

# CCSI

Carbon Capture Simulation Initiative

## Advanced Computational Tools for Optimization and Quantification of Uncertainty

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U.S. Department of Energy

National Energy Technology Laboratory

Pittsburgh, PA

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U.S. DEPARTMENT OF  
**ENERGY**

# Acknowledgements

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# Outline

- **DOE/FE Mission & Carbon Capture Challenge**
- Carbon Capture Simulation Initiative
- Optimization & Quantification of Uncertainty
- Summary





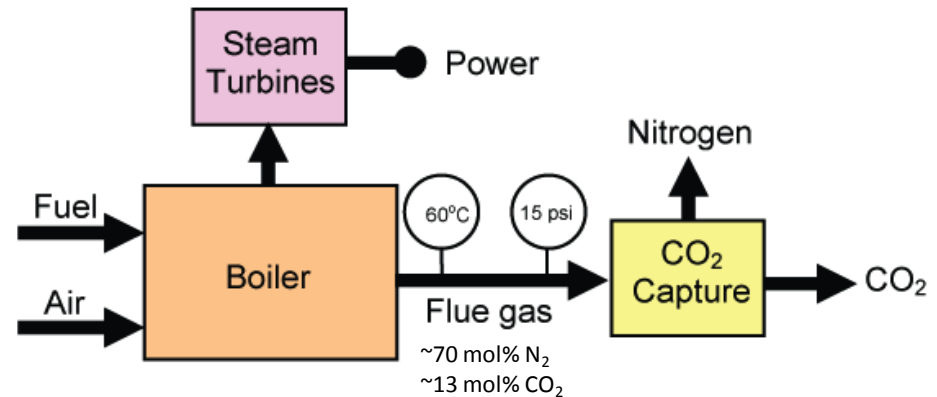
# Mission

- RD&D of cutting edge fossil energy technologies to ensure a secure, affordable, low-carbon energy future.
  - Collaborative partnerships with industry
  - International collaboration
- The Cross-cutting Research activity serves as a bridge between basic and applied research by fostering the development and deployment of innovative systems for improving efficiency and environmental performance.
  - Includes development of computation, simulation, and modeling tools focused on optimizing plant design and shortening developmental timelines.



# Challenges associated with carbon capture

- Fossil energy generates:
  - 67% of the US electricity
    - 42% coal, 25% natural gas
  - 65% worldwide electricity



- Large-scale problem
  - 2 billion tons/year CO<sub>2</sub> emitted from coal by 2020 in US
  - Flue gas: 5 million lb/hr for 550MW PC plant (630 kg/s)
- No existing economical solution
  - Cost of capture plant
  - De-rate of power plant
- Need for system optimization to fully evaluate technology options

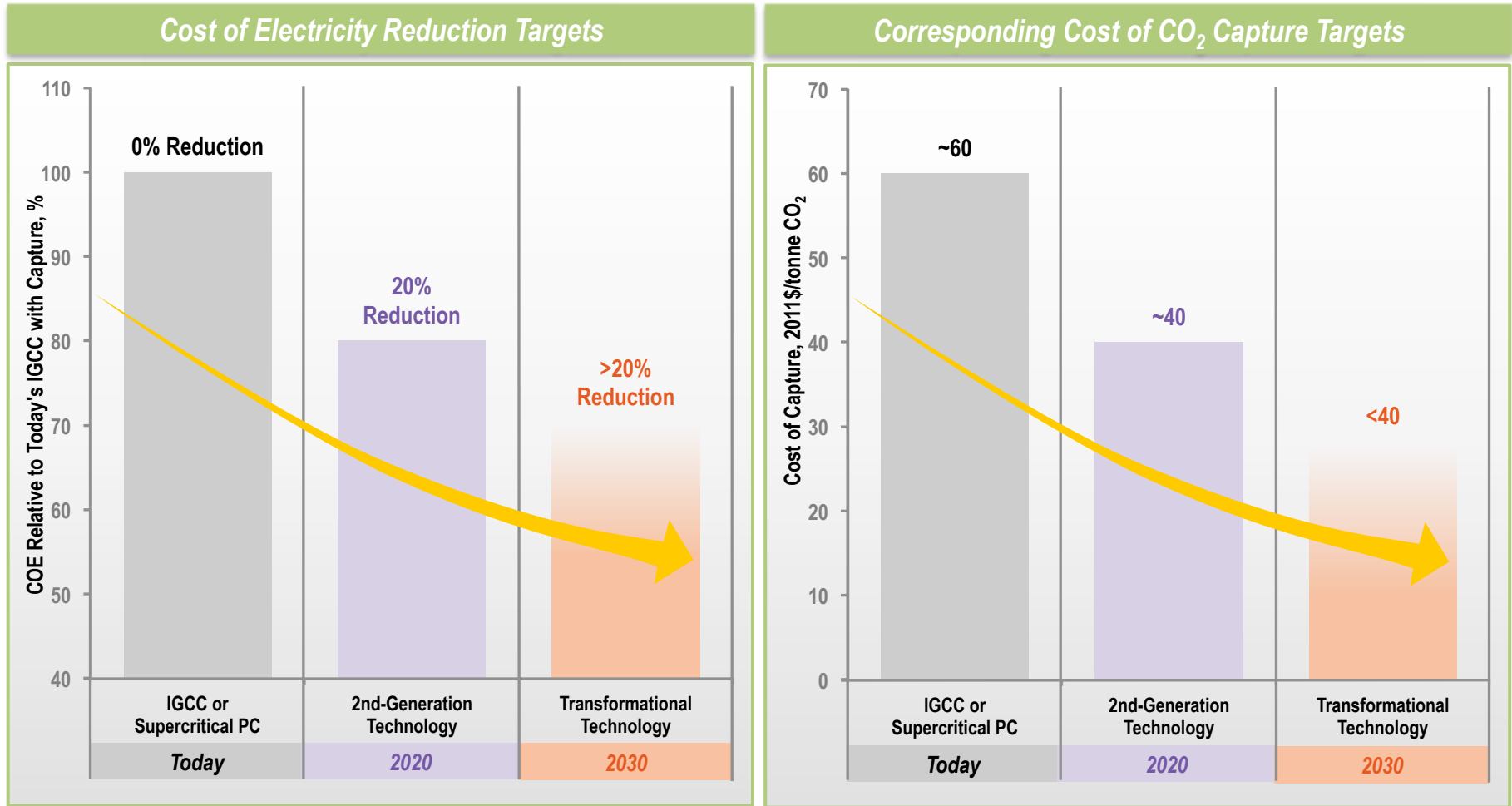
U.S. Energy Information Administration, Annual Energy Outlook 2013 Early Release Overview. Report Number: DOE/EIA-0383ER(2013), <http://www.eia.gov/forecasts/aeo/er/pdf/0383er%282013%29.pdf> (2013).

U.S. Energy Information Administration, International Energy Outlook 2011. Report Number: DOE/EIA-0484(2011), <http://www.eia.gov/forecasts/ieo/pdf/0484%282011%29.pdf> (2011).



# Clean Coal Research Program Goals

*Driving Down the COE and Cost of Coal Power CCS*

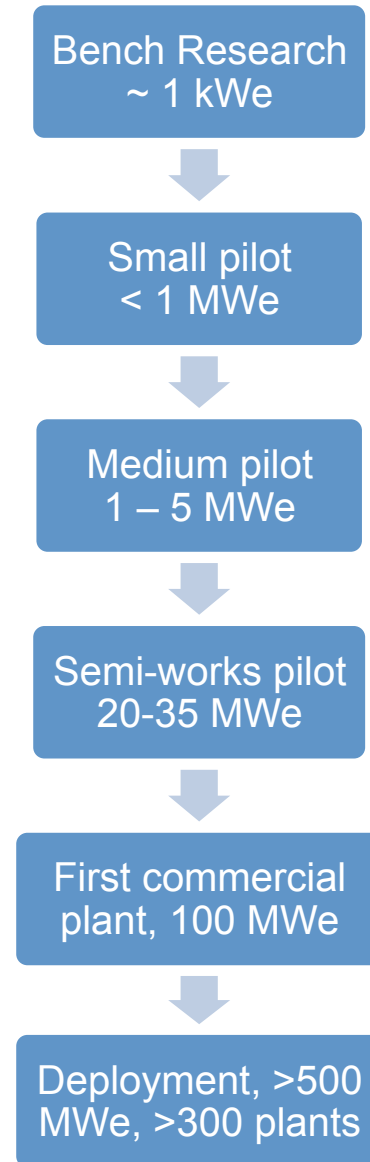


Goals shown are for greenfield plants. Costs are for nth-of-a-kind plants, during first year of plant operation, and include compression to 2215 psia but exclude CO<sub>2</sub> transport and storage costs. Today's capture costs are relative to Today's SCPC without CO<sub>2</sub> capture. 2020 and 2030 capture costs are relative to an A-USC PC without CO<sub>2</sub> capture.

