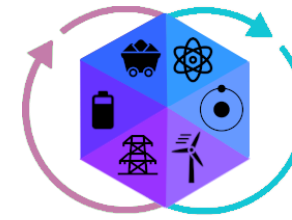


IDAES[®]
Institute for the Design of
Advanced Energy Systems



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



Beyond Price Taker: Optimizing Integrated Energy Systems Considering Market/Grid Interactions

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University of Notre Dame

With contributions from: Xinhe Chen (UND), Daniel Laký (UND), Radhakrishna Gooty (NETL), Tony Burgard (NETL), John Sirola (SNL), J. Kyle Skolfield (SNL), Darice Guittet (NREL), Bernard Knueven (NREL)



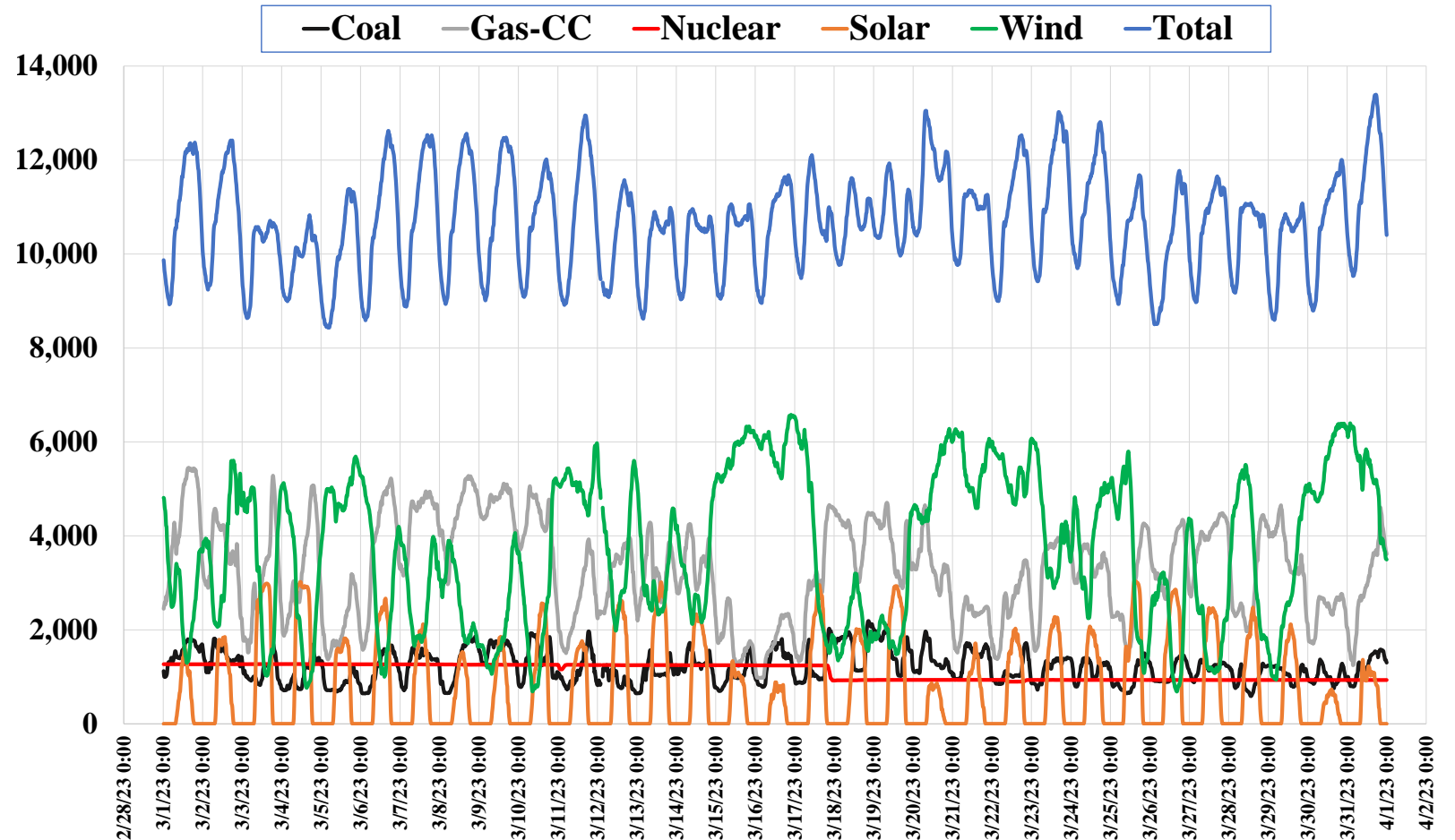
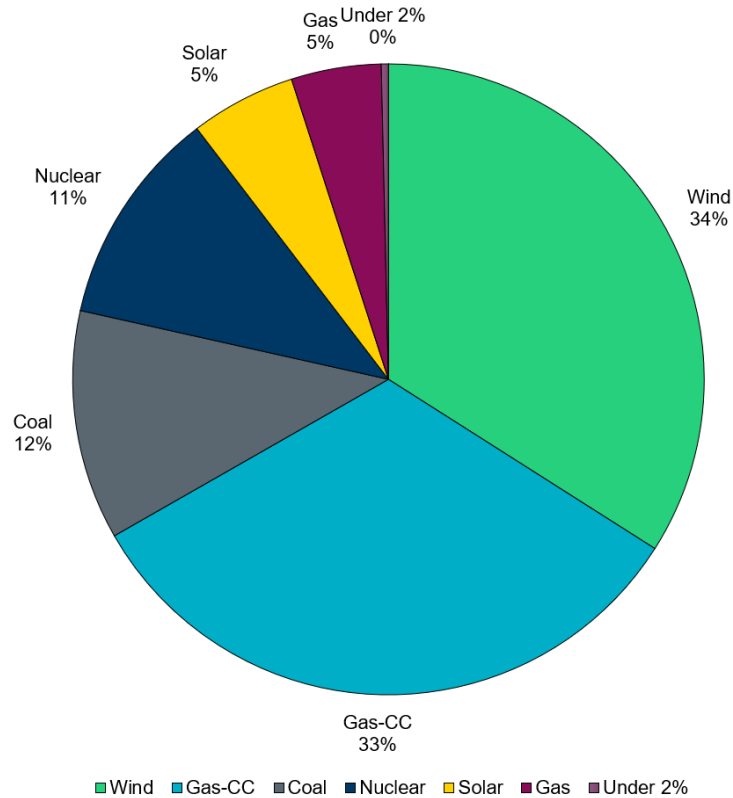
U.S. DEPARTMENT
of **ENERGY**

Evolving Grid Increasingly Requires Flexibility

Data for Electric Reliability Council of Texas (ERCOT) ISO

ERCOT Generation Mix - March 2023

Energy by Fuel for 2023



Source: <https://www.ercot.com/gridinfo/generation>

Multiple inputs and technologies:



Electricity
Energy Storage
Ancillary Services
Heating/Cooling
Chemicals

- Provide operation flexibility
- Facilitate integration of multiple energy sources
- Reduce grid operation costs
- Increases grid reliability and resiliency

How to **co-optimize** IES design and operation considering **dynamic market interactions**?

Presentation Outline

How to **co-optimize** IES design and operation considering **dynamic markets**

Price Taker

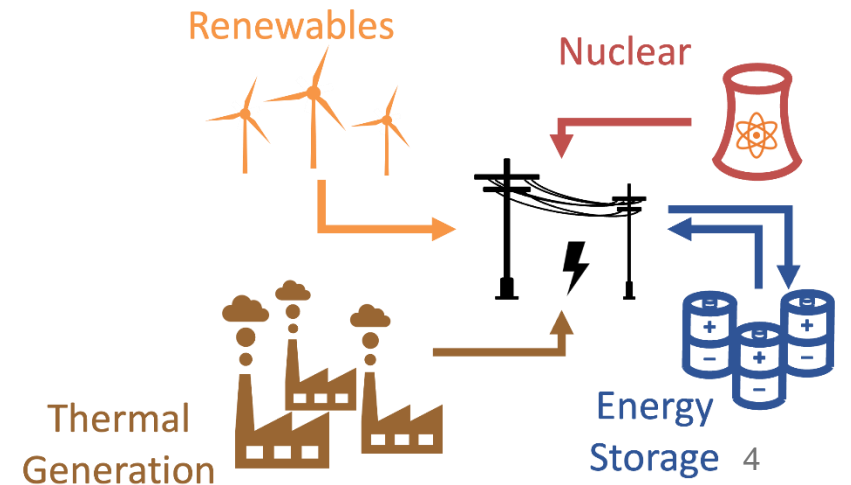
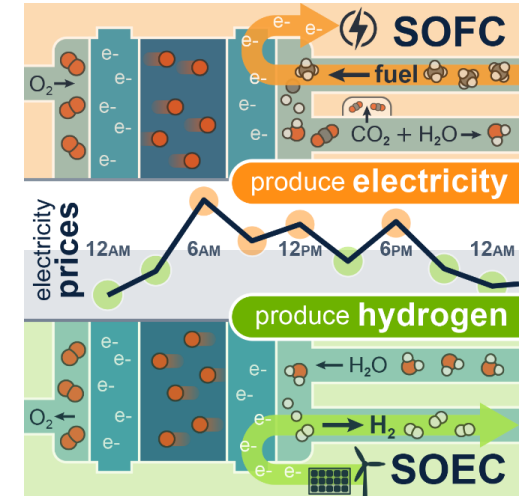
Solid oxide fuel cell IESs that co-produce H_2 and electricity

