

GREEN POWER PLANTS OF THE FUTURE

Using rolling-horizon optimization to achieve load-following grid power with near-zero emissions from next generation power plants.

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May 20, 2014. Fields Lecture, University of Toronto.

Motivation: The Toronto Problem

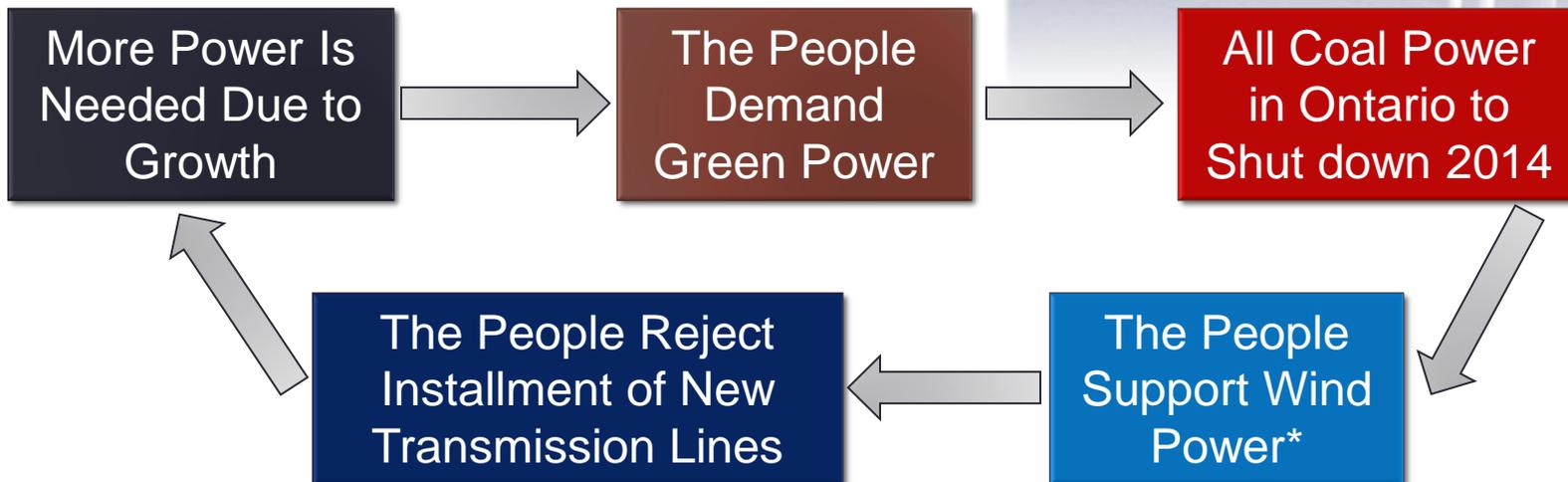


Clean Energy – A set of conflicting goals?

- People want clean energy, whatever that means to them.
- But they don't want to pay for it, in **money or inconvenience**.



Example: The Toronto Problem



* Some people oppose wind power due to bird deaths. Example, March 31, 2013, Wind farm in Nevada faces \$200,000 fine after the death of a gold eagle.

Triple Bottom Line of Sustainability

ECONOMICS

- ❖ Capital
- ❖ Operating
- ❖ Supply chain, materials
- ❖ Job creation and losses
- ❖ Profitability
- ❖ Uncertainty and Risk

This talk:
Profitability analysis

ENVIRONMENT

- ❖ Particulates
- ❖ CO₂, NO_x, SO_x
- ❖ Deforestation & Land Use
- ❖ Mining & Resource Extraction
- ❖ Water consumption
- ❖ Resource Depletion
- ❖ Toxicity
- ❖ Wildlife impact
- ❖ Noise

This talk:
Life Cycle Analysis

SOCIETY

- ❖ Public acceptance
- ❖ NIMBYs / BANANAs
- ❖ Health Impacts
- ❖ Safety of workers and community
- ❖ Accidents
- ❖ Public policy
- ❖ Elections and Politics

This talk: How CO₂
Tax Policy affects
design choices

Sources Jimenez-Gonzales and Constable, Green Chemistry and Engineering. 2012. And others.

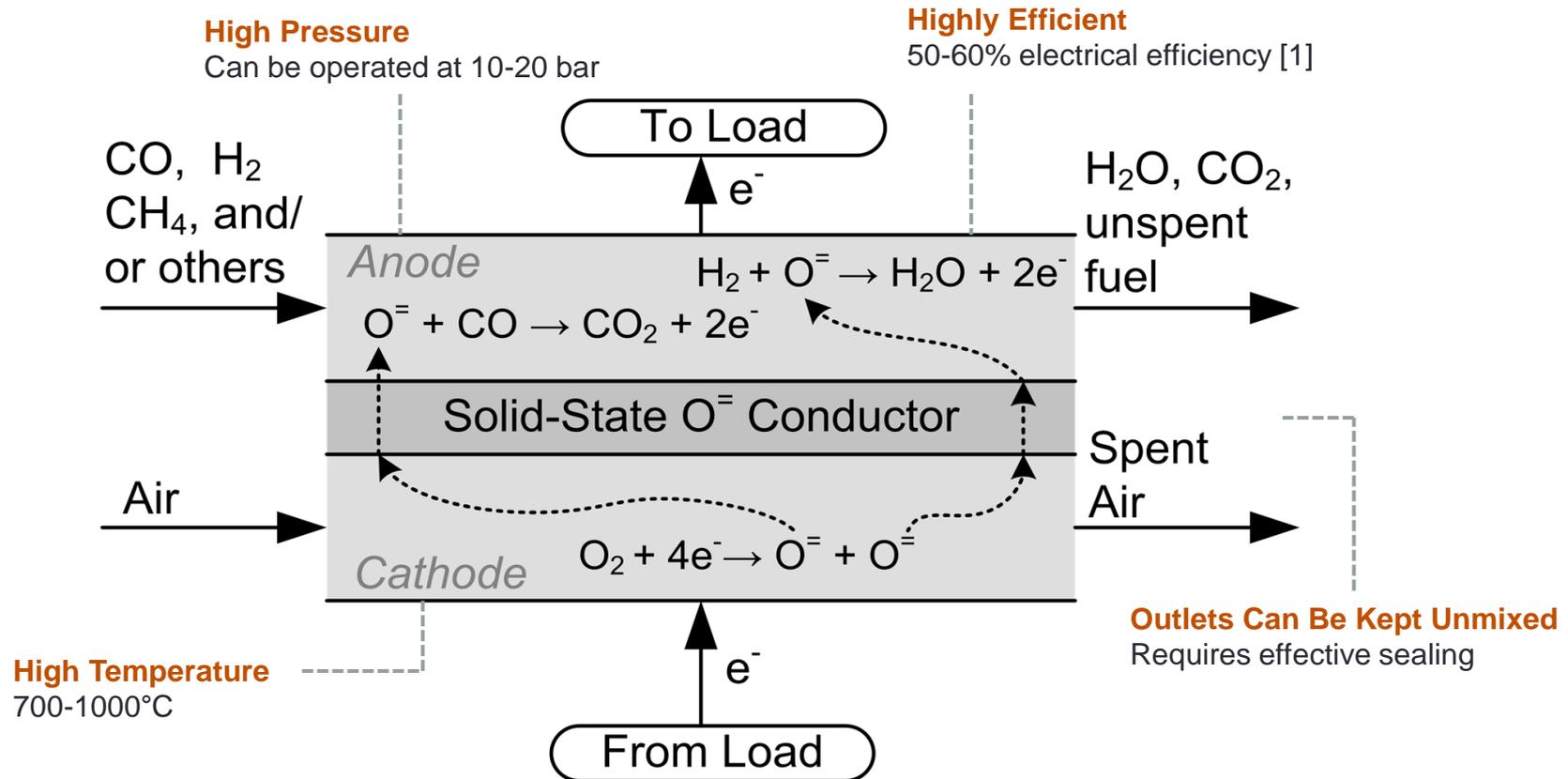


1. BULK SCALE POWER

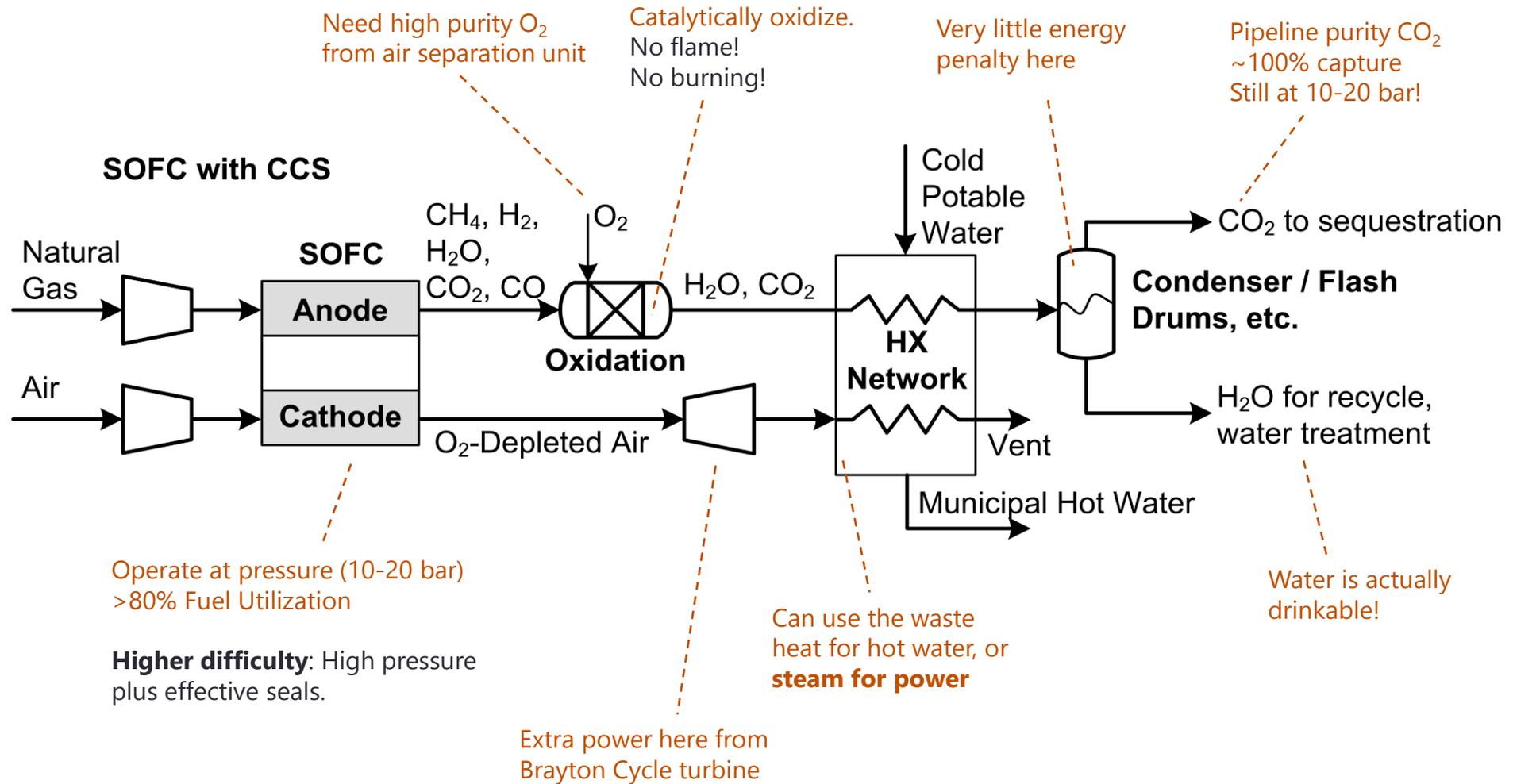
Integrates SOFCs and CAES, controlled by a real time optimizer.

Solid Oxide Fuel Cells

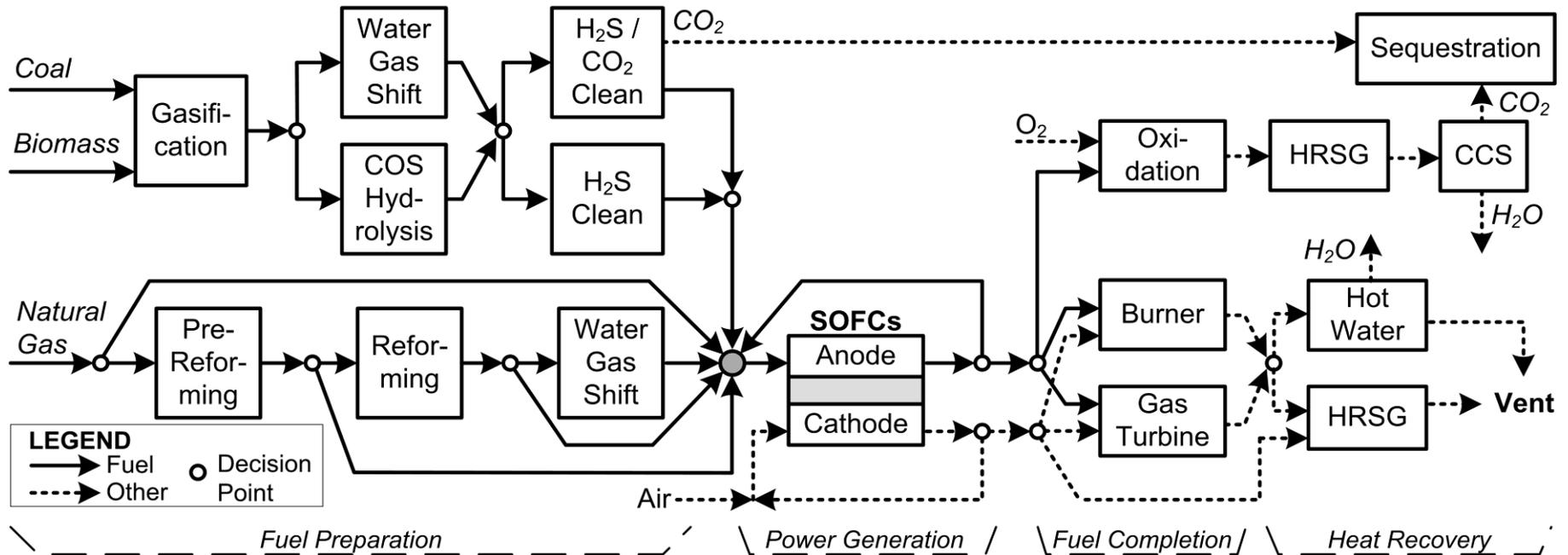
Electrochemical reactions between O_2 and a fuel gas occur across an impermeable oxide barrier, producing current



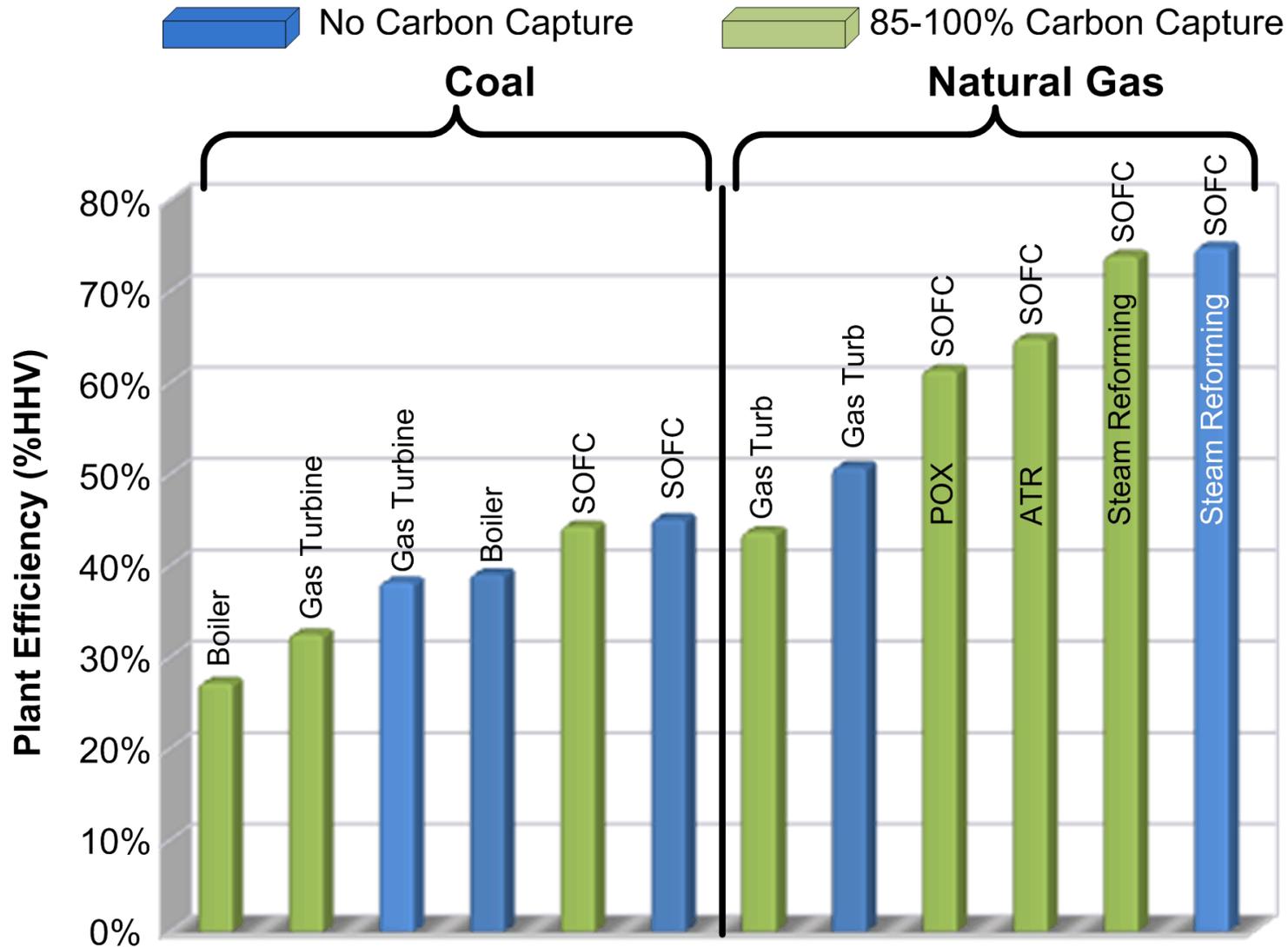
Vision: Long Term Bulk Power (NGFC)



