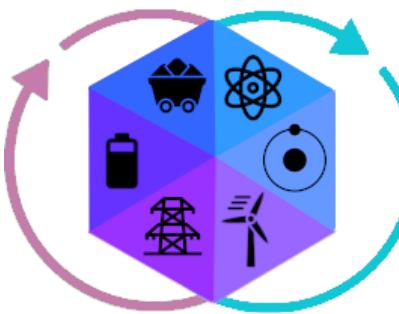


# IDAES

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 10<sup>th</sup>, 2021, Boston, MA

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Carnegie Mellon

West Virginia University

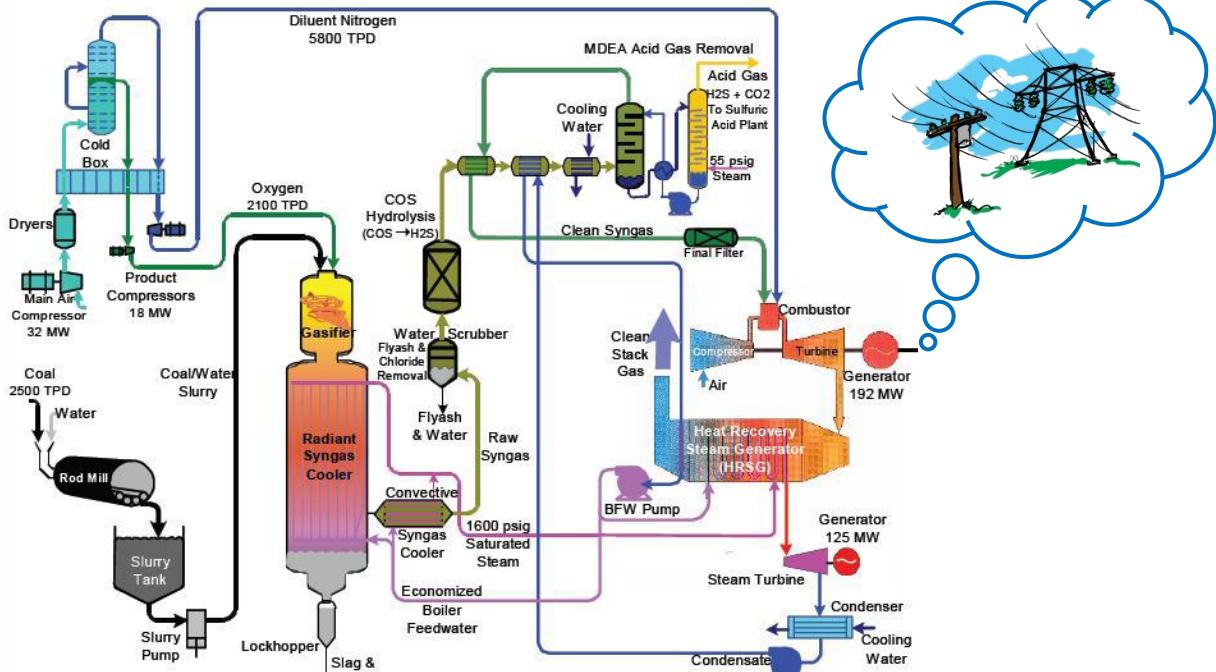


U.S. DEPARTMENT OF  
ENERGY

# Energy system analysis capabilities are applied in isolation

## Process-centric Modeling

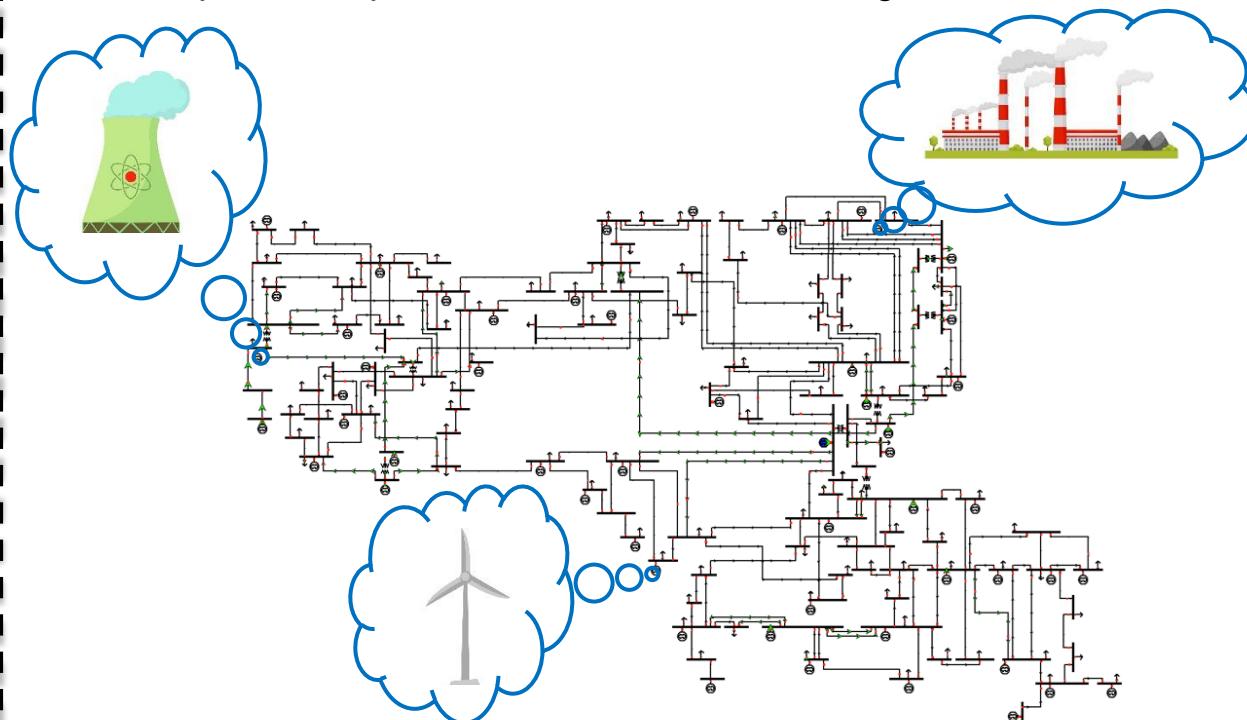
Detailed steady state or dynamic process models, with the grid modeled as an infinite capacity bus



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

## Grid-centric Modeling

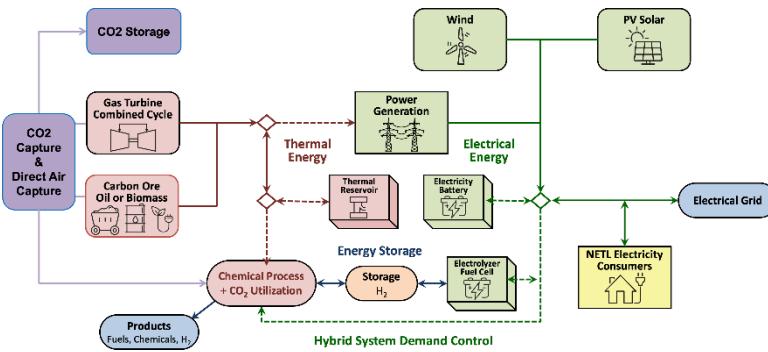
Detailed power flow models, with individual generators modeled as either dispatchable point sources or stochastic "negative loads"



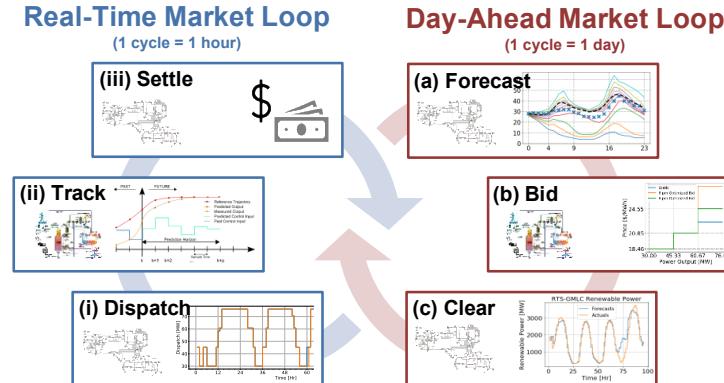
<https://icseg.iti.illinois.edu/files/2013/10/IEEE118.png>