Beyond Price Taker

Nuclear and electrolyzer IESs that co-produce H_2 and electricity

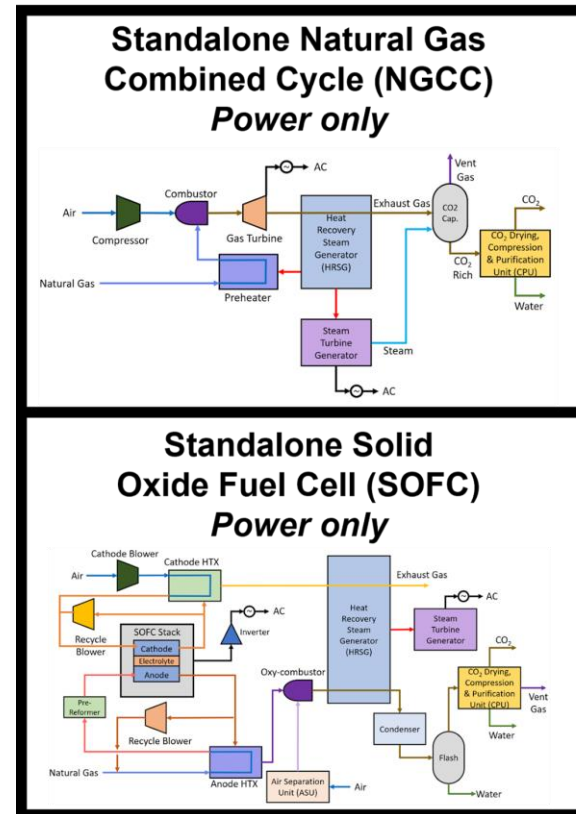
Wind and electrolyzer IESs that co-produce H_2 and electricity

Wind and battery IESs



Compare Solid Oxide Cells (Emerging Technology) with Legacy Technology (NGCC)

- Analyze the viability of SOFCs against legacy technology (NGCCs)

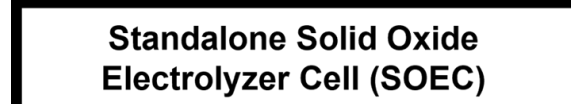
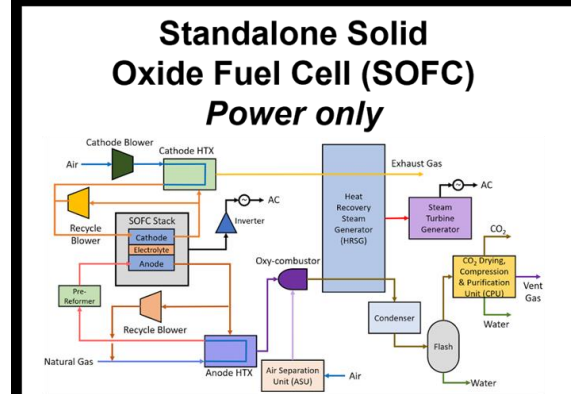
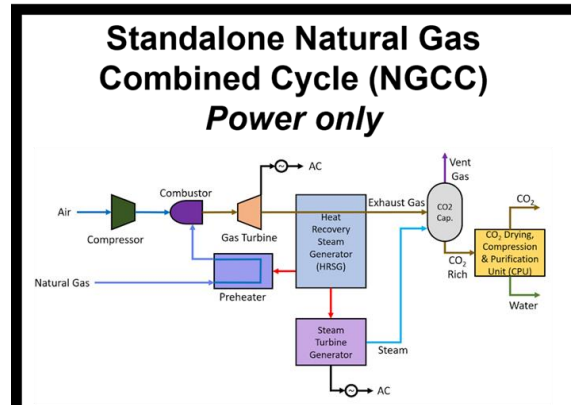


Which generators/technologies should be retired?

Which emerging technologies are worth research/scale-up investment?

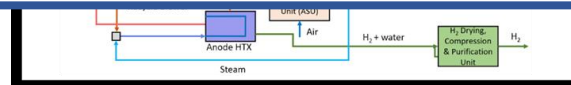
Evaluate Production of Alternative Fuels

- Analyze the viability of SOFCs against legacy technology (NGCCs)
- Evaluate the viability of alternative fuel production (e.g., Hydrogen from SOEC)



Which technologies should produce hydrogen?

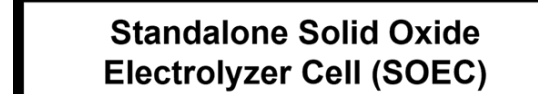
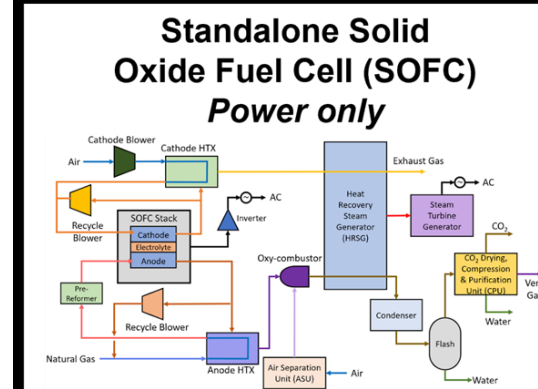
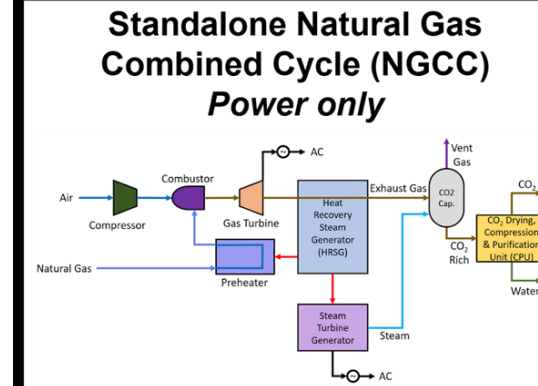
What price of hydrogen is economical?



Evaluate Coproduction

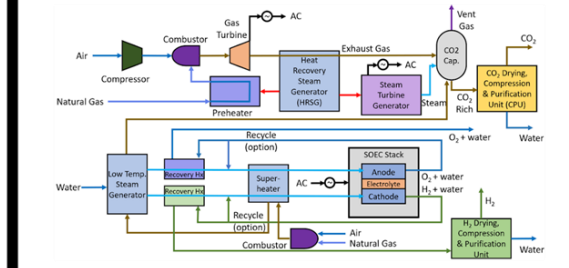
- Analyze the viability of SOFCs against legacy technology (NGCCs)
- Evaluate the viability of alternative fuel production (e.g., hydrogen from SOEC)
- Evaluate integrated energy systems (IESs)
 - Coproduction:
 - Power (NGCC)
 - Hydrogen (SOEC)
 - Reversible sys

Single Product Systems

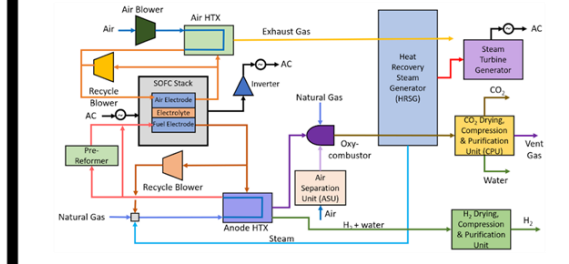


Integrated Systems

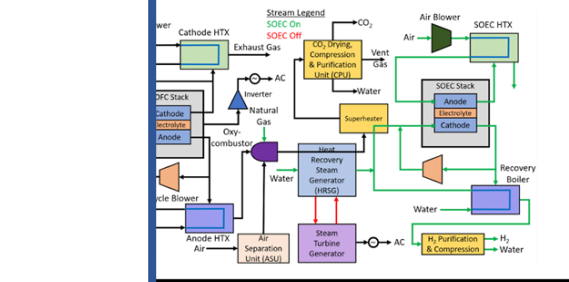
NGCC + SOEC Power, Hydrogen, Coproduction



Reversible Solid Oxide Cell (rSOC) Power, Hydrogen



SOFC + SOEC Power, Hydrogen, Coproduction



Natural Gas
Combined Cycle-
Based

Solid Oxide Fuel Cell-Based

Which technologies should be coupled to comprise an integrated energy system?

Which emerging technologies are worth research/scale-up investment?



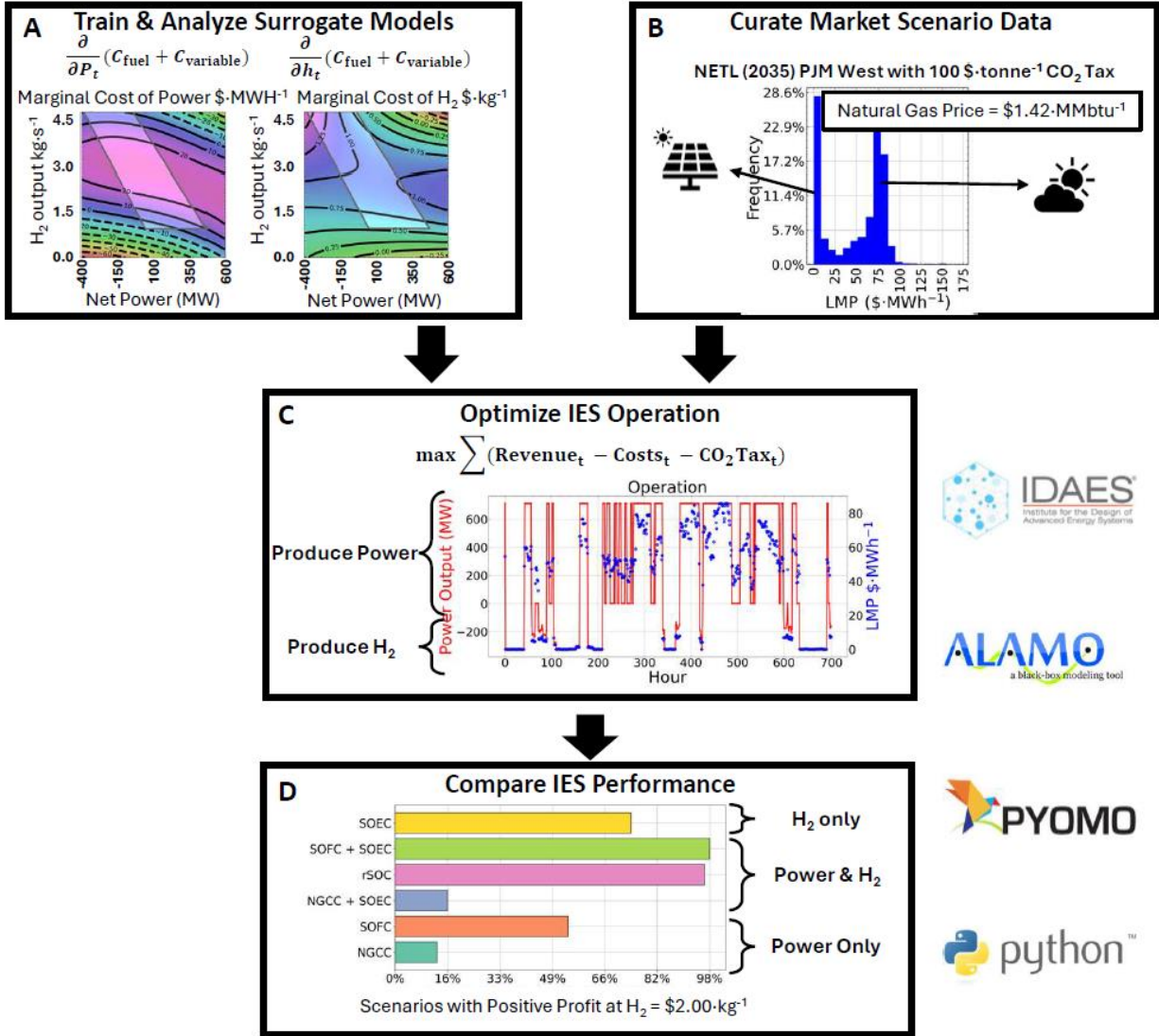
Grid-Wide Decisions for Locational Markets Require a Streamlined Evaluation Framework

Which technologies should produce hydrogen?