1st and 2nd Generation Superstructure

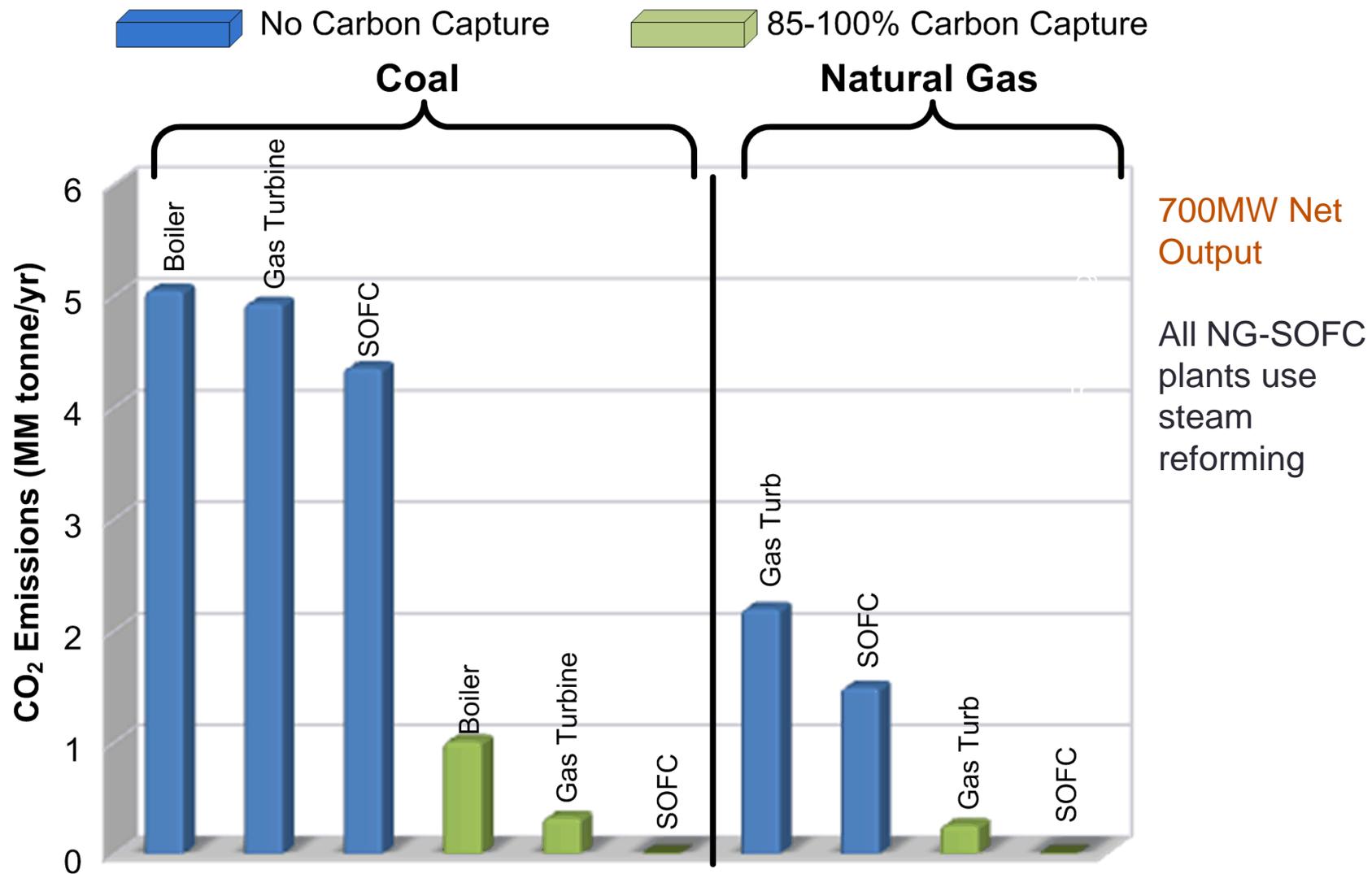


Efficiencies



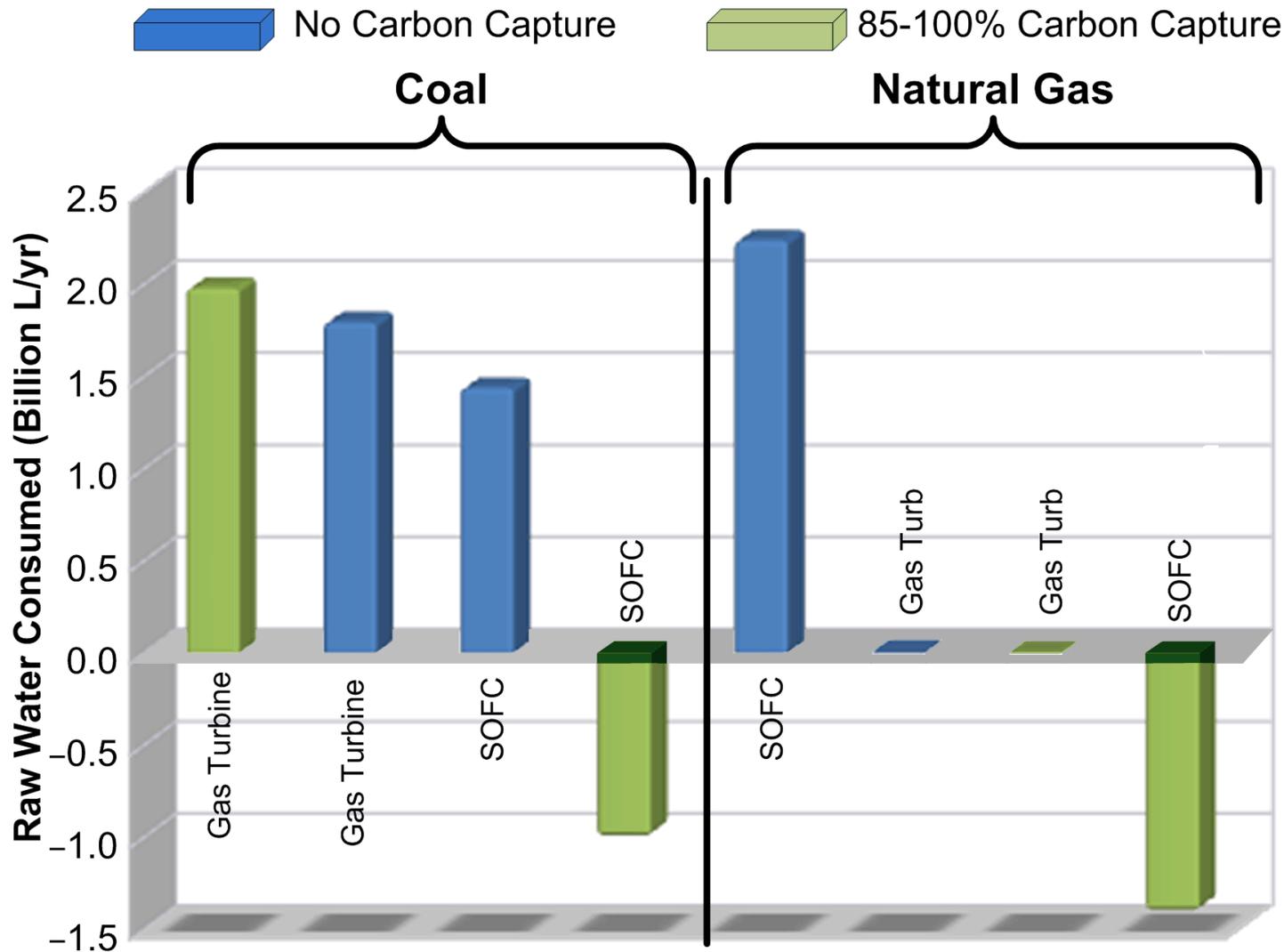
Sources: Adams & Barton. J Power Sources (2010).
Adams & Barton, AIChE J (2010)

CO₂ Emissions



Sources: Adams & Barton, J Power Sources (2010).
 Adams & Barton, AIChE J (2010)

Water Consumption



700MW Net Output

All NG-SOFC plants use steam reforming

Dry cooling used (no water losses from cooling)

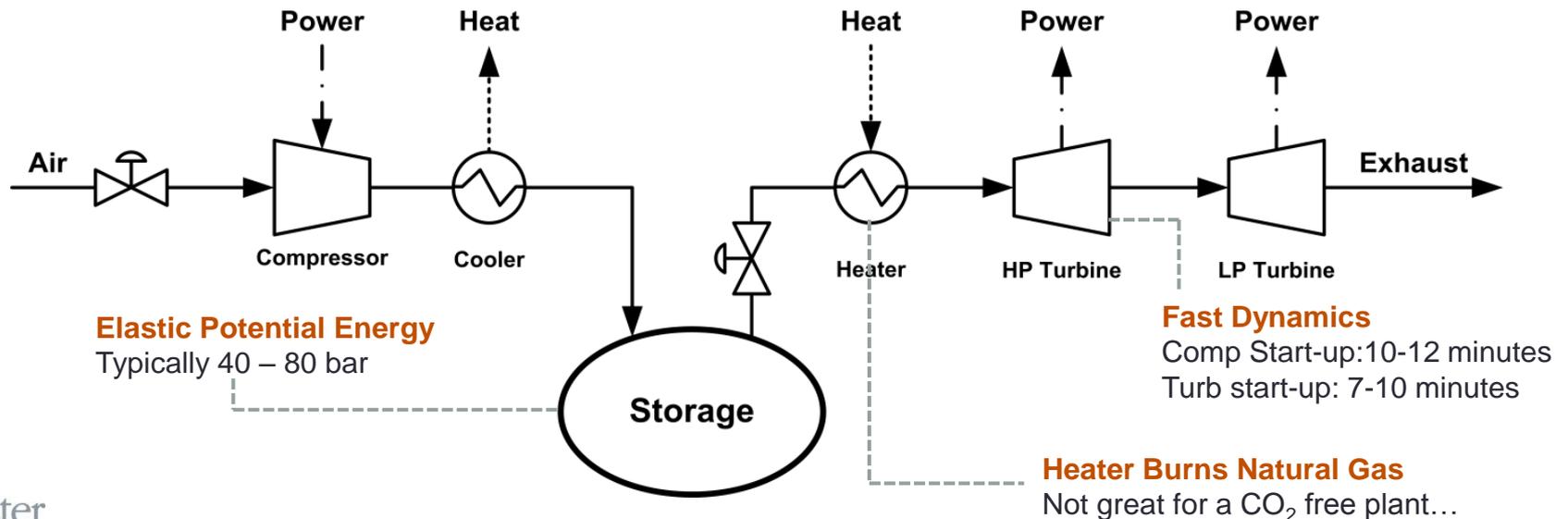
Compressed Air Energy Storage

❖ CAES: an intermittent source or sink

- ❖ Consumes power to compress and store air as elastic potential energy, which may be released as needed

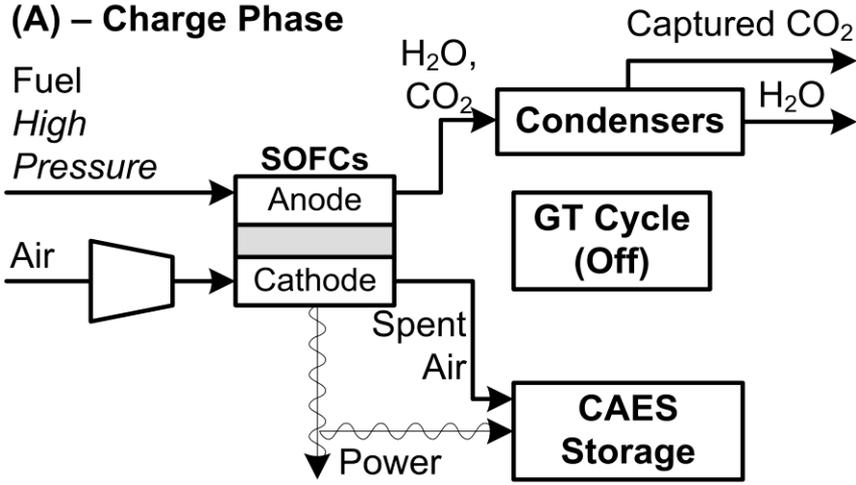
❖ Two CAES plants already operational

- ❖ Alabama Electric Co (110 MW)
- ❖ Apex Energy (317 MW in 2014)
- ❖ E.N. Kraftwerke [8] (290 MW)
- ❖ Chamisa Energy (270 MW, planned)

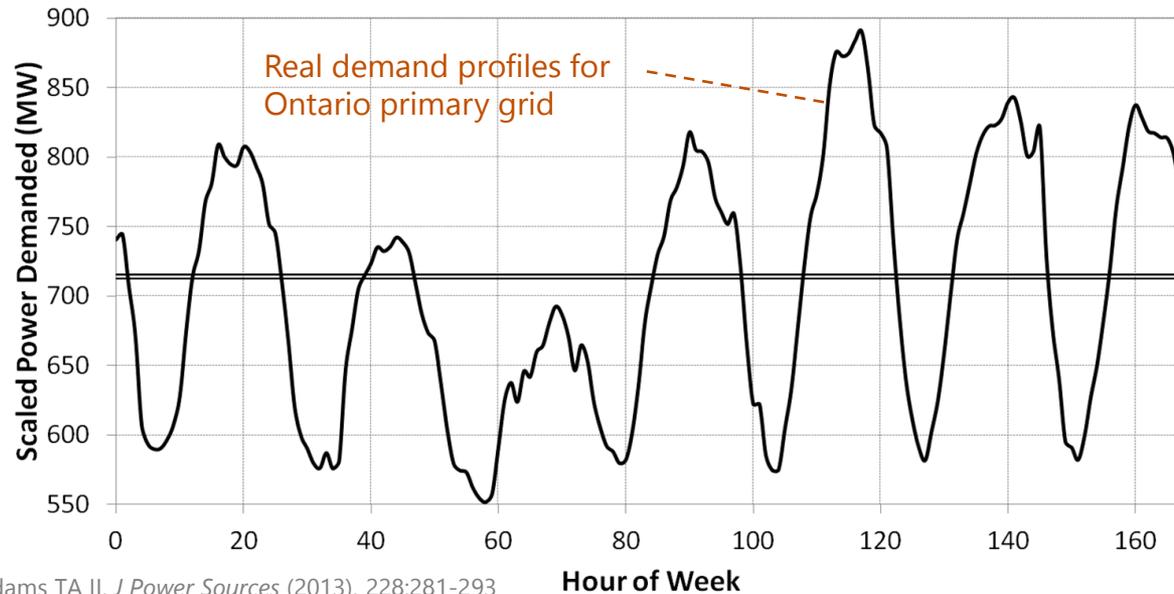
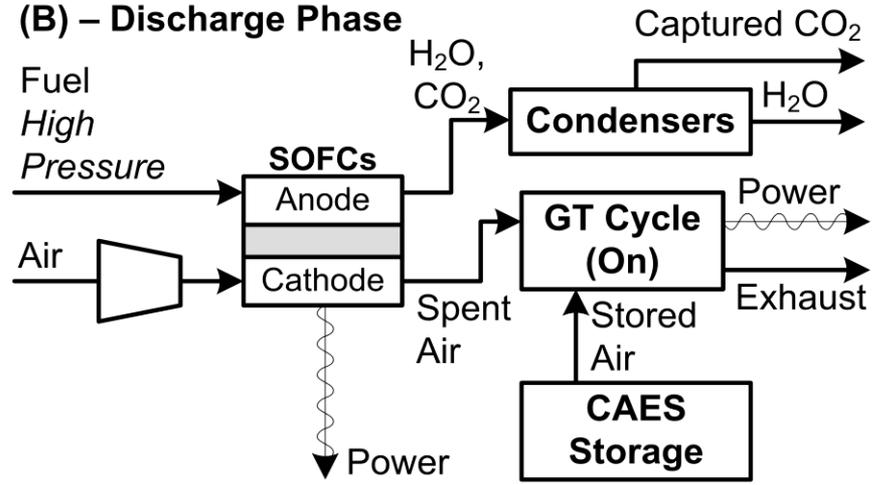


