



Accelerating Technology Development: Post Combustion Capture Sorbents

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National Energy Technology Laboratory

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U.S. DEPARTMENT OF
ENERGY

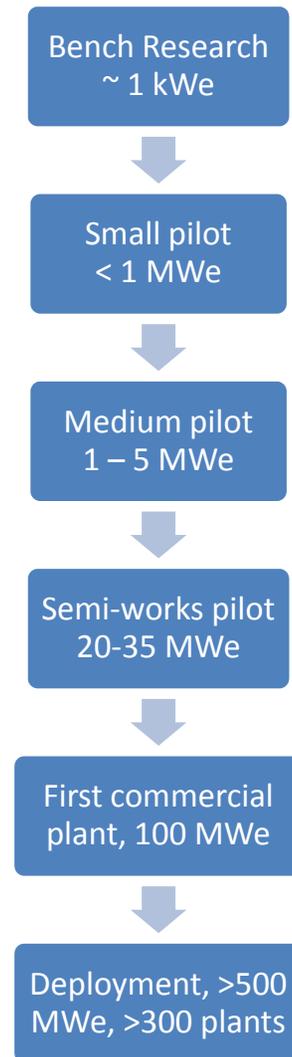
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Laboratory

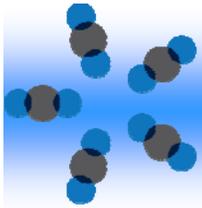
Outline

- **Carbon Capture Challenge**
- **Computational Tools to Accelerate Technology Development**
- **Experimental Carbon Capture Research @ NETL**
- **Multiscale Model Development, Simulation & Optimization**
- **Experimental Validation**
- **Conclusions**

Carbon Capture Challenge

- The traditional pathway from discovery to commercialization of energy technologies can be quite long, i.e., 20-30 years
- New approaches are needed for taking carbon capture concepts from lab to power plant, quickly, and at low cost and risk
- Science-based simulations will accelerate the development of carbon capture technology, from discovery through deployment

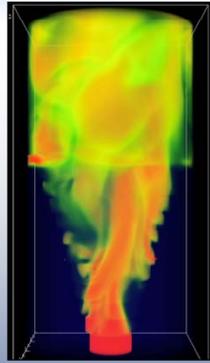
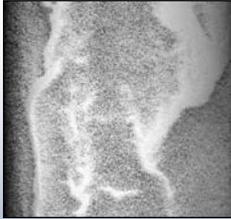
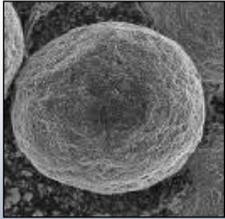




CCSI

Carbon Capture Simulation Initiative

For Accelerating Technology Development



Identify promising concepts



Reduce the time for design & troubleshooting



Quantify the technical risk, to enable reaching larger scales, earlier



Stabilize the cost during commercial deployment

National Labs



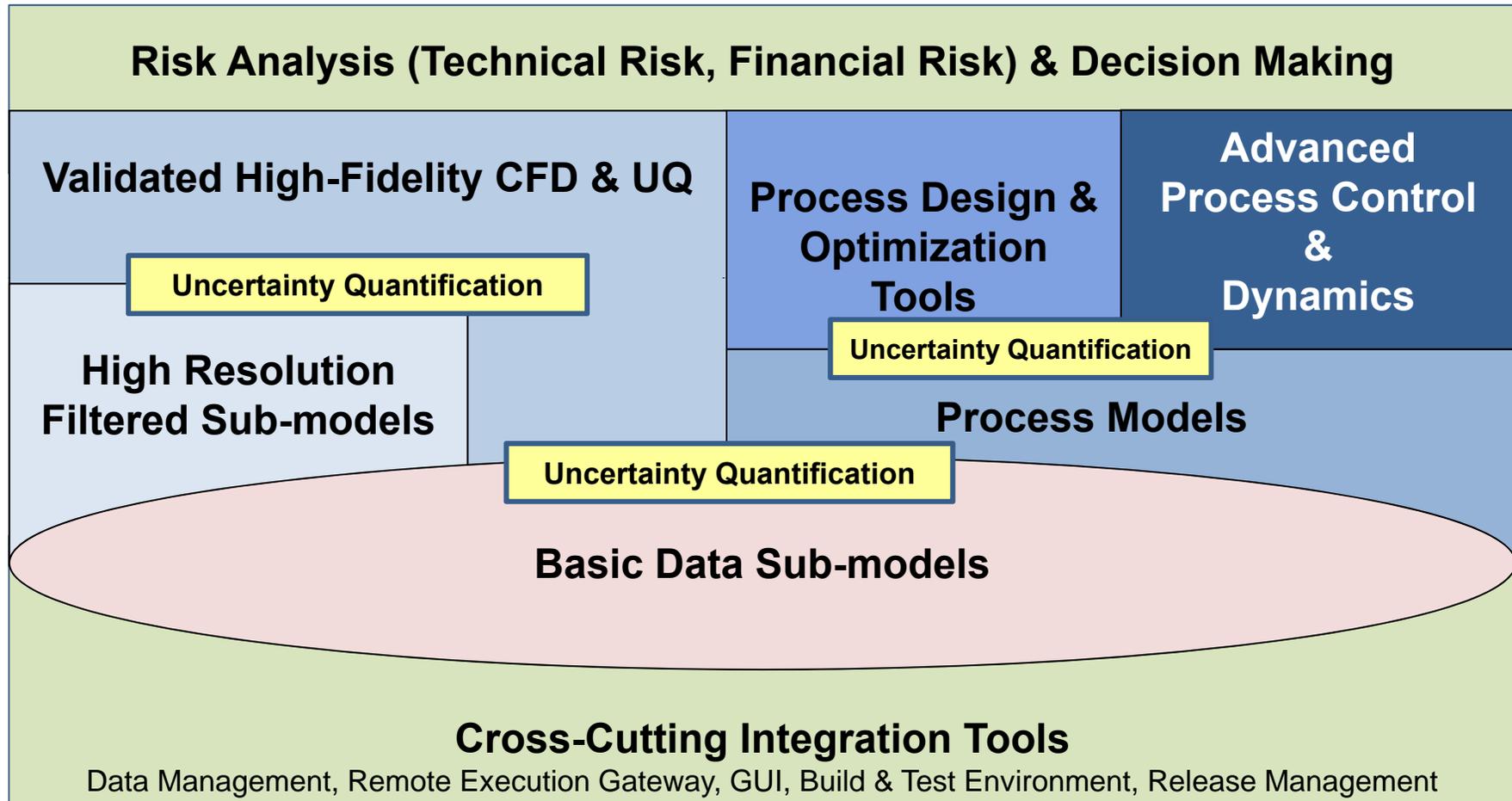
Academia



Industry



Advanced Computational Tools to Accelerate Next Generation Technology Development