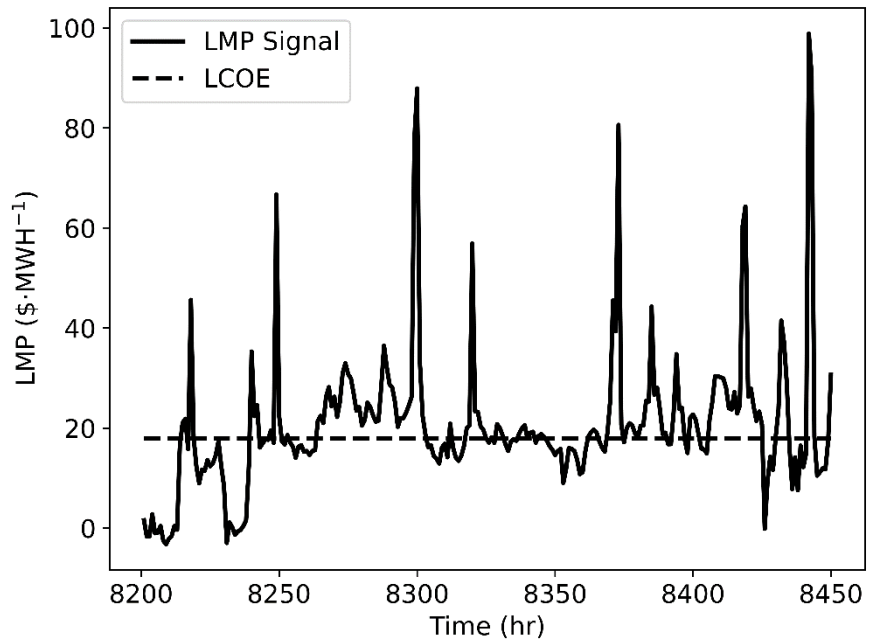
Which technologies should be coupled to comprise an integrated energy system?

Which generators/technologies should be retired?

Which emerging technologies are worth research/scale-up investment?



Price Taker Model Considers Dynamic Prices



Locational marginal price (LMP) changes dynamically, where traditional low-fidelity modeling only uses a static levelized cost of electricity (LCOE)

Objective options: NPV, Annualized NPV, Net Profit

Profit expression at each time period (e.g., hour)

$$f_t^{\text{profit}} = \pi_t^e p_t - \frac{\pi^g}{\pi^g} f^{\text{fuel}}(p_t) - f^{\text{var}}(p_t) - \pi^c f^{\text{carbon}}(p_t) - f_t^{\text{fixed}}(P^{\text{max}}) \quad \forall t \in \mathcal{T}$$

Capacity Constraints

$$P^{\min} y_t \leq P_t \leq P^{\max} y_t \quad \forall t \in \mathcal{T}$$

Startup and Shutdown Constraints

$$\begin{aligned} y_t &\leq z_{\text{build}} & \forall t \in \mathcal{T} \\ \sum_{t=\tau^u+1}^t v_j &= y_t & \{t \mid t > \tau^u\} \\ \sum_{t=\tau^d+1}^t w_j &= (1 - y_t) & \{t \mid t > \tau^d\} \\ y_t - y_{t-1} &= v_t - w_t & \{t \mid t > 1\} \end{aligned}$$

Ramping Constraints

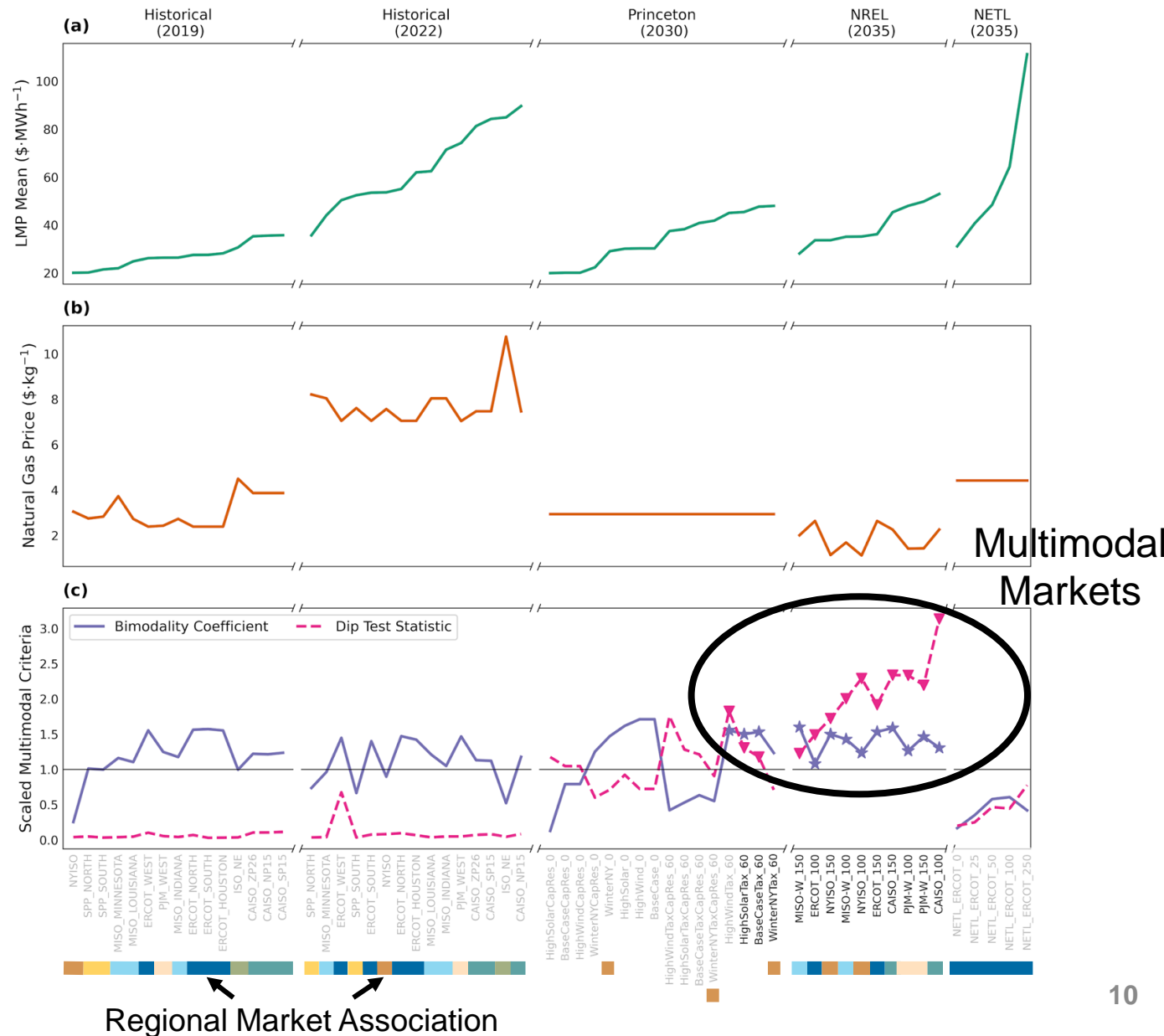
$$\begin{aligned} \frac{(P_t - P_{t-1})}{P^{\max}} &\leq (r_{\text{su}} - r_{\text{op,u}}) v_t + r_{\text{op,u}} y_t & \forall t \in \mathcal{T} \\ \frac{(P_{t-1} - P_t)}{P^{\max}} &\leq r_{\text{sd}} w_t + r_{\text{op,d}} y_t & \forall t \in \mathcal{T} \end{aligned}$$

61 Markets Used to Evaluate Emerging Technologies



Figure from FERC (<https://www.ferc.gov/electric-power-markets>)

- 15 historical markets (2019)
- 15 “current” markets (2022)
- 16 forecasted scenarios (2030)
 - “Princeton”
- 10 forecasted scenarios (2035)
 - “NREL”
- 5 forecasted scenarios (2035)
 - “NETL”



