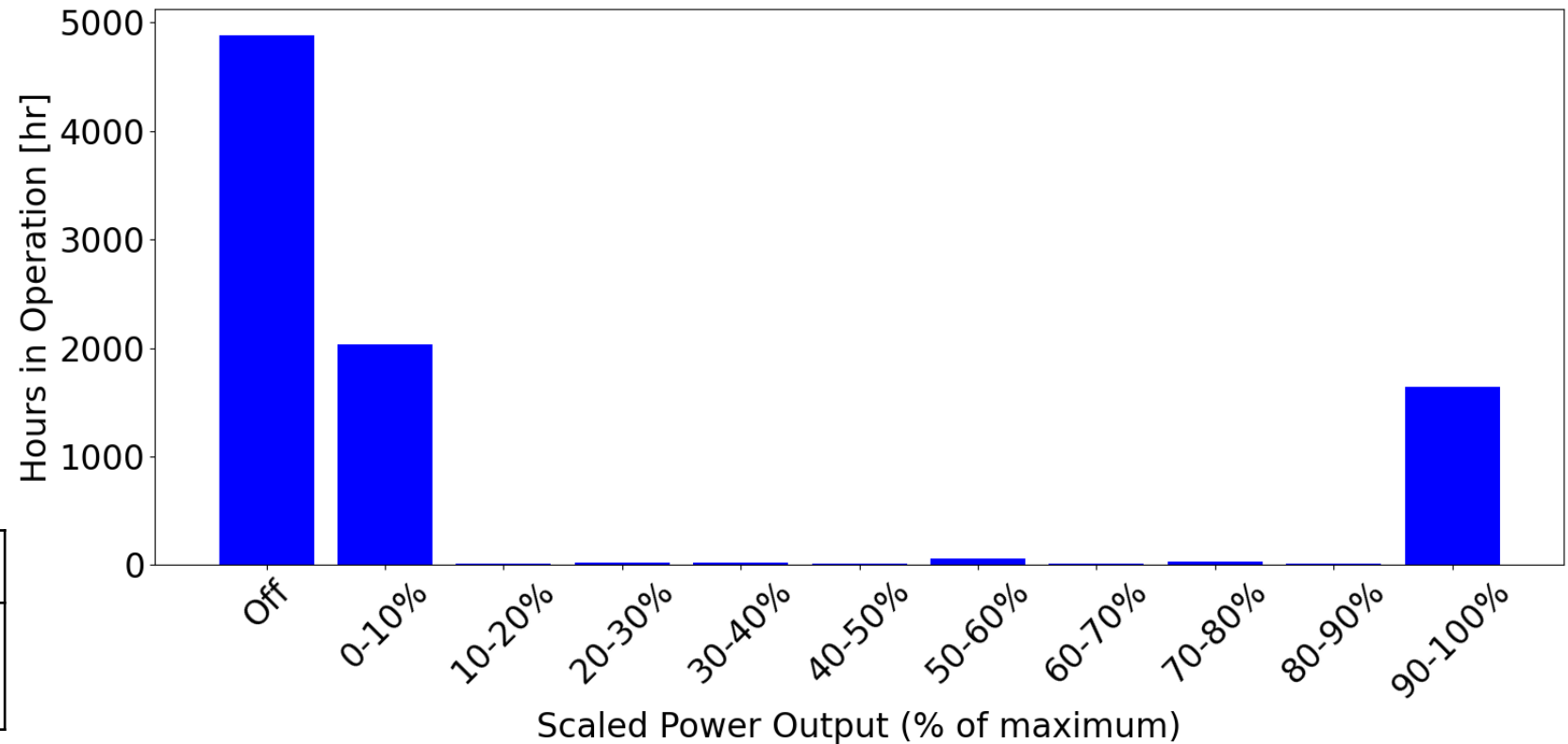
- Build small plant, operate at minimum power unless LMP exceeds operating cost
- Investment would require better prices (not surprising for RTS-GMLC system)

Surrogate Results



P_{\max}	177.5 MW
Marginal Cost	24 \$/MWh
Revenue	13.5 MM\$/yr
Opex	12 MM\$/yr
Capex	408 MM\$
20 year return	-382 MM\$