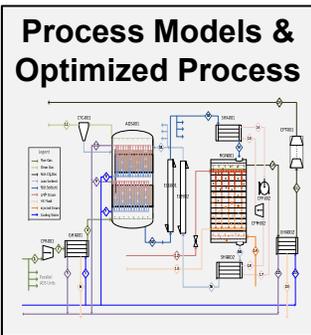


Simulation & Experiments to reduce time for design/troubleshooting

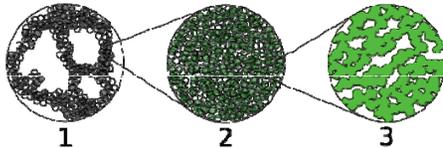
Experimental Validation



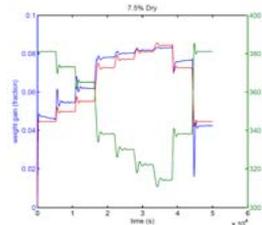
Process Models & Optimized Process



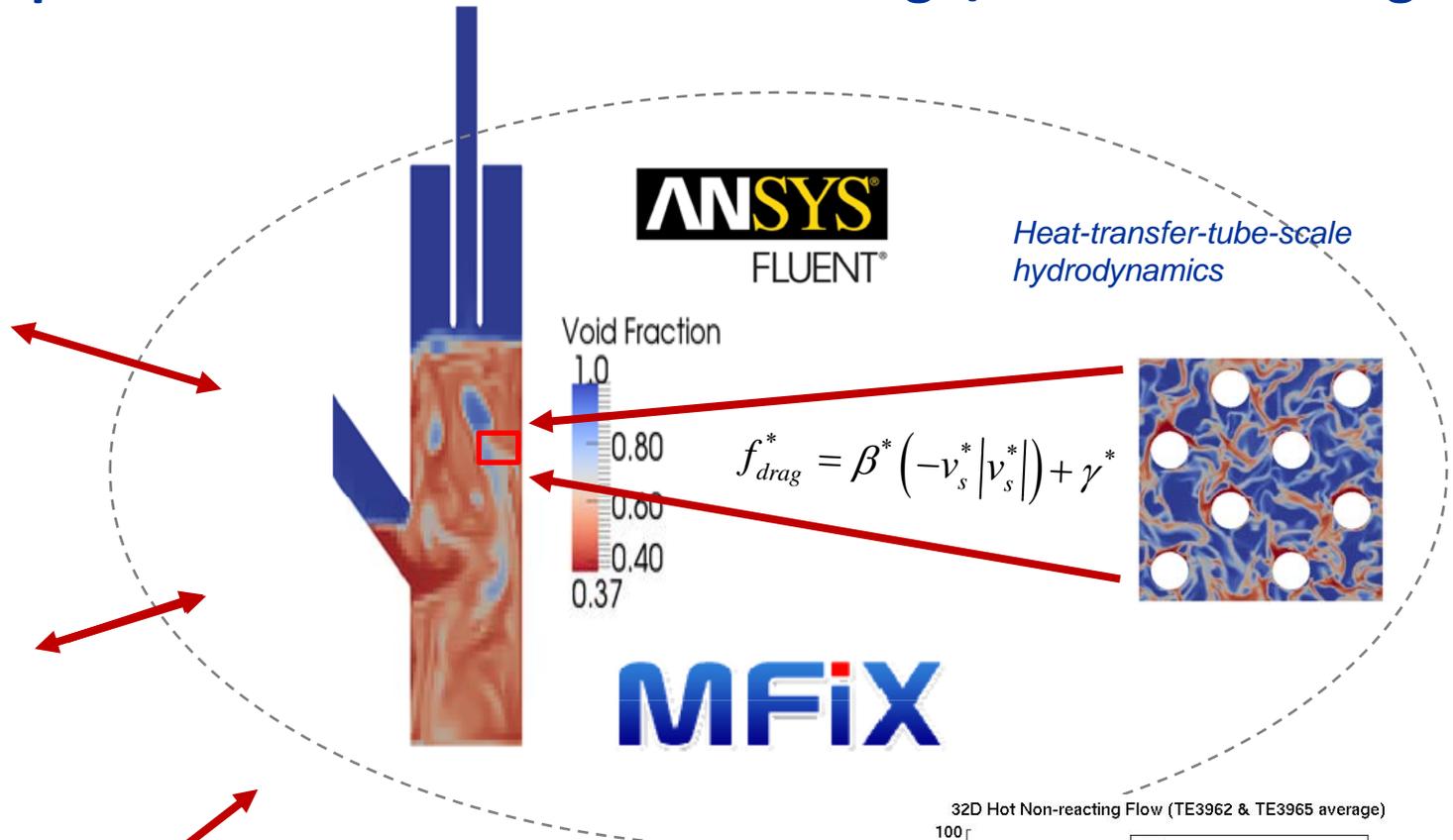
SORBENTFIT



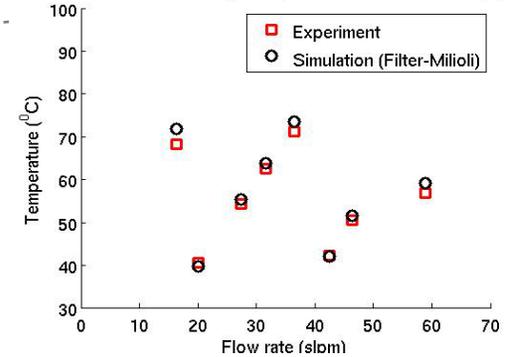
Experimental Kinetic/Mass Transfer Data



Void Fraction along vertical center plane



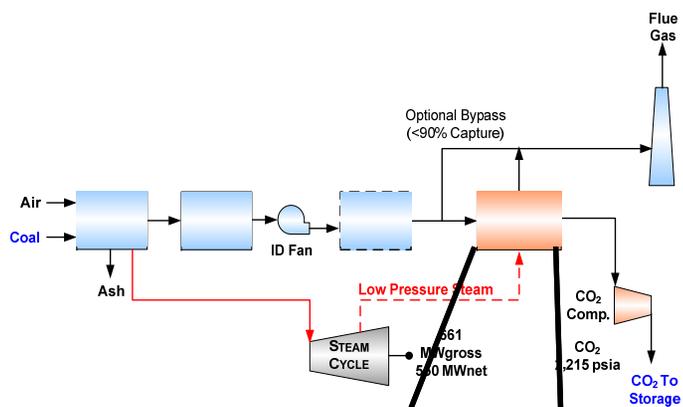
32D Hot Non-reacting Flow (TE3962 & TE3965 average)



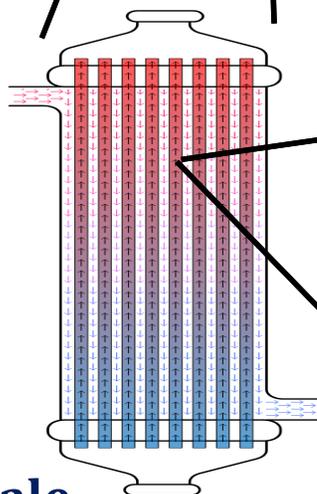
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Carbon Capture

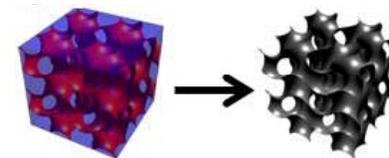


Plant Scale



Device Scale

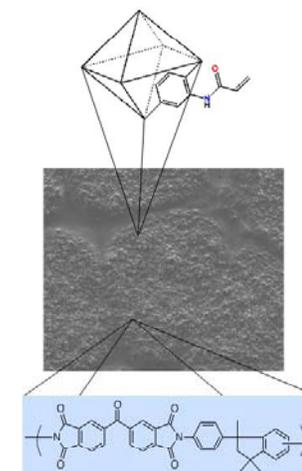
Development of efficient and economic carbon capture strategies applicable for post-, pre- and oxy-combustion schemes



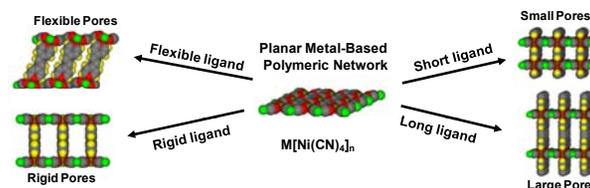
Polymers and Derivatives



Ionic Liquids

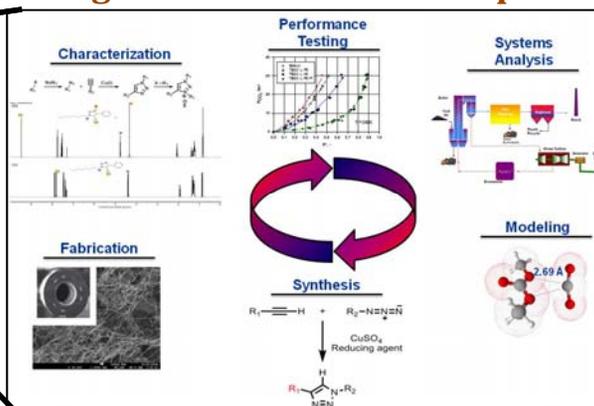


Composites



Metal-Organic Frameworks

Integrated Materials Development



Materials Scale



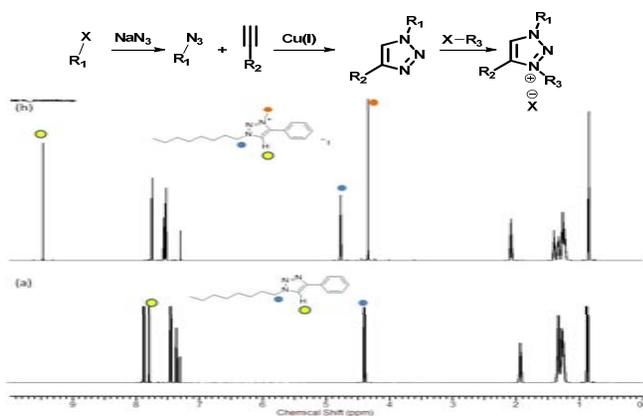
Carnegie Mellon



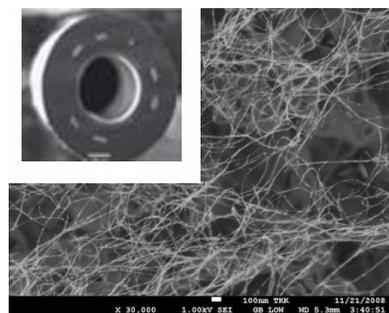
University of Pittsburgh

Integrated Materials Development

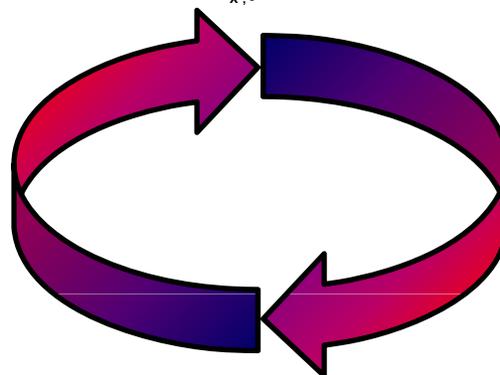
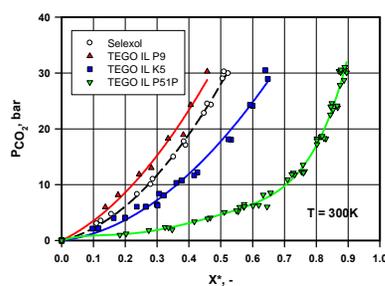
Characterization