Emerging Coproduction Technologies Make a Profit in Most Scenarios

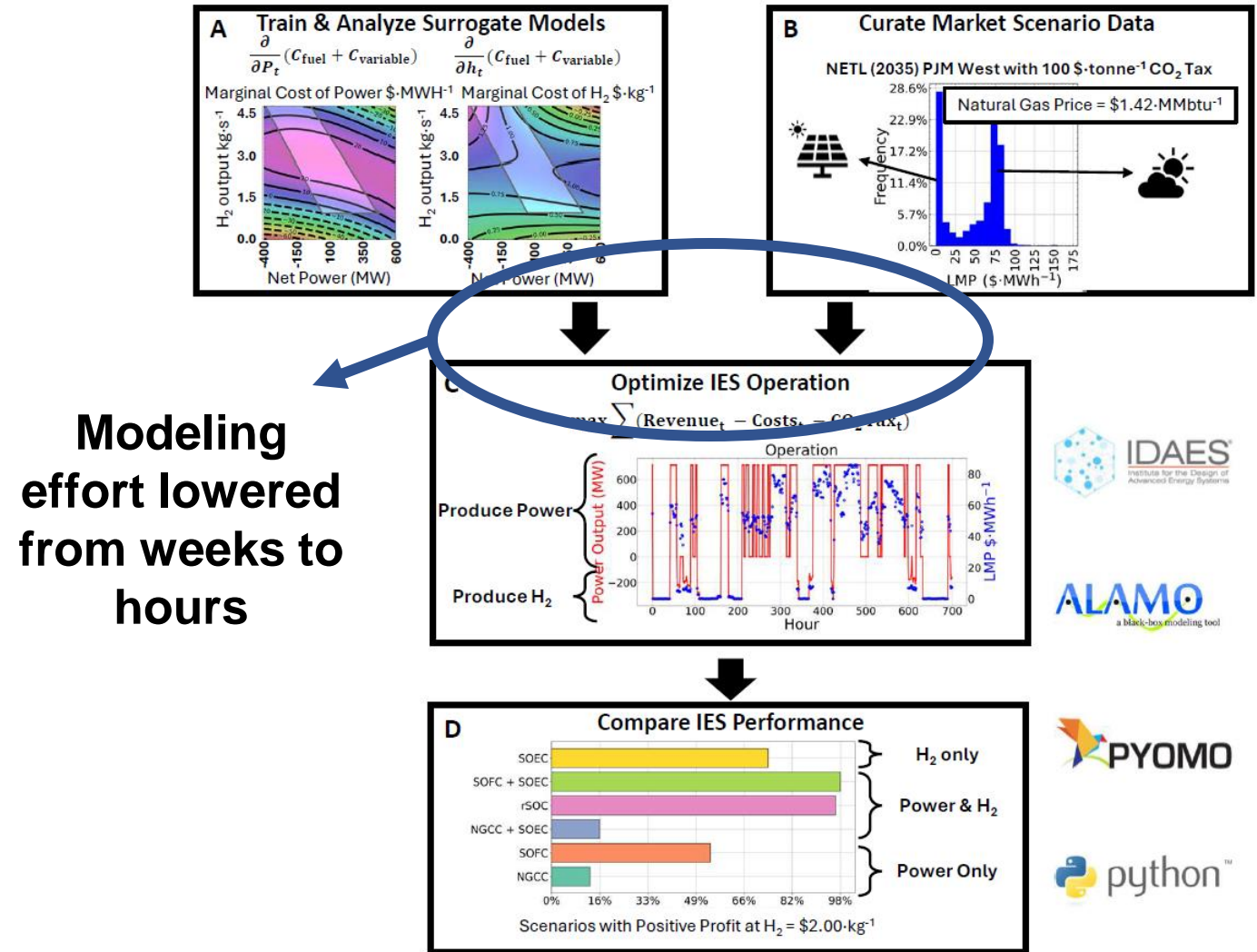
Percentage of scenarios that make profit at each **hydrogen selling price** (\$1.00 kg⁻¹ to \$3.00 kg⁻¹)

Process Concept	1.00\$.kg ⁻¹	1.50\$.kg ⁻¹	2.00\$.kg ⁻¹	2.50\$.kg ⁻¹	3.00\$.kg ⁻¹
NGCC	13%	13%	13%	13%	13%
SOFC	54%	54%	54%	54%	54%
NGCC + SOEC	8%	11%	16%	62%	80%
rSOC	54%	77%	97%	100%	100%
SOFC + SOEC	46%	79%	98%	100%	100%
SOEC	10%	49%	74%	87%	98%

Takeaway: At sufficient hydrogen price (\$2.50+), even the existing thermal generation technology with co-production (**NGCC +SOEC**) sees profit in over half of the market scenarios.

Price Taker Class Streamlines Optimal Operation Scheduling

- Automatically populates LMP (market data) in model
- Only need to specify costing equations and choose operational constraints for a representative time period
- Easily automated/scriptable using the IDAES ecosystem (Python, open-source)



IDAES Documentation:

https://idaes-pse.readthedocs.io/en/main/reference_guides/apps/grid_integration/index.html

Presentation Outline

How to **co-optimize** IES design and operation considering **dynamic markets**.

Price Taker

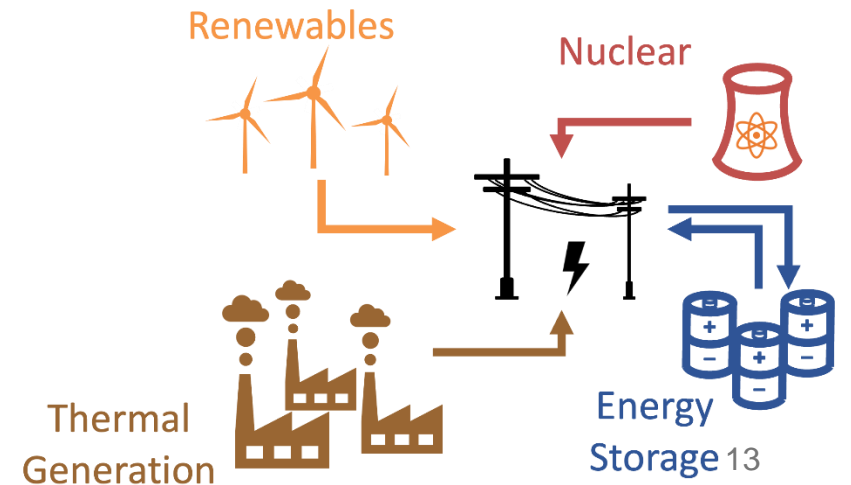
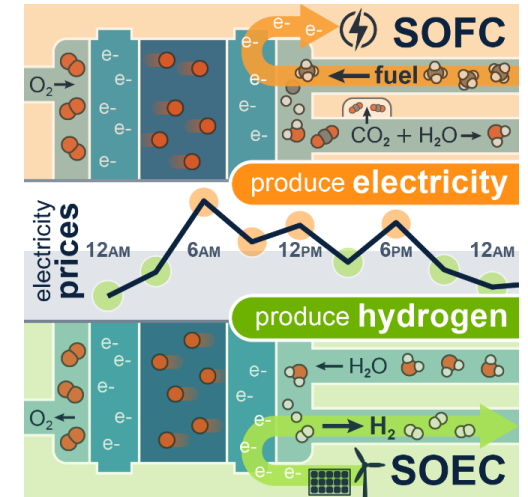
Solid oxide fuel cell IESs that co-produce H_2 and electricity

Beyond Price Taker

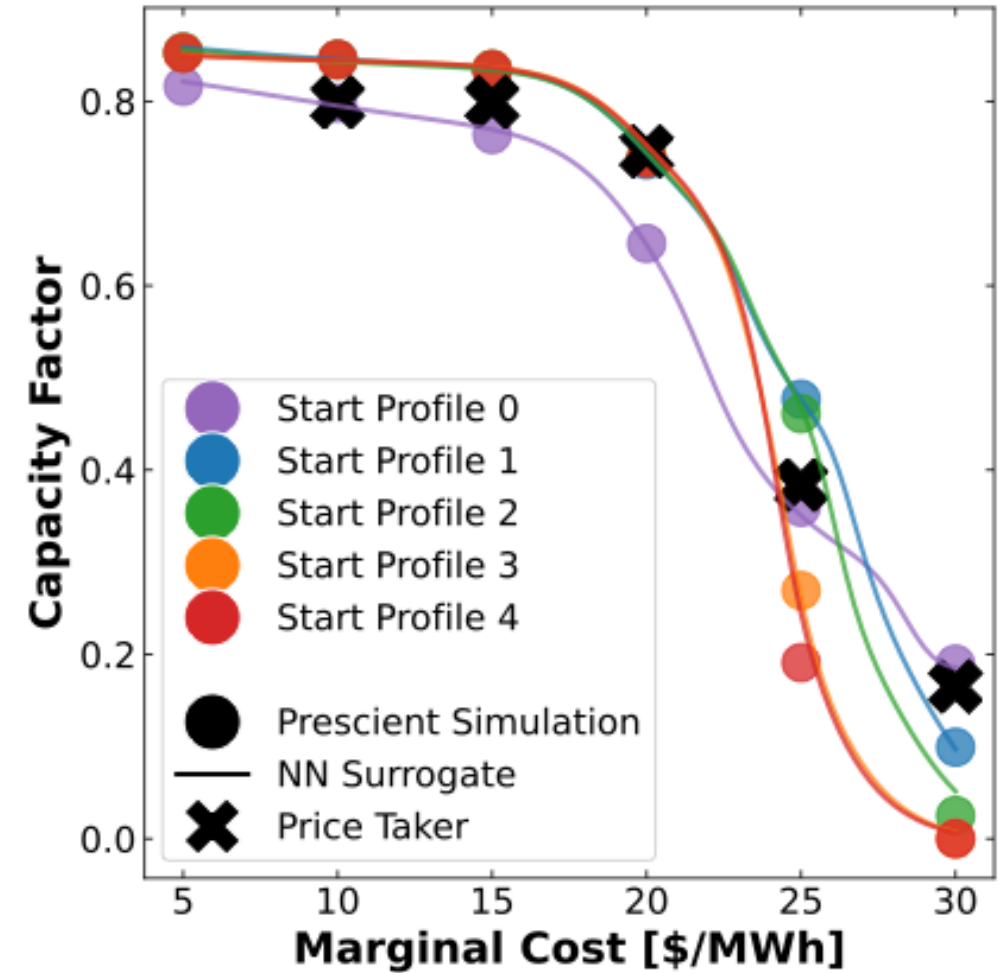
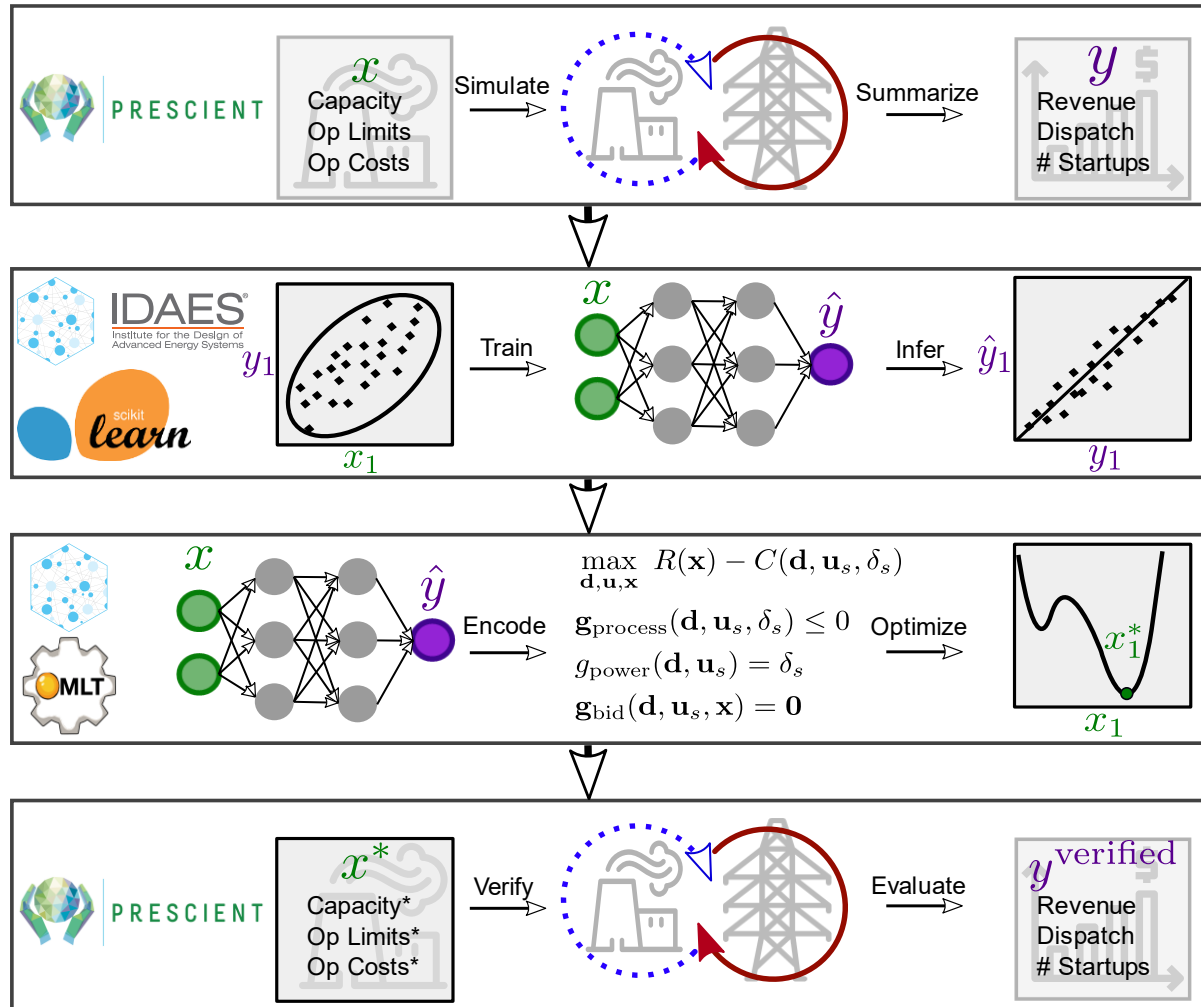
Nuclear and electrolyzer IESs that co-produce H_2 and electricity