SOFC / CAES Integrated Systems

(A) – Charge Phase



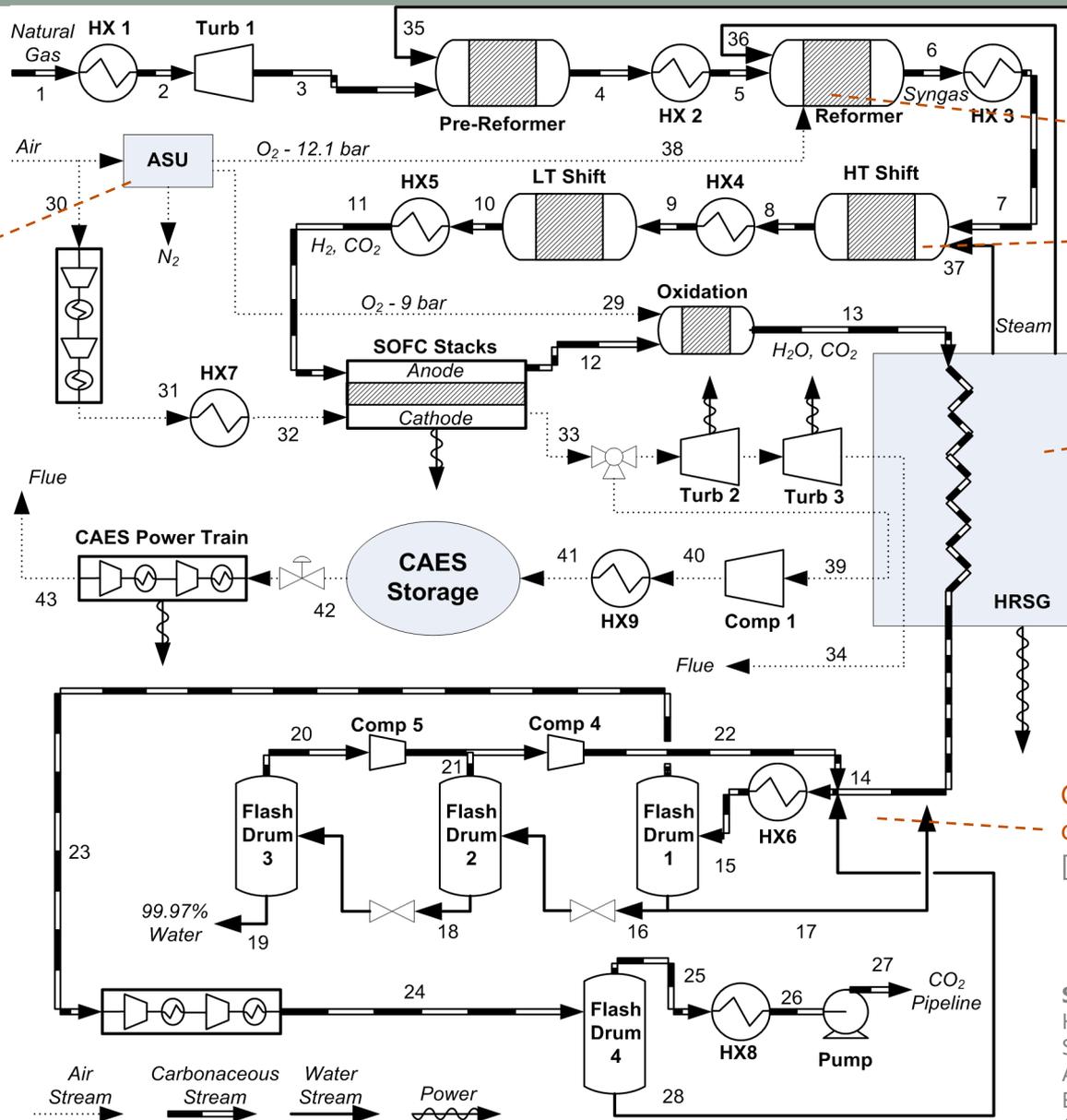
(B) – Discharge Phase



Sources: Nease J, Adams TA II. *J Power Sources* (2013). 228:281-293
 Adams TA II, Nease J, Tucker D, & Barton PI. *Ind Eng Chem Res.* (2013) 52:3089-3111

System Details

Small "coldbox" needed for fuel completion



Gas reforming steps are heat-integrated with SOFCs (planar design)

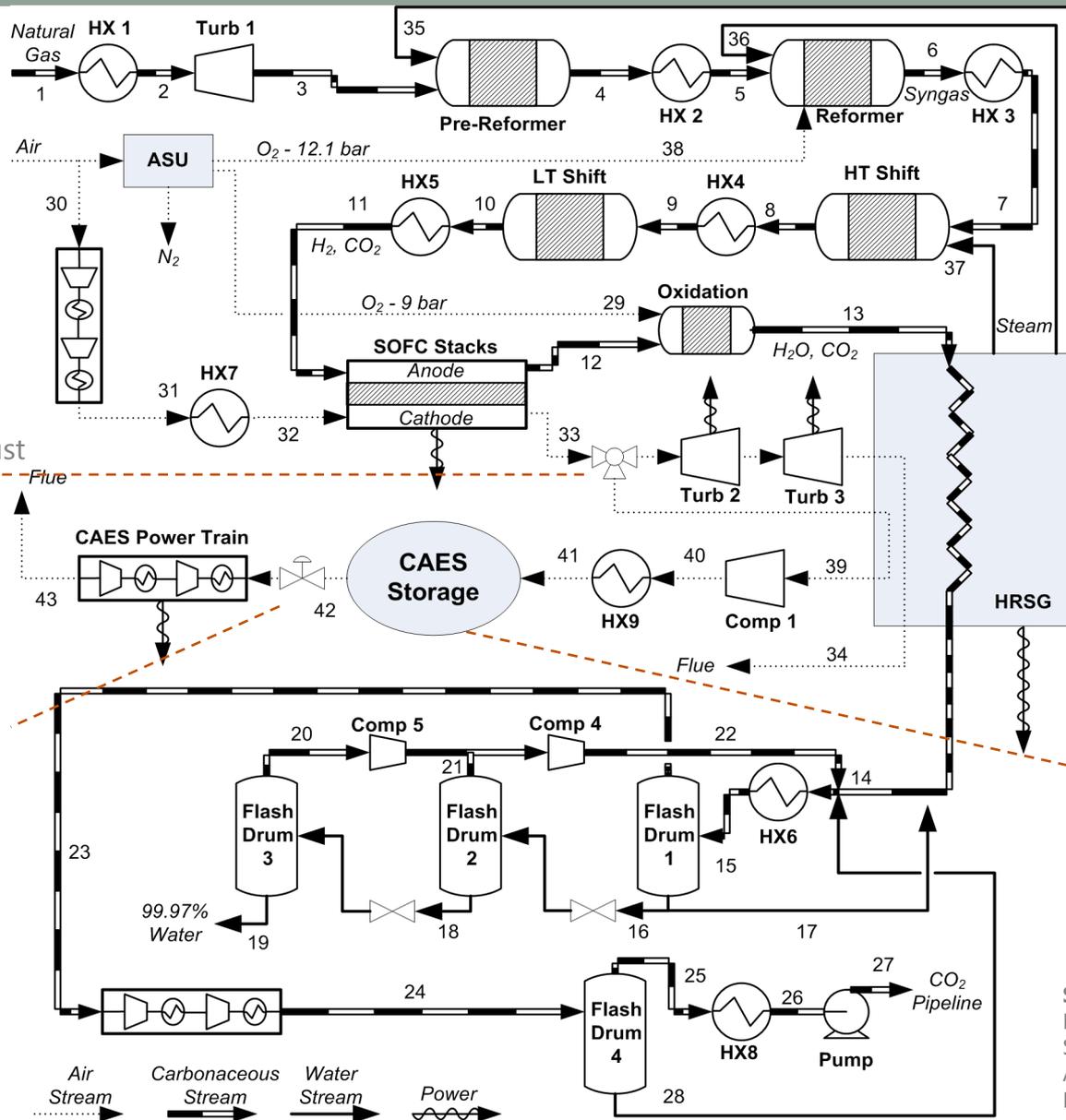
WGS is an optional step (we found it better to use)

Complete plant heat integrations considered

CO₂ capture system uses flash cascade for efficient capture [US Patent 8,500,868 (2013)]

Source: Nease J, Adams TA II. Rolling-Horizon Optimization of Integrated Solid-Oxide Fuel Cell and Compressed Air Energy Storage Plant for Zero-Emissions Peaking Power. Submitted (2014)

Optimal Performance Strategy



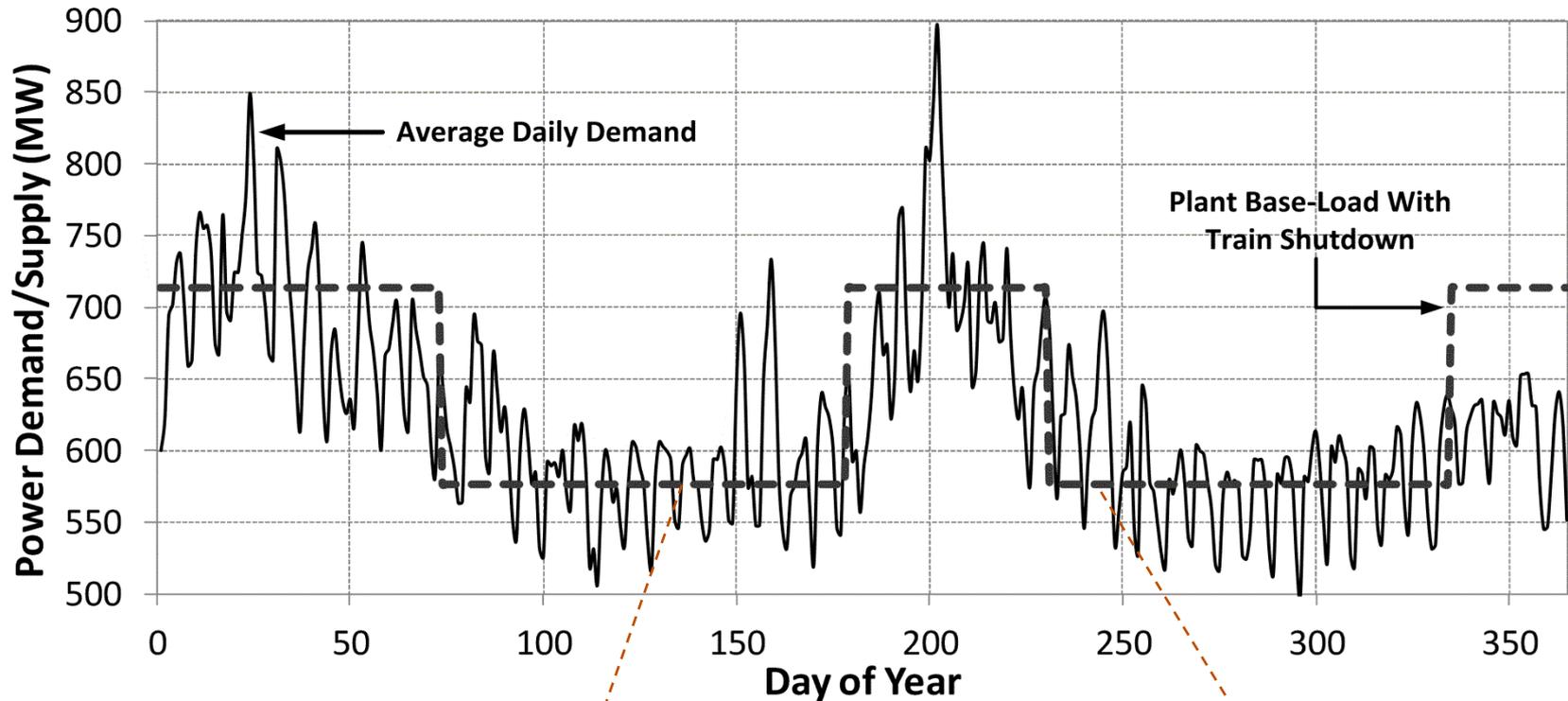
Manipulated:
% of Cathode Exhaust
diverted to CAES

Manipulated:
Air Release Valve %
opening

Storage Pressure:
Major impact on
downstream power
generated

Source: Nease J, Adams TA II. Rolling-Horizon Optimization of Integrated Solid-Oxide Fuel Cell and Compressed Air Energy Storage Plant for Zero-Emissions Peaking Power. Submitted (2014)

Note: Seasonal Variability



1 of 6 SOFC Modules turned off Each Spring/Fall
Each gets 3 month break for repairs every 3 years
Fits the real demand curve quite well.

For our study, we pre-selected the maintenance schedule ahead of time.

Rolling Horizon Optimization

- ❖ How do we best use the storage capability in real time in order to match real market demand?
 - ❖ We have access to **excellent predictive models for demand**
 - ❖ We have access to **less excellent predictive models for price**
 - ❖ We have access to **our own models of plant performance**

Problem Definition

How can rolling horizon optimization be used to achieve better system performance? Two approaches:

OBJECTIVE 1: Load Matching

$$\min_{\delta_{i,t}, F_{i,t}} SSE_i = \sum_{t=1}^N (E_{i,t} - D_{i,t})^2$$

Decision variables: The hourly schedule of how much we store or withdraw from the cavern for the next N hours

Power Produced
Hourly schedule for next N hours

Predicted Power Demand
Hourly schedule for next N hours

OBJECTIVE 2: Maximize Profit

$$\max_{\delta_{i,t}, F_{i,t}} \mathcal{R}_i = \sum_{t=1}^N (E_{i,t} \omega_{i,t}),$$

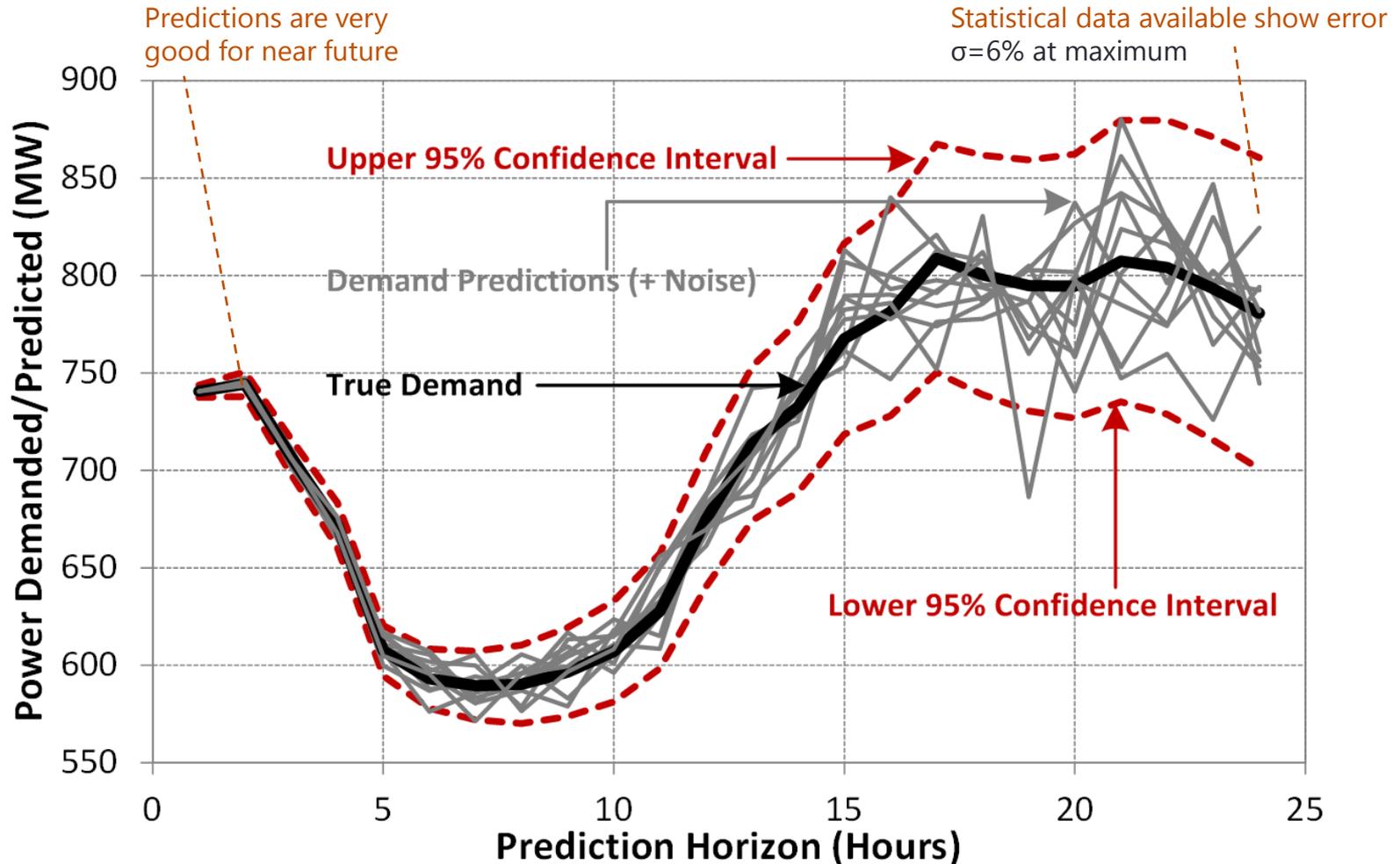
Predicted Market Price
Hourly schedule for next N hours