# Challenge of increasingly integrated & dynamic grid/generation

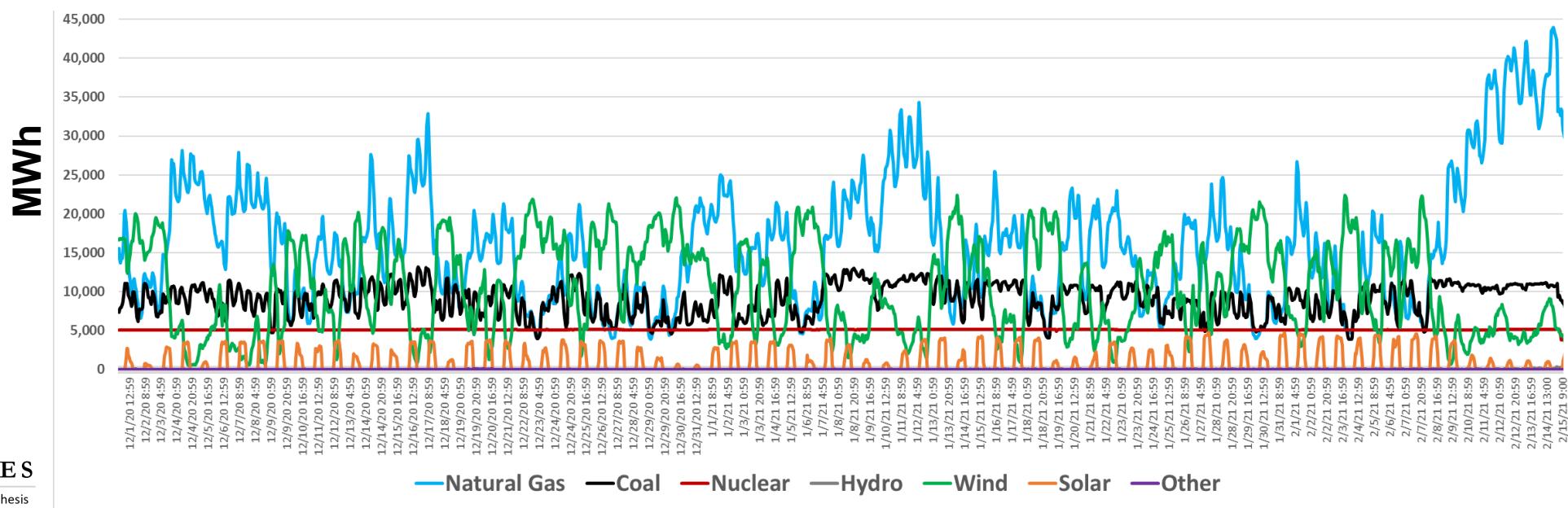
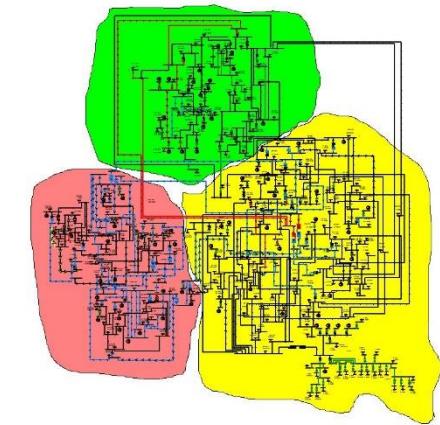
## Generation & Process Modeling



## Integrated Resource-Grid Model

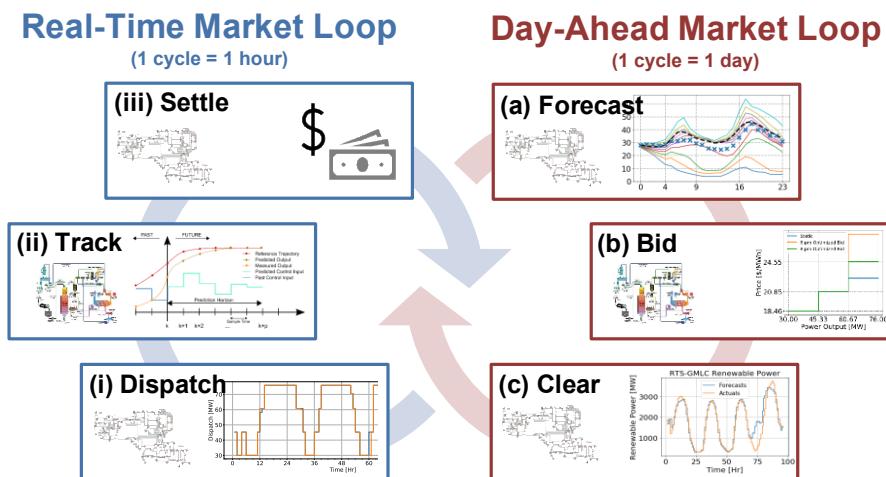


## Grid Modeling



# Multiscale Market-Based Optimization of IES

## Multiscale Simulation to Quantify Grid/IES Interactions

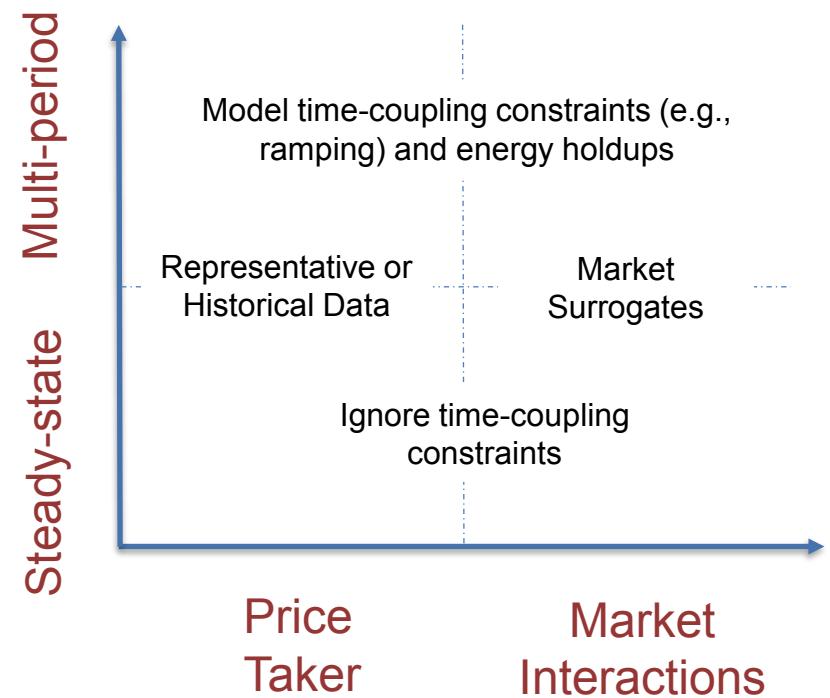


*Data to refine market representation (e.g., prices, surrogates)*

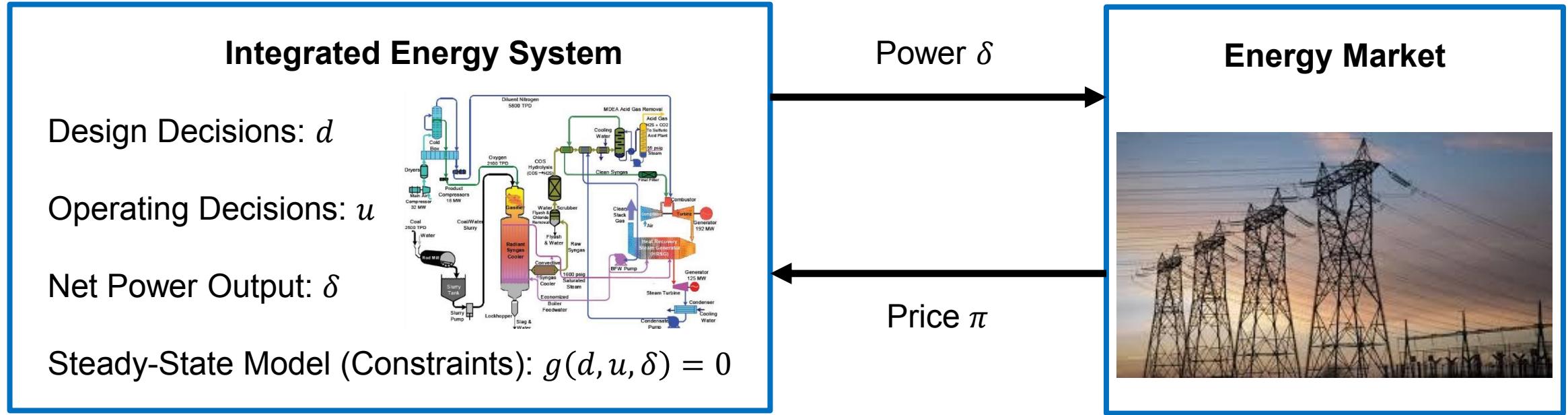
*Candidate IES designs (i.e., IDAES models) to evaluate*