# Development Challenges

- The traditional pathway from discovery to commercialization of energy technologies can be quite long, i.e., **~ 2-3 decades<sup>1</sup>**
- President's plan requires that barriers to the widespread, safe, and cost-effective deployment of CCS be overcome **within 10 years<sup>2</sup>**
- To help realize the President's objectives, new approaches are needed for taking CCS concepts **from lab to power plant, quickly, and at low cost and risk**
- Accelerate the development of CCS technology, from discovery through deployment, with the help of **advanced computations tools and models**

1. International Energy Agency Report: Experience Curves for Energy Technology Policy," 2000. 2. <http://www.whitehouse.gov/the-press-office/presidentialmemorandum-a-comprehensive-federal-strategy-carbon-capture-and-storage>

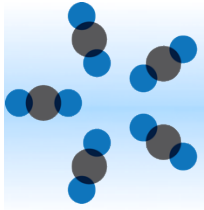


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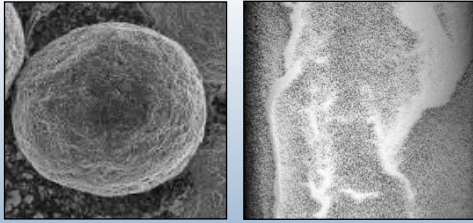




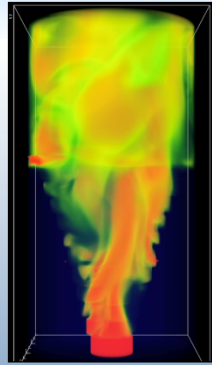


# CCSI | For Accelerating Technology Development

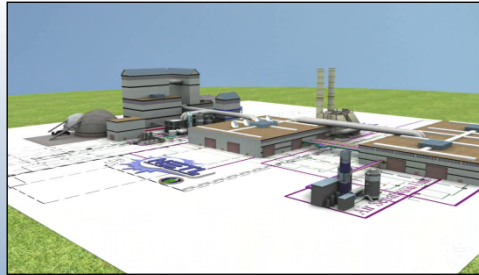
Carbon Capture Simulation Initiative



Rapidly synthesize optimized processes to identify promising concepts



Better understand internal behavior to reduce time for troubleshooting



Quantify sources and effects of uncertainty to guide testing & reach larger scales faster



Stabilize the cost during commercial deployment

## National Labs



## Academia



## Industry



# Goals & Objectives of CCSI

- **Develop** new computational tools and models to enable industry to more rapidly develop and deploy new advanced energy technologies
  - Base development on industry needs/constraints
- **Demonstrate** the capabilities of the CCSI Toolset on non-proprietary case studies
  - Examples of how new capabilities improve ability to develop capture technology
- **Deploy** the CCSI Toolset to industry
  - Support initial industry users
  - Obtain feedback on features and capabilities
  - Arrange for long term commercial licensing

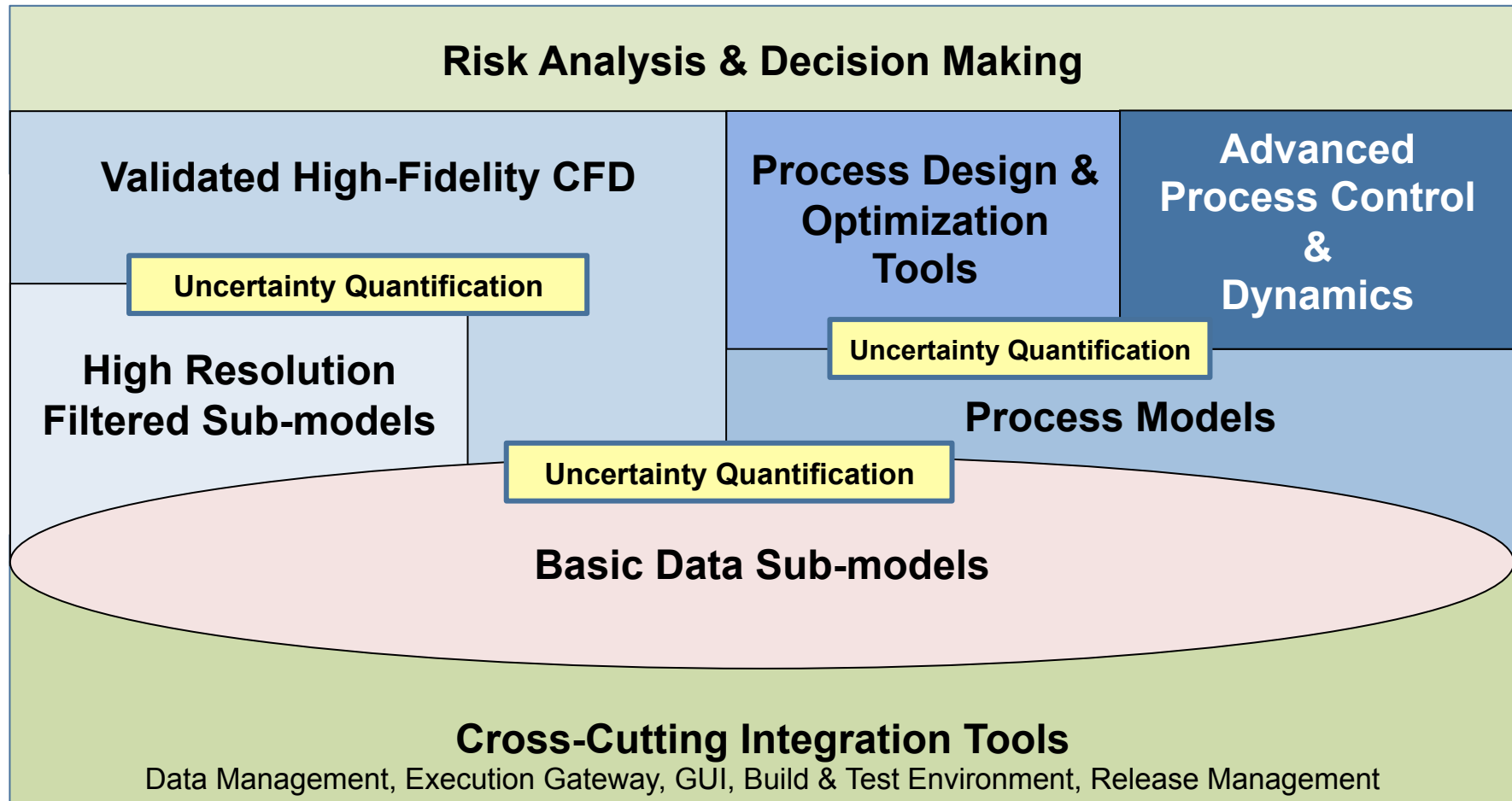


# CCSI Timeline

- Organizational Meetings: March 2010 - October 2010
- **Technical work initiated: Feb. 1, 2011**
- **Preliminary Release of CCSI Toolset: September 2012**
  - Initial licenses signed
- CCSI Year 3 starts Feb. 1, 2013
  - Began solvent modeling/demonstration component
- **2013 Toolset Release: October 31, 2013**
  - **Multiple tools and models released and being used by industry**
- **Future**
  - **2014 Toolset Release: October 31, 2014 – planned**
  - **Final release and workshop late 2015**
  - **Commercial licensing late 2015 or early 2016**



# Advanced Computational Tools to Accelerate Carbon Capture Technology Development



# Carbon Capture (and other process) Simulation Grand Challenges

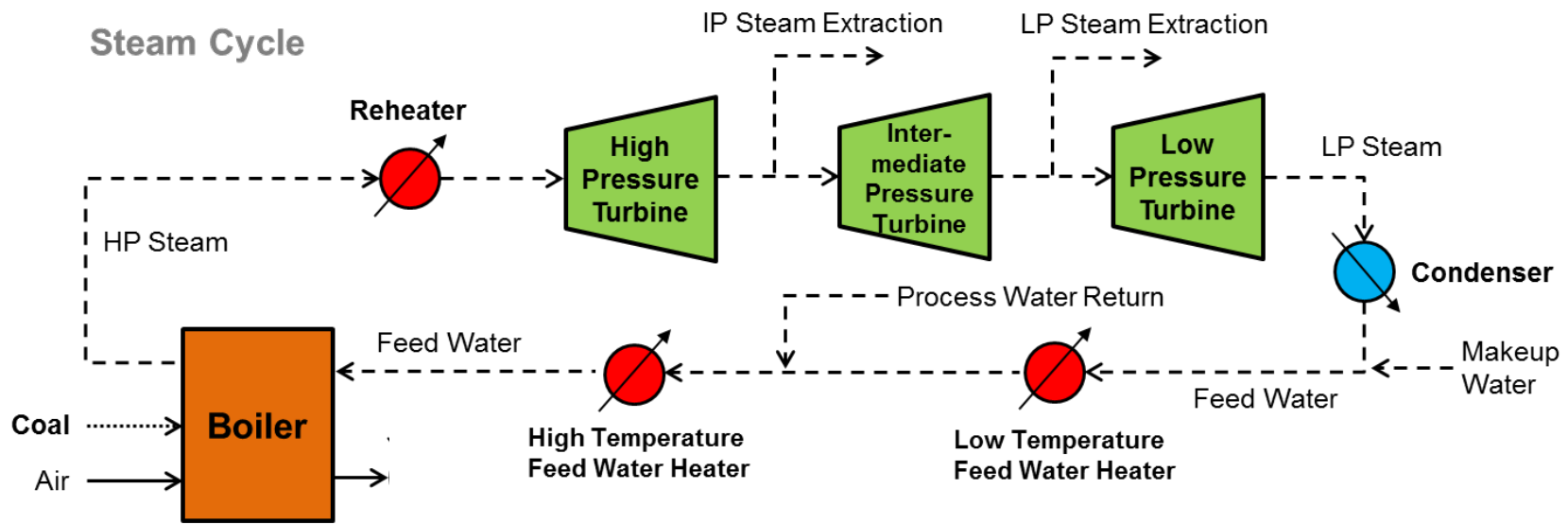
- **Multiple Scales**
  - **Particle:** individual adsorbent behavior, kinetics and transport
  - **Device:** fluid and heat flows within a sorbent bed
  - **Process:** integration of devices for a design of a complete sorbent system
- **Integration across scales**
  - Effective simplifications: Detailed tools too complex to integrate/optimize
- **Verification/Validation/Uncertainty**
  - Create confidence in predictions of models
- **Decision support**
  - Evaluate key process performance issues affecting choices of technology deployment/investment