CONSTRAINTS

- ❖ Model equations for the system
- ❖ Pressure limits for the cavern ($40 \text{ bar} \leq P_{i,t} \leq 72 \text{ bar}$)

Our Approach for Predictions

Problem: Only actual demands and prices are kept
Have to create our own predictive curves to test the RTO



Step 1: Create Reduced Models

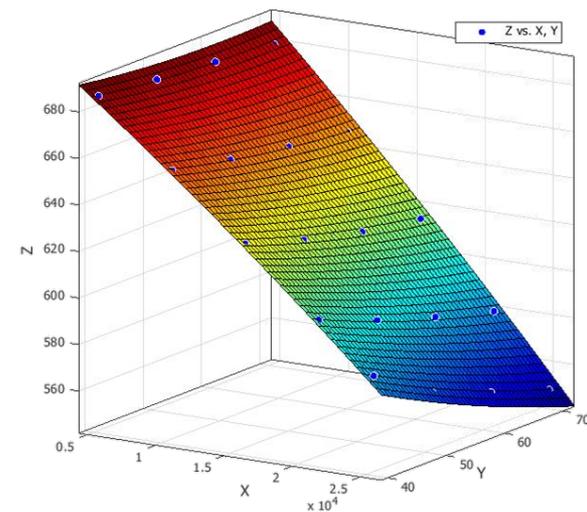
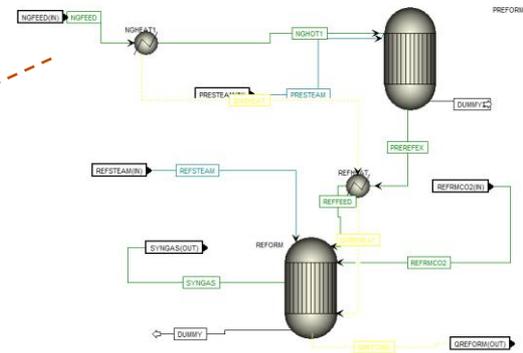
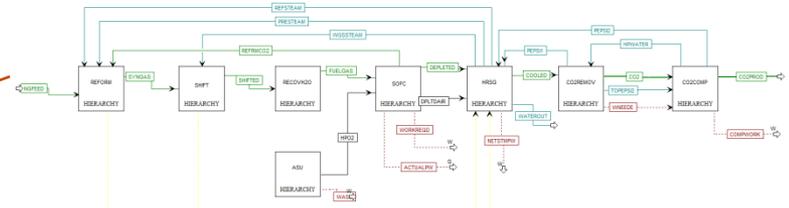
❖ Detailed models in **Aspen Plus**

❖ **Steady-state parts** need only 1 model

❖ **Dynamic parts** modeled with pseudo-steady-state approach:

- 1000s of **Aspen Plus models** for different potential combinations of cavern inlet/outlet flows and cavern pressure.
- **Reduced model for the dynamic system** created by linear-in-the-parameters regression (polynomial basis functions)

❖ **Cavern behaviour modelled separately** using PSRK equation of state



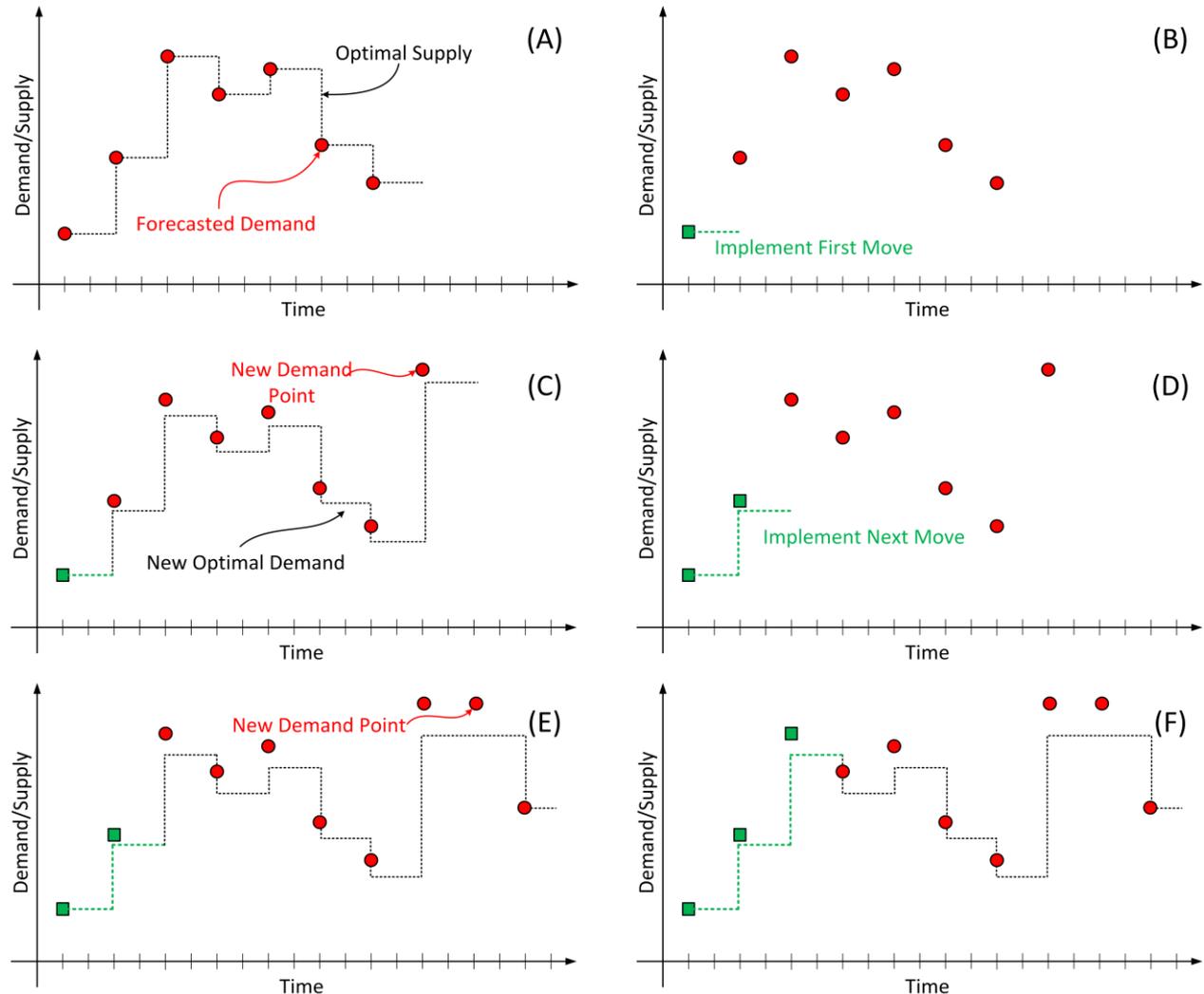
Step 2: Optimize in GAMS

The optimization and reduced models are implemented in **GAMS**

Solved as a **series of 8760 problems**

Once each hour, for the entire year

(Only the first **timestep** result is actually implemented from reach result)



Challenges and Methods

- ❖ Need good initial guesses
 - ❖ **avoid the locally optimal trivial solution:** “don’t use the CAES”
 - ❖ The results from the previous problem used as initial guesses for the next problem.
- ❖ Per each of 8760 problems:
 - ❖ 217 variables (including 143 in nonlinear terms, 24 discrete)
 - ❖ 169 constraints
 - ❖ **1.9 million total variables solved** per “yearlong run”
- ❖ DICOPT → Finds global optimum about 98% of problems
 - ❖ → If DICOPT fails, use BONMIN
 - ❖ → If BONMIN fails, use KNITRO
 - ❖ → BARON was terrible, slower than real time
 - ❖ **Global optimal found in 99.7% of cases** eventually.
 - ❖ Fast enough to use in real time.

Objective 1: Try to Match Profiles

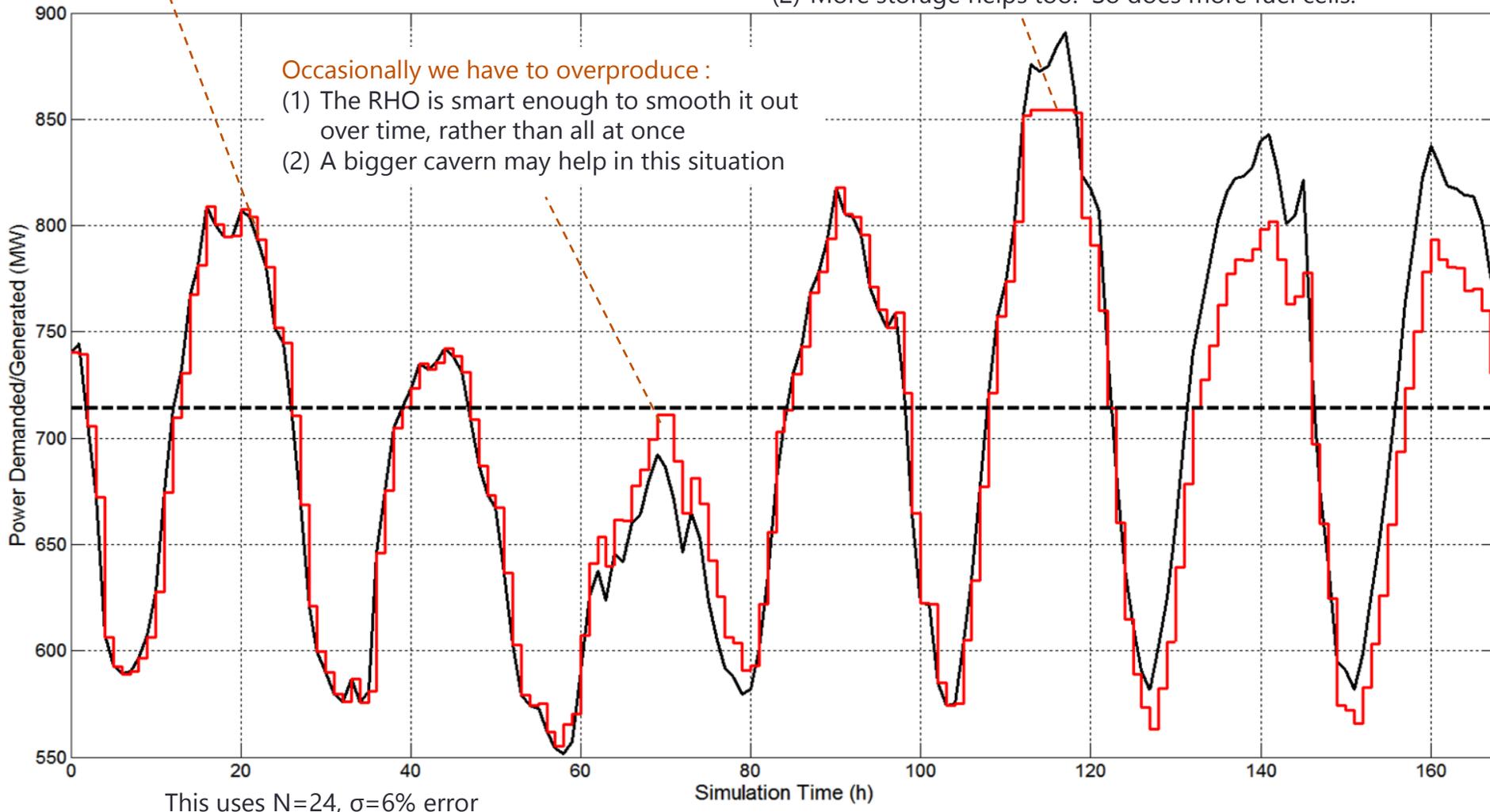
Matching is quite excellent in general, even with uncertain predictions accounted for

Occasionally we underproduce a bit, but:

- (1) That's either less NG firing that's needed, or:
- (2) More storage helps too. So does more fuel cells.

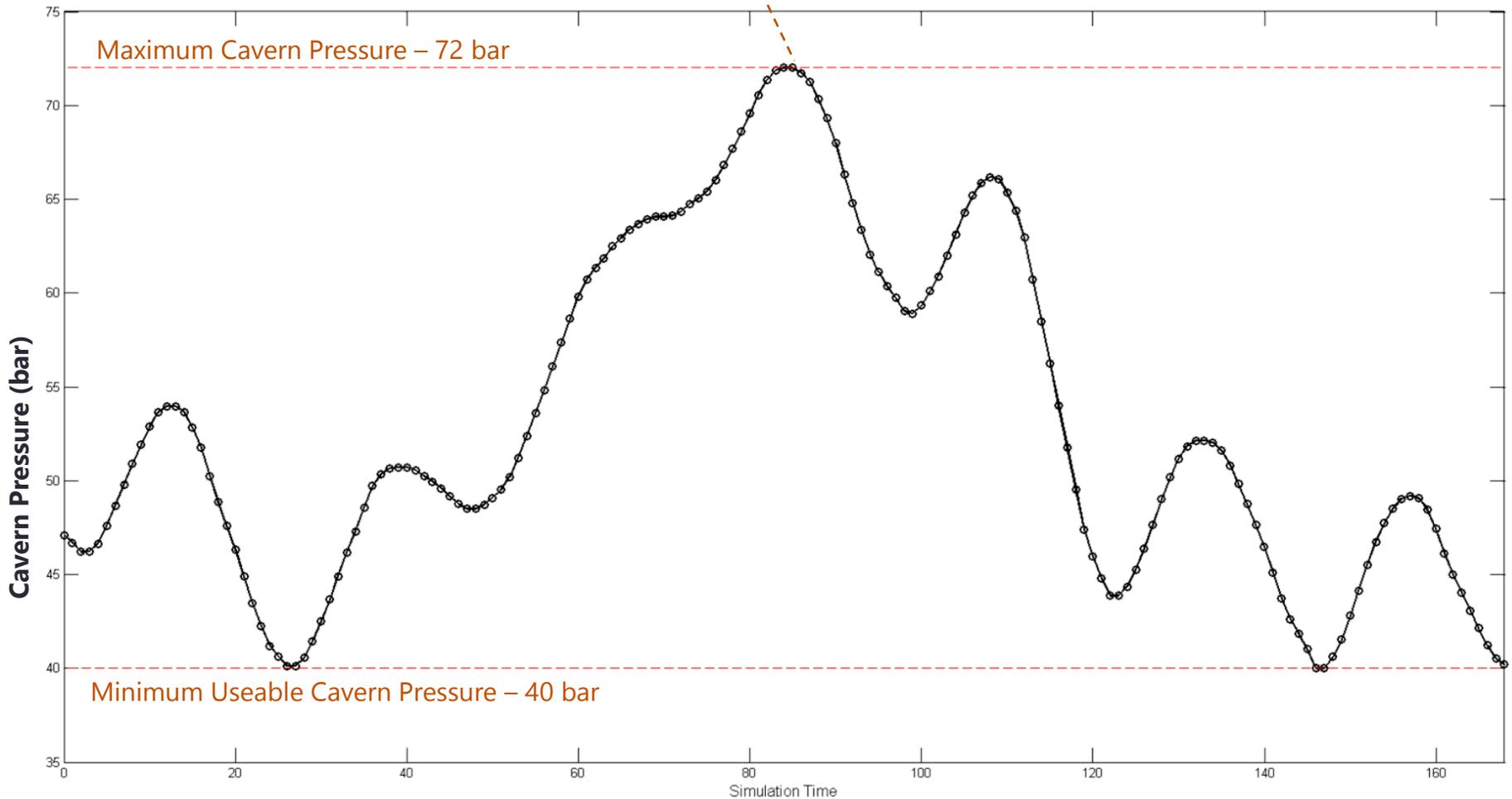
Occasionally we have to overproduce:

- (1) The RHO is smart enough to smooth it out over time, rather than all at once
- (2) A bigger cavern may help in this situation