## Superstructure Optimization

Maximize Net Present Value



# Abstract Integrated Energy System



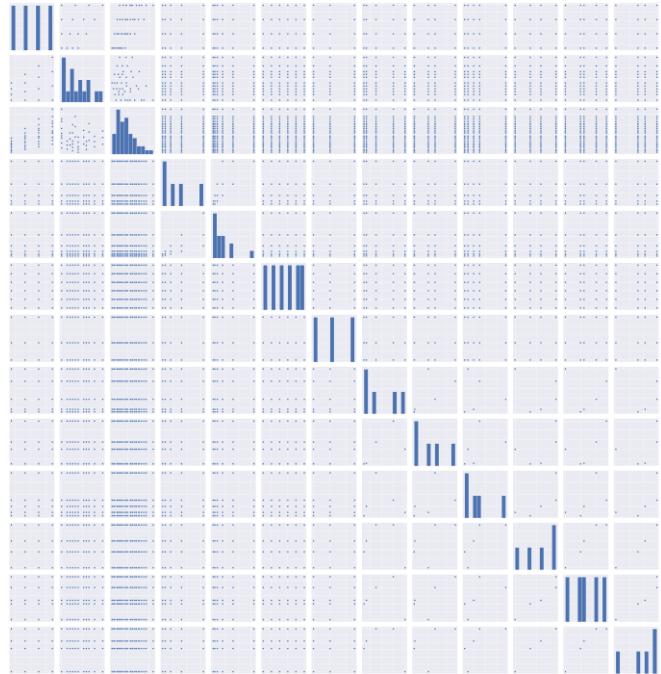
**This abstraction is easy to extend to...**

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

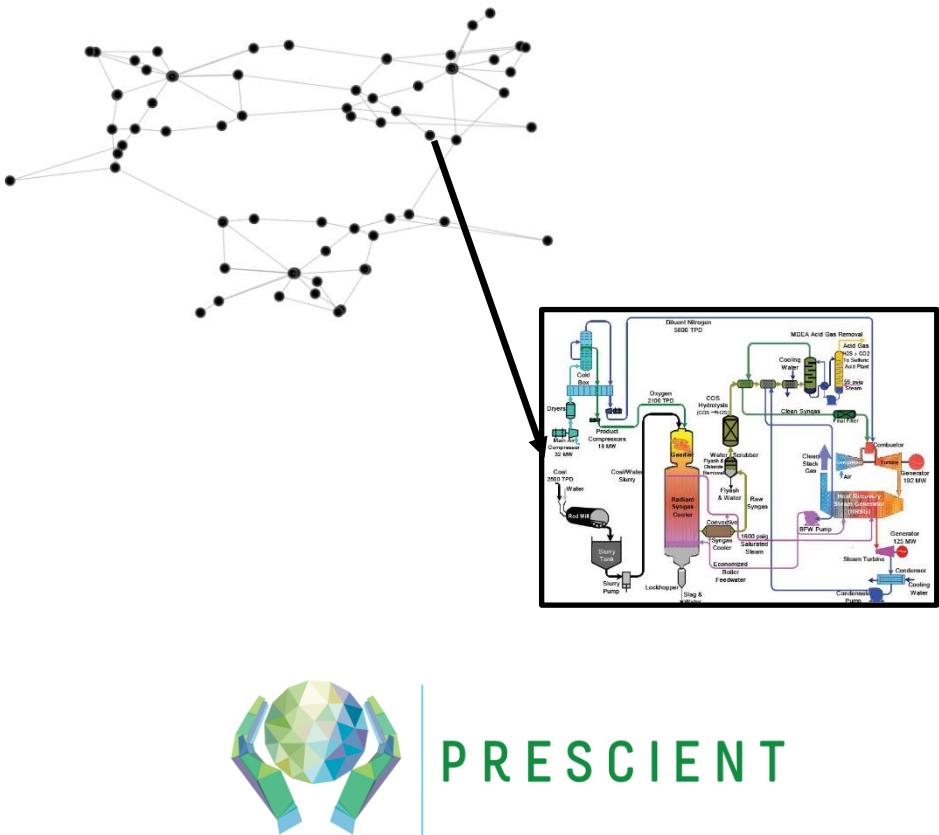
# How would a new generator change market outcomes?

## Simulation Design

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

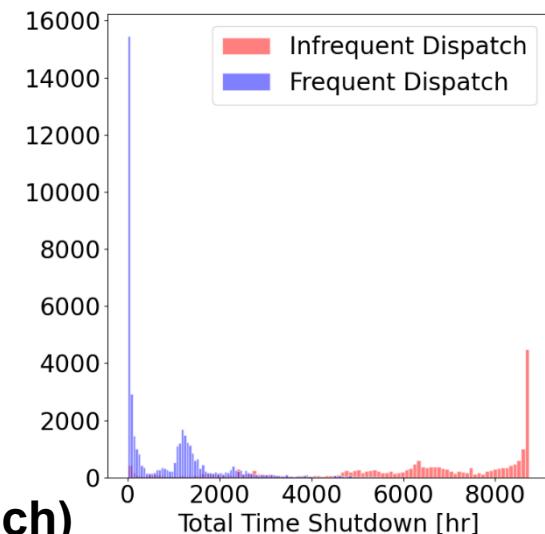
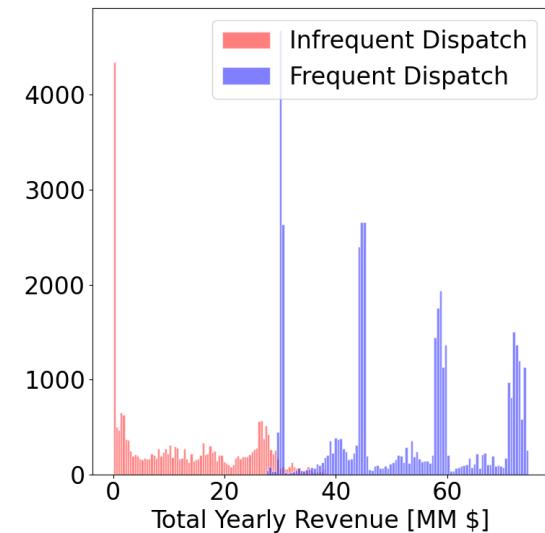