# Outline

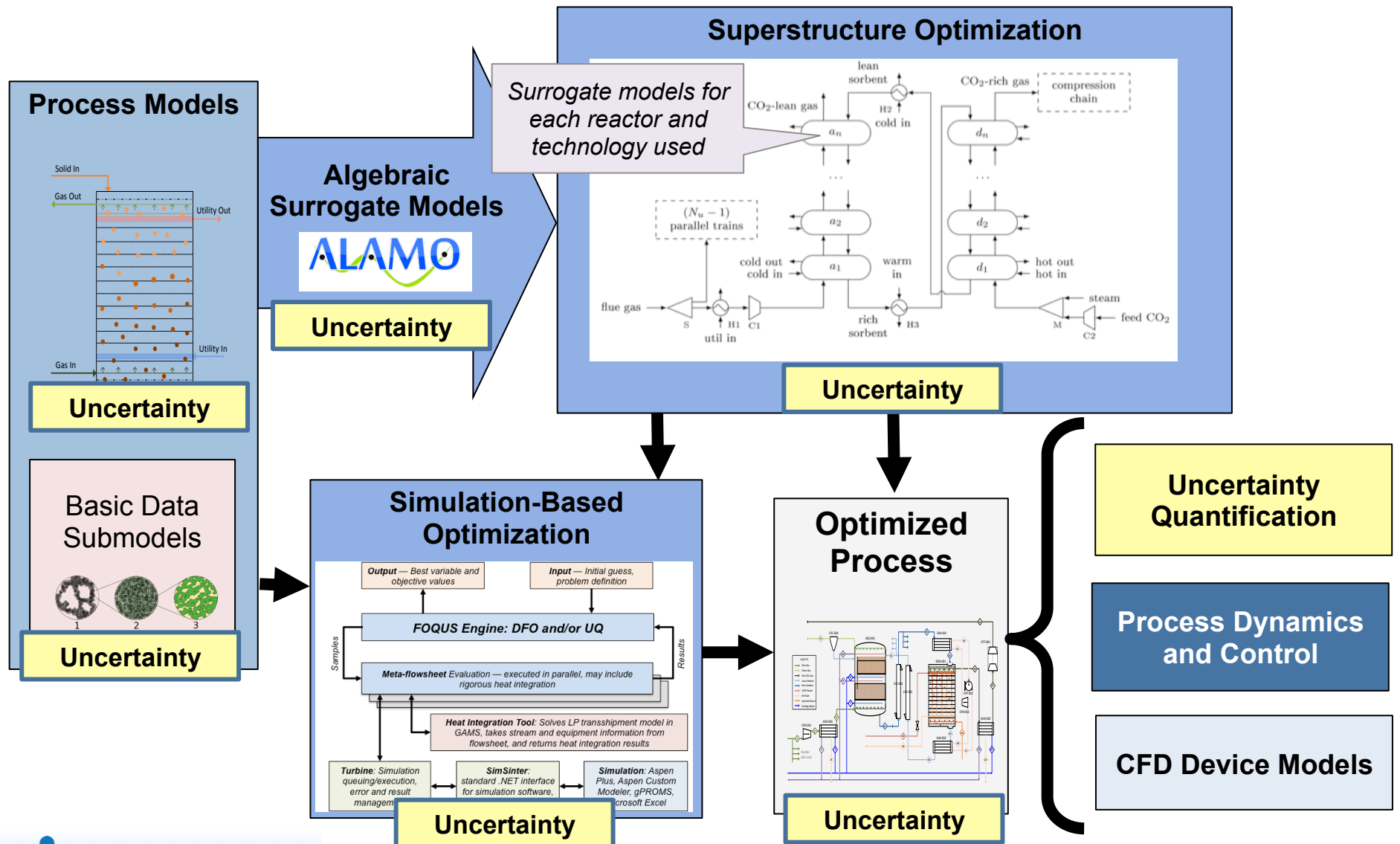
- DOE/FE Mission & Carbon Capture Challenge
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# Tools to develop an optimized process using rigorous models

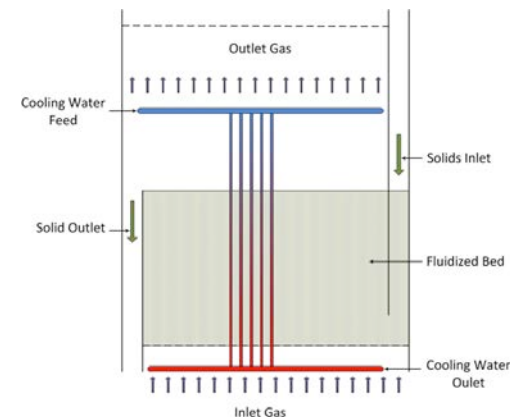
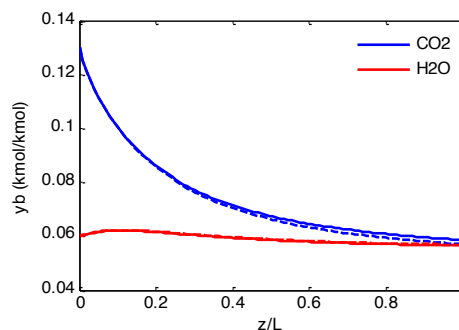




# Process Models

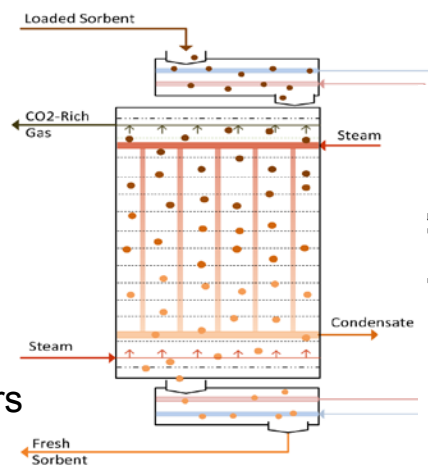
## Bubbling Fluidized Bed (BFB) Model

- 1-D, nonisothermal with heat exchange
- Unified steady-state and dynamic
- Adsorber and Regenerator
- Variable solids inlet and outlet location
- Modular for multiple bed configurations

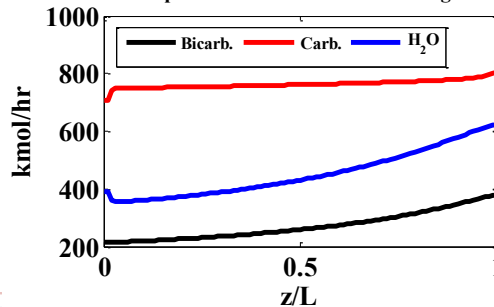


## Moving Bed (MB) Model

- 1-D, nonisothermal with heat exchange
- Unified steady-state and dynamic
- Adsorber and Regenerator
- Heat recovery system

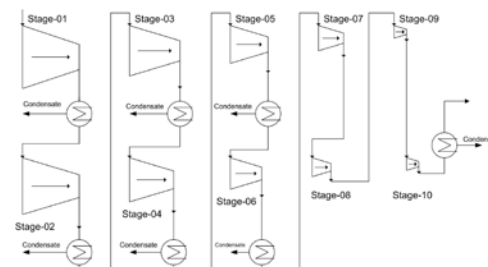


Solid Component Flow Profile of MB Regenerator

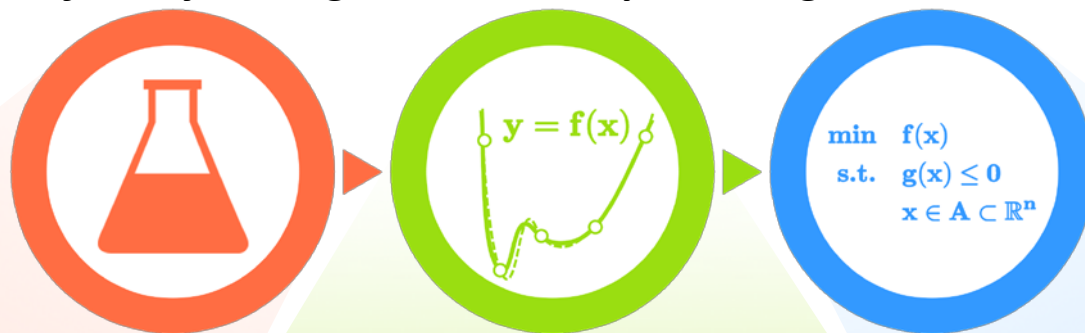


## Compression System Model

- Integral-gear and inline compressors
- Determines stage required stages, intercoolers
- Based on impeller speed limitations
- Estimates stage efficiency
- CO<sub>2</sub> drying (TEG absorption system)
- Off-design performance.
- Includes surge control algorithm



## Simplifying the balance between optimal decision-making and model fidelity through tailored simple surrogate models



### High-fidelity simulations and experiments

```

AREA = SQ
WRITE(6,60)
GO TO 10
50 WRITE(6,60)
STOP
90 WRITE(6,60)
STOP
    
```

### Algebraic surrogate models

$$\hat{f}(x, y) = 2 + y + 5e^x$$

### Superstructure optimization

**Technology selection**

A. Cozad, N. V. Sahinidis and D. C. Miller, 2014, "Learning surrogate models for simulation-based optimization." *AICHE Journal* 60(6): 2211-2227.