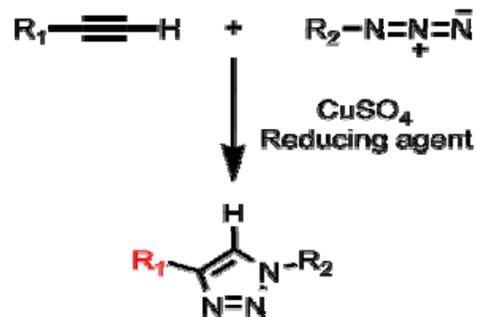
Fabrication



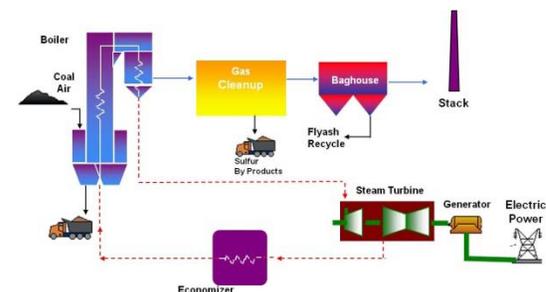
Performance Testing



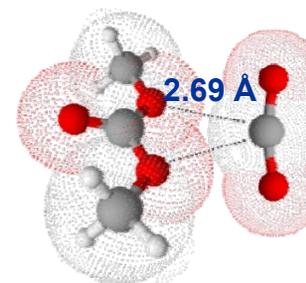
Synthesis



CCSI Simulation & Analysis

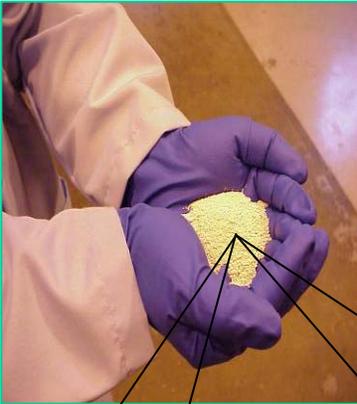


Molecular Modeling

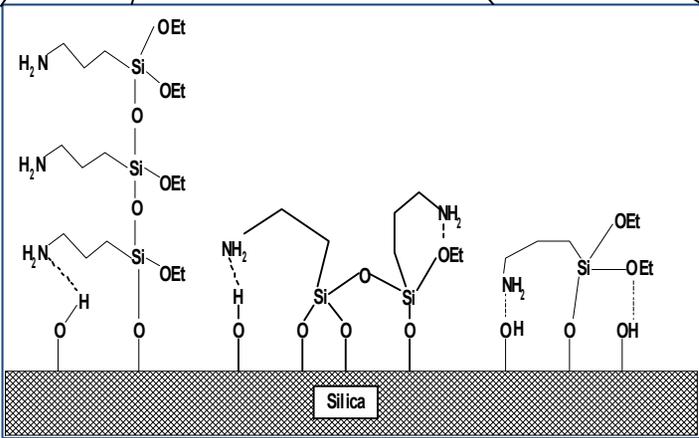
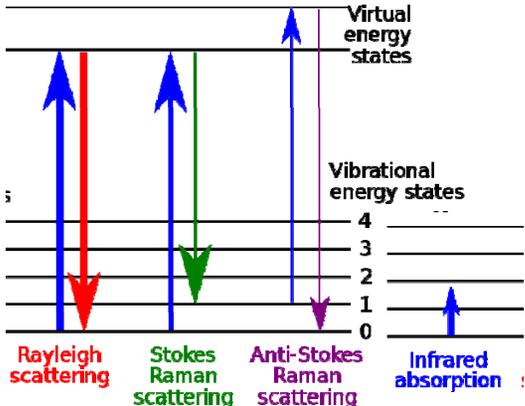


Carbon Capture

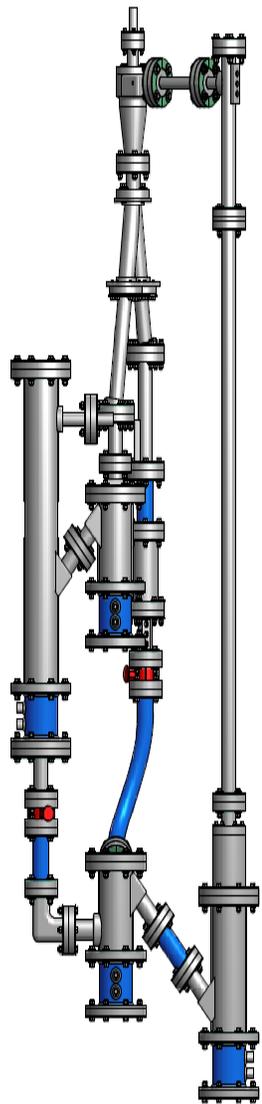
Supported Amine Sorbents



Characterization

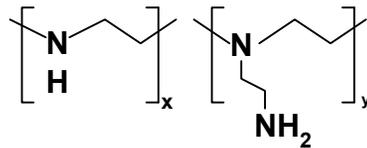


Scale-up



Components of Basic Immobilized Amine Sorbents (BIAS)

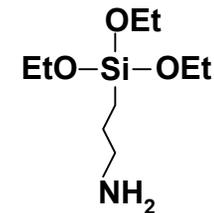
PQ 2129



Polyethyleneimine

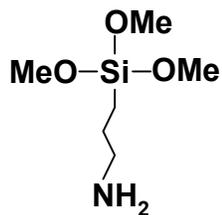
PEI

Mn 423-2000



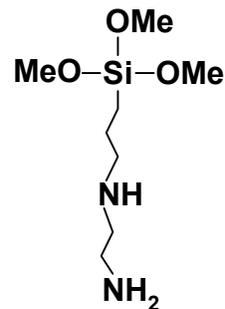
(3-Aminopropyl) triethoxysilane

APTES



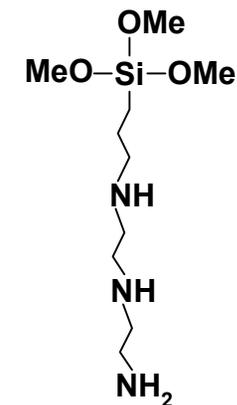
(3-Aminopropyl) trimethoxysilane

APTMS



N1-(3-Trimethoxysilypropyl)
diethylenediamine

DEDA



N1-(3-Trimethoxysilypropyl)
diethylenetriamine

DETA

Modification of US 7,288,136 High Capacity
Immobilized Amine Sorbents

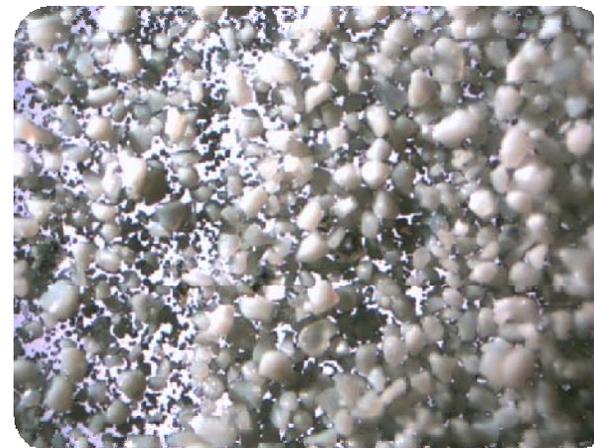
US Patent Application 13212284 filed 8/11

Sorbent AX

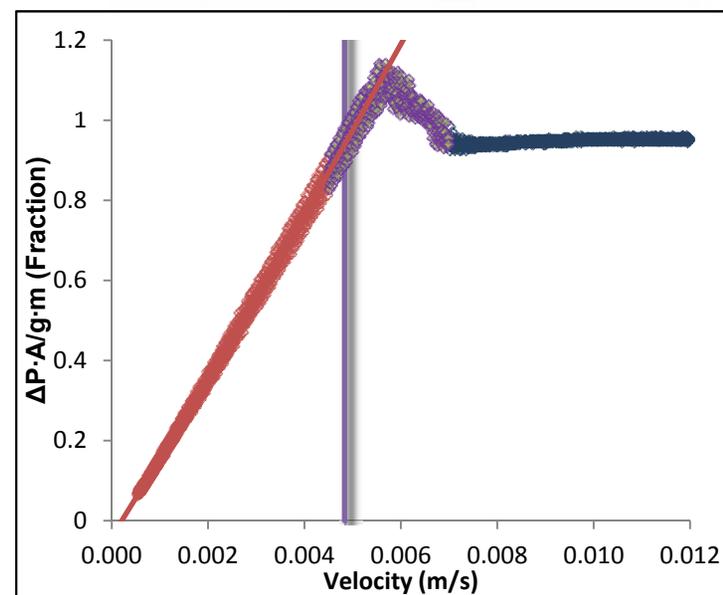
Two BIAS sorbents are being tested: AX and 32D