Wind and electrolyzer IESs that co-produce H_2 and electricity

Wind and battery IESs



Moving Beyond Price Taker



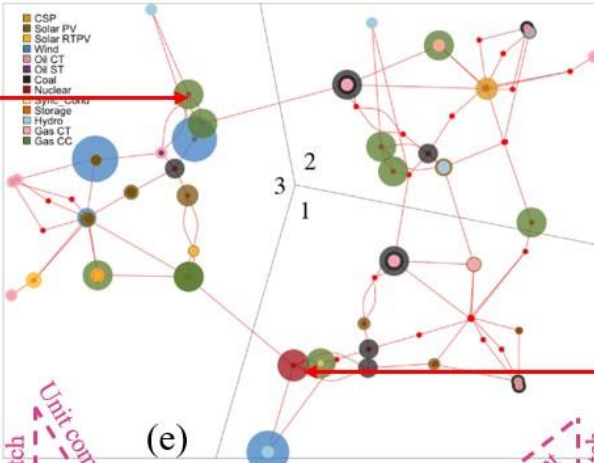
price
taker
breaks
down

Jalving et al, *Applied Energy* (2023)

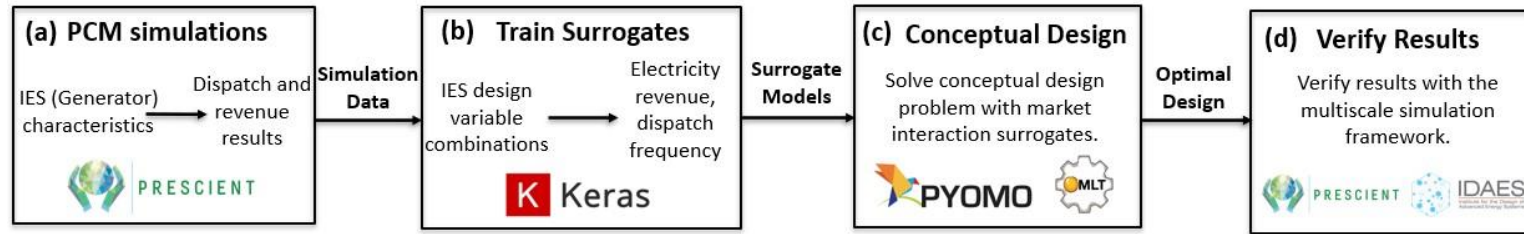
Beyond Price Taker: Co-Production of Electricity and H₂

Problem Statement

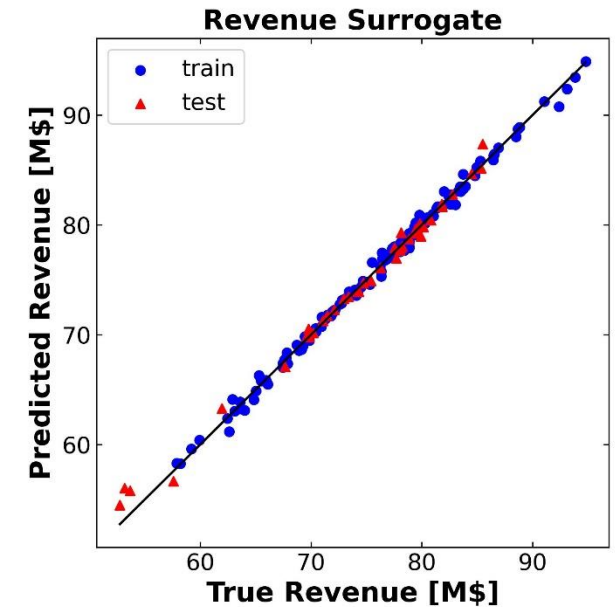
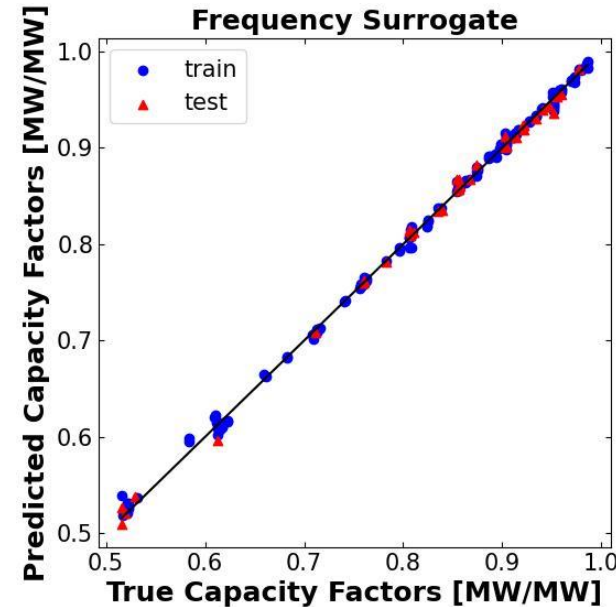
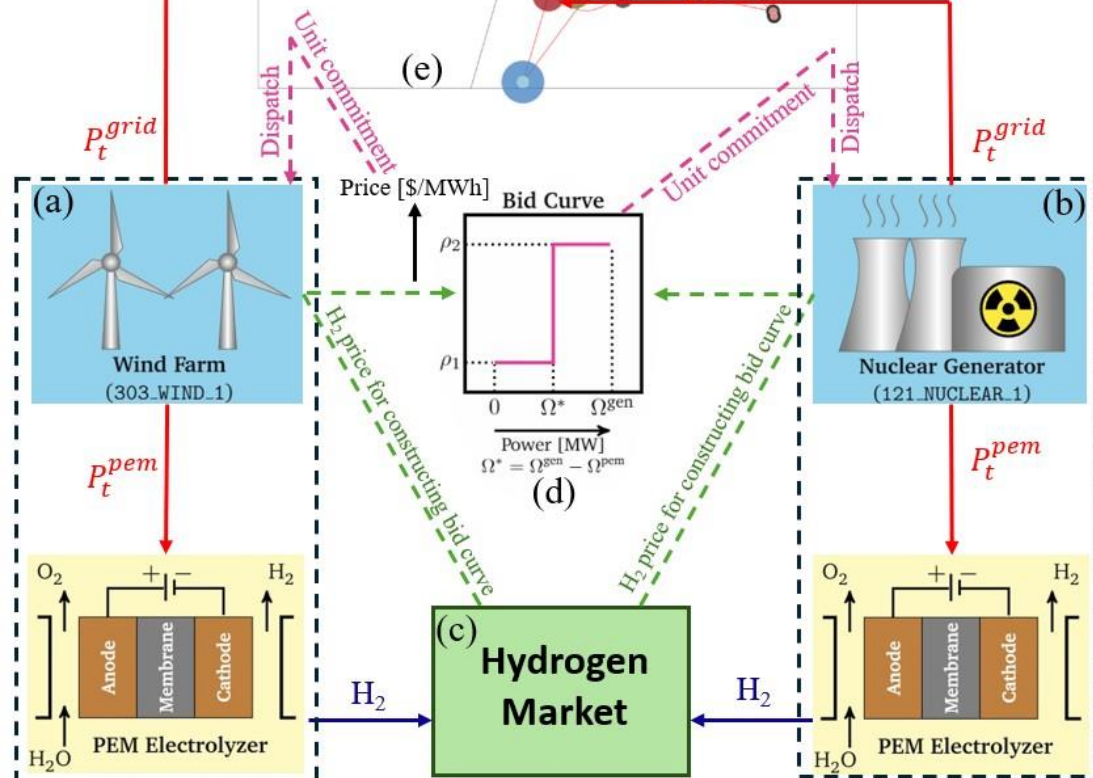
Electric Grid (RTS-GMLC)