# Variables	738
Solution Time	13 sec (build) 336 sec (solve)

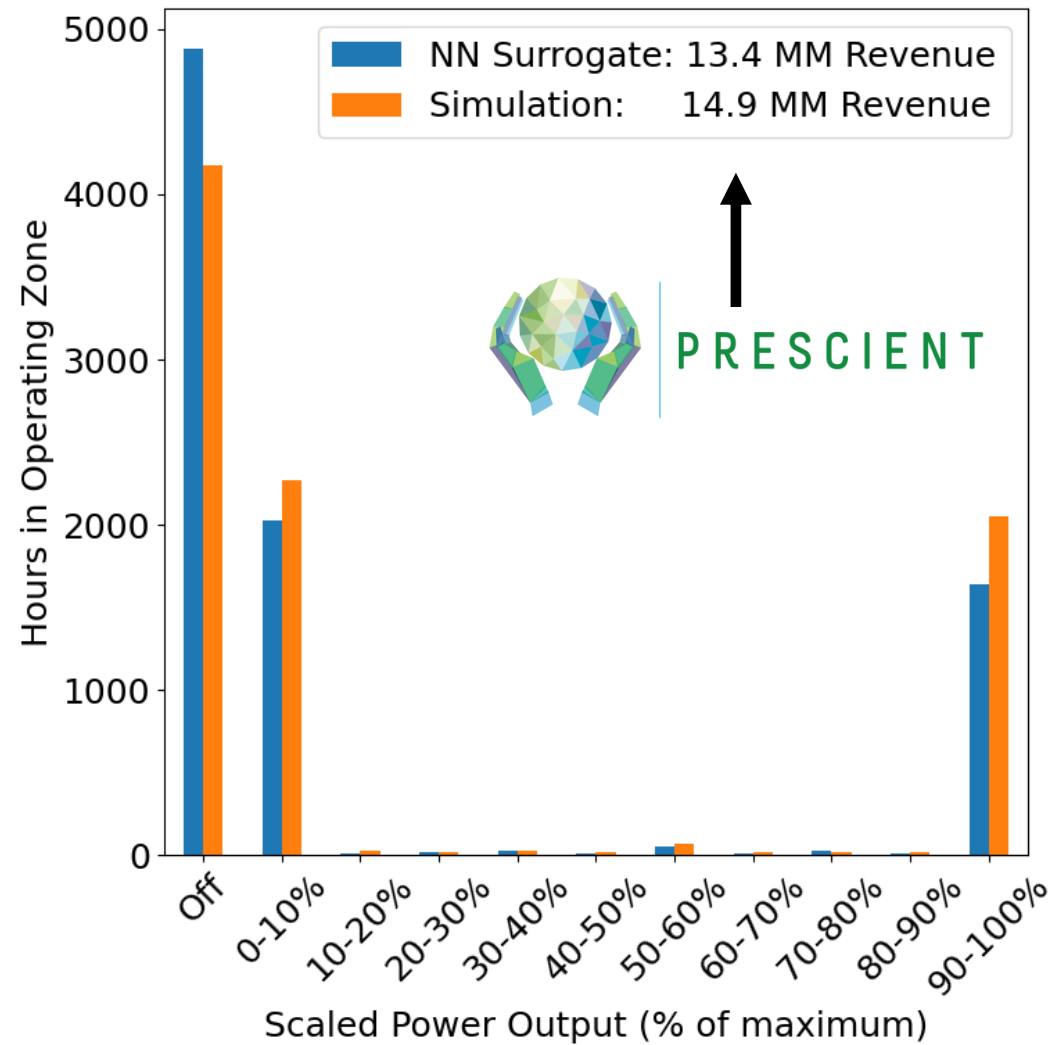


- Surrogate also builds small plant, but effectively captures shutdown operating modes
- Investment would still require better prices (but capturing shutdown helps)

Comparison of Results

Verification of Surrogate Solution

	Price Taker	Surrogate	Verification
P_{\max} [MW]	175	177.5	177.5
Marginal Cost [\$ /MWh]	24	24	24
Revenue [\$MM/yr]	29.6	13.5	14.9
Opex [\$MM/yr]	30.8	12	14.5 (Rankine) 16.2 (Prescient)
Capex [\$MM]	404	408	408
20 year return [\$MM]	-428.8	-382	-401