Sorbent AX

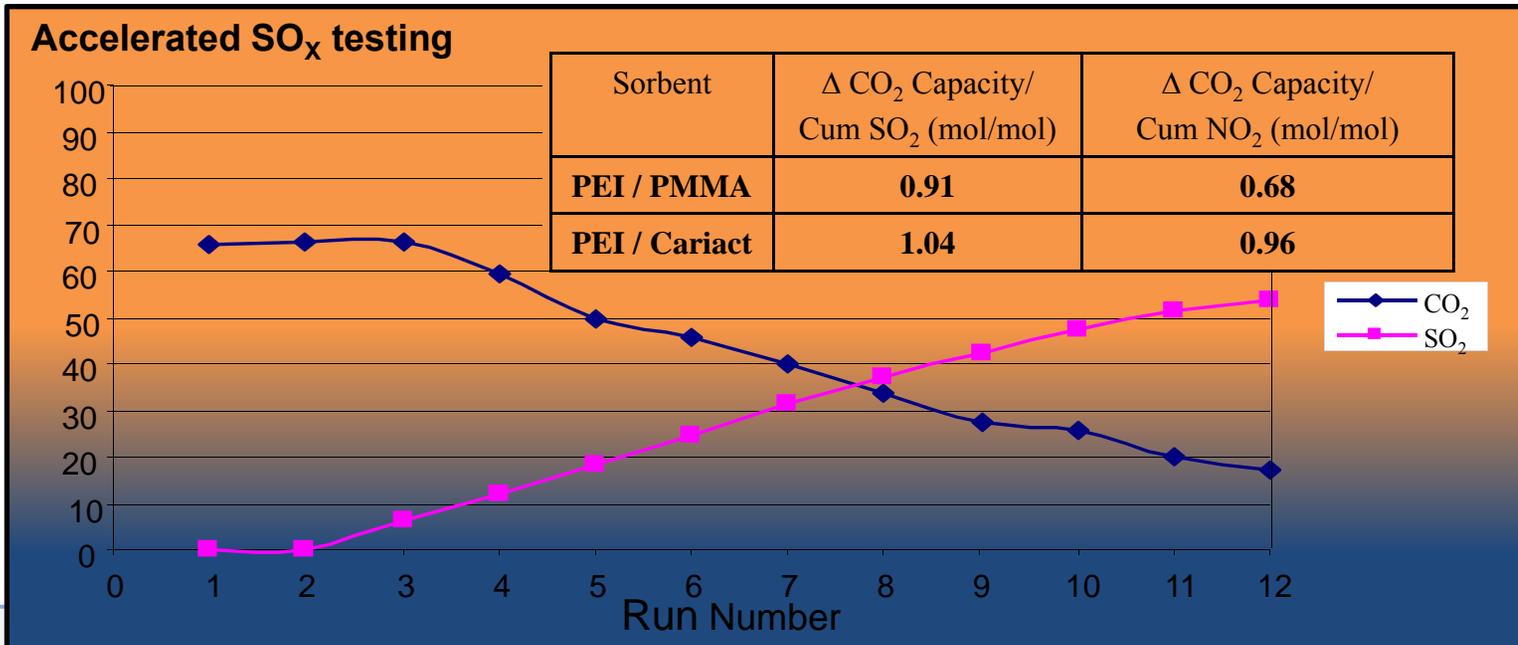
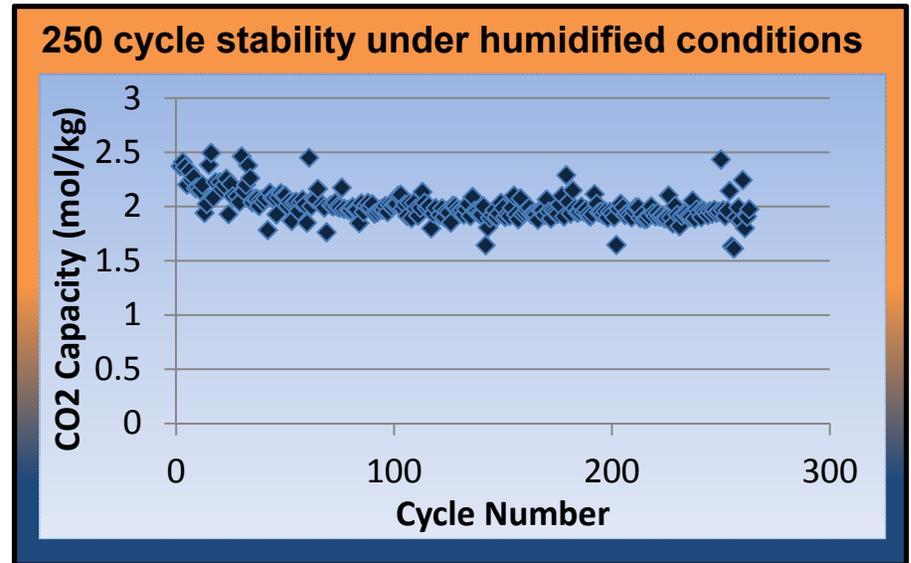
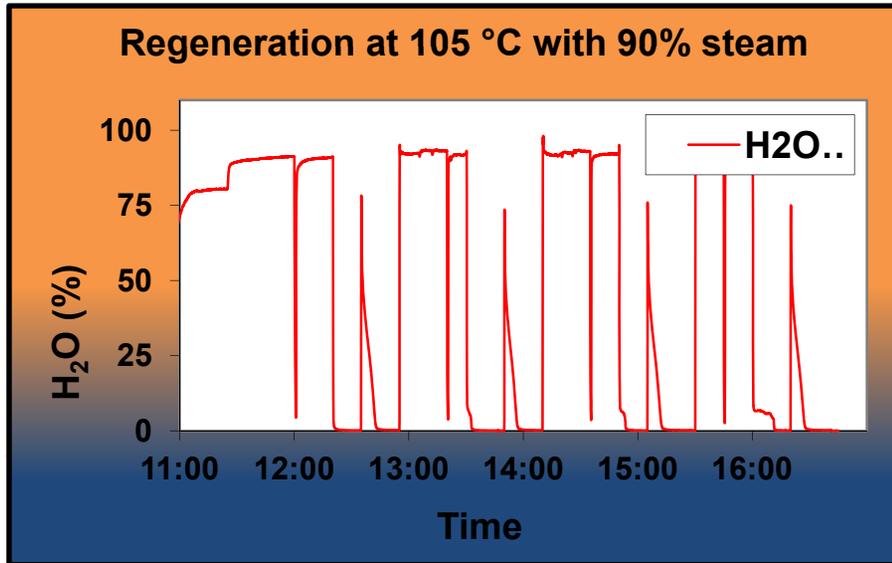
- 40% PEI - BASF M_n 2000
- Mesoporous Silica support- PQ Inc 2129
- 8 - 15 gal. drums from ADA
- Capacity (avg.) = 2.82 mmole CO₂/g adsorbent ranging from 2.60-2.87 for n=7
- ADA packed bed, 13-14% CO₂, ~55°C Capacity (avg.) =1.25 avg. mmole CO₂/g adsorbent
- $u_{mf}=0.48$ cm/s
- Particles behavior is Geldart Group A



Sauter mean particle diameter	(μm)	114
Spericity	(UNITLESS)	0.86
Particle porosity	(UNITLESS)	0.39
Particle skeletal density	(g/cc)	1.50
Particle density	(g/cc)	0.91