# Production Cost Modeling: RTS-GMLC Test System



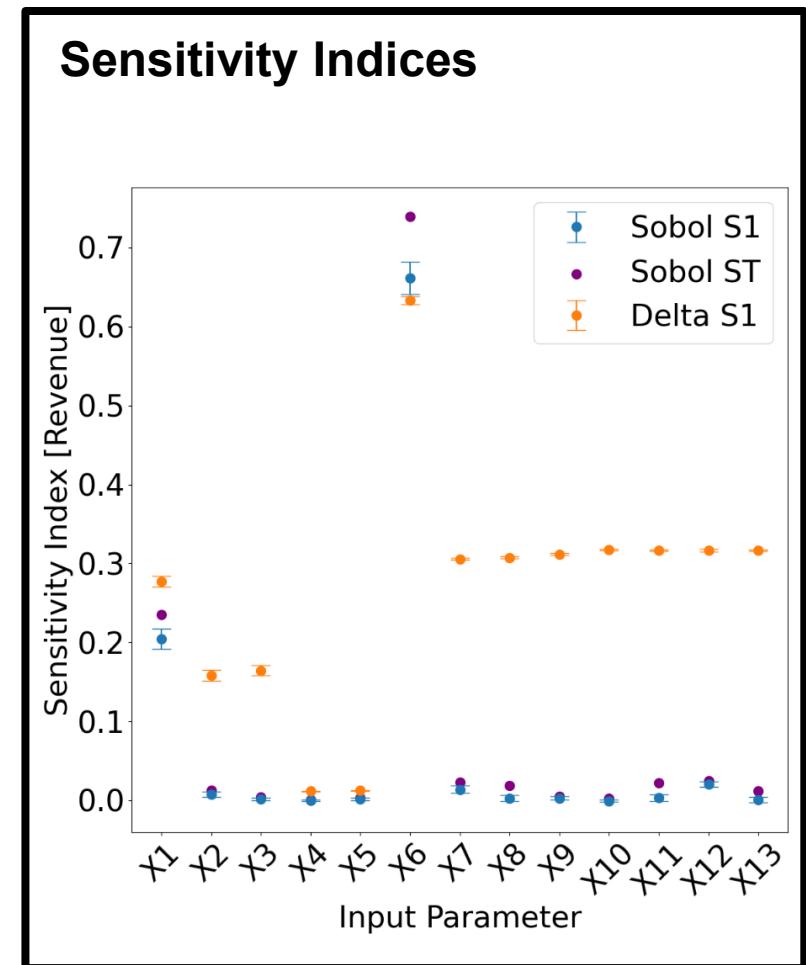
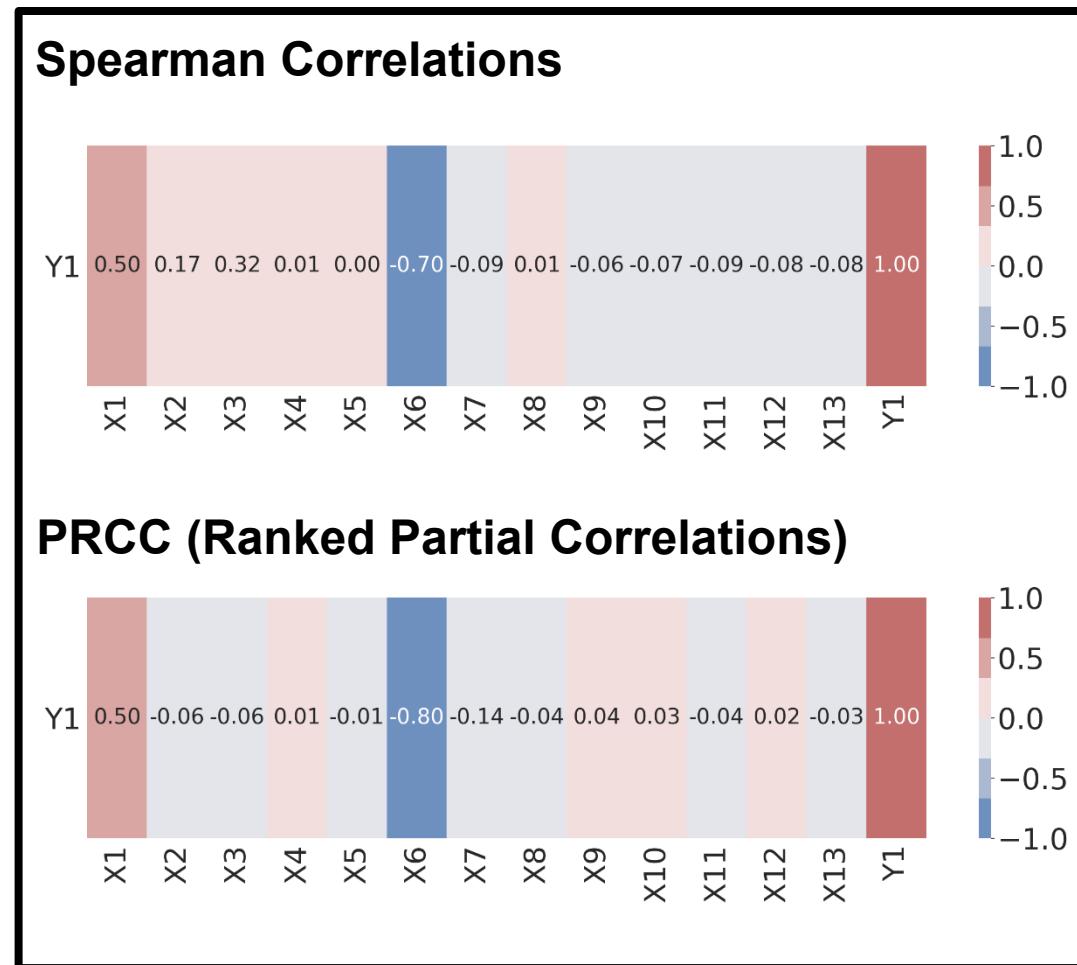
**64,800 1-year Prescient simulations (@ 3 hr each)**

## Revenue & Dispatch Results



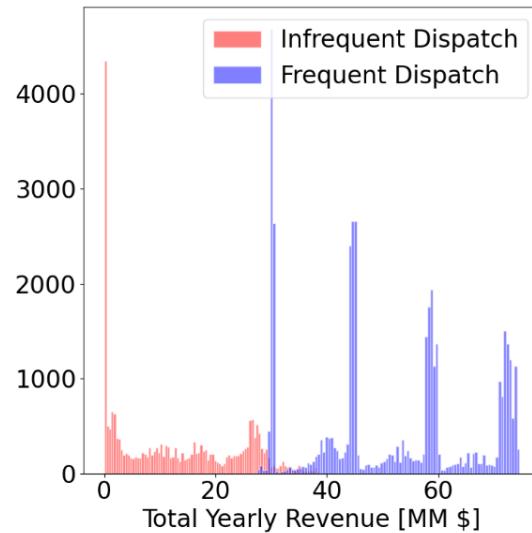
# Global Sensitivity Analysis

	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\\$]



- 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



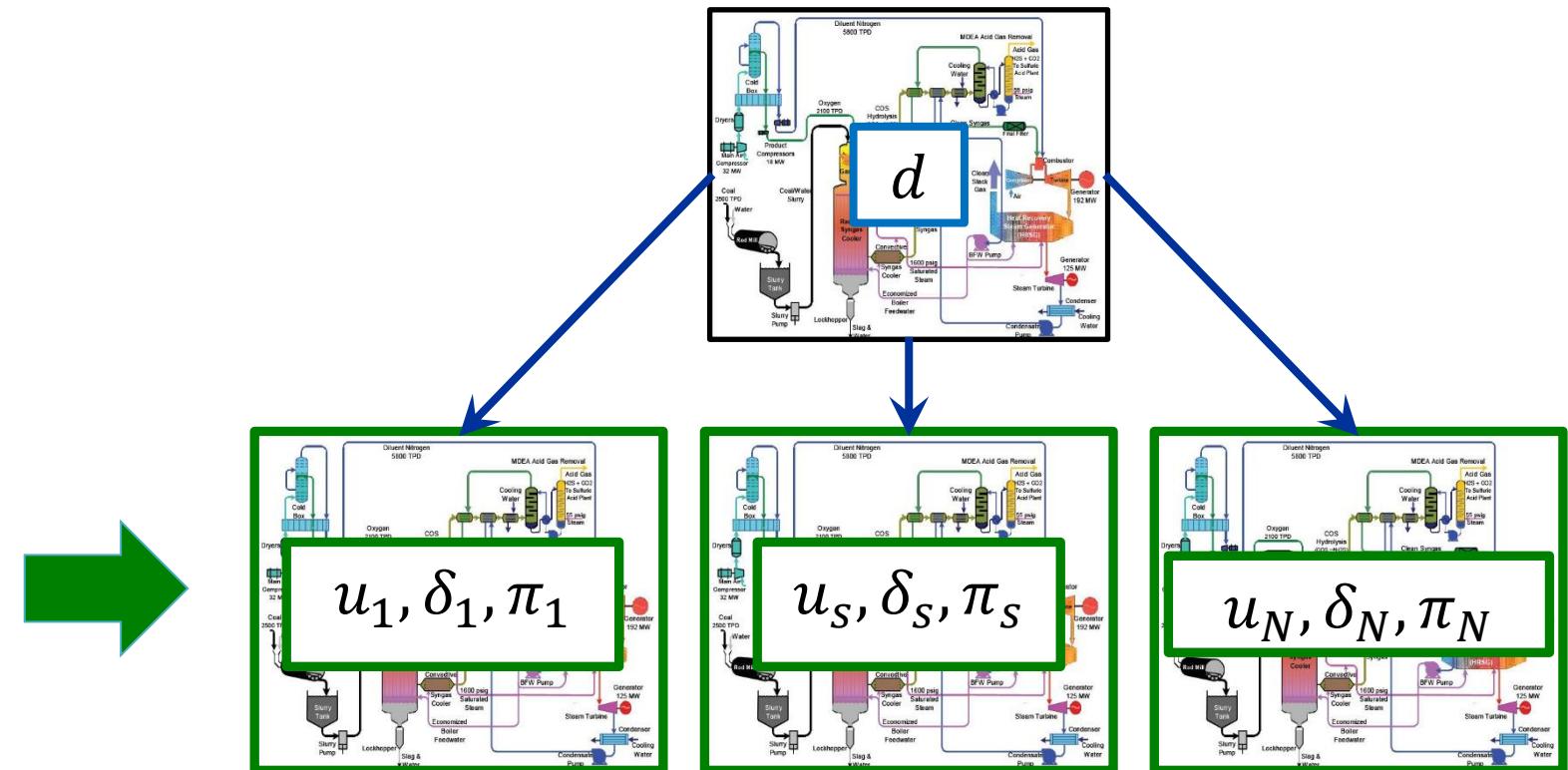
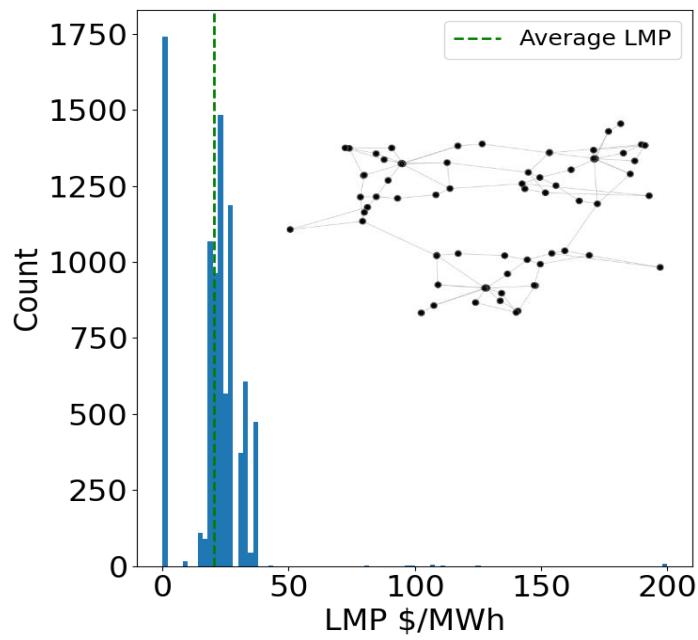
# Steady-State Price Taker (Self-Schedule)