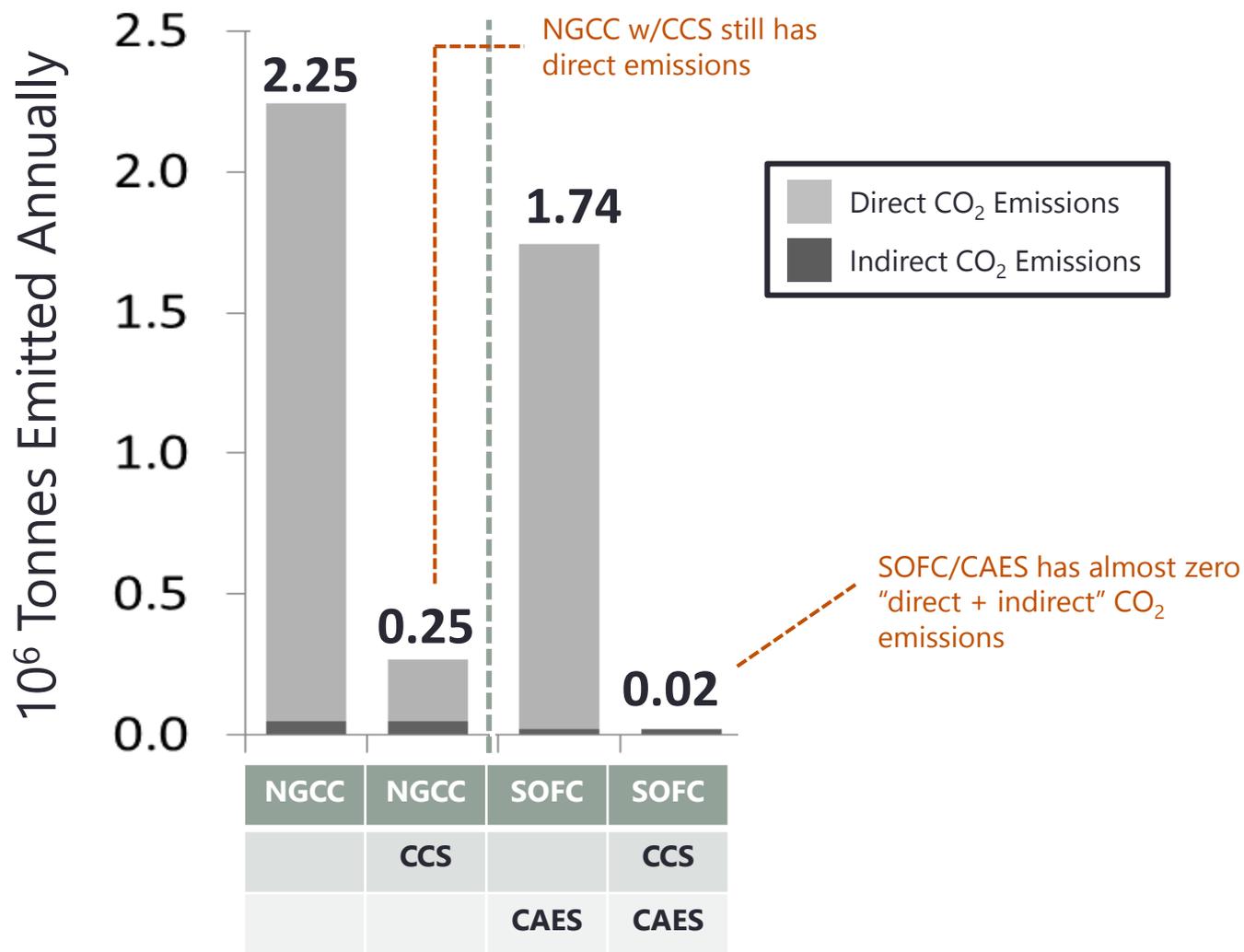


Cavern Pressures

An effective RTO will ensure that the pressure profile only touches the bounds momentarily

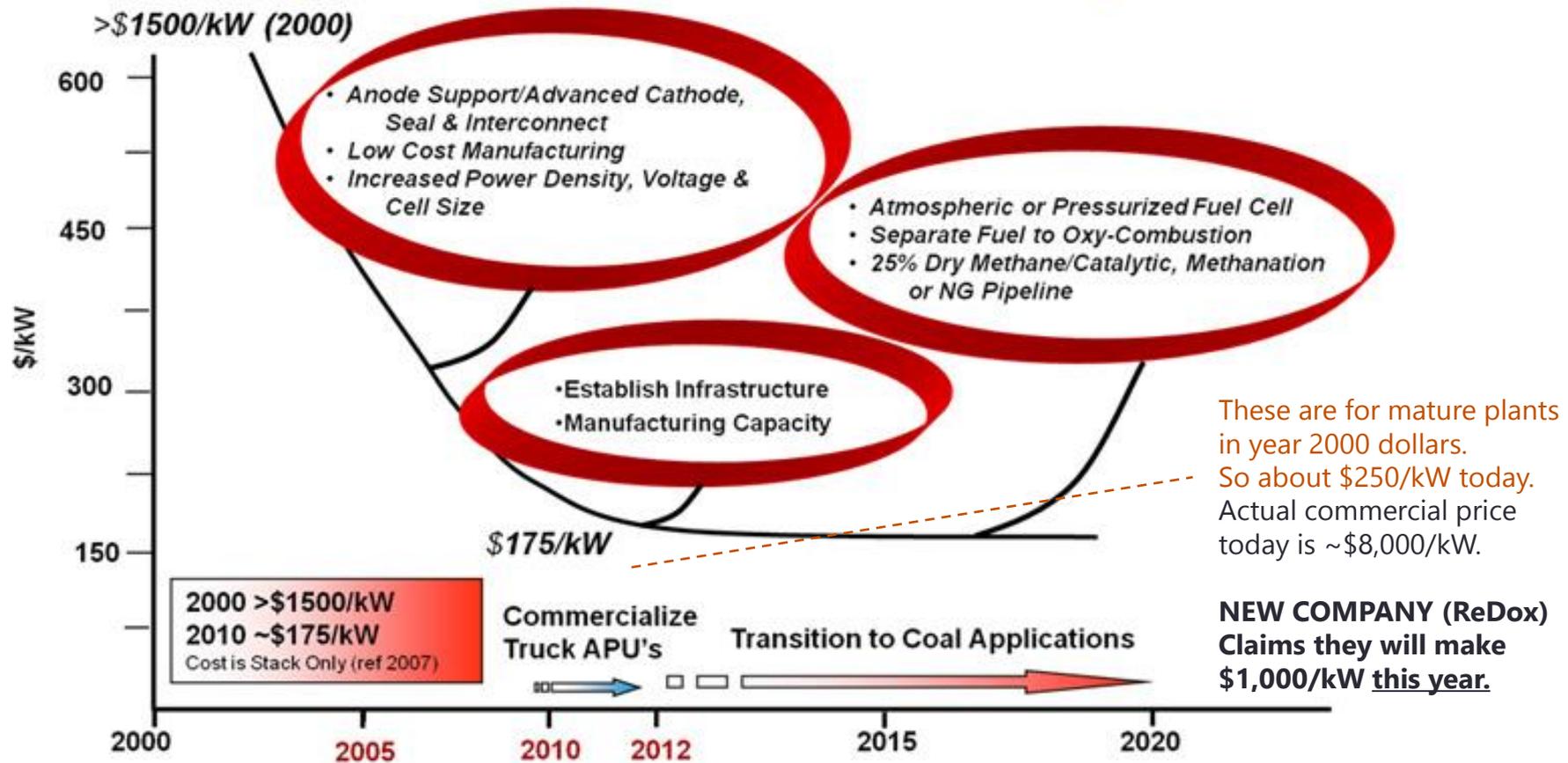


CO₂ Emissions



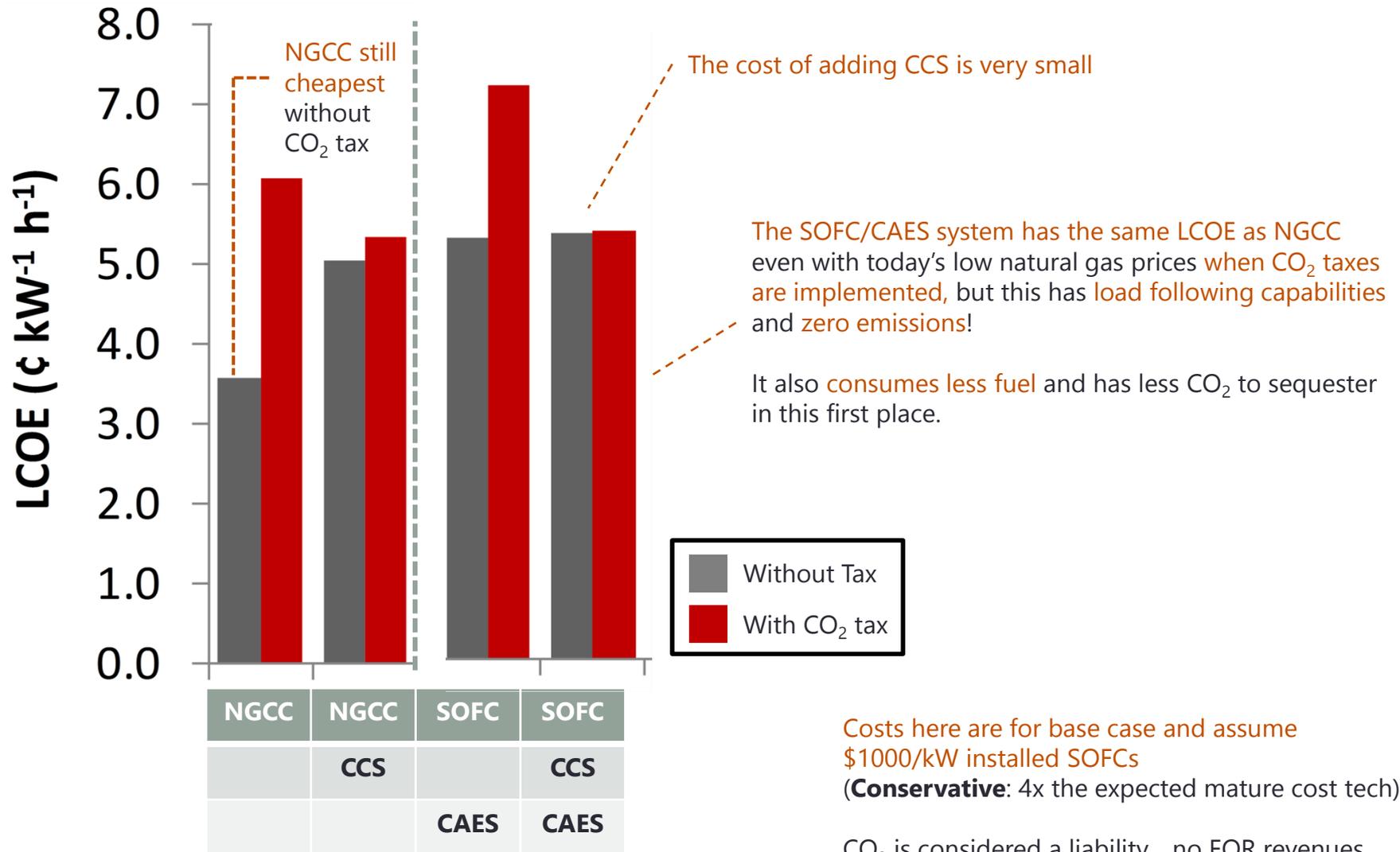
Costs

Driving Down Costs For Fuels Cells (Order of Magnitude Cost Reduction)



Sources: NETL / SECA Homepage. <http://www.netl.doe.gov/technologies/coalpower/fuelcells/seca/>

Levelized Costs of Electricity



Costs here are for base case and assume \$1000/kW installed SOFCs
 (**Conservative:** 4x the expected mature cost tech)

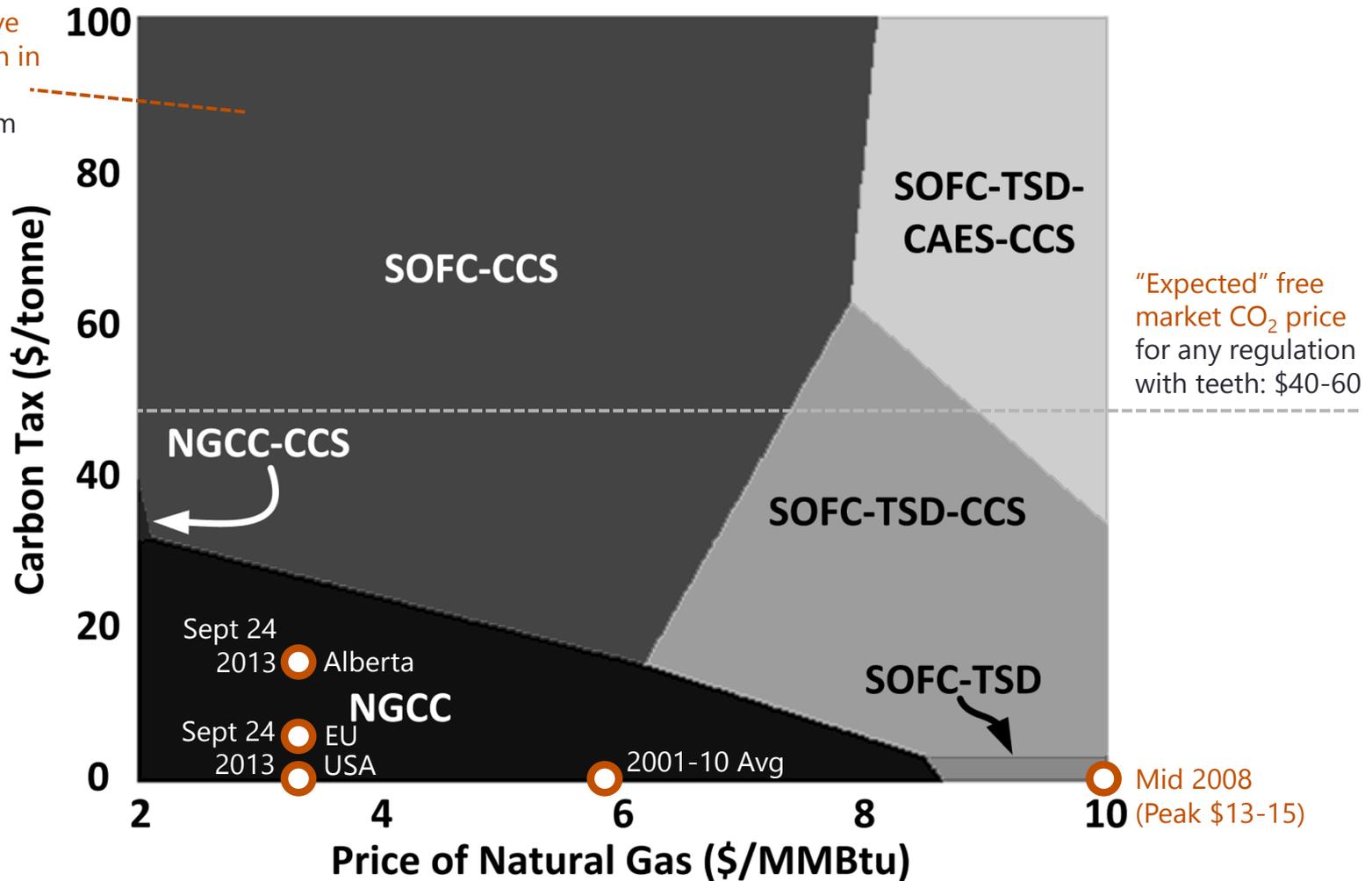
CO₂ is considered a liability... no EOR revenues considered.

Sources: Nease J, Adams TA. J Power Sources, 228:281-293 (2013)

Market Impacts

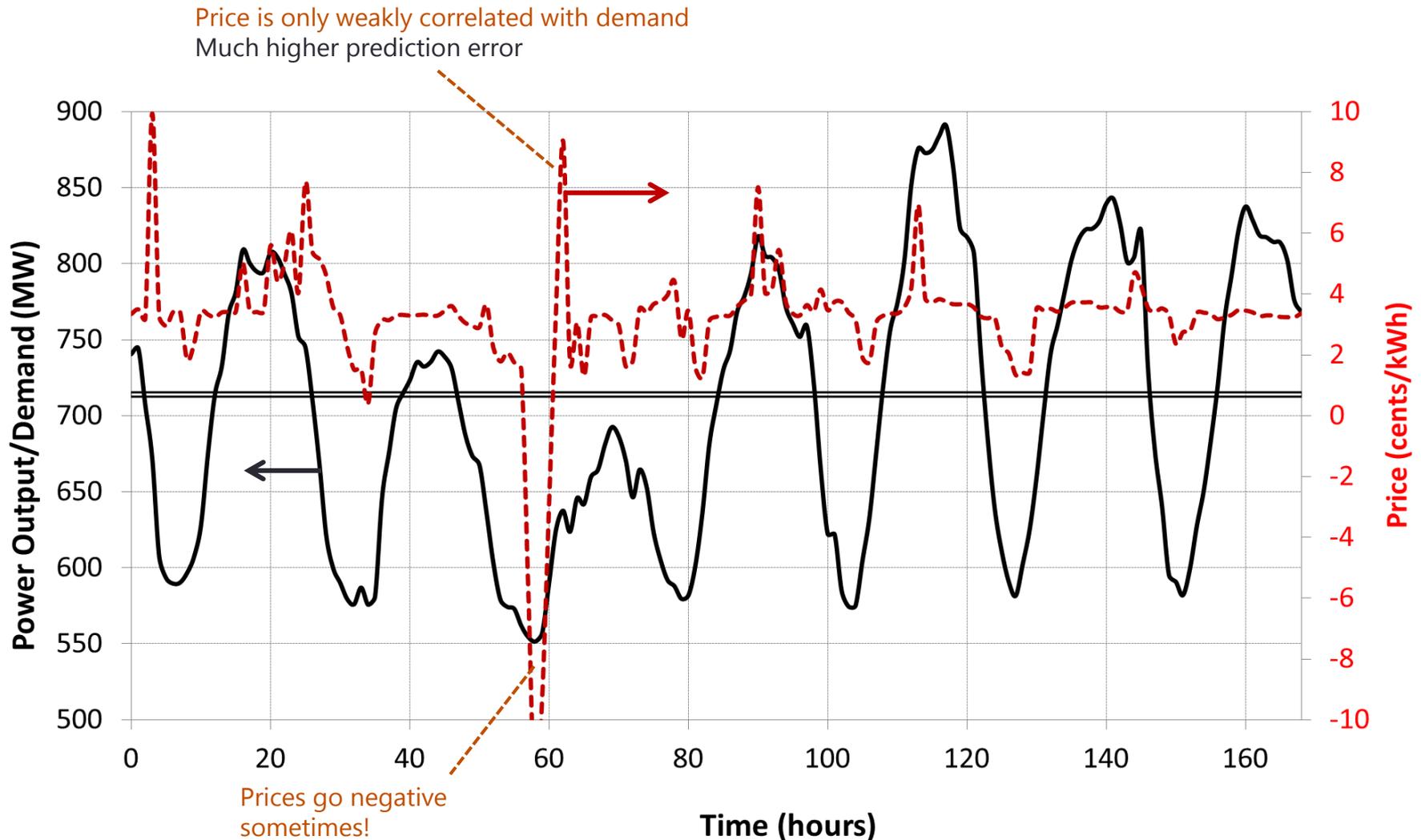
Lowest LCOE depending on fuel price and CO₂ tax