BIAS Sorbent Testing



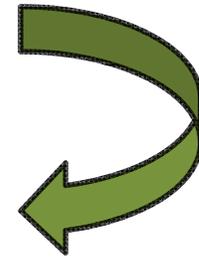
CCSI communication



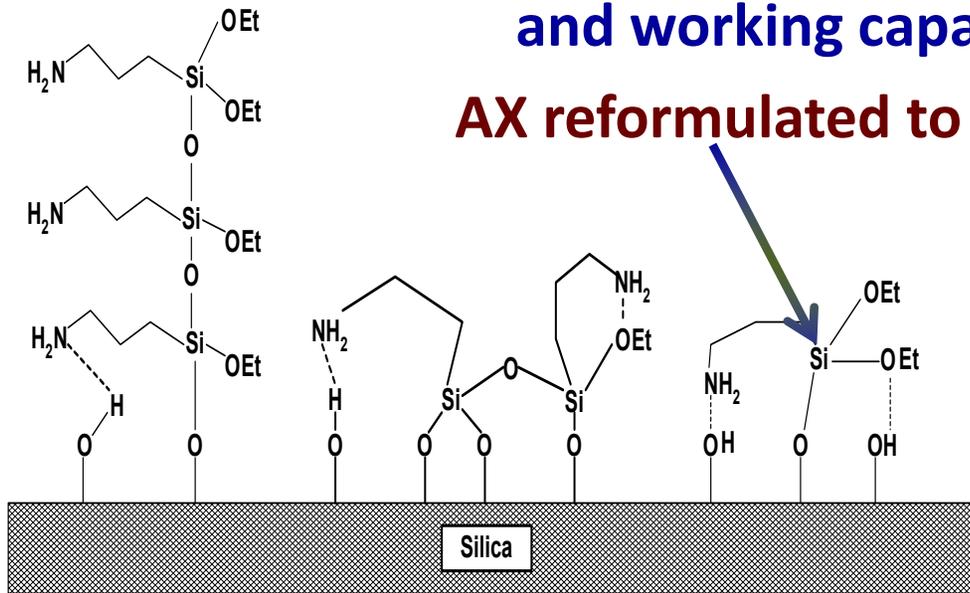
Provided AX sorbent properties to CCSI

Feedback on working capacity and moisture requirements

Reformulates sorbent based moisture and working capacity requirements

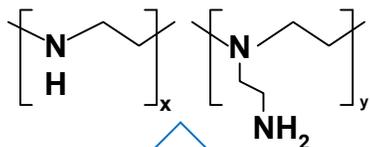


AX reformulated to 32D sorbent



Polyethyleneimine Silane Coupling

Polyethyleneimine Mn 423-2000

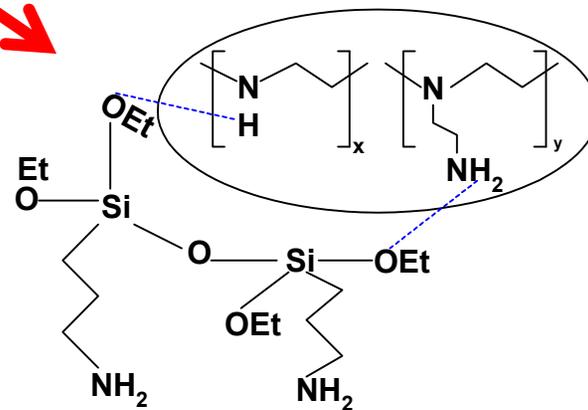


Simple
Scalable
Acceptable Capacity
Moisture Resistance
Stability
Saleable

Methanol 80 C
Vacuum atm- 10 mm Hg

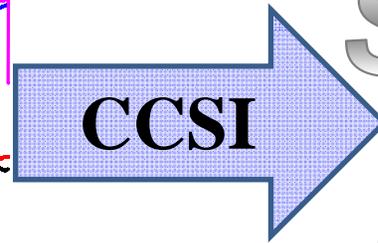
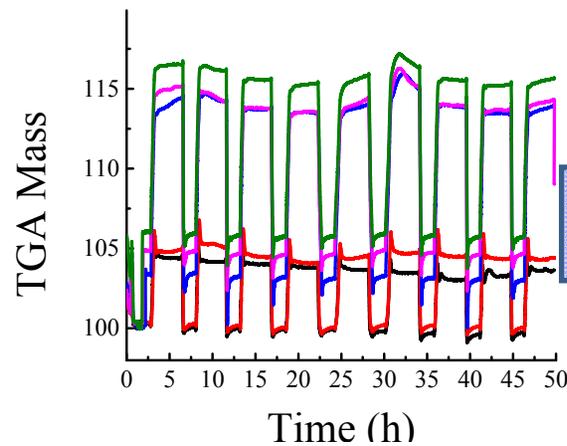
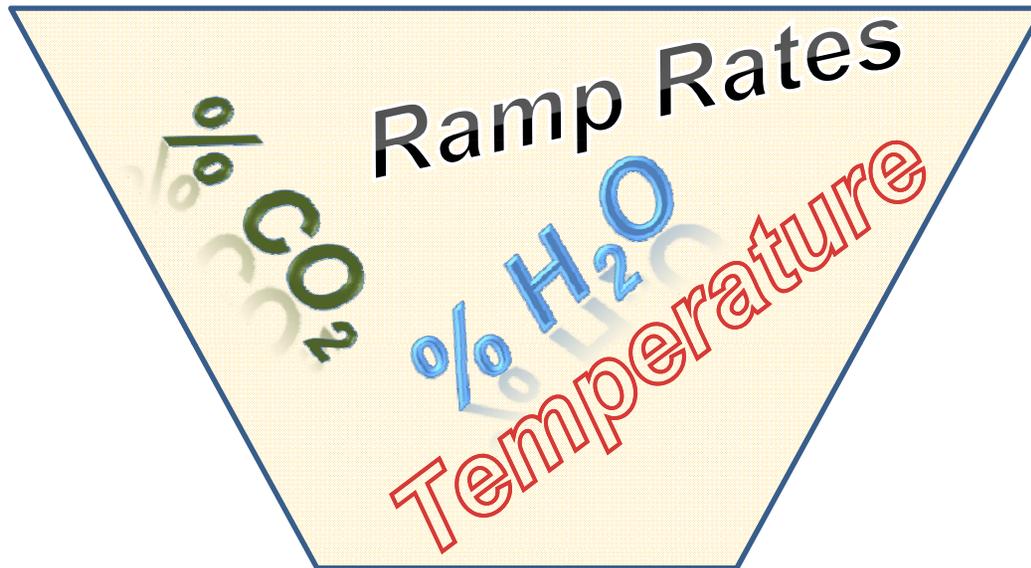


Pressure Chemical – Pan Dyer



Synthesis was scaled to 1,000 lb range

TGA Data Collection



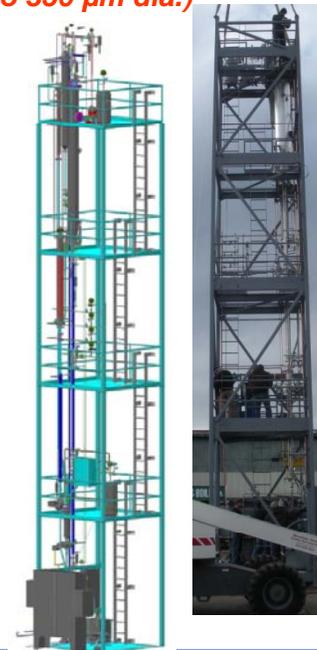
Sorption Kinetics

Summary for Basic Immobilized Amine Sorbent

- High delta loadings in the 3-4mol/kg range
- CO₂ regeneration improbable
- Stable at elevated temperatures
- Silica substrate candidate of choice
- Loading results confirmed by TVA and ADA-ES
- Moisture adsorption may impact energetics
- Susceptible to poisoning with SO₂ and NO₂; upstream cleaning required in process
- Kinetic study conducted



PEI on CARiACT Q10
(100 to 350 μm dia.)



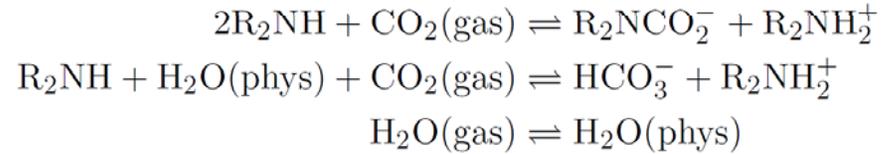
17

Outline

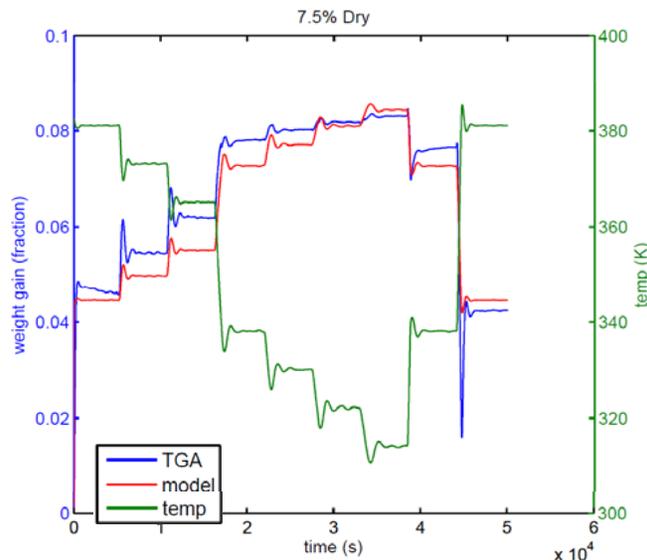
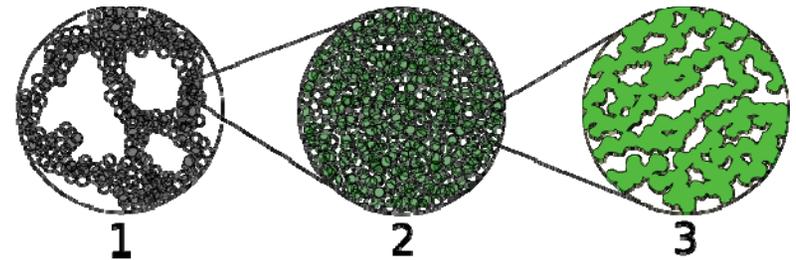
- Carbon Capture Challenge
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PEI-Impregnated Silica Sorbent Reaction Model

- **A general lumped kinetic model, quantitatively fit to TDA data, needed for initial CFD and process simulations**

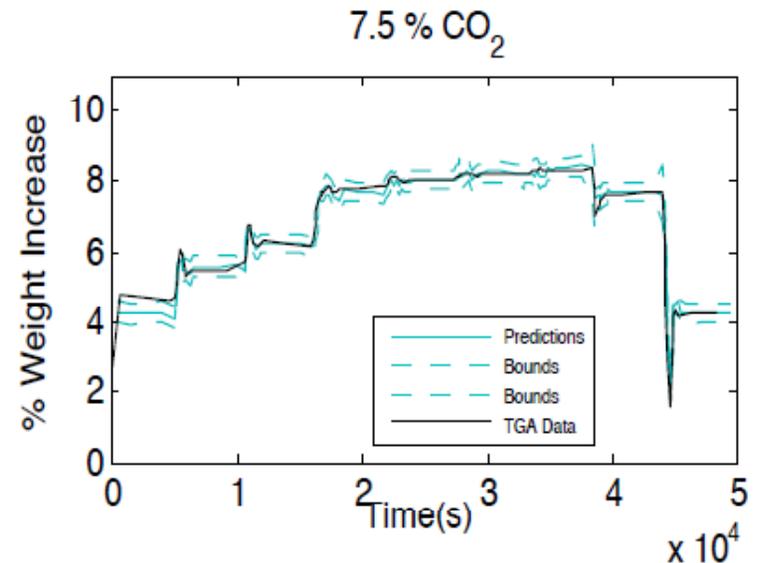


- **High-fidelity model:**
 - Sorbent microstructure broken down into three length scales
 - Rate of reaction controlled by the diffusion of CO₂ through the amine polymer
 - Ab initio calculations indicate dependence of the diffusion process on water.