- Surrogate can capture startup/shutdown in steady state model, more realistic opex
- Surrogate solution is reasonably verified with Prescient simulation

Conclusions and Future Work

Integrated Energy System

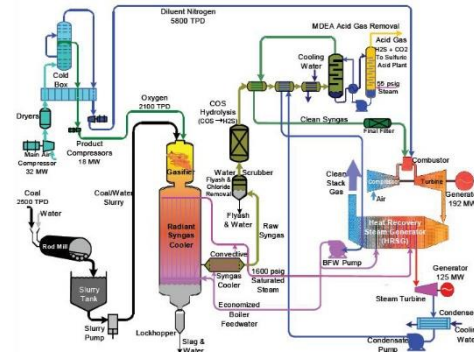
Design Decisions: d

Operating Decisions: $u_{t_1..t_N}$

Net Power Output: $\delta_{t_1..t_N}$

Multi-Period Model:

$$h(d, u_{s,t}, \delta_{s,t}, u_{s,t+1}, \delta_{s,t+1}) = 0$$



Dispatch Schedule $\delta_{t_1..t_N}$



Price Forecast $\pi_{t_1..t_N}$

Energy Market

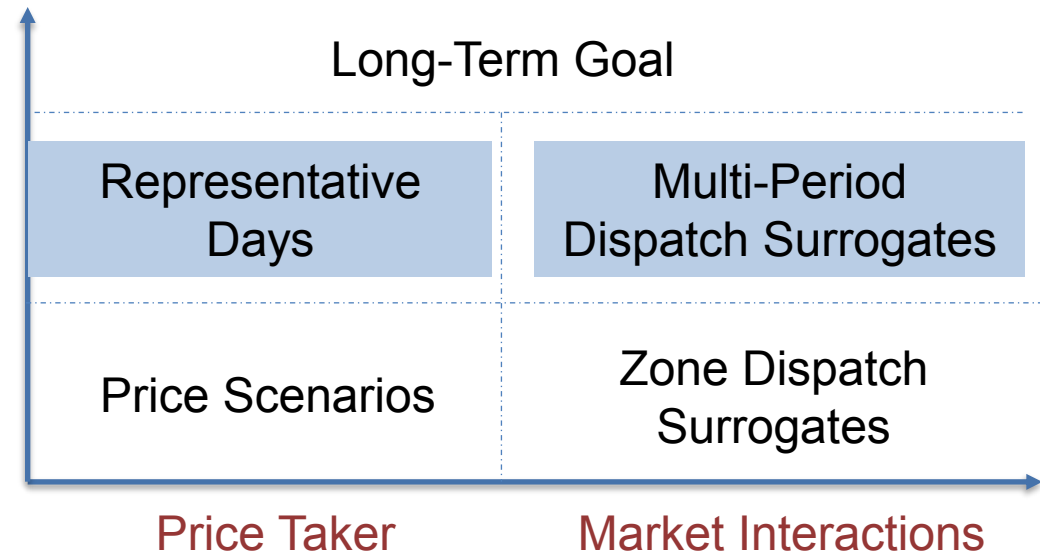


- Surrogates can incorporate exogenous market uncertainty into conceptual design problems
- Steady-State problems can capture startup/shutdown effects using surrogate methodology
- Future work is developing **multi-period** formulations for IES design