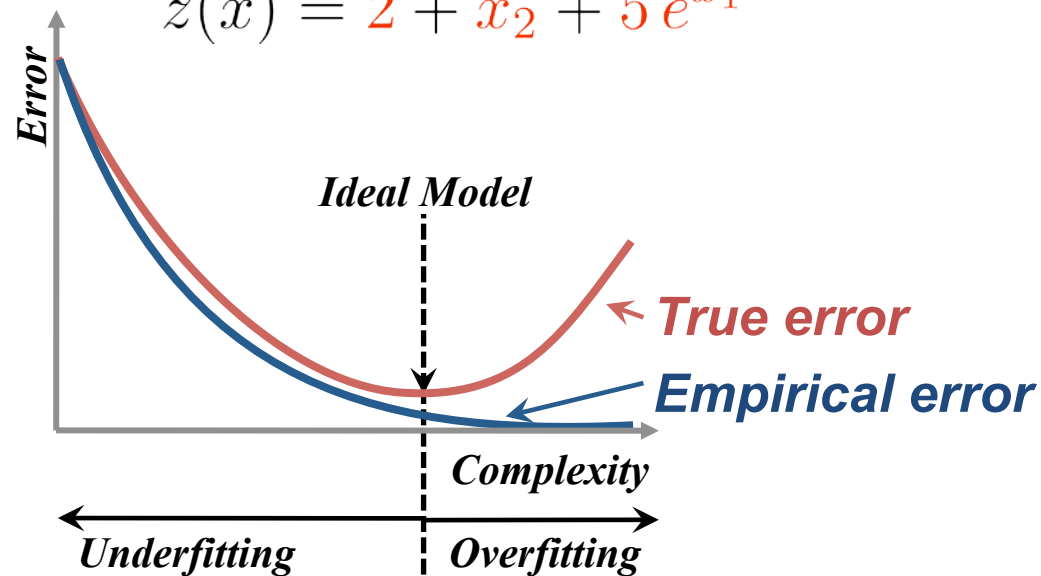
# ALAMO: Model Development & Overfitting

- **Step 1:** Define a large set of potential basis functions

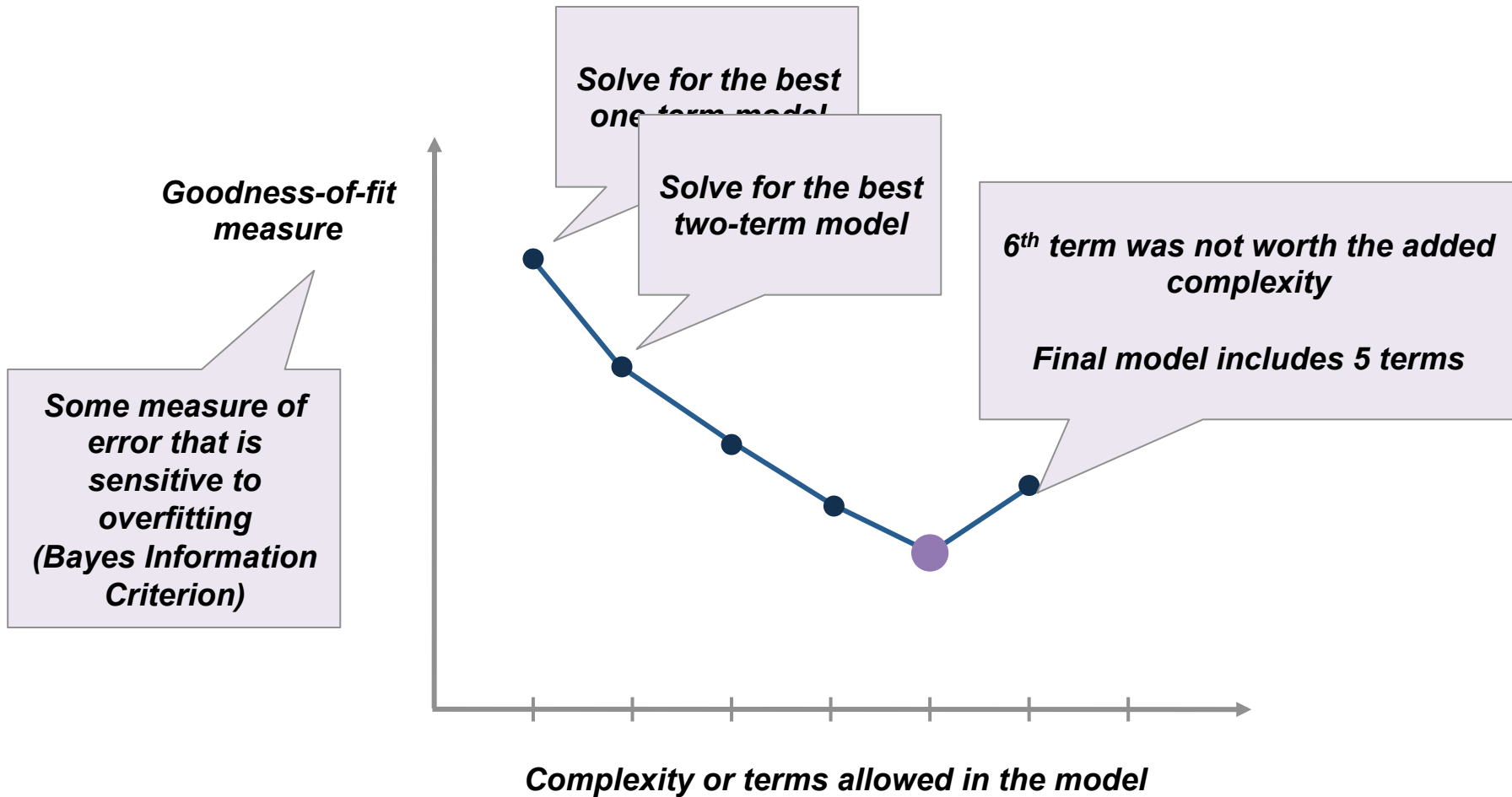
$$\hat{z}(x) = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_1 x_2 + \beta_4 e^{x_1} + \beta_5 e^{x_2} + \dots$$

- **Step 2:** Model reduction

$$\hat{z}(x) = 2 + x_2 + 5 e^{x_1}$$

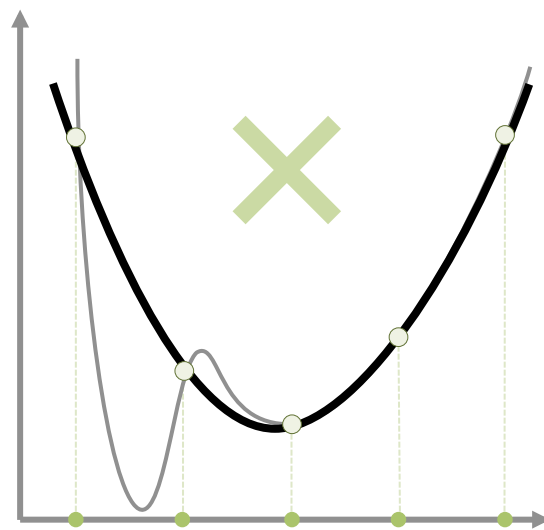


# ALAMO: Surrogate Model Development

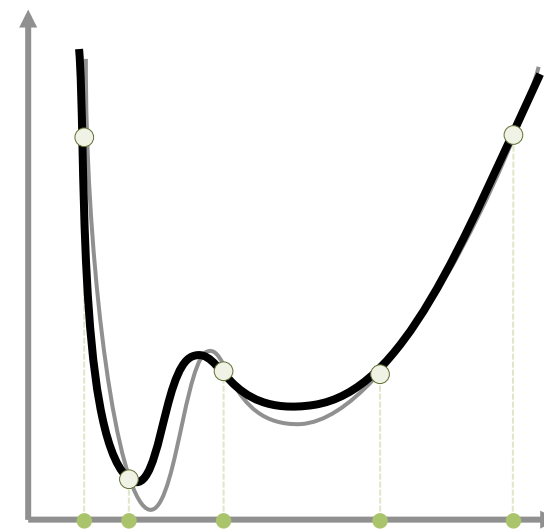


# Adaptive Sampling Improves Surrogate Model

- We use an iterative design of experiments to
  - Sample better or sample fewer data points
- Two models given the same data set size:



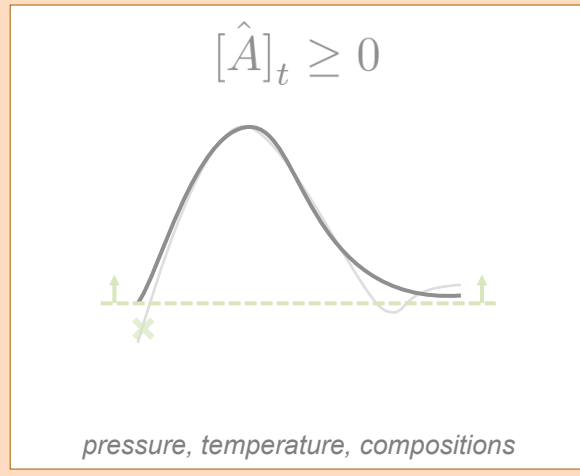
*Even sampling*



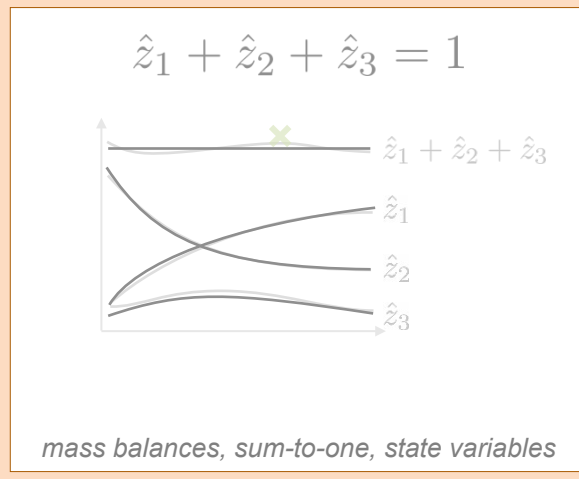
*Stronger sampling*

# Constrained Regression Improves Surrogates

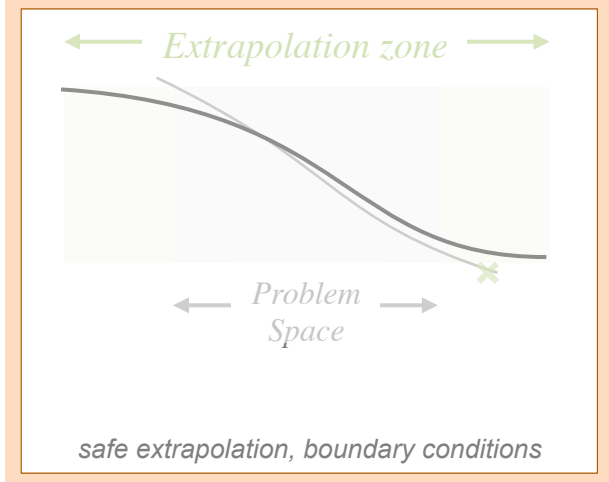
## Response bounds



## Multiple responses



## Alternative domains

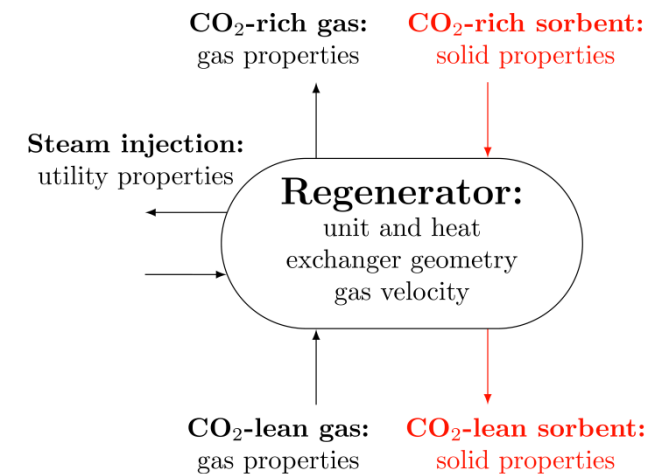
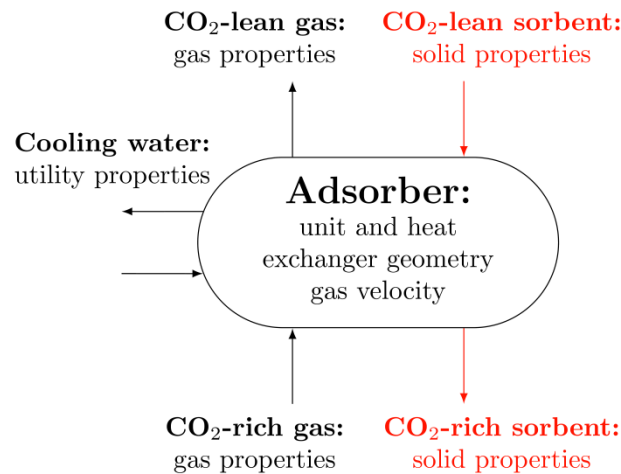


# Carbon Capture Reactors

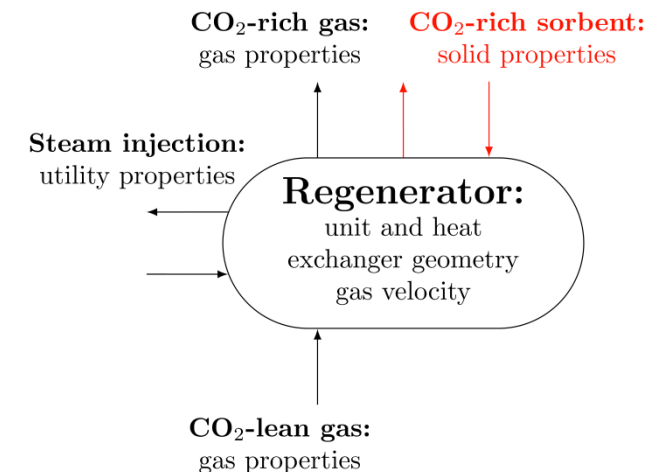
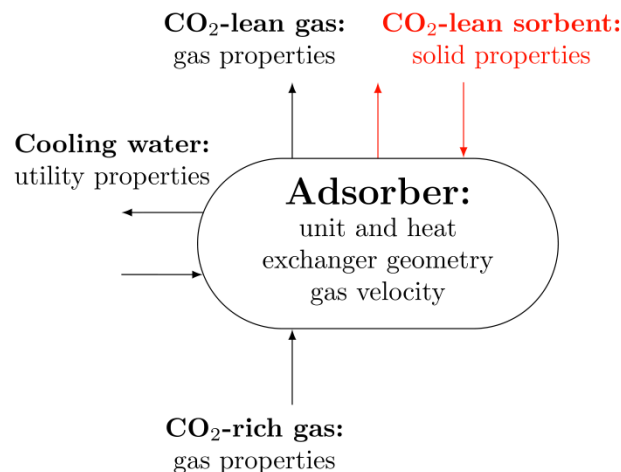
## Adsorber