(left) lumped kinetic fit to experimental TGA for NETL-32D sorbent

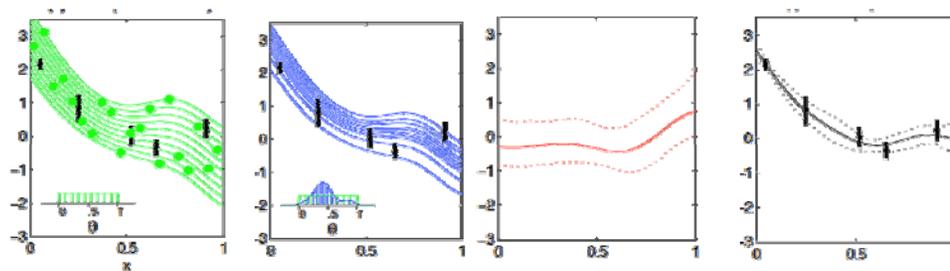
(right) calibrated model with discrepancy and error bounds



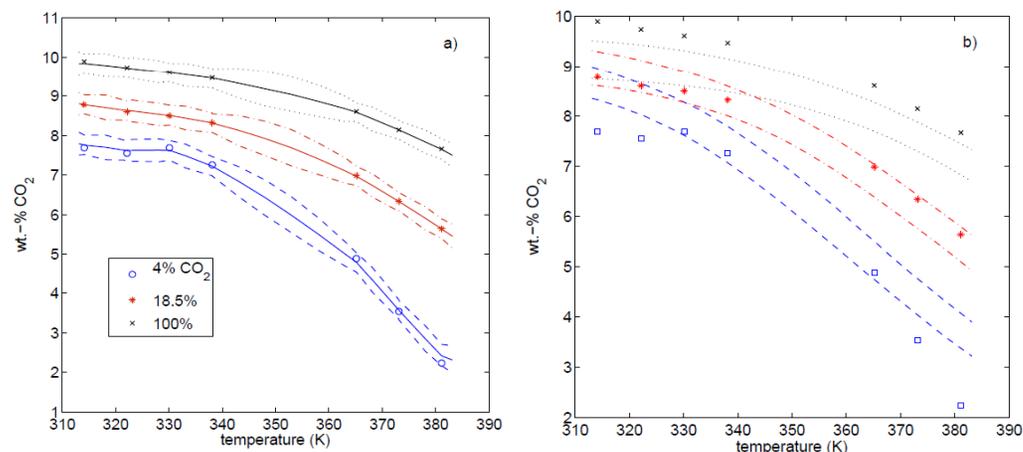
Bayesian Methods in Parameter Estimation

- *Experimental data tends to constrict the prior distribution, resulting in a experiment-based estimate influenced by theoretical calculations.*
- *A stochastic function representing the model error can also be estimated in this way.*

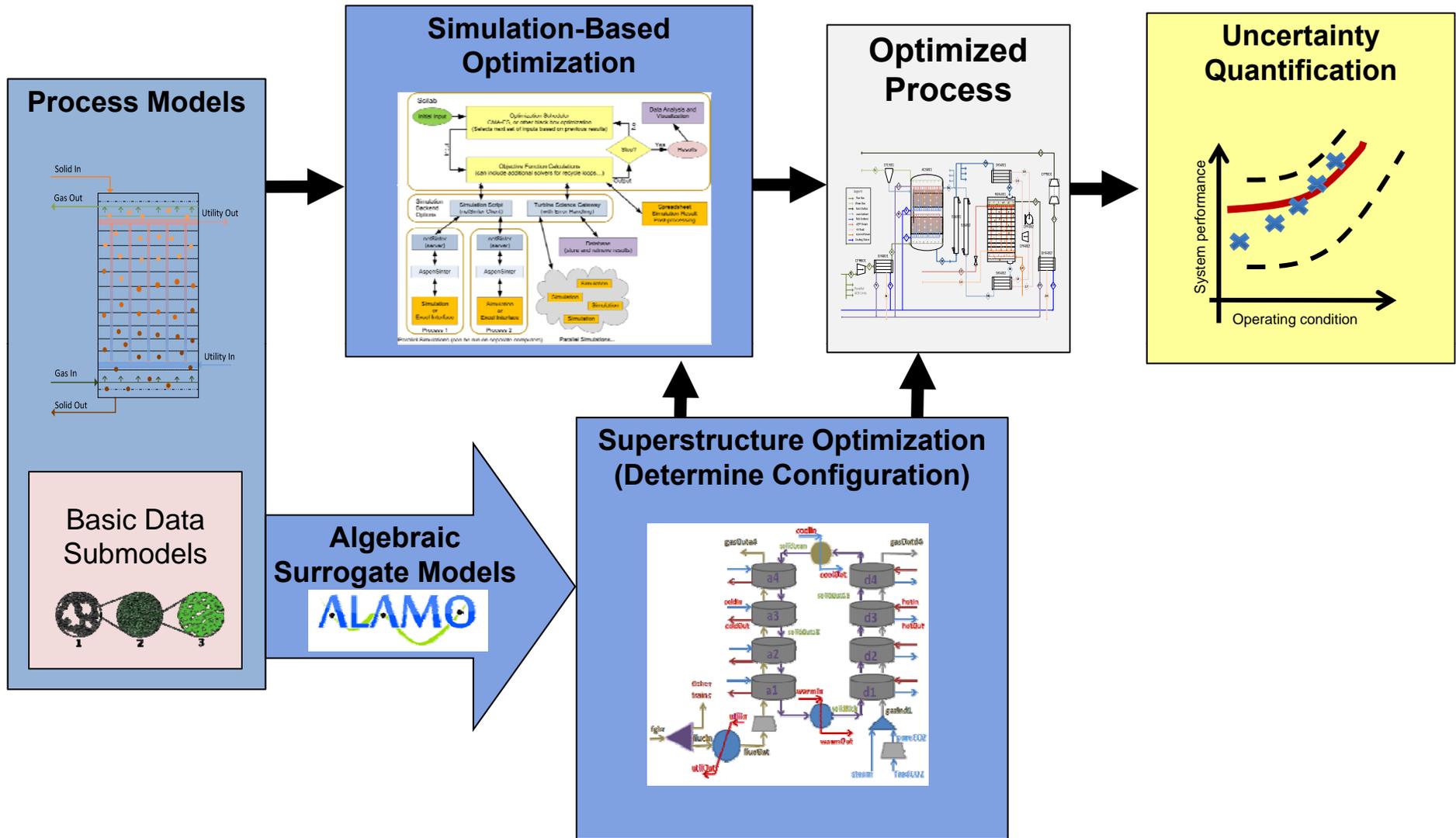
Right: model-plus-discrepancy (a) and model-only predictions (b), with confidence bounds.



Above: schematic of the calibration process. Left to right: draws from the prior, draws from the posterior, discrepancy, and predictions.