CAES more expensive by 0.08 – 0.3 ¢/kW-h in this area
(Small price premium for flexibility)



Sources: Nease J, Adams TA. J Power Sources, 228:281-293 (2013)
Prices from Offsetters.ca & Bloomberg.com

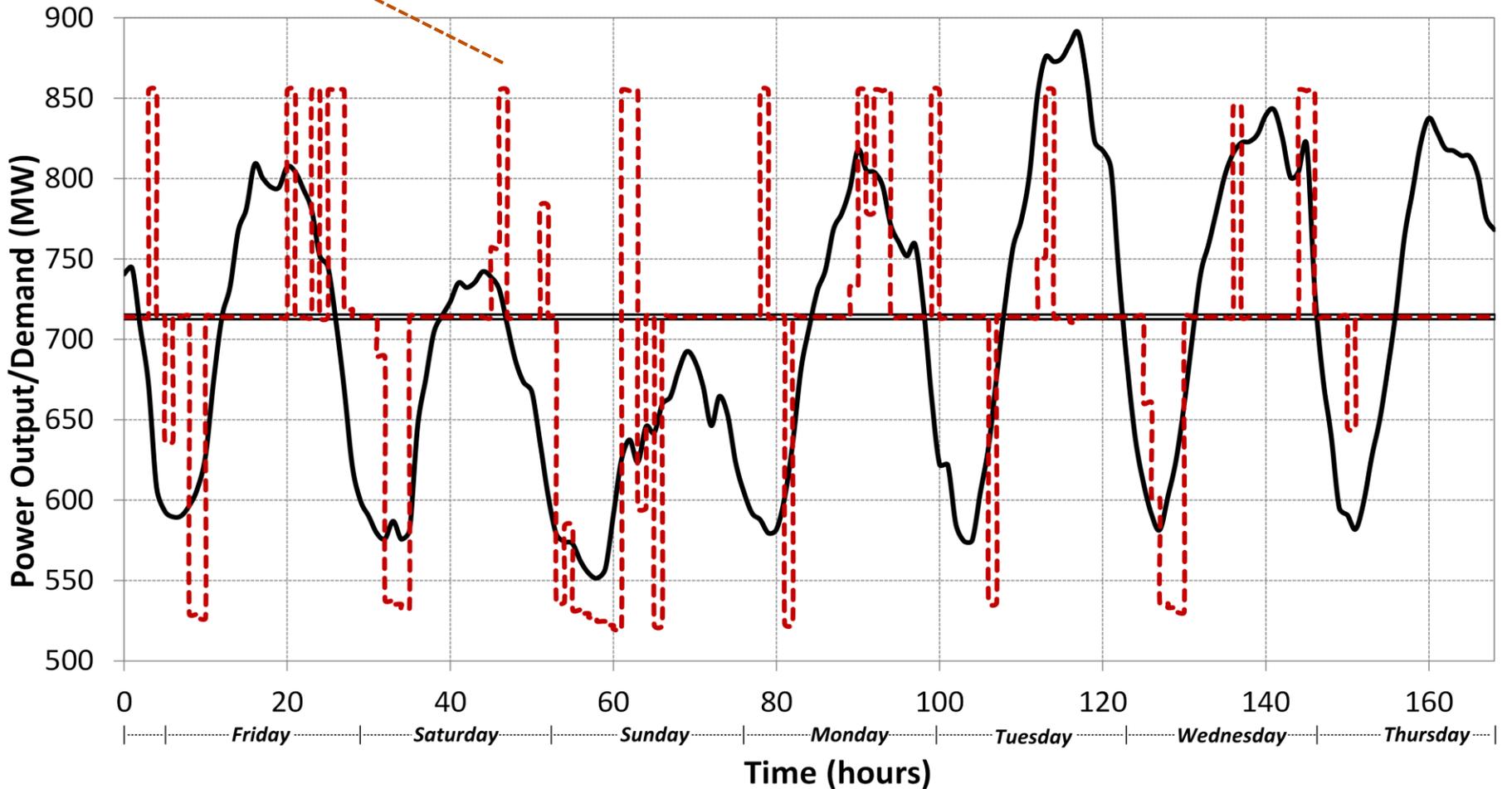
Objective 2: Maximize Profits



Objective 2: Maximize Profits

Very weakly correlated with demand
Almost all sudden swings between store and release

Only about 4% revenue increase
Does not justify the cost of building CAES for purely economic purposes.



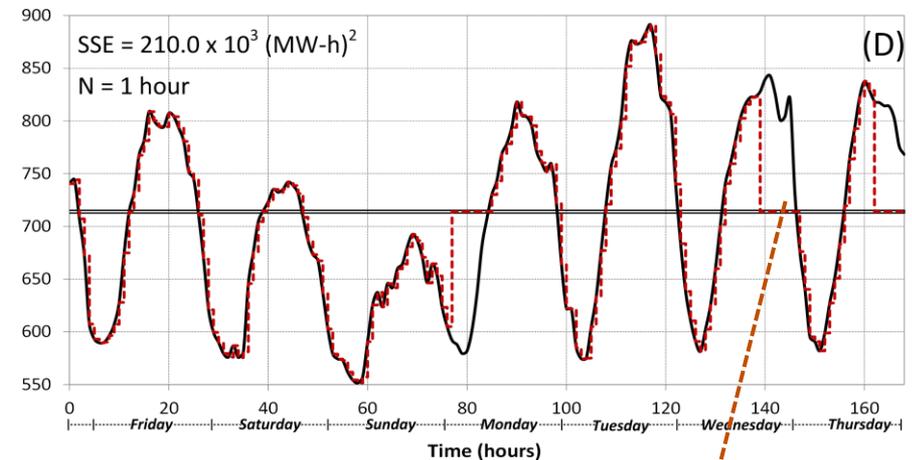
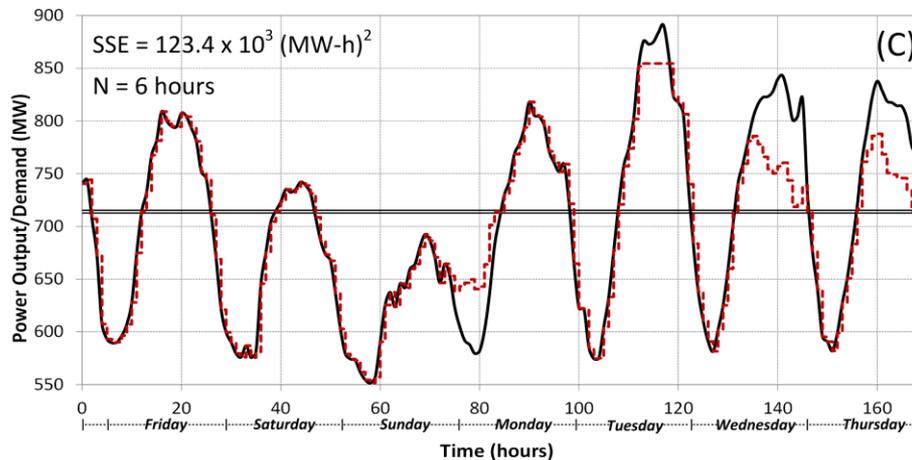
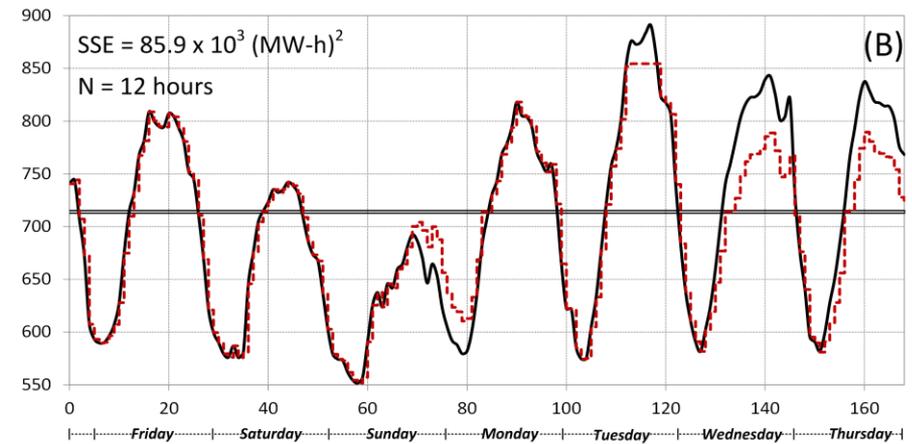
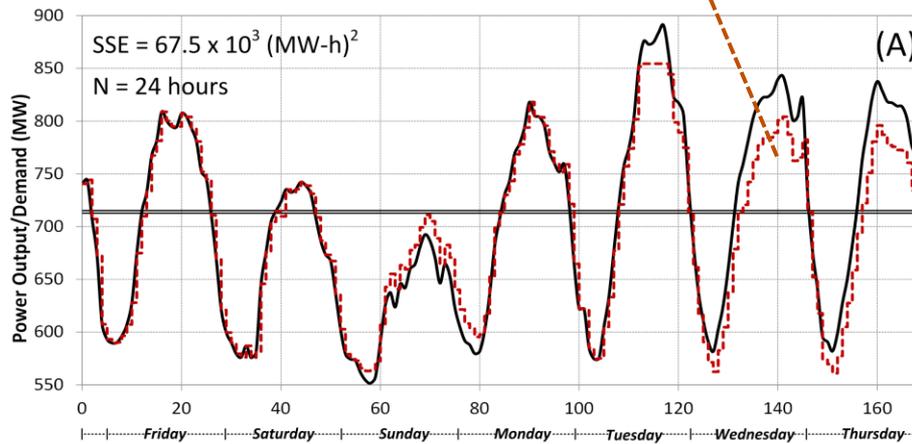
Base Load Output

Power Demanded

Integrated SOFC/CAES Output

Examples: Effect of Prediction Horizon

The bigger your prediction horizon
The smoother the curves



Base Load Output
 Power Demanded
 Integrated SOFC/CAES Output

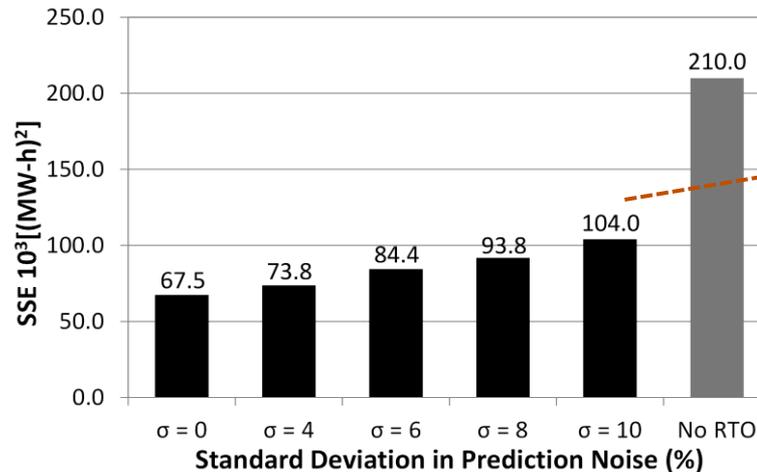
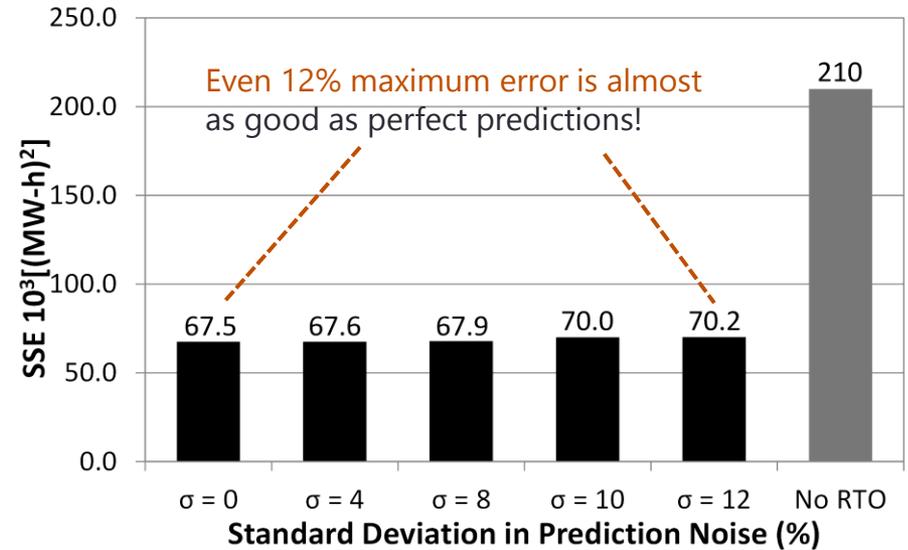
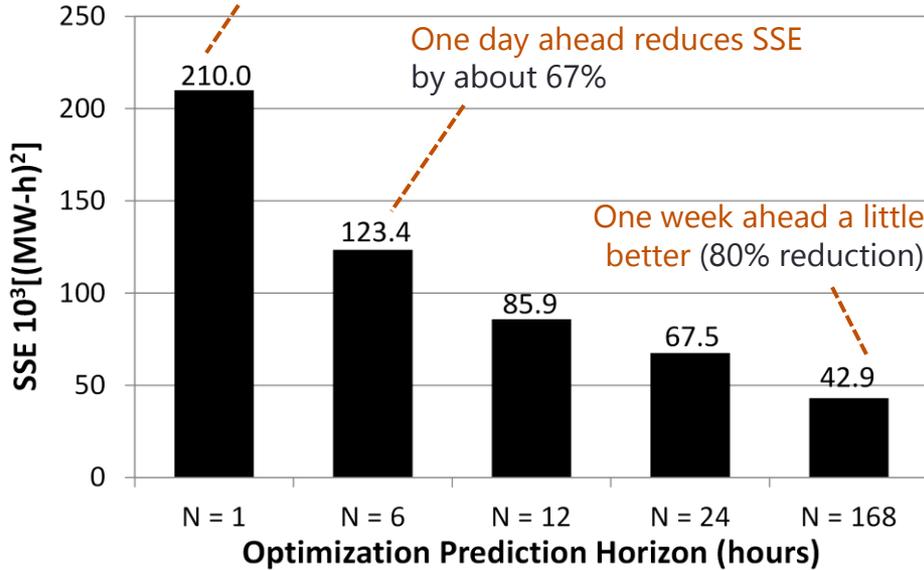
With no horizon, we experience sudden shutoffs due to lack of cavern pressure.

Sensitivity Analysis

Monte Carlo Methods:

Ran RHO repeatedly for 1000s of different random instances of the prediction errors over an entire year

Greedy algorithm: Just try to match the current load



Worst case: even when we always under-predict demand, it is still very good!

Is this actually better for the Earth?