Fully Dynamic

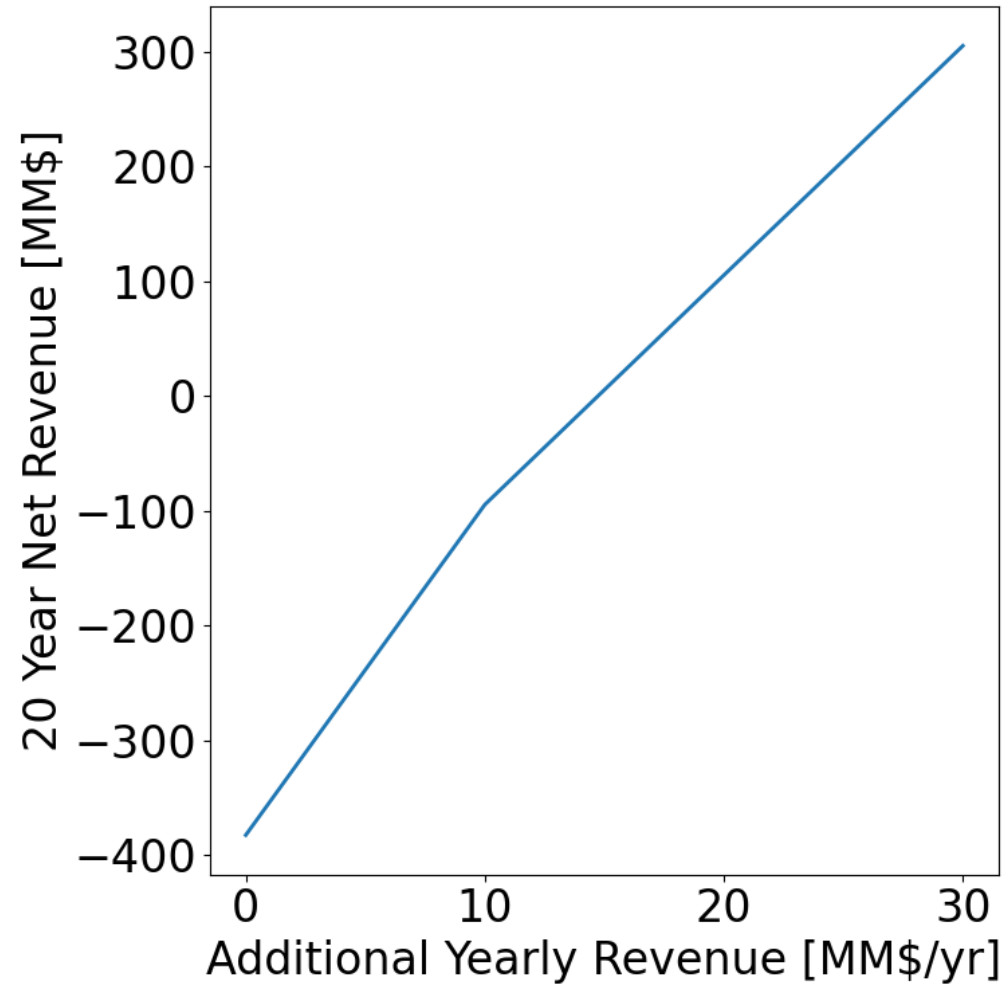
Multi-period

Steady-state

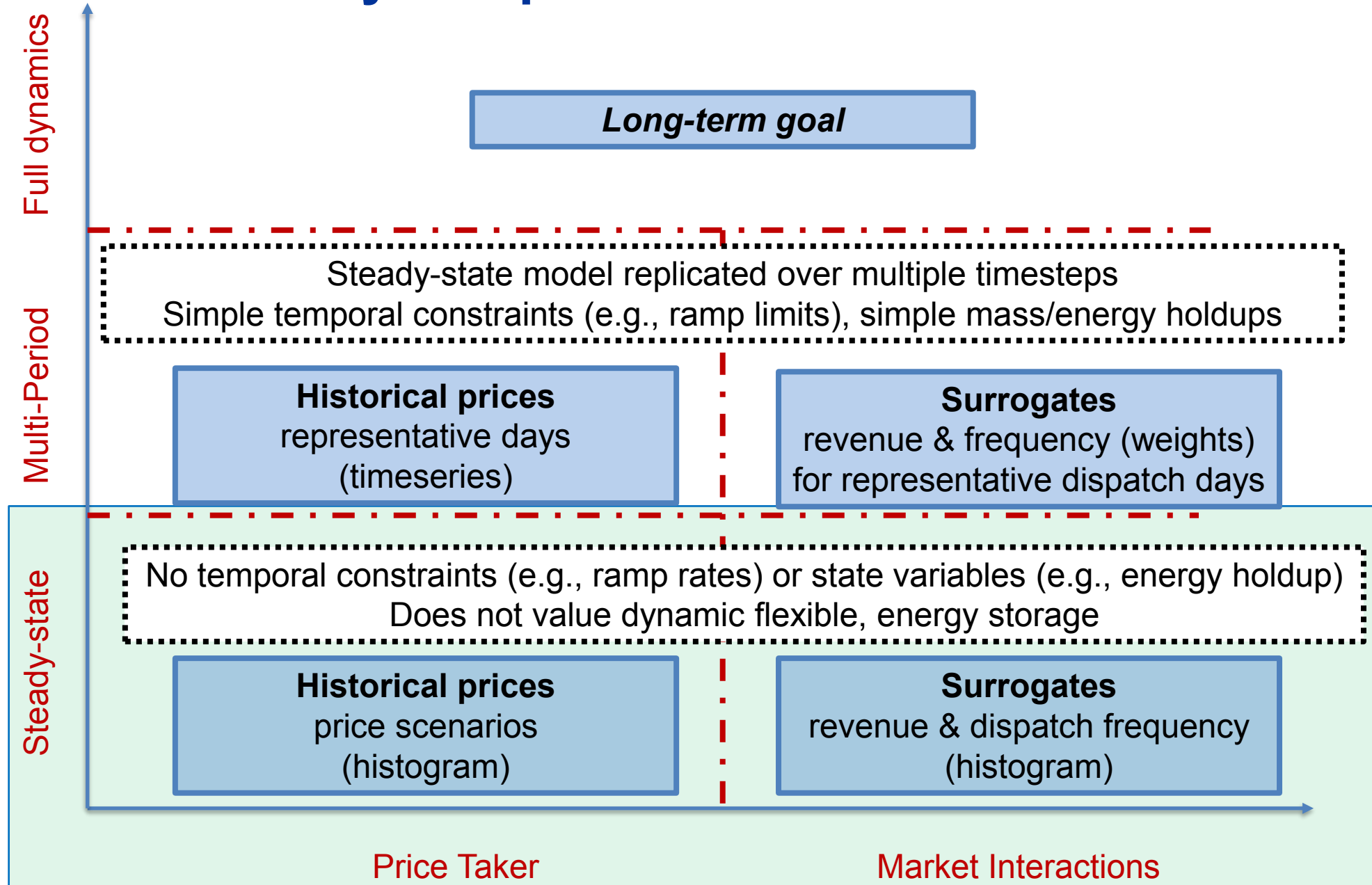


Additional Slides

Offset Results



Summary of Optimization Formulations



Multi-Period Price Taker (Self-Schedule)

Scenarios are timeseries, e.g., representative days

$$\max_{d,u,\delta} \sum_{s \in S} \sum_{t \in T} \overbrace{w_s}^{\text{Weight / Frequency}} \left[\overbrace{R(d, u_{s,t}, \delta_{s,t}, \pi_{s,t})}^{\text{Revenue}} - \overbrace{C(d, u_{s,t}, \delta_{s,t})}^{\text{Cost (Operating + Capital)}} \right]$$

Process
Model

$$g(d, u_{s,t}, \delta_{s,t}) = 0, \quad \forall s \in S, t \in T$$

Temporal
Constraints

$$h(d, u_{s,t}, \delta_{s,t}, u_{s,t+1}, \delta_{s,t+1}) = 0, \quad \forall s \in S, t \in T$$

Multi-Period with Market Surrogates (Bid)

$$\max_{d, u, x} \overbrace{R(x)}^{\text{Revenue}} - \sum_{s \in S} \sum_{t \in T} w_s(x) \overbrace{\left[C(d, u_{s,t}, \delta_{s,t}) \right]}^{\text{Cost (Operating + Capital)}}$$

Process Model

$$g(d, u_{s,t}, \delta_{s,t}) = 0, \quad \forall s \in S$$

“Bid Rules”

$$h(d, x) = 0$$

Temporal Constraints

$$h(d, u_{s,t}, \delta_{s,t}, u_{s,t+1}, \delta_{s,t+1}) = 0, \dots$$

Revenue Surrogate

$$R(x) = f_{rev}(x)$$

Multi-Period

$$w_s(x) = f_s(x), \quad \forall s \in S$$

Dispatch Surrogate

Note: δ are representative days for market dispatch and are parameters (scenario data)