CCSI Tools to develop an optimized process using rigorous models



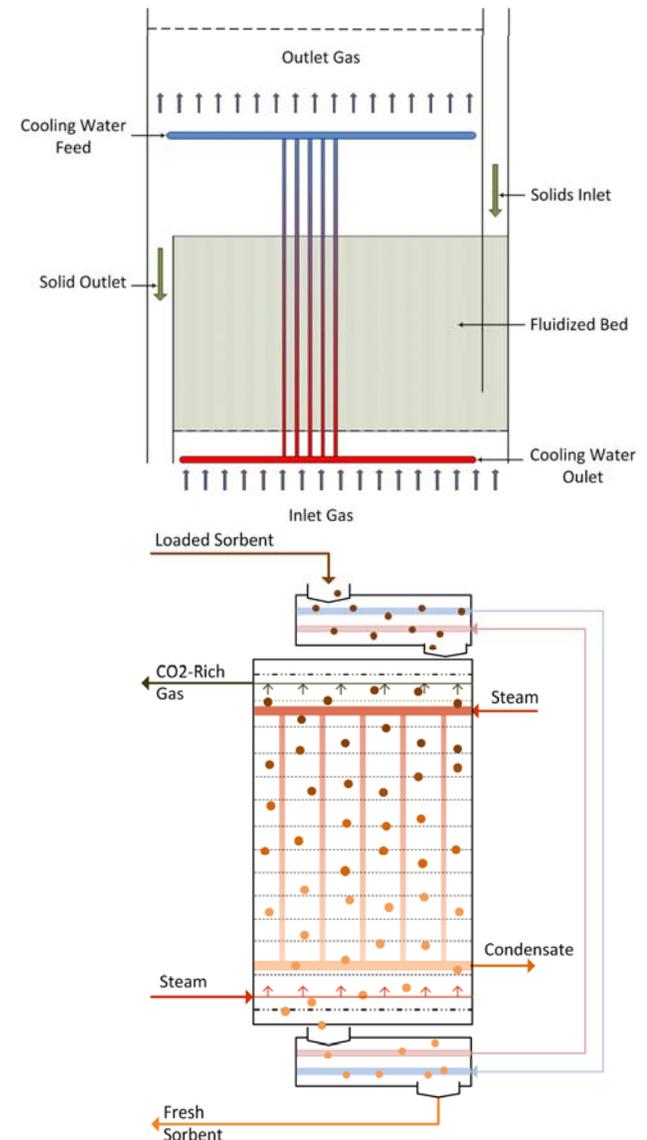
Solid Sorbent System Models

Bubbling Fluidized Bed (BFB) Models

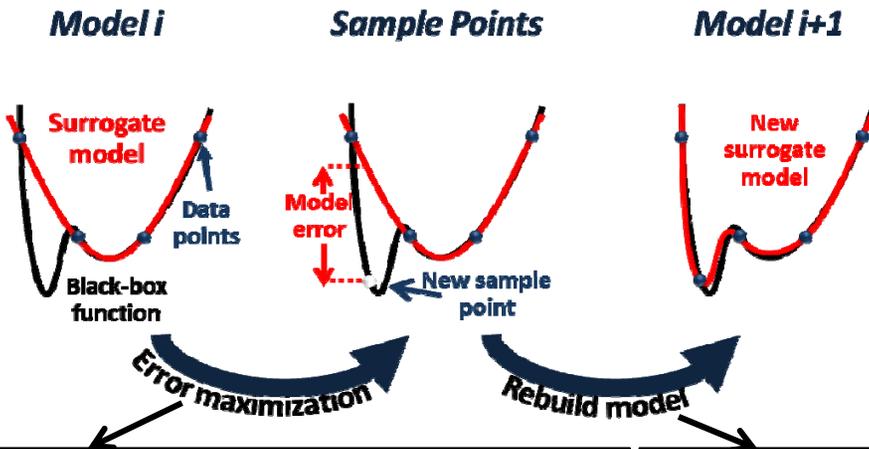
- Flexible BFB models with immersed heat exchangers have been developed to be used as adsorber or regenerator, as needed, with varying locations for solids inlet and outlet streams
- Any number of BFB adsorbers and/or regenerators can be connected in series and/or in parallel depending on the user requirements
- A 2-stage adsorption model with customized variables suitable for incorporating UQ has been developed

Moving Bed (MB) Models

- External resistance to mass transfer has been modeled. This is particularly important in the regenerator model due to the high operating temperature.
- Heat exchanger model, mass and heat transfer coefficients, boundary conditions, temperature specifications, and properties models are revisited for better model accuracy.



Automated Learning of Algebraic Models for Optimization



For building accurate, simple algebraic surrogate models of simulated processes

$$\max_x \left(\frac{z(x) - \hat{z}(x)}{z(x)} \right)^2$$

← **Simulation/black-box**

← **Surrogate model**

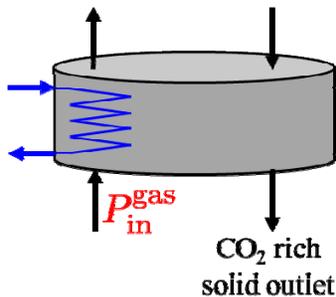
Step 1: Define a large set of potential basis functions

$$\hat{z}(x) = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_1 x_2 + \beta_4 \frac{x_1}{x_2} + \beta_5 \frac{x_2}{x_1} + \beta_6 e^{x_1} + \beta_7 e^{x_2} + \dots$$

Step 2: Model reduction

$$\hat{z}(x) = \beta_0 + \beta_2 x_2 + \beta_5 \frac{x_2}{x_1} + \beta_7 e^{x_2}$$