## Regenerator

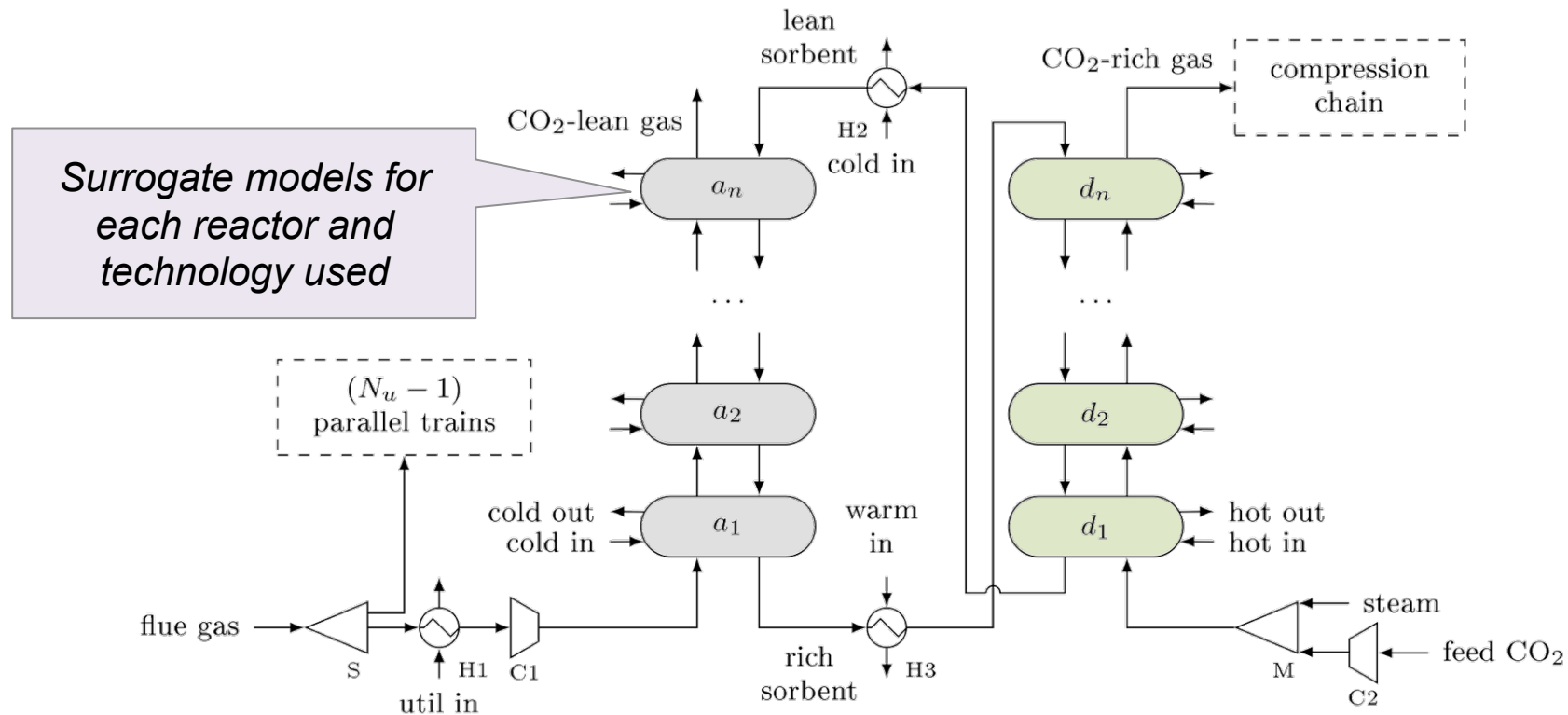
### Underflow configuration



### Overflow configuration



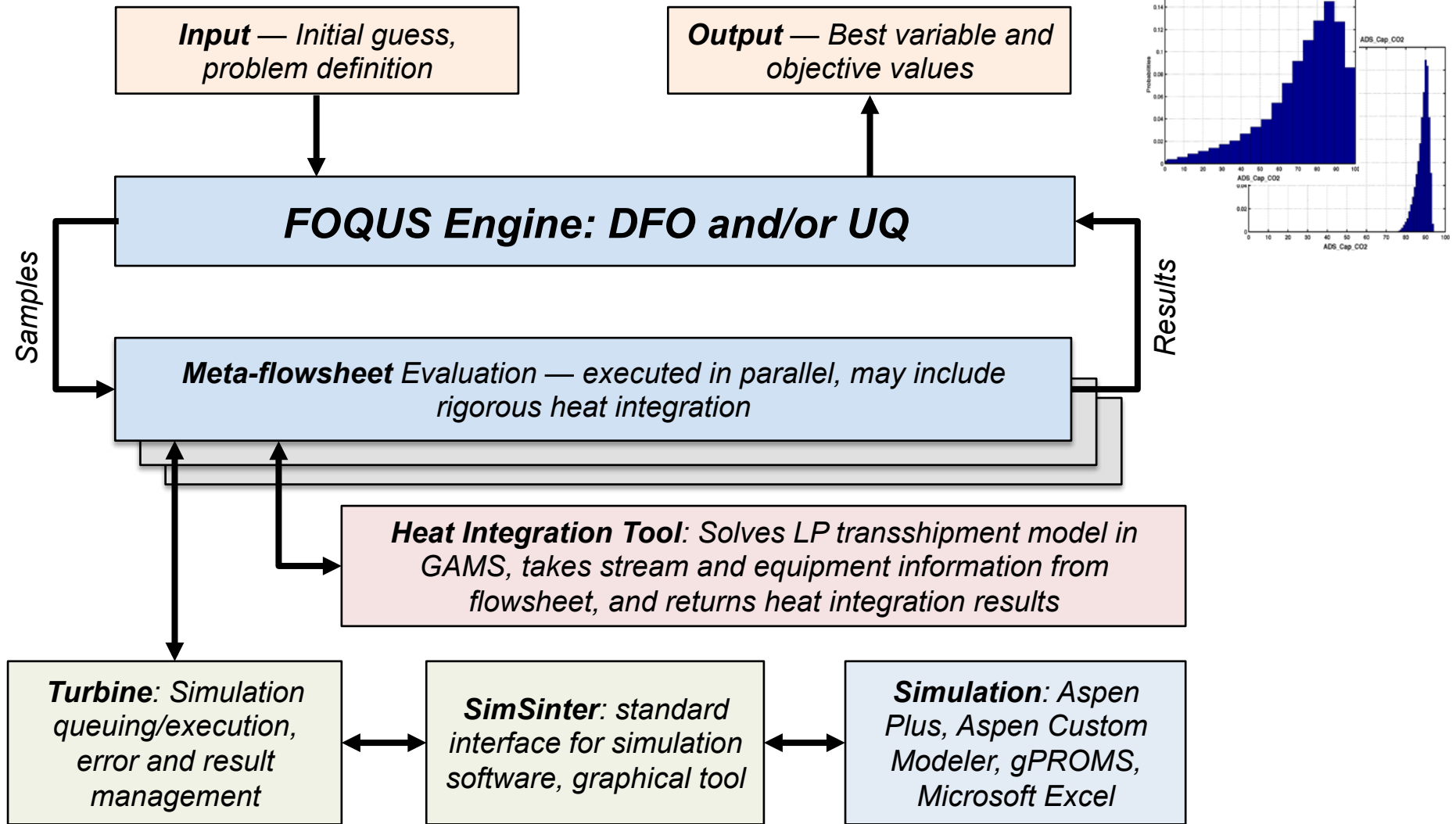
# Carbon Capture System Configuration



- Discrete decisions: How many units? Parallel trains?  
What technology used for each reactor?
- Continuous decisions: Unit geometries
- Operating conditions: Vessel temperature and pressure, flow rates, compositions



# Framework for Optimization, Quantification of Uncertainty and Sensitivity



D. C. Miller, B. Ng, J. C. Eslick, C. Tong and Y. Chen, 2014, Advanced Computational Tools for Optimization and Uncertainty Quantification of Carbon Capture Processes. In *Proceedings of the 8th Foundations of Computer Aided Process Design Conference – FOCAPD 2014*. M. R. Eden, J. D. Sirola and G. P. Towler Elsevier.

# Sinter: Generic Connectivity for Simulations

- Links simulations to FOQUS/Turbine
- Support for
  - gPROMS
  - Aspen Plus
  - Aspen Custom Modeler
  - Excel
    - Costing
    - Generic connectivity (i.e., Thermoflex)
- Explicitly links to any input/output variables via GUI
  - JSON structure



# Turbine: Simulation Execution Management

- Works with Aspen, gPROMS, Excel
- Supports large number of simulation executions
  - Manages failed simulations
- Scalable
  - Amazon Cloud
    - Up to 500 parallel simulations tested
    - Over 100,000 simulations run
  - Cluster at NETL (coming online 2014)
  - Local desktop
- Challenge: Licensing of commercial simulation tools



# FOQUS: DFO Example with Heat Integration

**Objective Function:** Maximize **Net efficiency**

**Constraint:** **CO<sub>2</sub> removal** ratio  $\geq 90\%$

Flowsheet evaluation (via **process simulators**)

Minimum utility target (via **heat integration tool**)

**Decision Variables (17):** Bed length, diameter, sorbent and steam feed rate



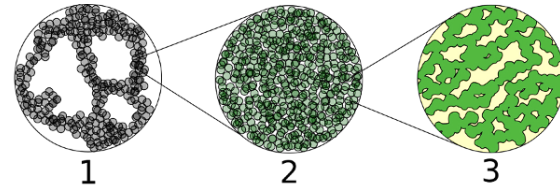
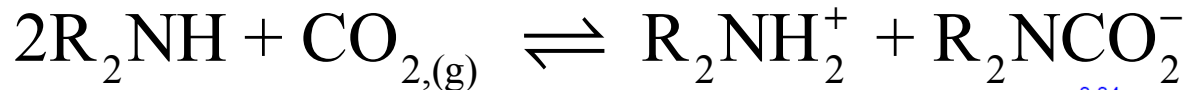
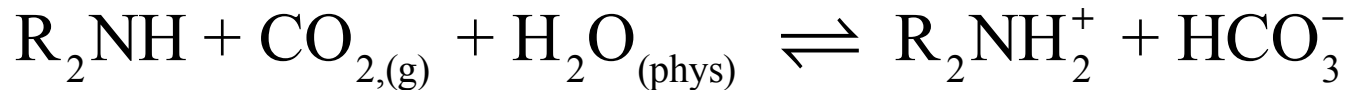
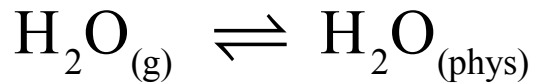
# FOQUS Optimization Results

	Simultaneous	Sequential	w/o heat integration
<b>Net power efficiency (%)</b>	<b>32.6</b>	<b>31.8</b>	<b>30.0</b>
Net power output ( $MW_e$ )	504.3	491.5	463.9
<b>CO<sub>2</sub> removal ratio (%)</b>	<b>91.9</b>	<b>90.2</b>	<b>90.2</b>
Electricity consumption ( $MW_e$ )	86.9	75.1	75.1
IP steam (kg/s)	0	0	0
LP steam (kg/s)	93.9	125.3	139.0
Cooling water consumption ( $m^3/s$ )	12.8	10.4	20.7
Heat addition to feed water ( $MW_{th}$ )	135.4	139.8	0

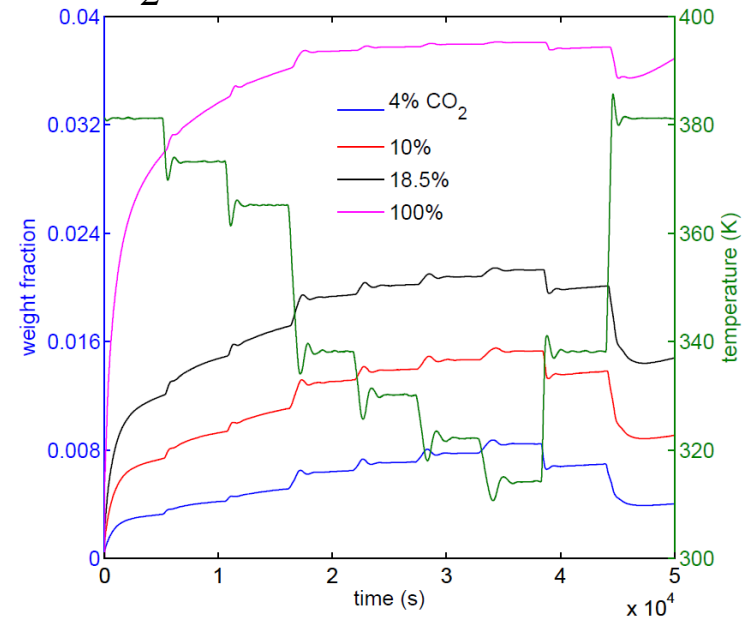
**Base case w/o CCS: 650  $MW_e$ , 42.1 %**