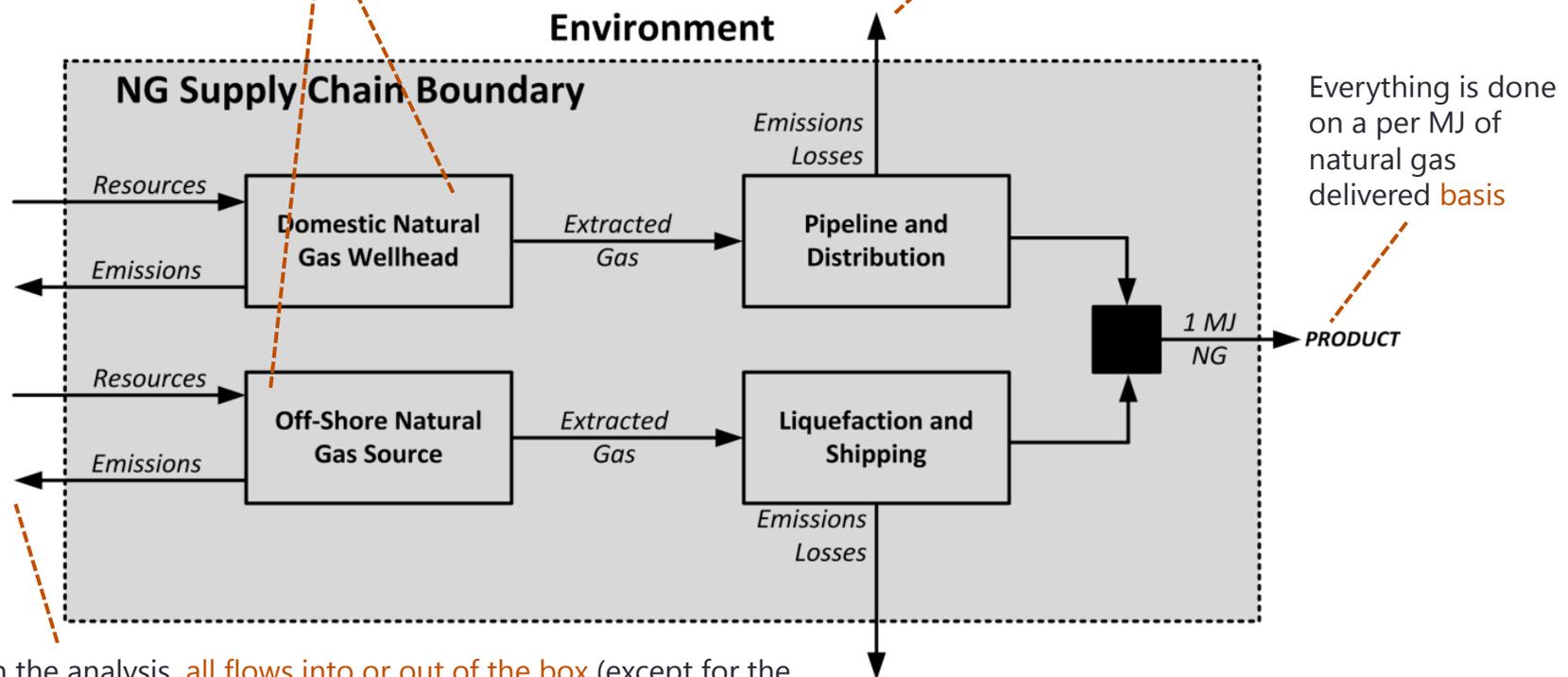
- ❖ So far, **this looks great!**
 - ❖ We can **hugely reduce water consumption**
 - ❖ **Remove almost all CO₂ emissions** from all power production,
 - ❖ We can **load follow** very effectively
 - ❖ We can do it all with only a **small price premium!!!**
- ❖ *But:*
 - ❖ Do we **cause other kinds of problems** instead?
 - ❖ What about the **rest of the supply chain**?
 - ❖ Is making SOFCs so bad that it **counteracts all of the global warming benefits**?
- ❖ So **how do we know if is actually better** for the Earth?
- ❖ Solution:
 - ❖ The ReCiPe **Life Cycle Analysis** methodology.

Step 1: Cradle-to-Grave Inventories

Determine how much comes in and out of your box for the entire supply chain. **Simplified example for natural gas production:**

Considered the US average mix of conventional, gas unconventional gas, shale gas, and LNG imports.

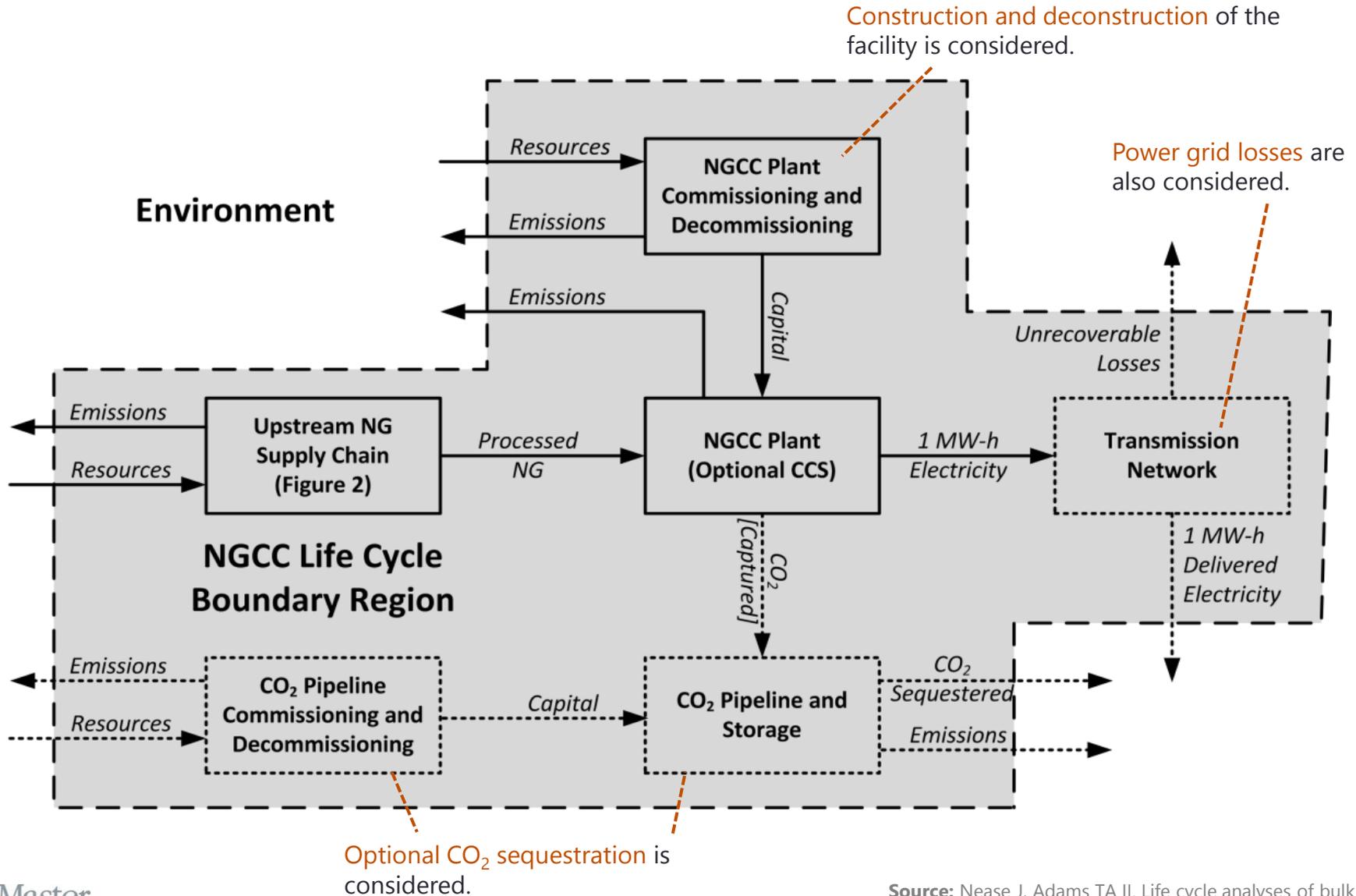
Methane leaks during transport a major source of global warming emissions.



In the analysis, all flows into or out of the box (except for the final delivered electricity) are either direct emissions to the air, water, soil, or resource pool, or direct removals from the air, water, soil or resource pool.

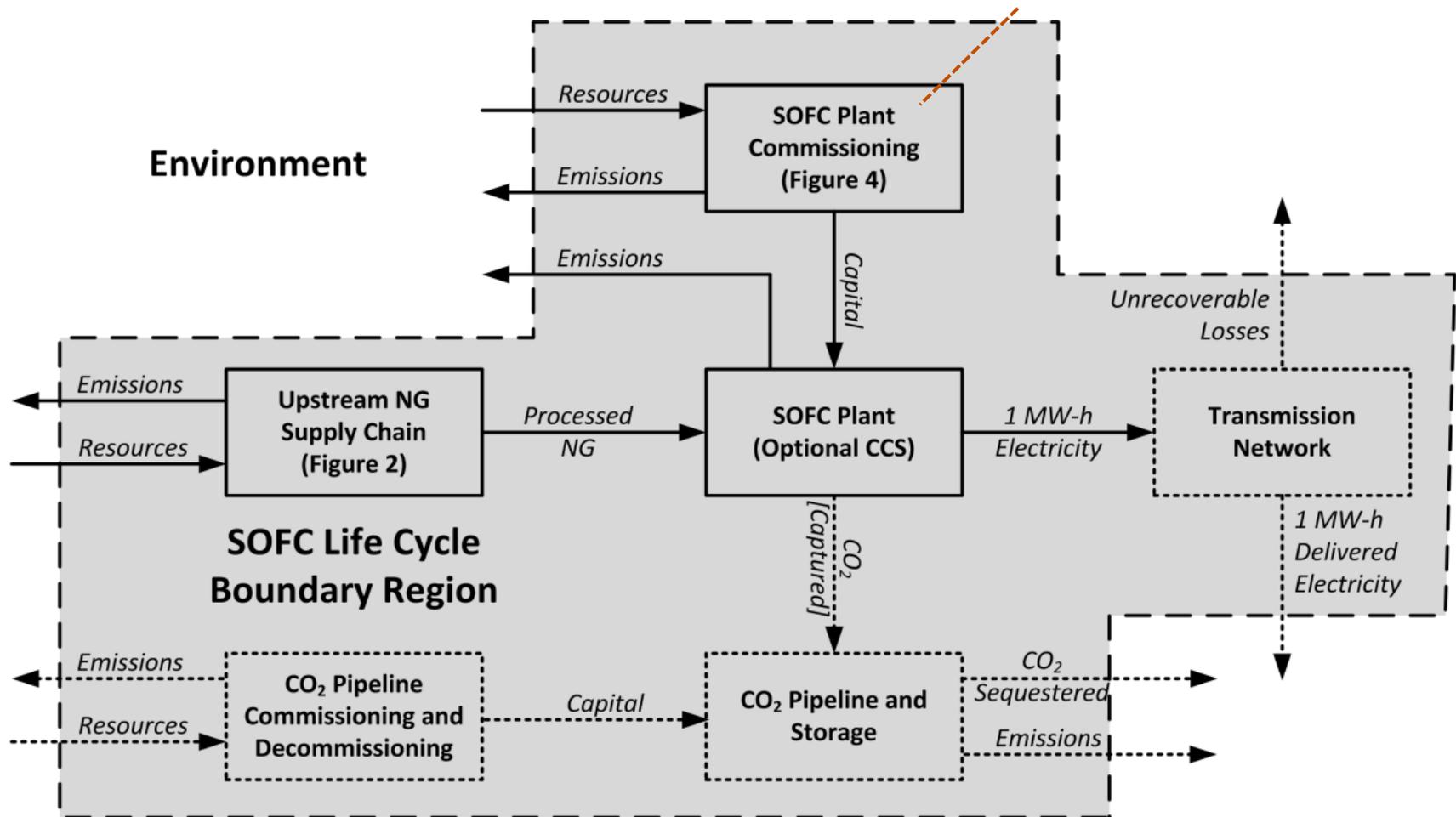
Source: Nease J, Adams TA II. Life cycle analyses of bulk-scale solid oxide fuel cell power plants. (in preparation 2014)

Boundaries for NGCC



Boundaries for SOFC

Basically the same boundaries, except the SOFC plant construction is a lot more impactful (short lifetime for cells... need to by them more often)

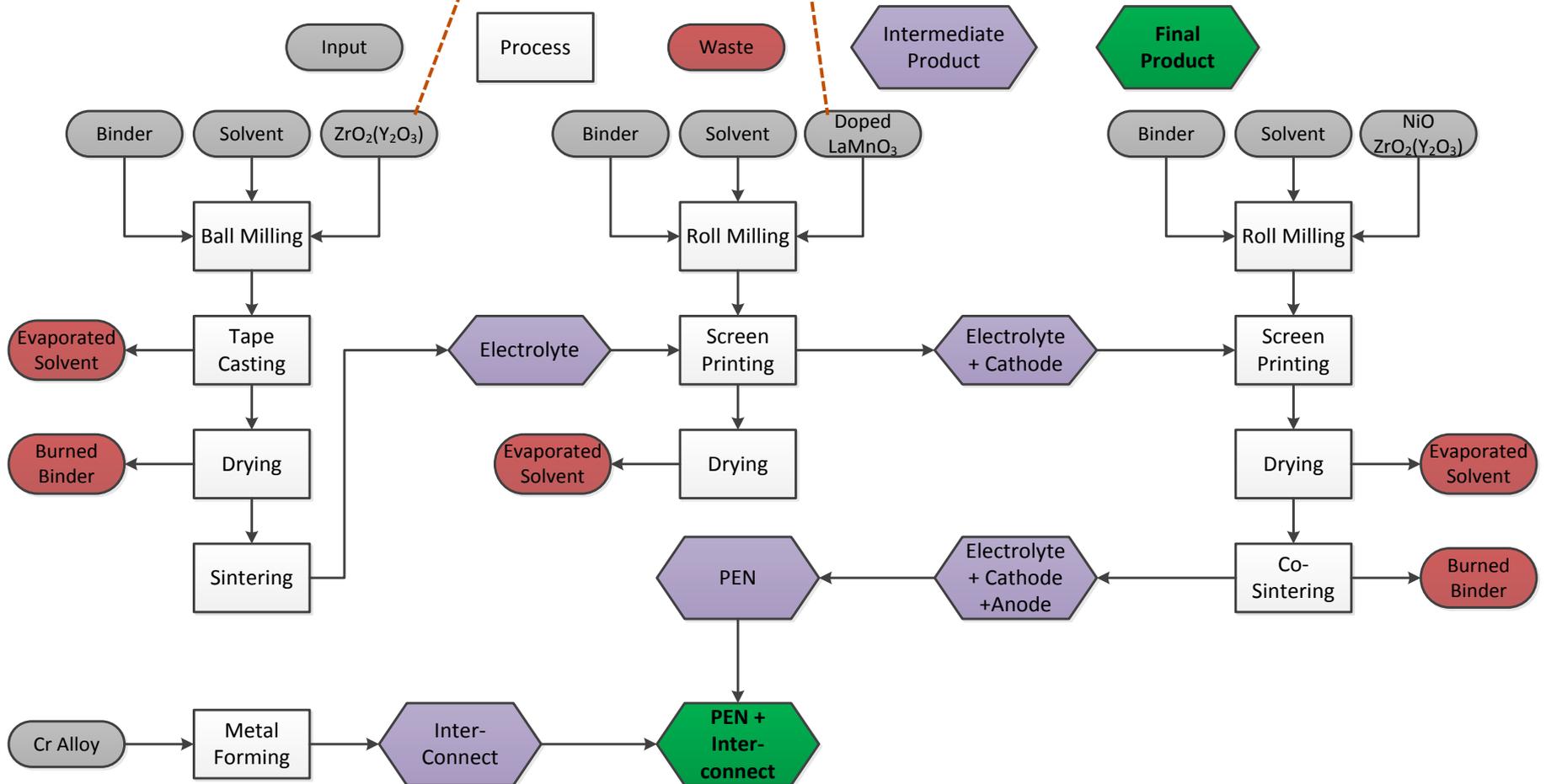


Source: Nease J, Adams TA II. Life cycle analyses of bulk-scale solid oxide fuel cell power plants. (in preparation 2014)

SOFC Stack Construction

Lots of details are factored into these boxes.

Here the difficult to get materials could contribute to large environmental impacts.



Cradle-To-Grave Life Cycle Inventory

Once we have constructed the boundaries, we get a nice table showing **everything that comes from the environment**, and **everything that goes out to it**, and where.