Overall Approach



Example Surrogates (Nuclear + PEM)

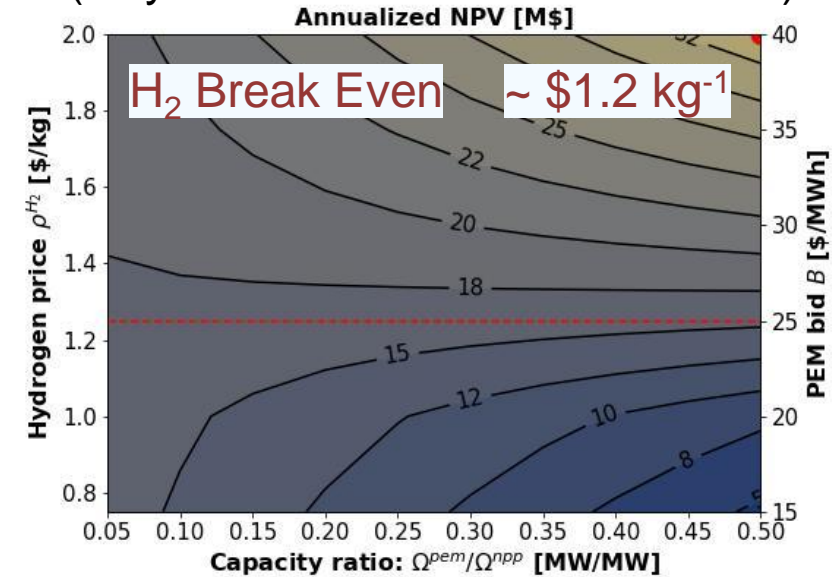


Chen et al, (2025) *in preparation*

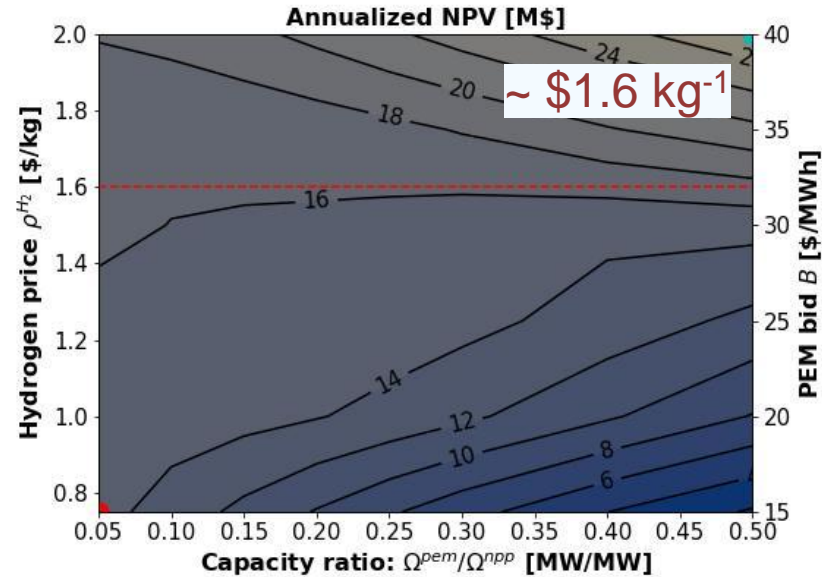
Nuclear + PEM Case Study

Price Taker

(Day-Ahead + Real-Time Markets)

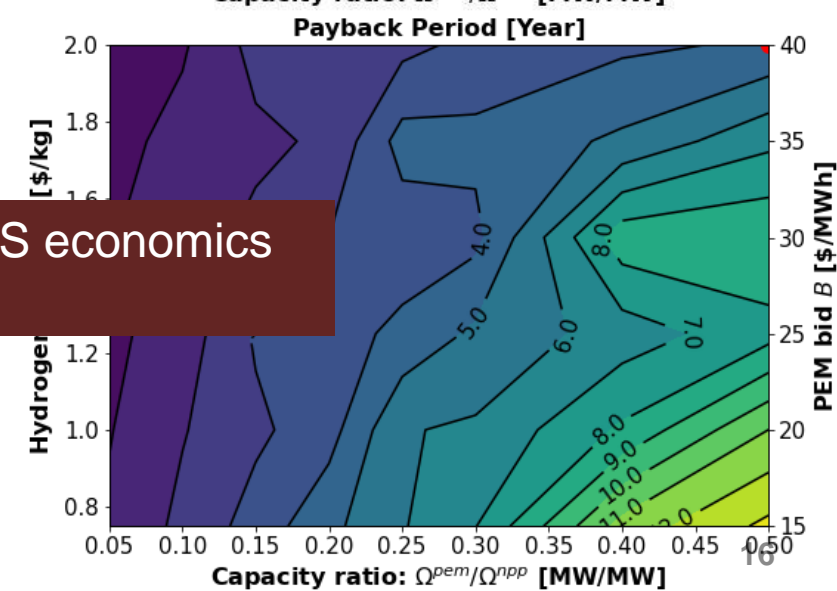
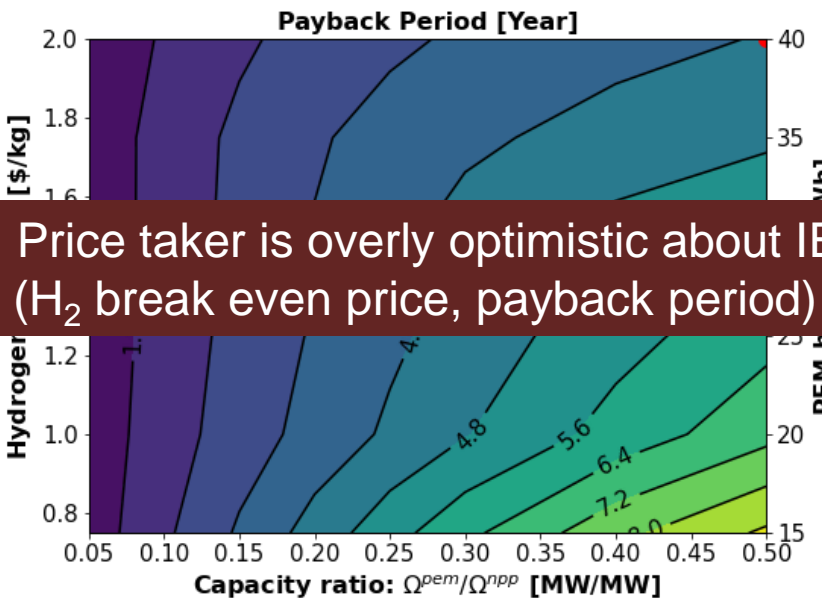
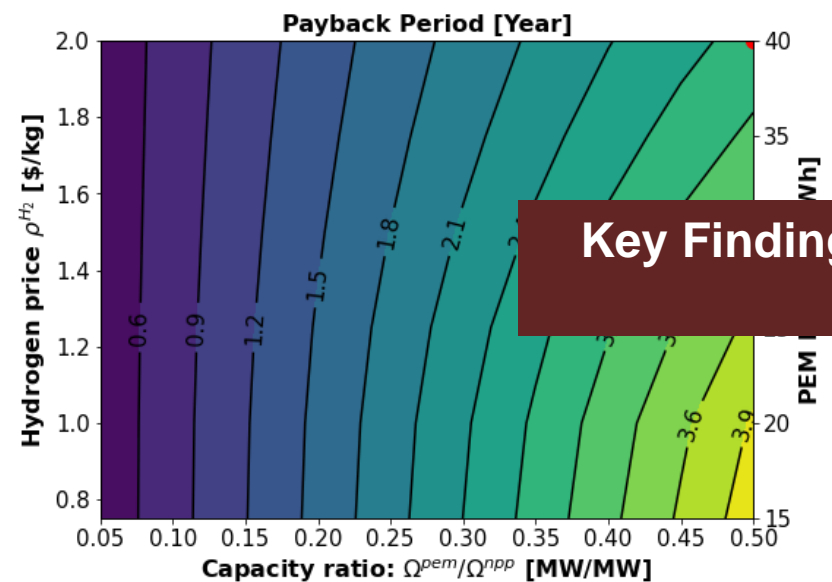
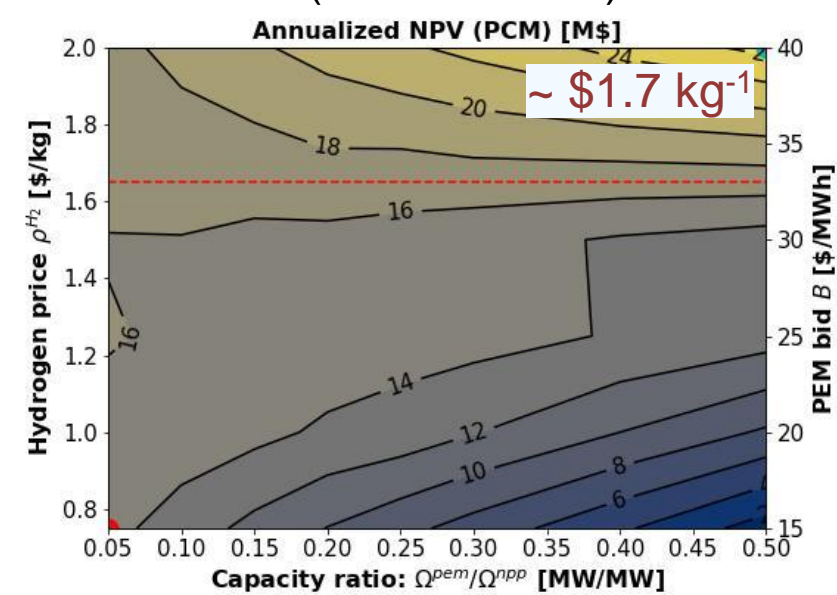