# Uncertainty in Kinetic Parameters



$$K_i = \exp\left(\frac{\Delta S_i}{R}\right) \exp\left(\frac{-\Delta H_i}{RT}\right) / P$$



# UQ Sampling

Simulation Ensemble Setup

Choose how to generate samples:

Choose sampling scheme

Load all samples from a single file

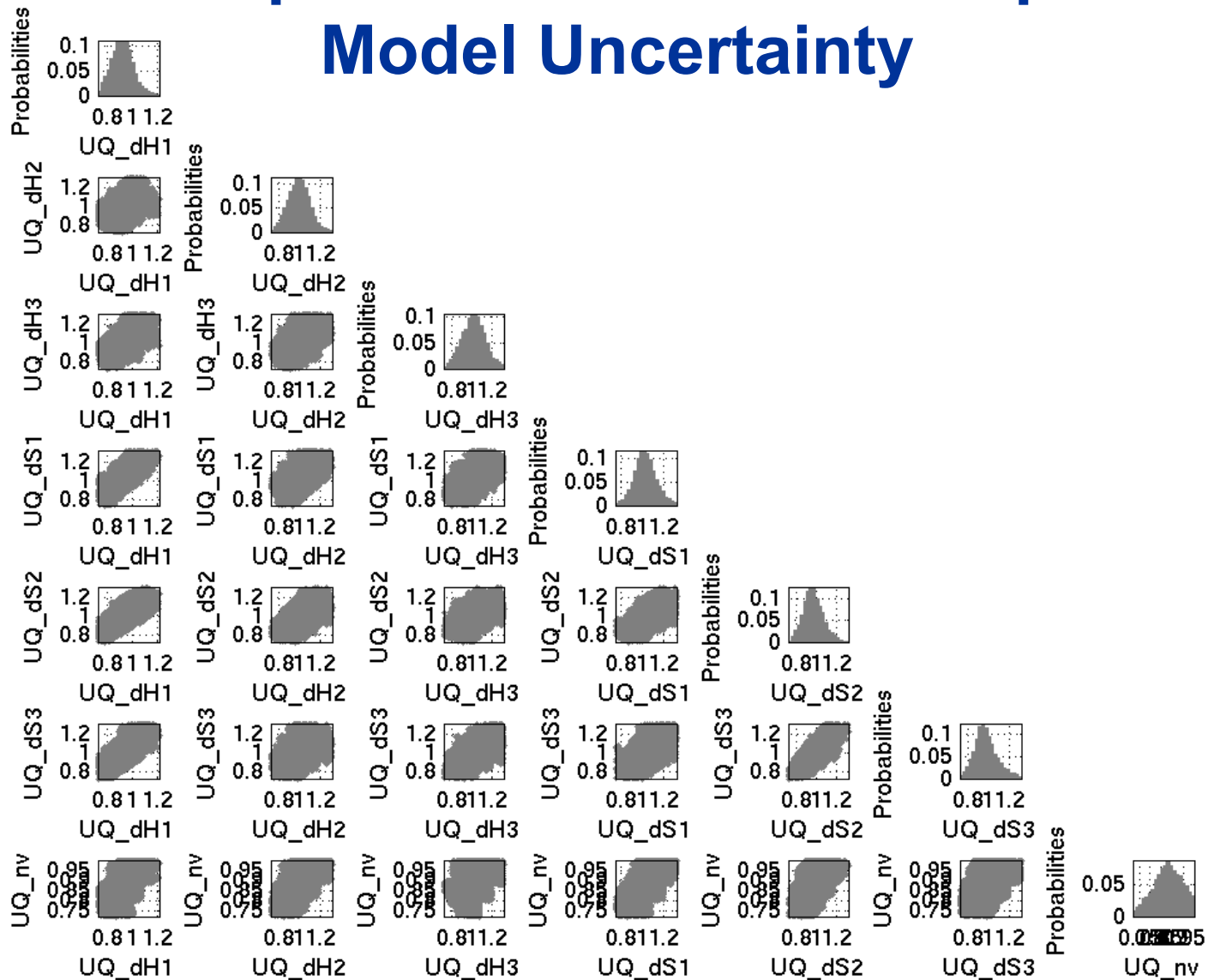
Warning: Only "Monte Carlo" sampling scheme is compatible with non-uniform input distributions (PDF).

Distributions | Sampling scheme

	Name	Type	Default	Min	Max	PDF	Param 1	Param 2
1	x1	Variable	4	-10	10	Uniform		
2	x2	Fixed	5	-10	10	Uniform		
3	x3	Variable	4	-10	10	Uniform		
4	x4	Variable	5	-10	10	Uniform		
5	x5	Variable	4	-10	10	Uniform		
6	x6	Variable	4	-10	10	Uniform		

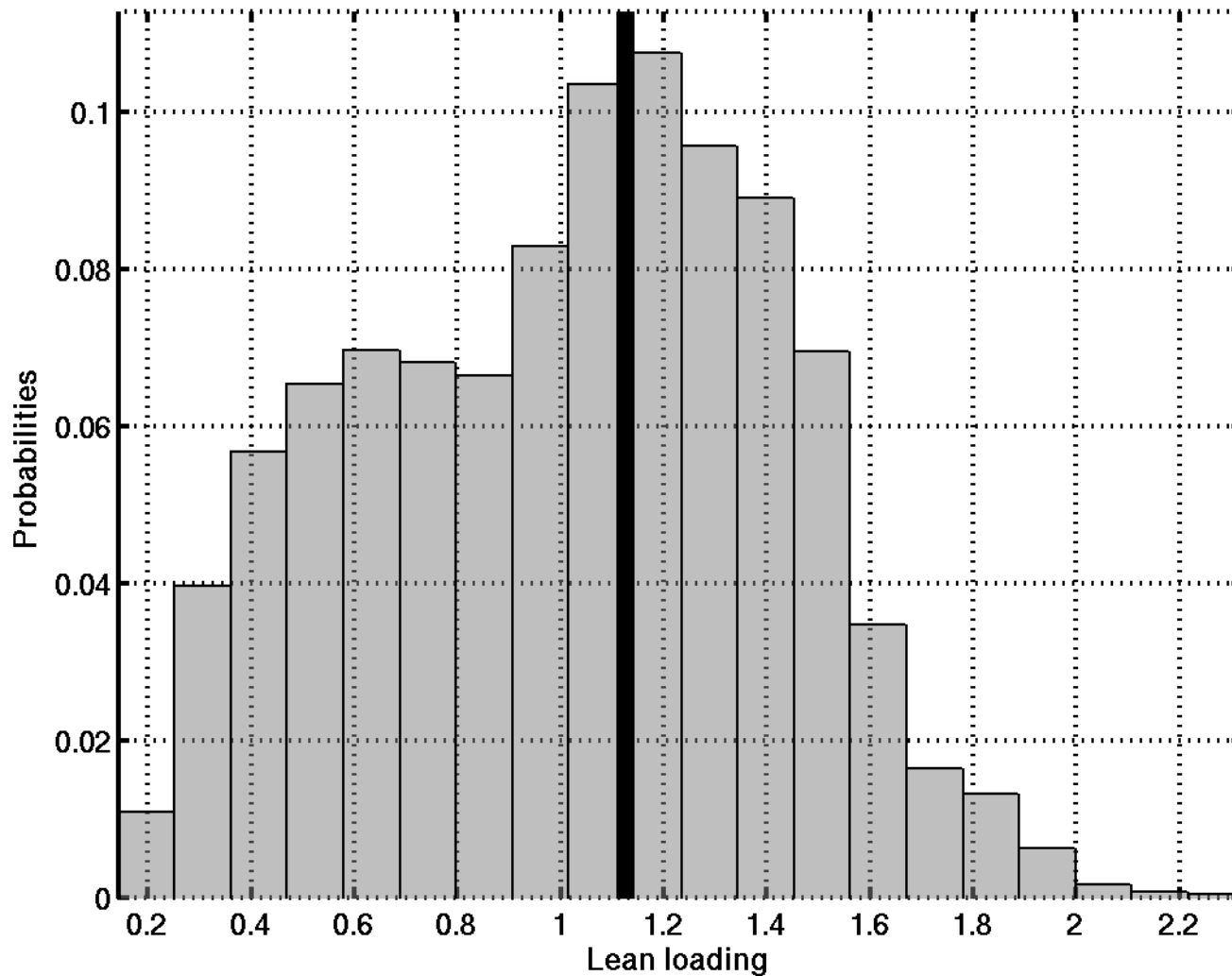
Browse for Sample file...