Weight / Frequency      Revenue      Cost (Operating + Capital)

$$\begin{aligned}
 & \max_{d, u, \delta} \sum_{s \in S} w_s [R(d, u_s, \delta_s, \pi_s) - C(d, u_s, \delta_s)] \\
 & g(d, u_s, \delta_s) = 0, \quad \forall s \in S
 \end{aligned}$$

$d$	Design decisions
$\delta_s$	Power output <b>decision</b> for scenario
$u_s$	Operating <b>decisions</b> for scenario
$\pi_s$	Scenario price ( <b>data</b> )
$R()$	Revenue: function of decisions and prices
$C()$	Cost: function of decisions

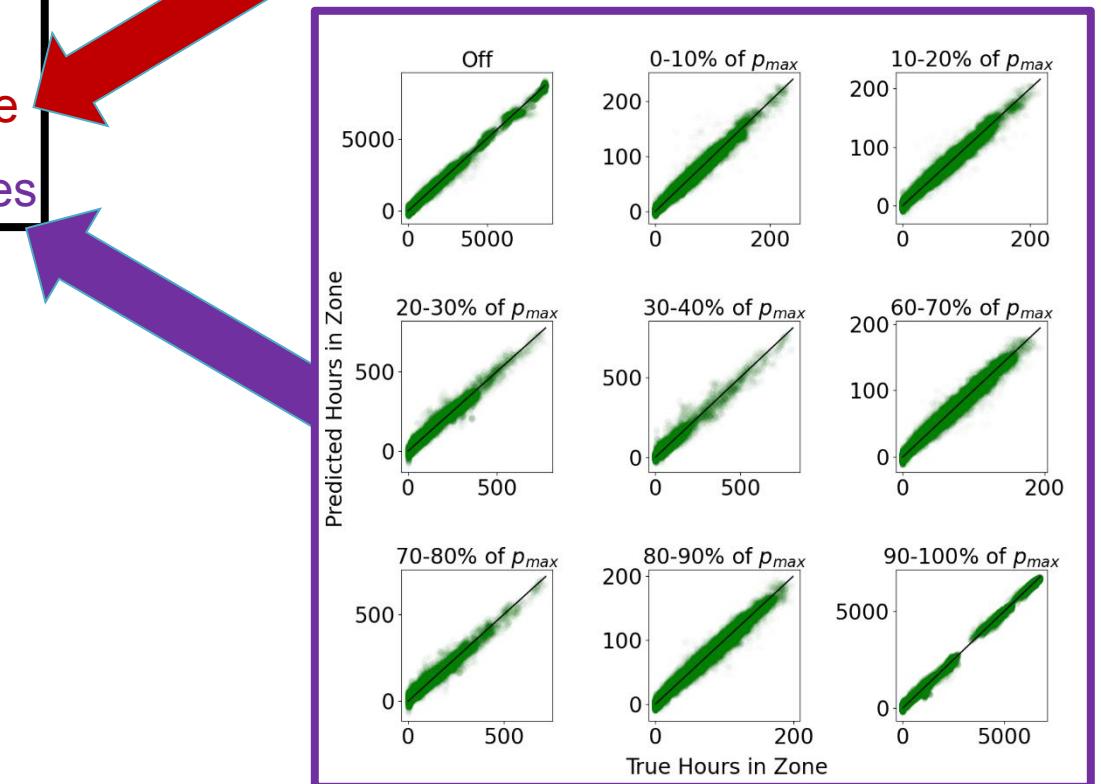
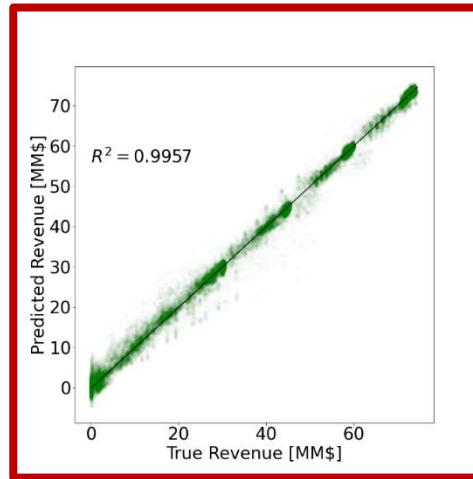
## Nominal RTS-GMLC Prices



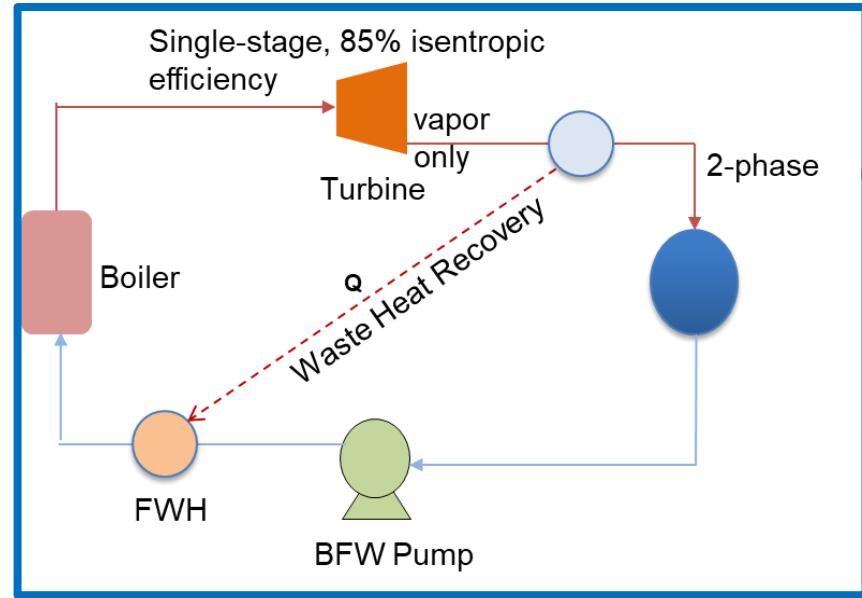
# Steady-State with Market Surrogates (Bid)