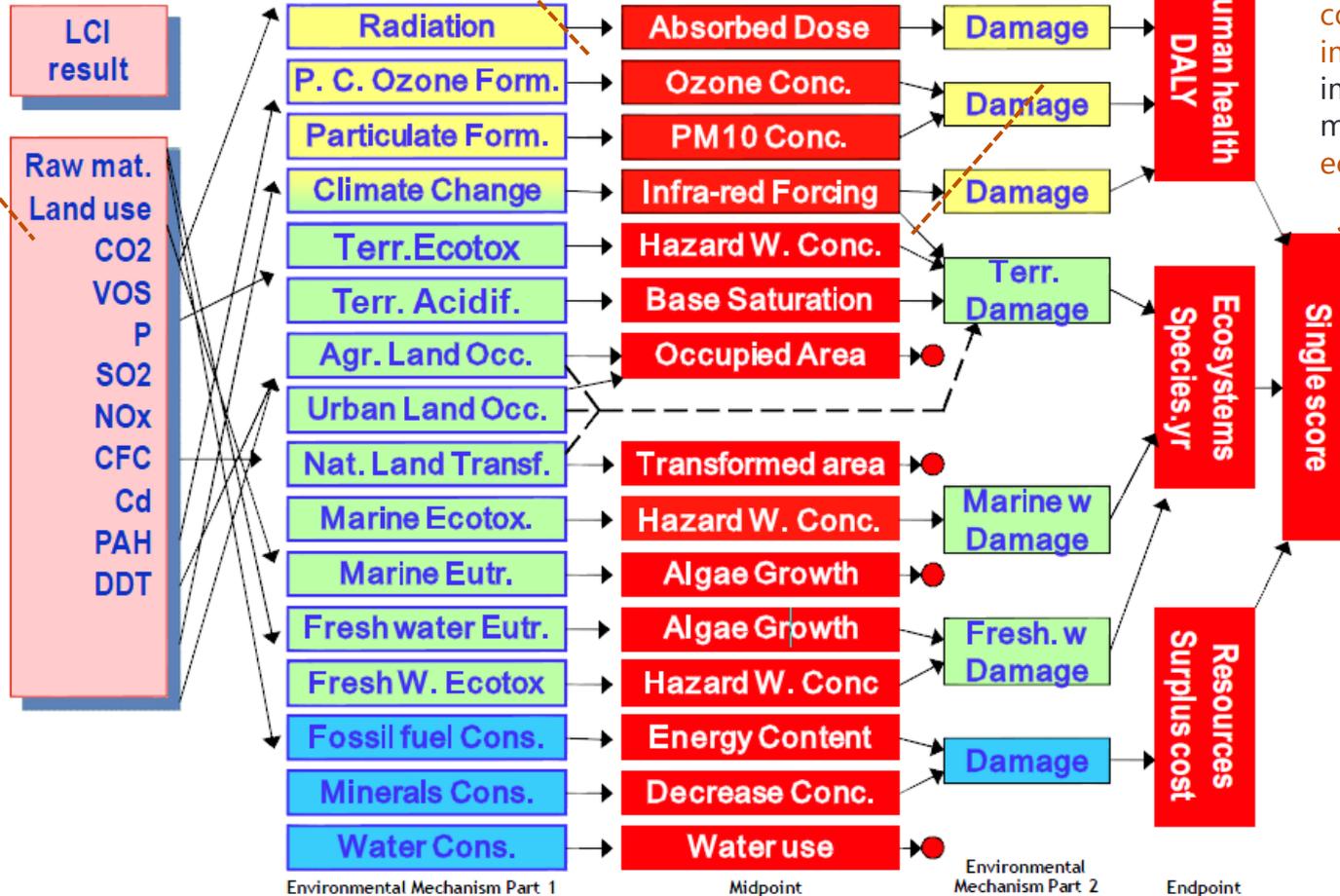
Inventory	NGCC	NGCC w/CCS	SOFC	SOFC w/CCS	
Input Flows (kg)					
Natural Gas (44.1 MJ/kg)	219.23	235.73	144.80	155.61	(I'm listing only a few things here for space)
Water (unspecified natural origin)	129.64	139.40	84.68	91.00	
Output Flows (kg)					
Emissions to air (kg; unspecified population density and height)					
Ammonia (NH ₃)	0.02	0.02	1.42 × 10 ⁻³	1.53 × 10 ⁻³	
Carbon Dioxide (CO ₂)	74.39	79.99	21.03	22.59	SOFC produces less CO₂ but more Nox and particulates . So what is better?
Carbon Monoxide (CO)	0.11	0.12	0.07	0.07	
Dinitrogen Monoxide (N ₂ O)	7.50 × 10 ⁻⁴	8.06 × 10 ⁻⁴	4.81 × 10 ⁻⁴	5.17 × 10 ⁻⁴	
Lead (Pb)	4.32 × 10 ⁻⁶	4.64 × 10 ⁻⁶	2.95 × 10 ⁻⁴	3.15 × 10 ⁻⁴	
Mercury (Hg)	1.02 × 10 ⁻⁷	1.09 × 10 ⁻⁷	9.53 × 10 ⁻⁷	1.02 × 10 ⁻⁶	
Methane (CH ₄)	3.10	3.33	4.58 × 10 ⁻⁸	4.92 × 10 ⁻⁸	
Nitrogen Oxides (NO _x)	0.43	0.47	2.05	2.20	
NMVOC (non-methane volatile organics)	0.02	0.03	0.27	0.28	
Particulates > 2.5 μm and < 10 μm	0.01	0.01	0.02	0.02	
Sulfur dioxide (SO ₂)	0.02	0.02	3.39 × 10 ⁻³	3.64 × 10 ⁻³	
Product Flows (MW-h)					
Electricity Delivered, AC, Grid Quality	1.00	1.00	1.00	1.00	

Step 2: Life Cycle Impact Assessment

These are converted into **midpoints**. Midpoints are **scientific** and **objective** ways of **quantifying** how different chemical affect **the same impact** (like global warming) **with one number**.

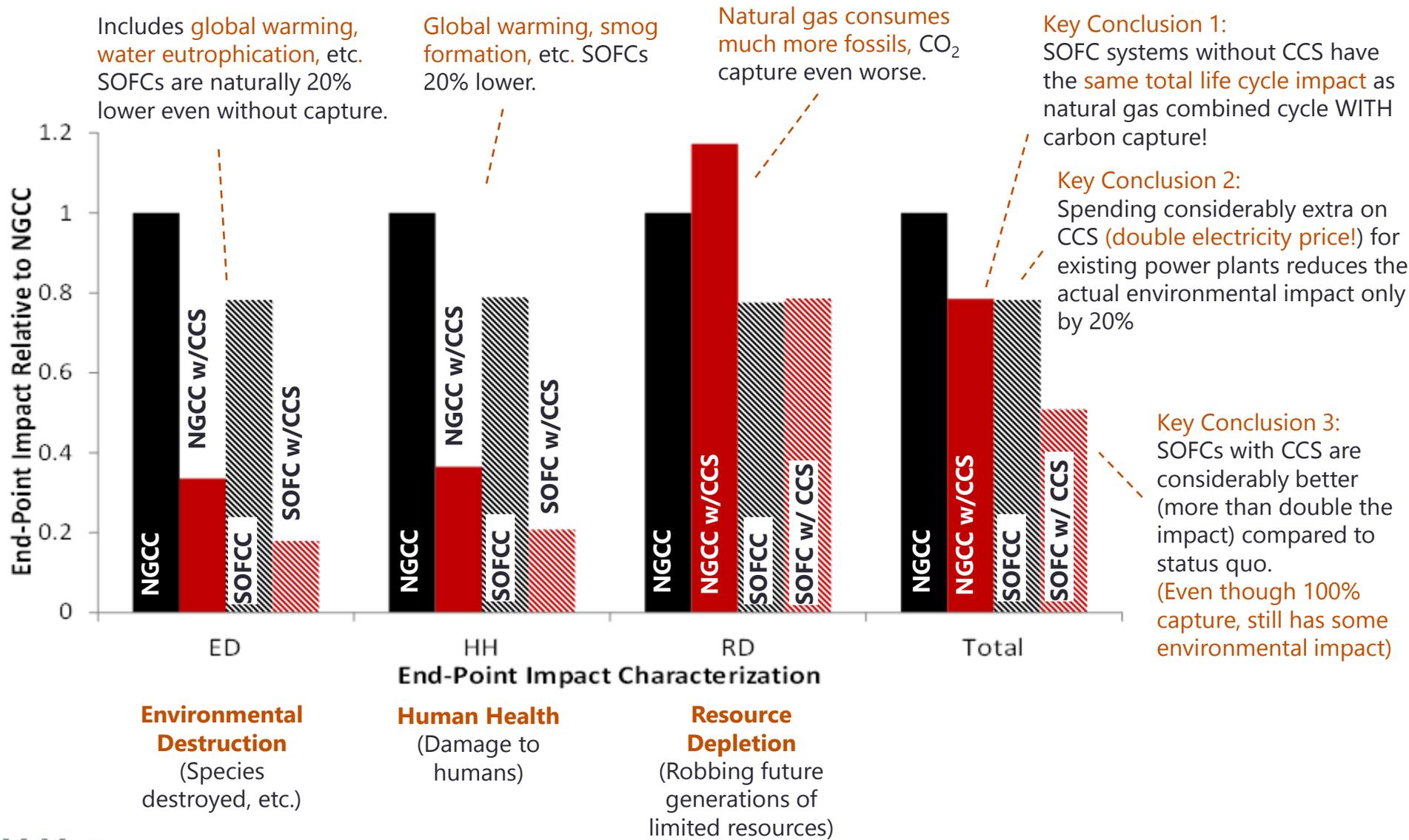
Midpoints are converted into **endpoints**. These are **scientific** but **partially subjective** weightings of how **comparatively important** each impact is. These are measured in **ecoPoints**.

The results of that table become the inputs to the **life cycle impact assessment**



Source: Goedkoop M, et al. ReCiPe 2008: Report I Characterization. (July 2012)

Results:



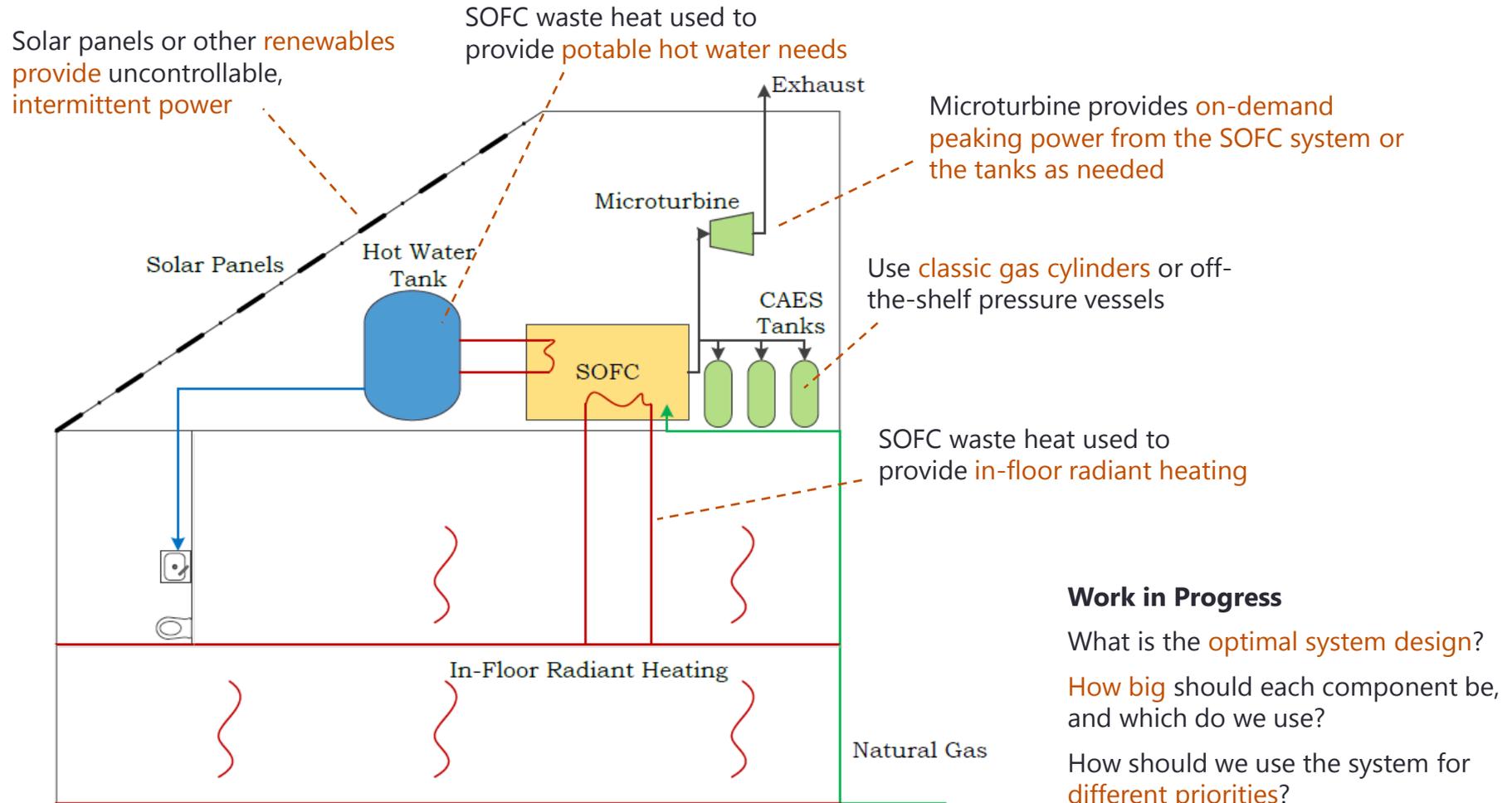


2. BUILDING SCALE

A new “green building” venture.
Student Researcher: Kyle Lefebvre



Building scale SOFC/CAES



Work in Progress

What is the optimal system design?

How big should each component be, and which do we use?

How should we use the system for different priorities?

- ❖ Economic objective?
- ❖ Environmental objective?
- ❖ Mix of two?

Case study for ExCeL building...

McMaster's "ExCEL" Building

Solar Panels on Roof
Direct DC circuits in walls ***

Grid connection for sale-back of excess power
(Or draw for additional power)



Integrated Power, Heat, Water Systems
***Experimental energy storage systems
***Experimental building-scale power generation

Pilot plant would:

- ❖ Demonstrate **first SOFC/CAES system**
- ❖ Provide **model validation** opportunities
- ❖ RHO uses real time occupancy/weather data

- ❖ Be adjustable for different "buildings" for different climates
- ❖ Integrate with subsets of other energy systems (**geothermal, solar, hot water, in-floor heating, steam-heating systems**) in order to experiment with different types of green buildings



3. MID-LEVEL SCALES

Medium term impacts.

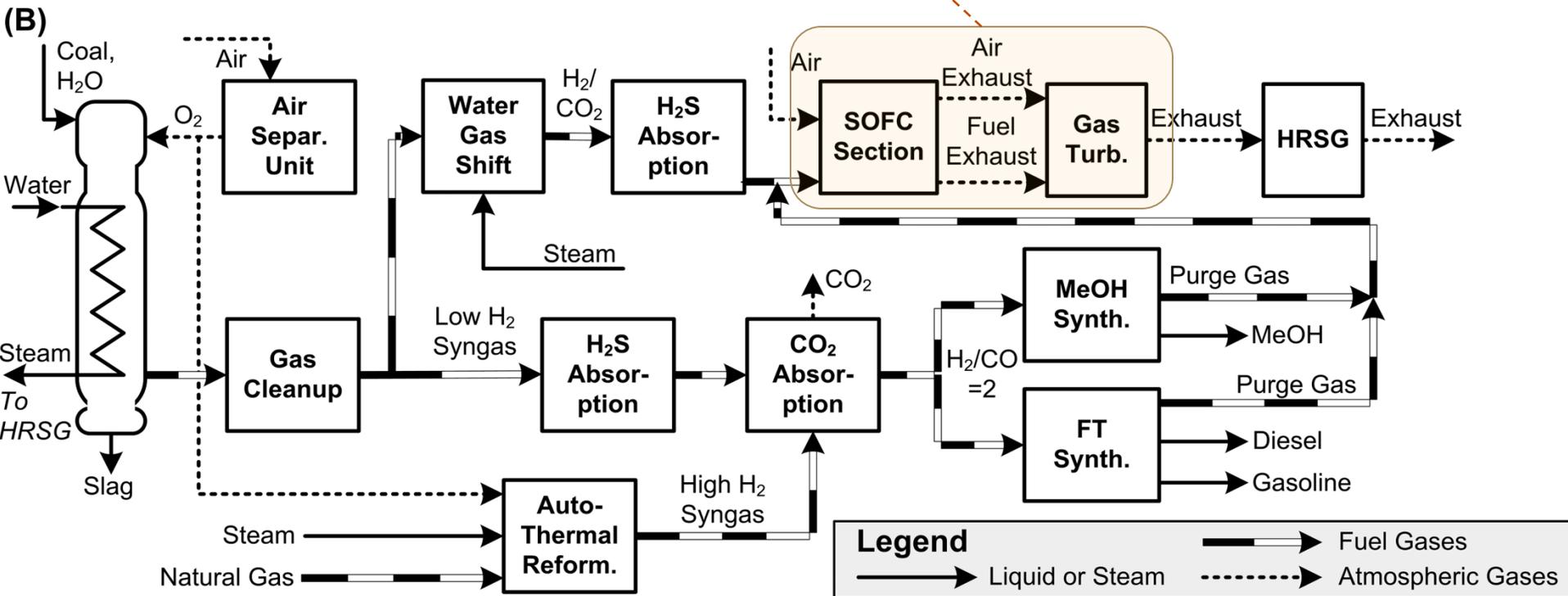
Student Researcher: Nor Farida Harun



SOFC/Gas Turbines (Medium Term)

This section exists as combined hardware software simulator

- (1) Real turbine, combustor, compressors, control, and heat exchange
- (2) Real-time simulated SOFC (1D spatial-temporal model)
- (3) Real SOFC exhaust gases generated based on model results in real-time



Current Team



Polygeneration Team



SOFC Systems Team



Optimization Team

Yaser Khojestah
PhD
Gas, Coal, and Nuclear to Liquids

Jaffer Ghouse
PhD
Integrated Coal & Gas Design

Dominik Seepersad
Master's
Integrated Coal & Gas Control

Jake Nease
PhD
SOFCs with Energy Storage

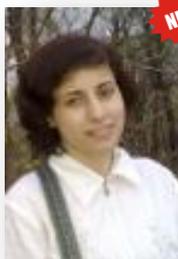
Farida Harun
PhD
Flexible Fuel SOFCs

Kyle Lefebvre
Master's
Building-scale SOFC Systems

Giancarlo Dalle Ave
Master's (Sept)
PBBBBB



Sustainable Energy Systems



Chinedu Okoli
PhD
Thermochemical BioButanol

Haoxiang Lai
Undergrad
Energy Storage for Concentrated Solar

Leila Hoseinzadeh
Research Associate
Waste Flare Gas to Butanol

Kalia Akkad
Undergrad
Tailing Pond Reduction

Vida Medianshahi
PhD
Novel SC Design & Control

Kushlani Wijesekera
Master's
Ultra-intensified SC

Sarah Ballinger
Master's (Sept)
PBBBBB