Surrogates



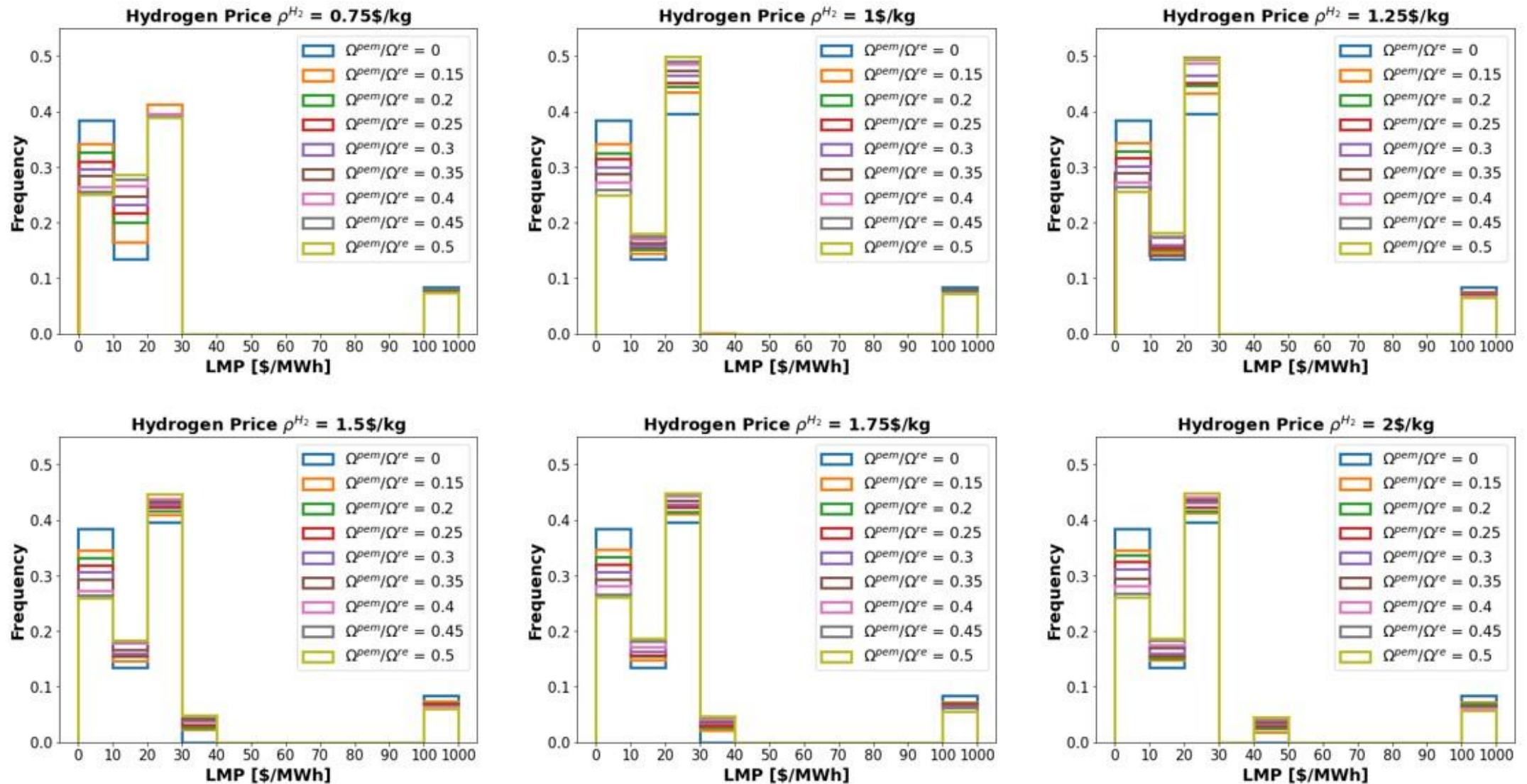
Production Cost Model

(Ground Truth)



Key Finding: Price taker is overly optimistic about IES economics (H₂ break even price, payback period)

Renewables + PEM Case Study: Key Finding



Key Finding: IES changes price (LMP) distribution

Presentation Outline

How to **co-optimize** IES design and operation considering **dynamic markets**

Price Taker

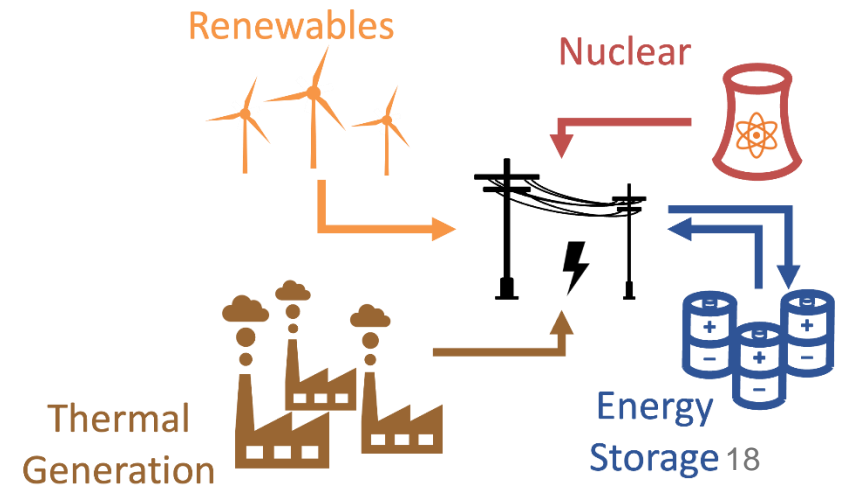
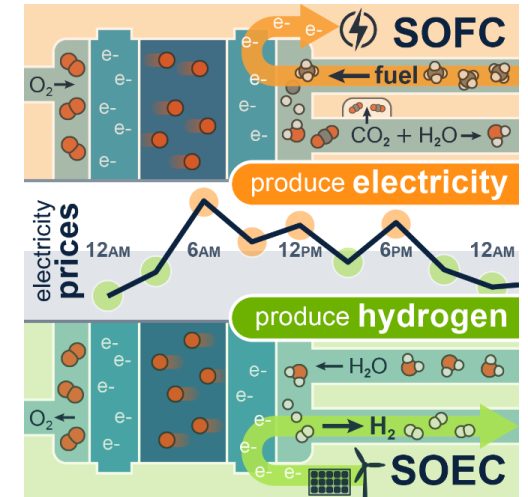
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Wind and electrolyzer IESs that co-produce H_2 and electricity

Wind and battery IESs



Multiperiod Optimization (MO) of the Wind-Battery IES

The market outcomes (LMP and dispatch schedule) are sent to the IES

(c) MO(PCM)

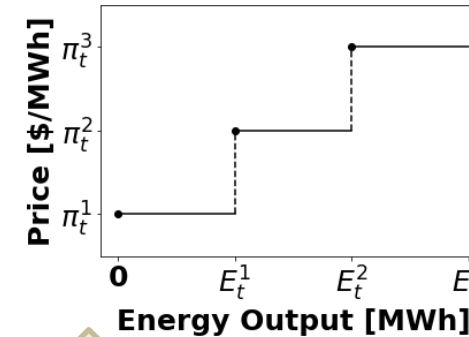


MO (PCM, product cost model) clears the market

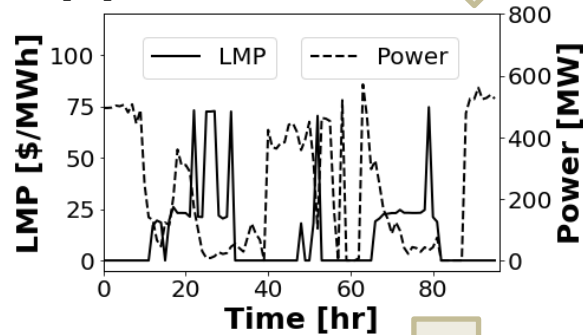
(e) Electric Grid



(b) Bids, $(E_{t,i}, \pi_{t,i})$



(d) Market Outcomes



Price Signal, $\hat{\pi}_t$
Dispatch Schedule, \hat{p}_t

IES follows the dispatch commitment and gets paid

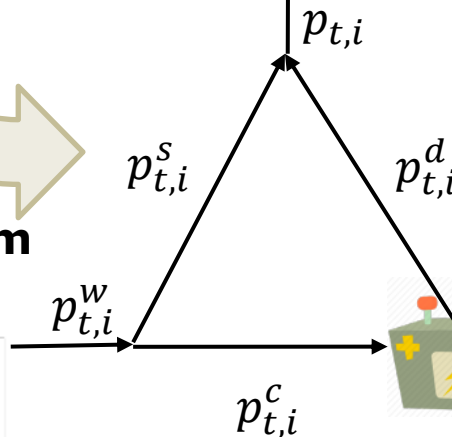
Wind Farm



Battery Storage

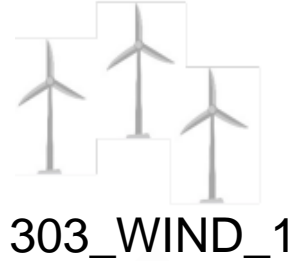
Solve the stochastic bidding problem to generate time-variant bids for the IES

(a) Wind-Battery IES

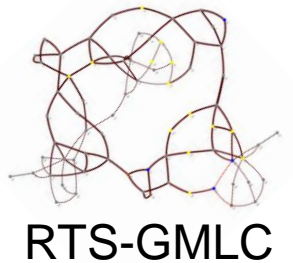


Price-taker (PT) with Perfect Information and Uncertainty

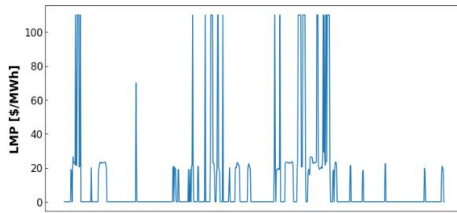
Choose wind farm
'303_WIND_1' in the
RTS-GMLC dataset



Product Cost
Model (PCM)



Obtain the real-
time LMP price
signals



Solve the price-taker optimization

Max Net Present Values of IES Investments

s. t. Wind farm and battery operation constraints

Perfect Information Mode:

LMP signals are deterministic

$$R_{t,i} = (\hat{\pi}_{t,i} + \varepsilon) \cdot p_{t,i} \cdot \Delta t$$

Where $I = \{0\}, T = \{0, 1, \dots, 8783\}$

$R_{t,i}$: Total revenue at time t , scenario i .