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

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



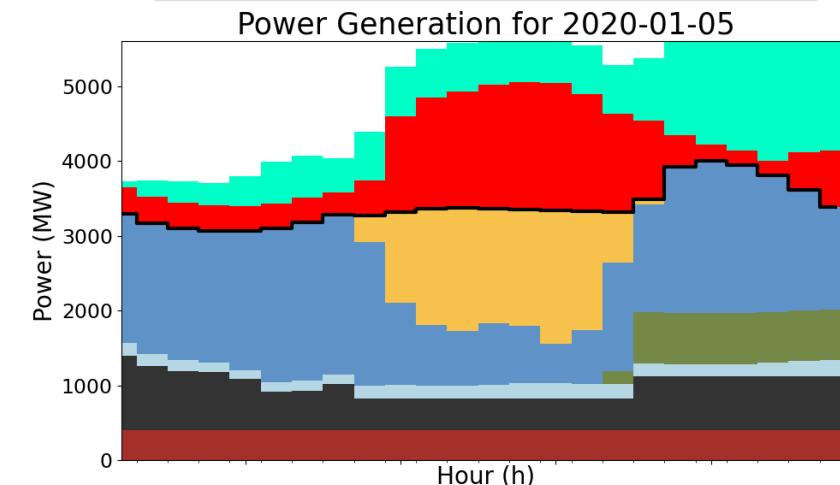
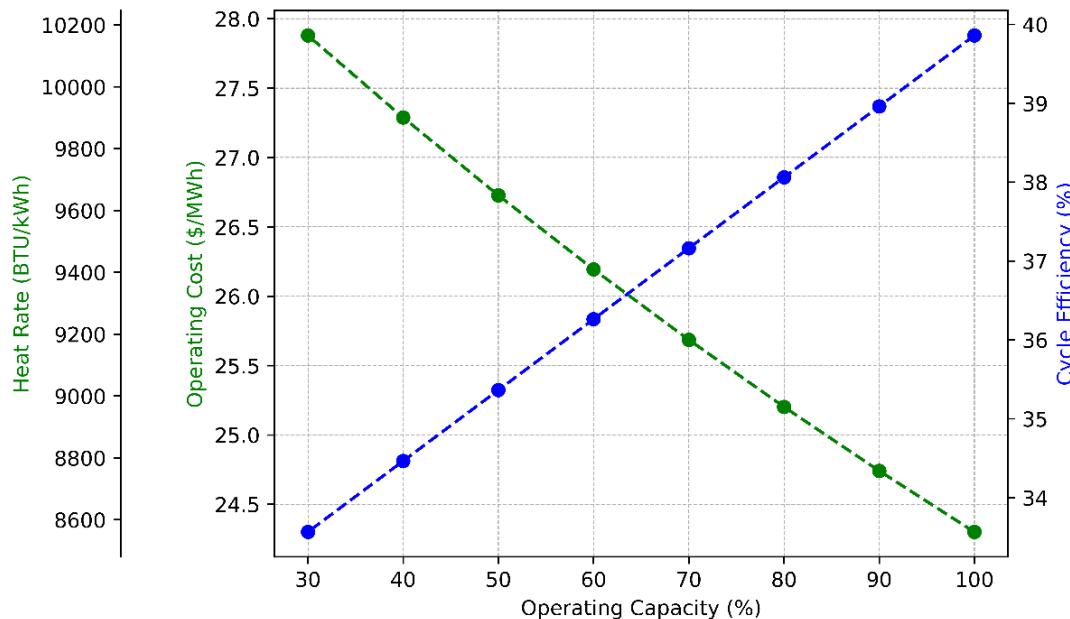
# Example: Optimal Design of Rankine Cycle for RTS-GMLC



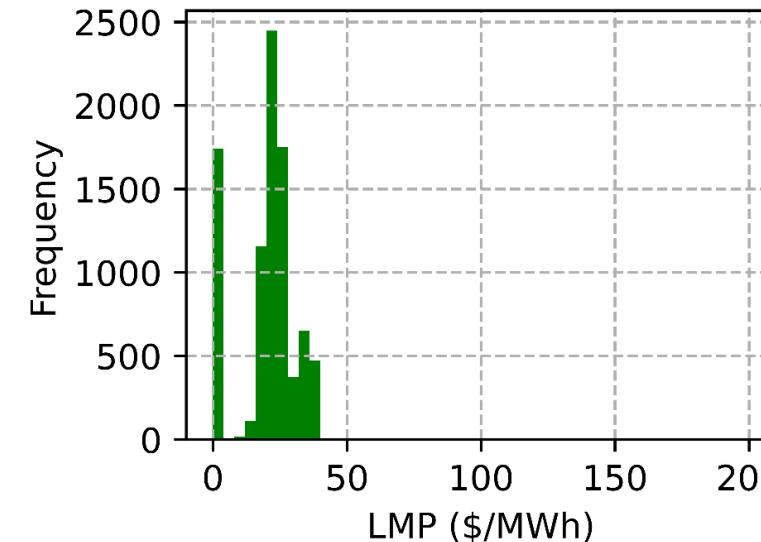
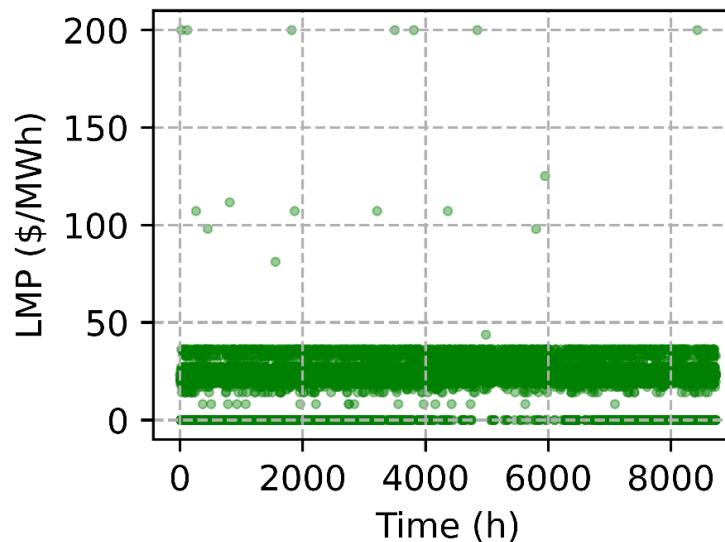
Power  $\delta$

Price  $\pi$

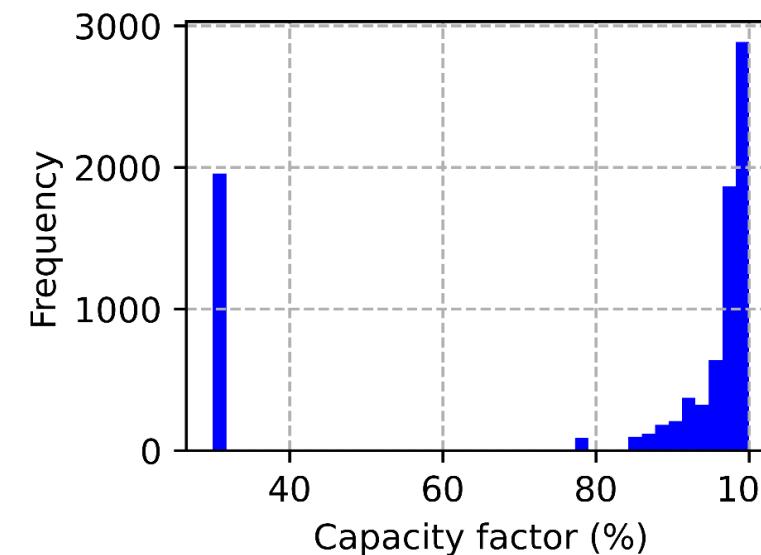
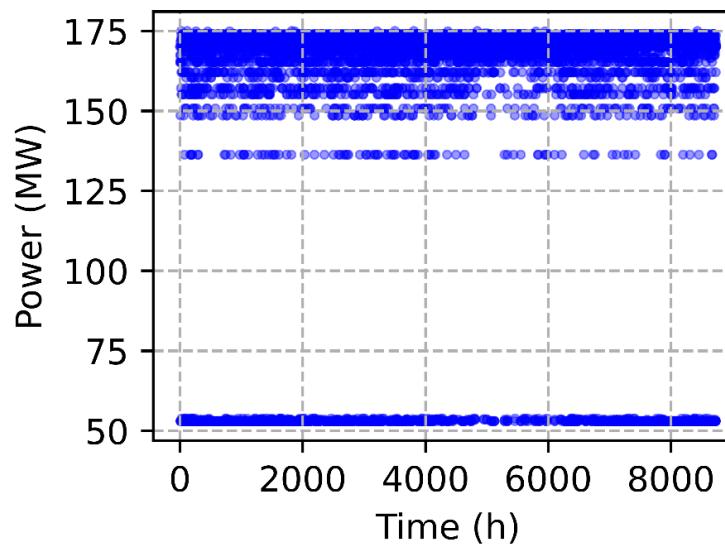
## Energy Market



# Steady-State Price Taker Results



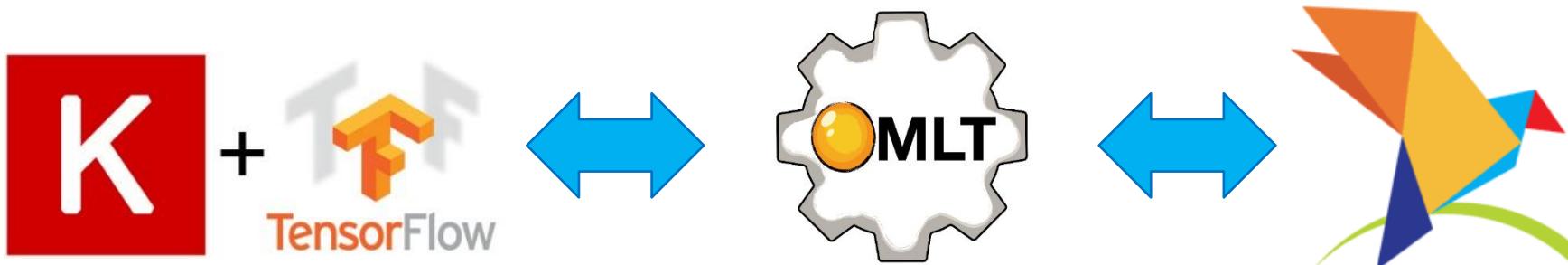
$P_{\max}$	175 MW
Marginal Cost	24 \$/MWh
<b>Revenue</b>	<b>29.6 MM\$/yr</b>
Opex	30.8 MM\$/yr
Capex	404 MM\$
20 year return	-428.8 MM\$