# Feasible Input Distributions for Equilibrium Model Uncertainty

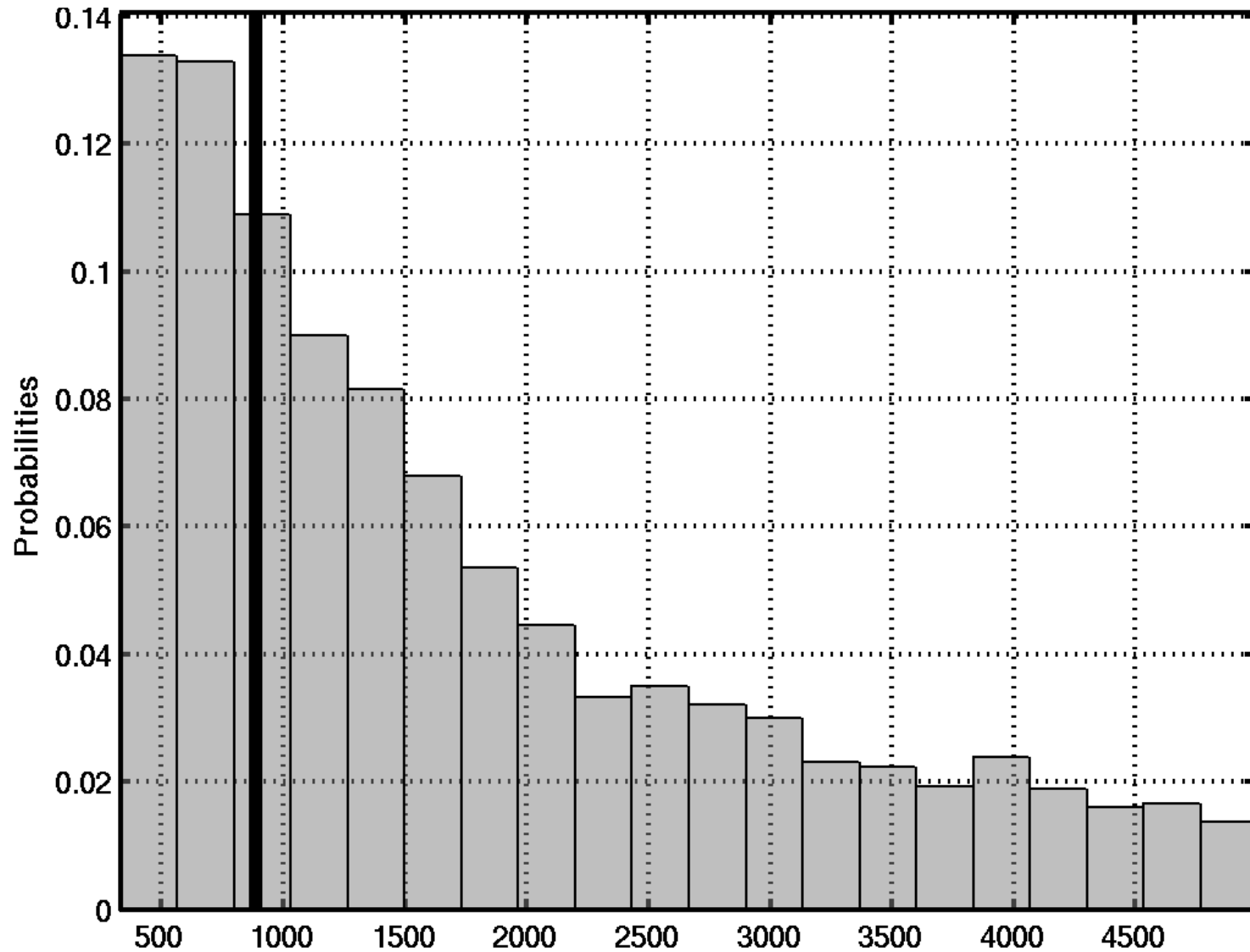




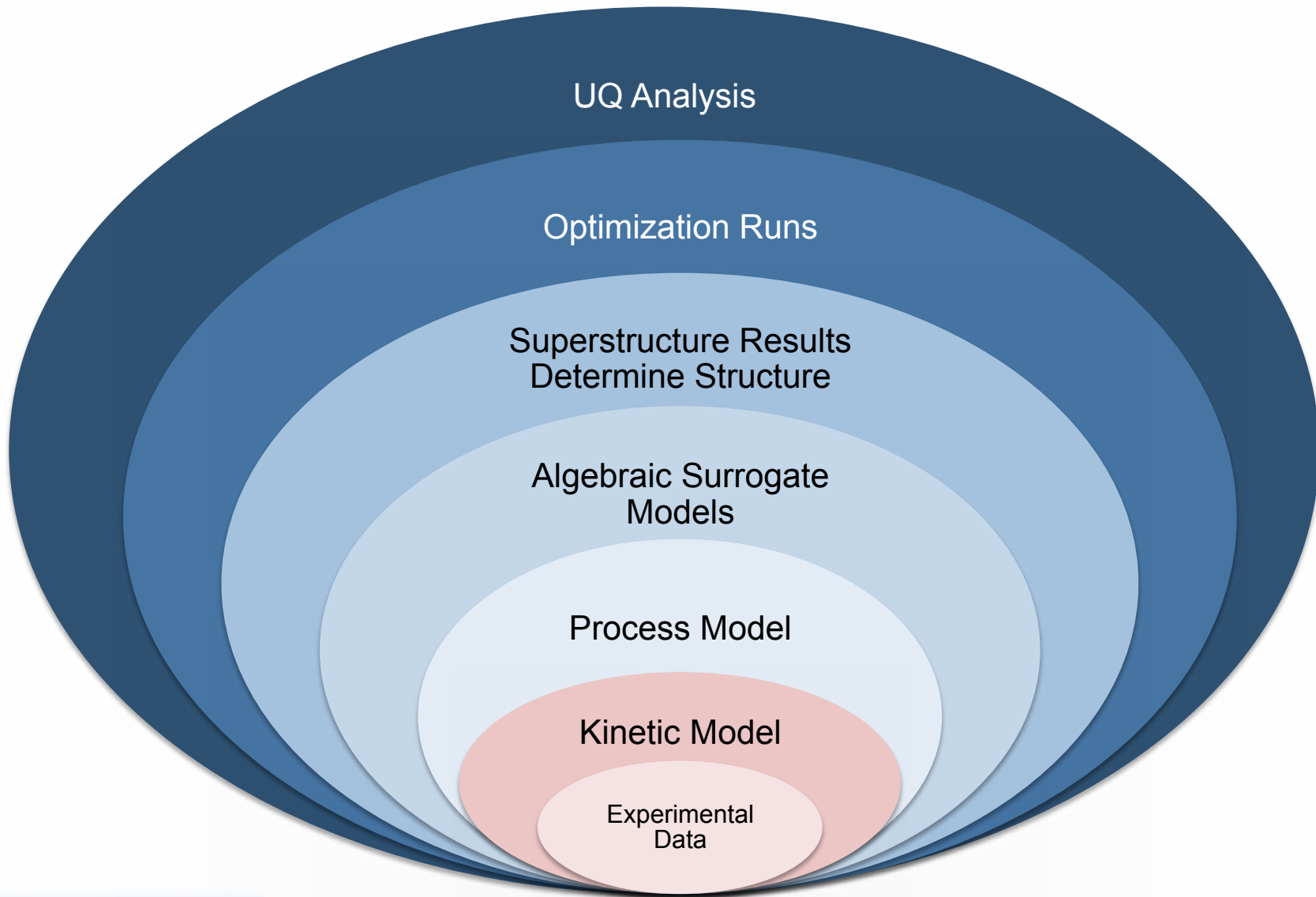
# Probability distribution for the loading exiting the regeneration system



# Required new steam flow to meet product specs



# Data Management: Provenance



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## CCSI Release 2013.10.2

<b>Basic Data Submodels</b>	High viscosity solvent model
	SorbentFit – Kinetic/diffusion basic data fitting tool with UQ
<b>High Resolution Filtered Submodels</b>	Attrition Model
	Cylinder Filtered Models with quantified uncertainty bounds
<b>Validated high-fidelity CFD models &amp; UQ tools</b>	1 MW Adsorber and Regenerator CFD Models (validated)
	Large scale adsorber and regenerator CFD Models
	Statistical Model Validation Tool for Quantifying Predictions
	REVEAL: Reduced Order Modeling Tools for CFD and ROM Integration Tools
<b>Process Models</b>	Bubbling Fluidized Bed Reactor Model
	Moving Bed Reactor Model
	Multi-stage Centrifugal Compressor Model
	Membrane CO <sub>2</sub> Separation Model
	Reference Power Plant Model
<b>Optimization and UQ Tools</b>	FOQUS – Optimization & Quantification of Uncertainty
	ALAMO – Surrogate models for optimization
	Process Synthesis Superstructure
	Oxy-Combustion Process Optimization Model
<b>Dynamics &amp; Control</b>	D-RM Builder
<b>Risk Analysis Tools</b>	Technical Risk Model
	Financial Risk Model
<b>Crosscutting Integration Tools</b>	SimSinter – Links simulation files to FOQUS/Turbine
	Turbine Science Gateway – Runs hundreds of thousands of simulations

# Toolset Deployment

- Initial licensees:   
- Additional licensees:  
- Others licenses in progress....    
- CRADA:  in development 





# Carbon Capture Simulation Initiative: A Case Study in Multiscale Modeling and New Challenges

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Debangsu Bhattacharyya,<sup>5</sup> Nikolaos V. Sahinidis,<sup>6</sup>  
Deb Agarwal,<sup>7</sup> Charles Tong,<sup>8</sup> Stephen E. Zitney,<sup>2</sup>  
Avik Sarkar,<sup>9</sup> Xin Sun,<sup>9</sup> Sankaran Sundaresan,<sup>10</sup>  
Emily Ryan,<sup>11</sup> Dave Engel,<sup>9</sup> and Crystal Dale<sup>4</sup>

*Annual Review of Chemical and Biomolecular Engineering* 5: 301-323.







56 National Lab researchers  
 35 Students/post-docs  
 8 Professors  
 5 National Labs  
 5 Universities  
 20 Companies on IAB

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