$\hat{\pi}_{t,i}$: LMP at time t , perfect information.

$p_{t,i}$: IES power output at time t ,
scenario i .

ε : Small incentive (0.001 \$/MWh) to
avoid degeneracy.

Δt : Time step, hour.

Uncertainty Mode:

Rolling horizon stochastic
optimization and use historical prices
as scenarios.

$$R_{t,i} = (\pi_{t,i} + \varepsilon) \cdot p_{t,i} \cdot \Delta t$$

Nonanticipativity constraints

$$p_{t,i} = p_{t,i'} \quad \forall t \in T'_1, \forall i' \in I \setminus i$$

Where $I = \{0, \dots, 9\}, T = \{T'_1 \cup T'_2\}$

$\pi_{t,i}$: LMP at time t and scenario i .

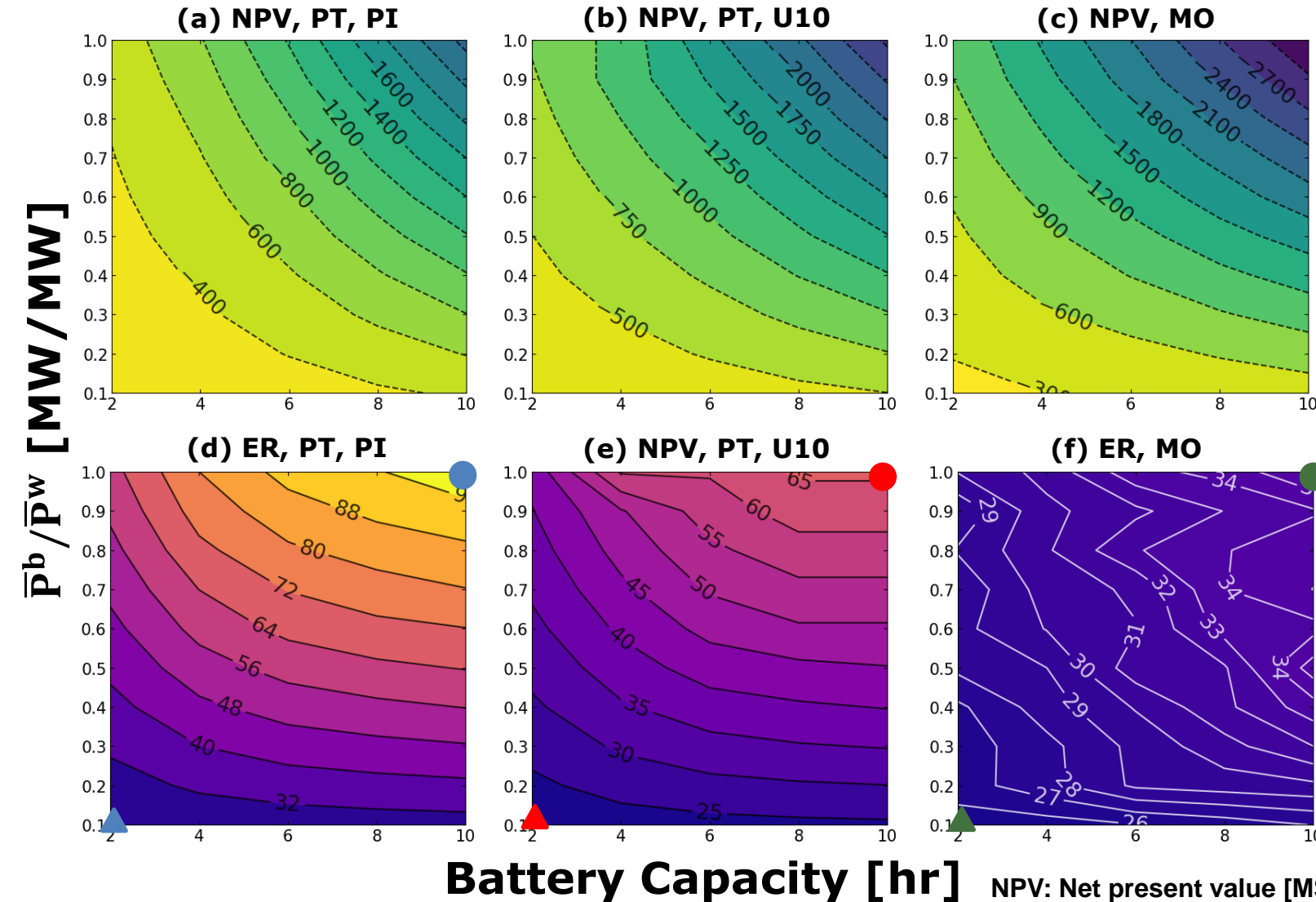
T'_1 : Stage 1 time set, $\{0, \dots, 23\}$.

T'_2 : Stage 2 time set, $\{24, \dots, 71\}$.

After optimization is solved:

$$\hat{R}_d = \sum_{t \in T'_1} \hat{\pi}_t \cdot p_{t,0} \cdot \Delta t$$

PT is Overly Optimistic on IES Economic Values



- All NPV values are negative in both PT and MO (overbuilt grid).
- Electricity revenue (ER), [M\$]
 - ▲ Case 1 (PT, PI, smallest battery): 24.2
 - Case 2 (PT, PI, largest battery): 100.4
 - ▲ Case 3 (PT, U10, smallest battery): 21.2
 - Case 4 (PT, U10, largest battery): 65.9
 - ▲ Case 5 (MO, smallest battery): 25.5
 - Case 6 (MO, largest battery): 36.0
- PT overestimates the NPV and ER.
- NPV and ER are more sensitive to the maximum battery power \bar{P}^b (\bar{P}^w : maximum wind power, parameter)

NPV: Net present value [M\$]
 ER: Annual electricity revenue [M\$]
 PI: Perfect information
 U10: Uncertainty with 10 scenarios

Take Away Messages

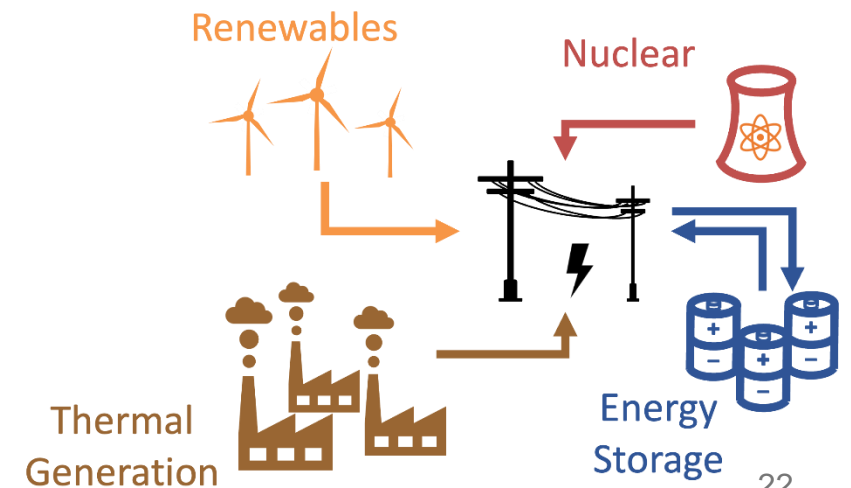
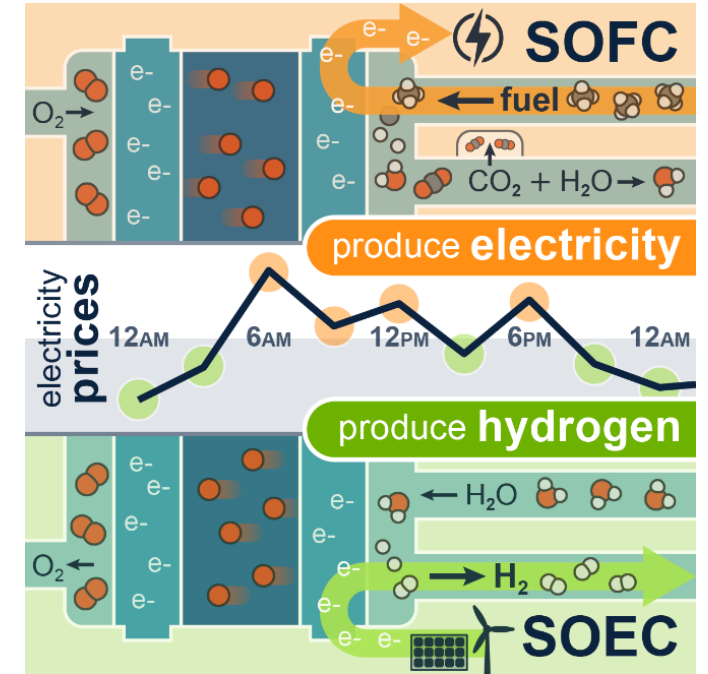
Optimization is a powerful tool to analyze integrated energy systems (IESs) in dynamic energy markets.

Price taker assumes IES decisions do not impact market prices.

- Surrogates and PriceTaker class in IDAES makes this analysis fast and easy.

Need to go beyond price taker (with IDEAS)!

- Price taker is often overly optimistic.
- IES decisions shift market prices.



Acknowledgements

The IDAES team gratefully acknowledges support from the U.S. DOE's **Hydrogen with Carbon Management** and **Simulation-Based Engineering Research Programs**.

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2025 Joint IDAES/CCSI₂/PrOMMiS/WaterTAP Technical Team Meeting
University of Notre Dame

<https://idaes.org/about/contact-us/>

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