<b># Variables</b>	419387
Solution Time	1846 sec (build) 49 sec (solve)

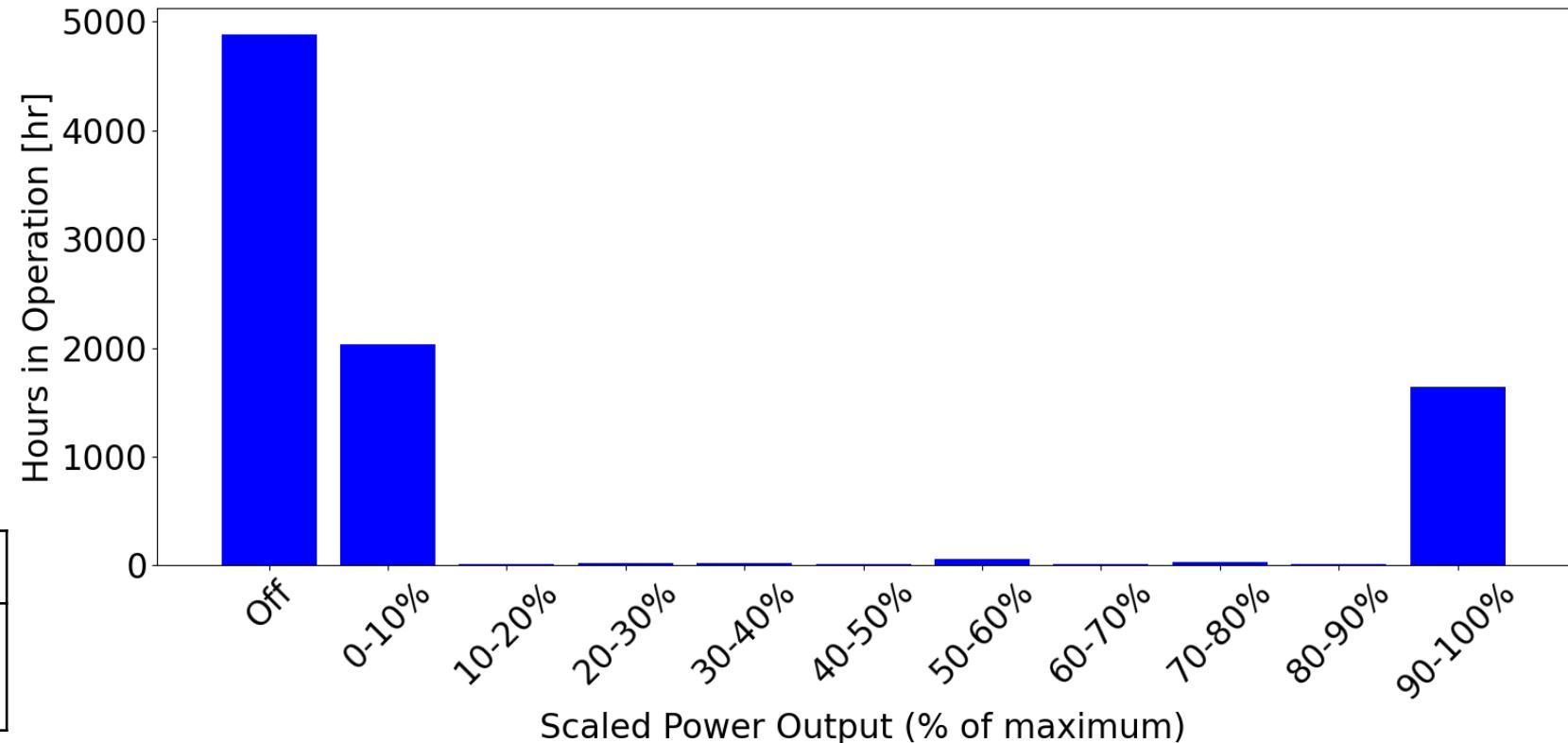
- Build small plant, operate at minimum power unless LMP exceeds operating cost
- Investment would require better prices (not surprising for RTS-GMLC system)

# Surrogate Results



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

<b># Variables</b>	738
Solution Time	13 sec (build) 336 sec (solve)

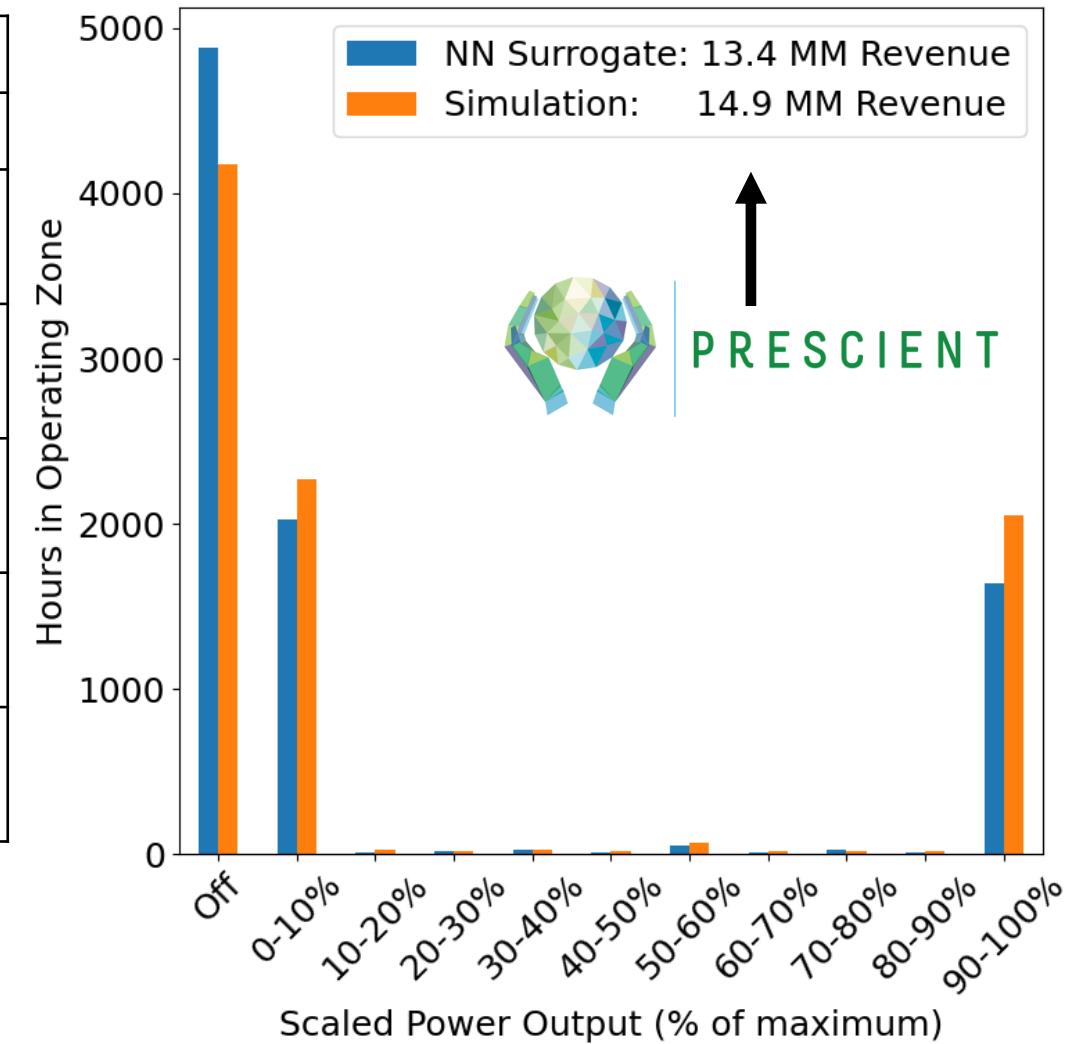


- Surrogate also builds small plant, but effectively captures shutdown operating modes
- Investment would still require better prices (but capturing shutdown helps)

# Comparison of Results

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

## Verification of Surrogate Solution



- 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

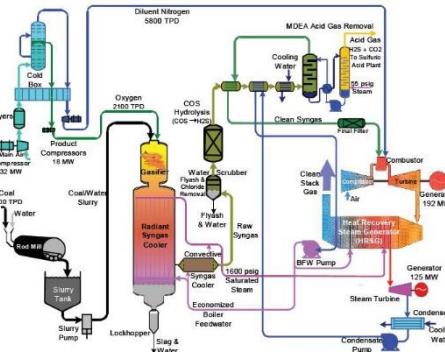
Design Decisions:  $d$

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

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

**Multi-Period Model:**

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



- 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

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

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

## Energy Market



Long-Term Goal

Representative Days

Price Scenarios

Price Taker

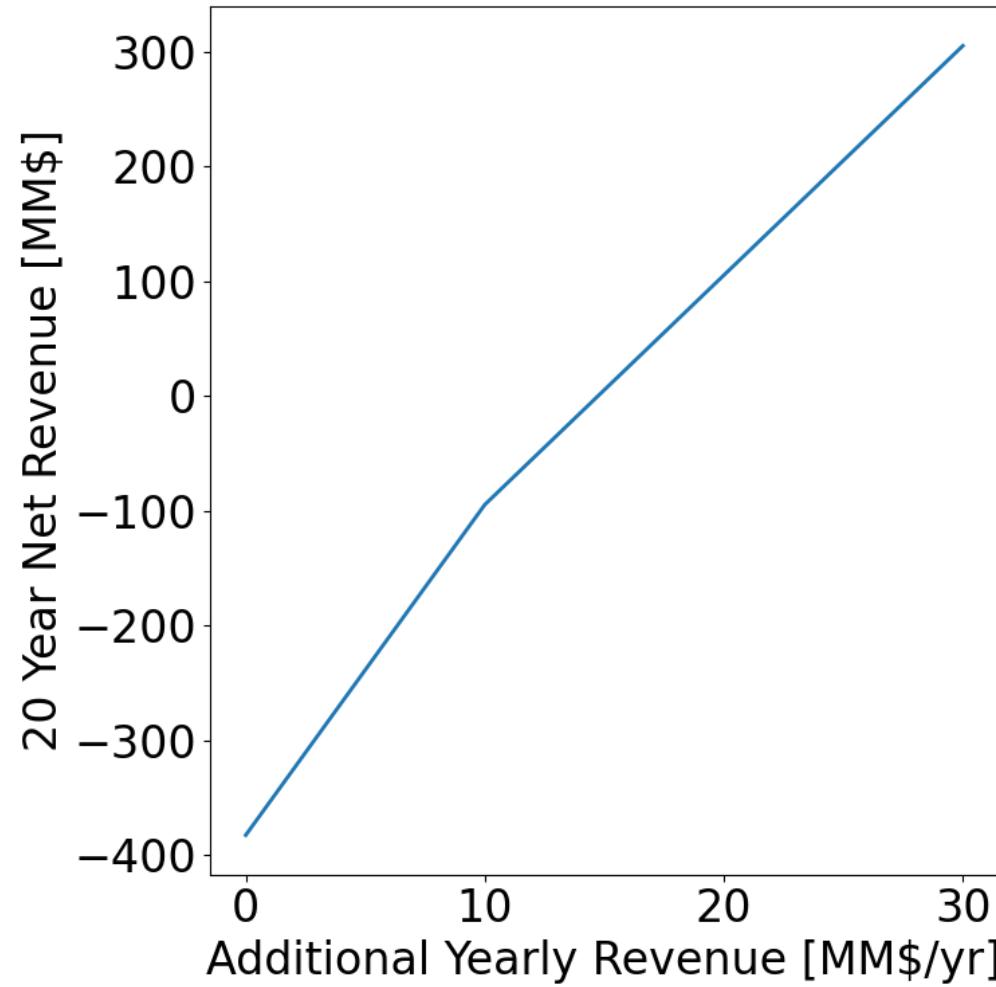
Multi-Period Dispatch Surrogates

Zone Dispatch Surrogates

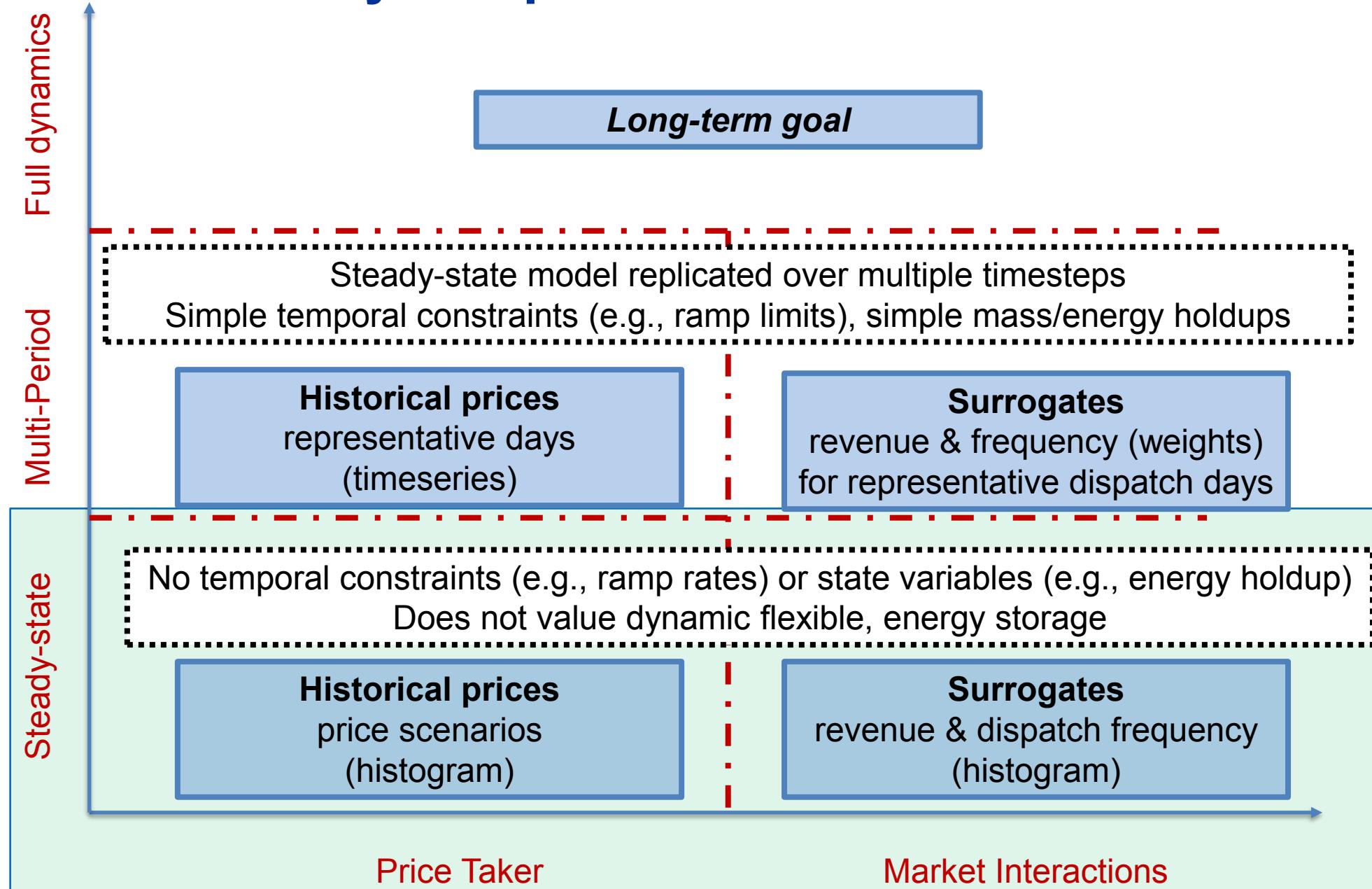
Market Interactions

# Additional Slides

# Offset Results



# Summary of Optimization Formulations



# Multi-Period Price Taker (Self-Schedule)

Scenarios are timeseries, e.g., representative days

	Weight / Frequency	Revenue	Cost (Operating + Capital)
Process Model	$\max_{d,u,\delta} \sum_{s \in S} \sum_{t \in T} w_s [R(d, u_{s,t}, \delta_{s,t}, \pi_{s,t}) - C(d, u_{s,t}, \delta_{s,t})]$		
Temporal Constraints	$g(d, u_{s,t}, \delta_{s,t}) = 0, \quad \forall s \in S, t \in T$		
	$h(d, u_{s,t}, \delta_{s,t}, u_{s,t+1}, \delta_{s,t+1}) = 0, \quad \forall s \in S, t \in T$		

# Multi-Period with Market Surrogates (Bid)

	Revenue	Cost (Operating + Capital)
	$\max_{d,u,x} R(x) - \sum_{s \in S} \sum_{t \in T} w_s(x) [C(d, u_{s,t}, \delta_{s,t})]$	
Process Model		
“Bid Rules”	$g(d, u_{s,t}, \delta_{s,t}) = 0, \quad \forall s \in S$	
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 Dispatch Surrogate	$w_s(x) = f_s(x), \quad \forall s \in S$	

**Note:**  $\delta$  are representative days for market dispatch and are parameters (scenario data)