Example Model: BFB Adsorber Inlet Gas Pressure



- ACM Simulation
- >900 terms possible
- 14 input variables
- 0.13% error

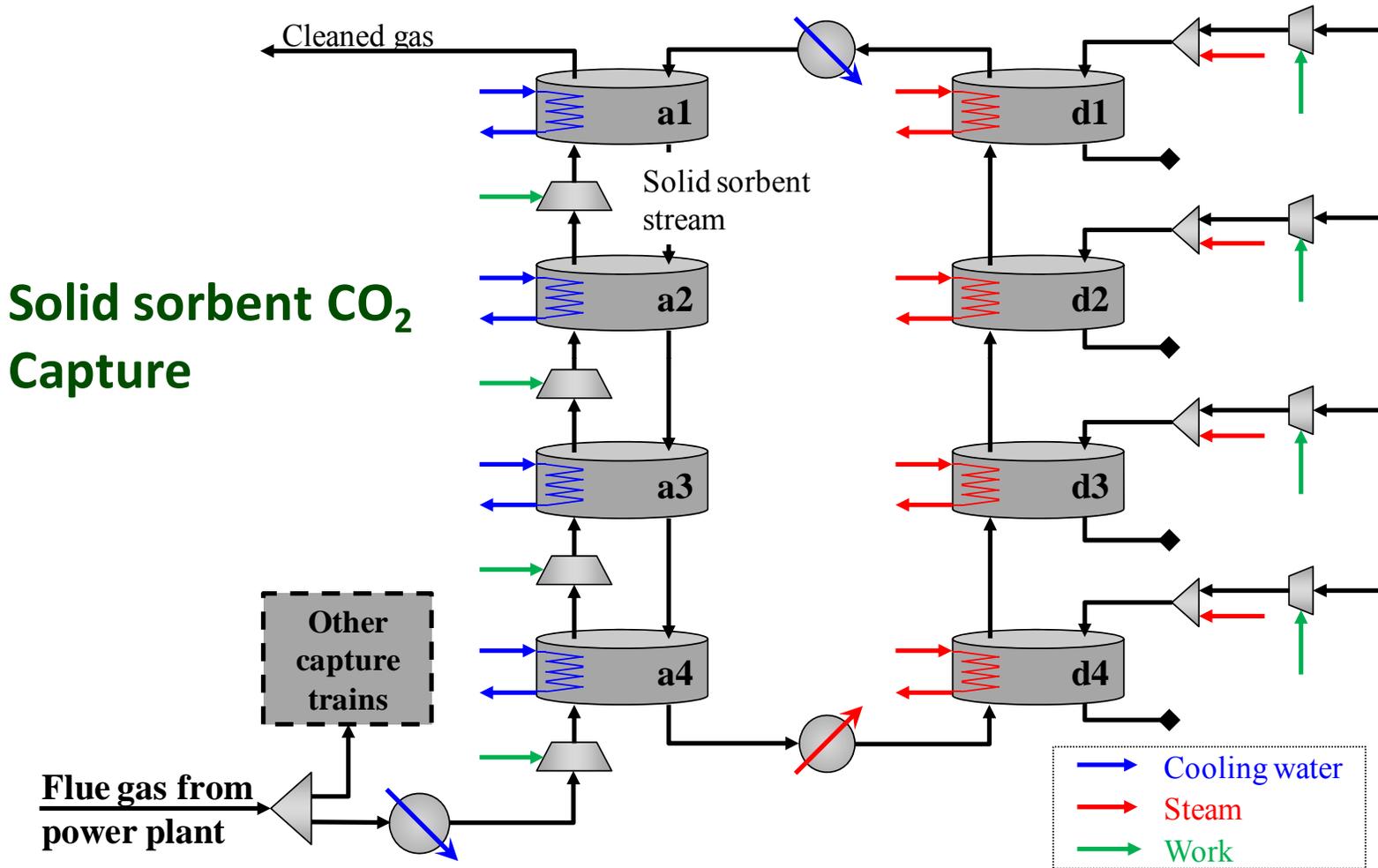
Pressure drop across length of bed

$$\hat{P}_{in}^{gas} = P_{out}^{gas} + 0.019 L_b + 0.0055 \sqrt{D_T}$$

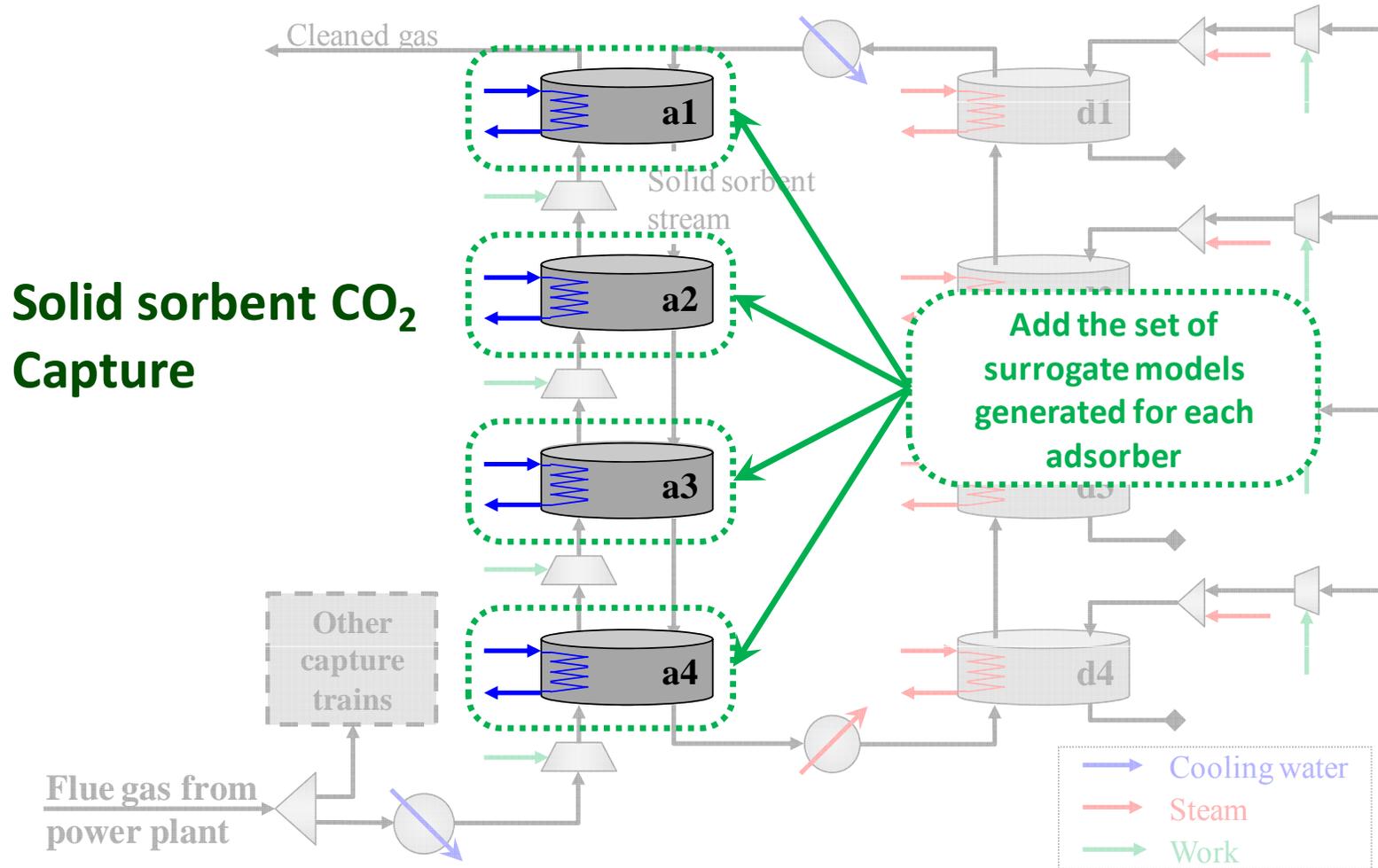
Proportional to outlet pressure

Pressure drop due to bed diameter

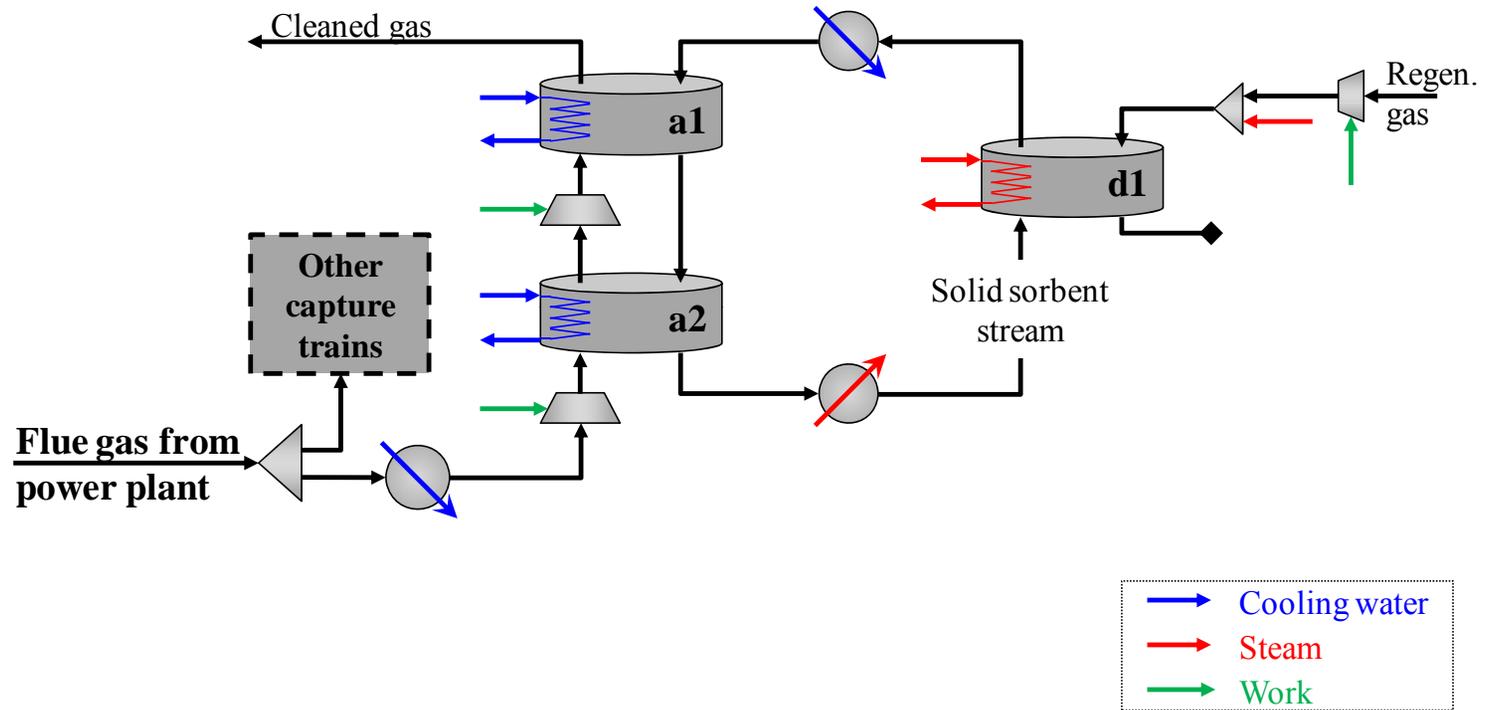
Superstructure Formulation & Optimization



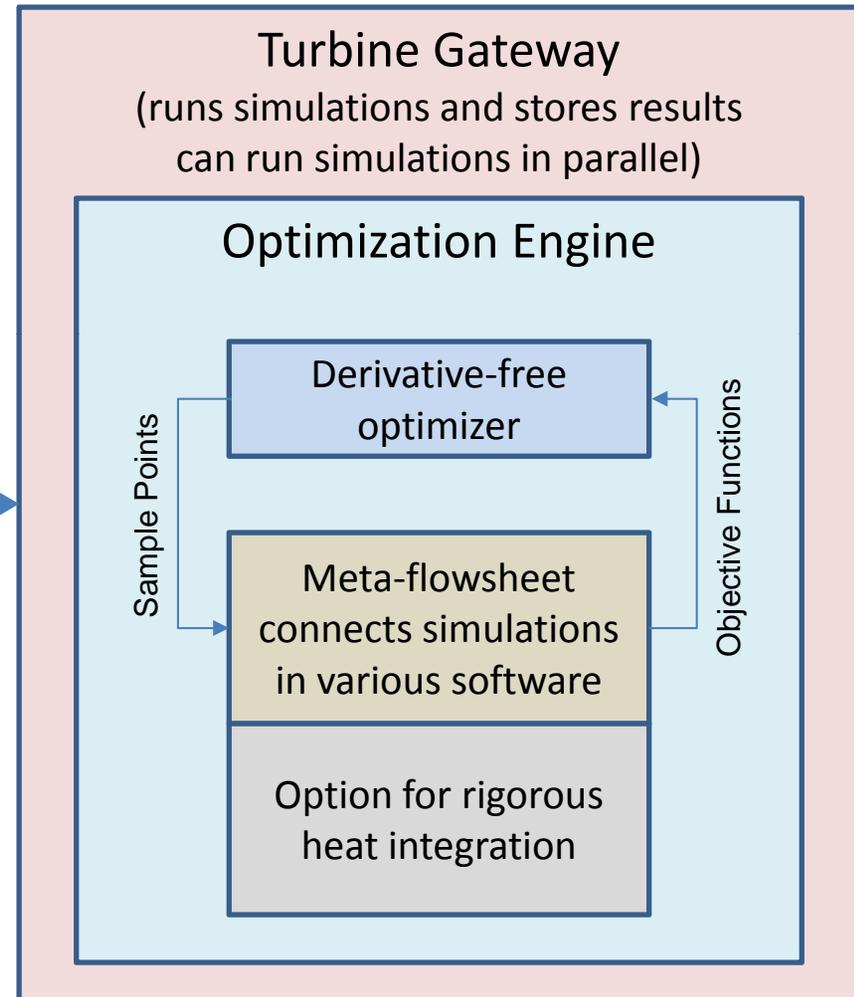
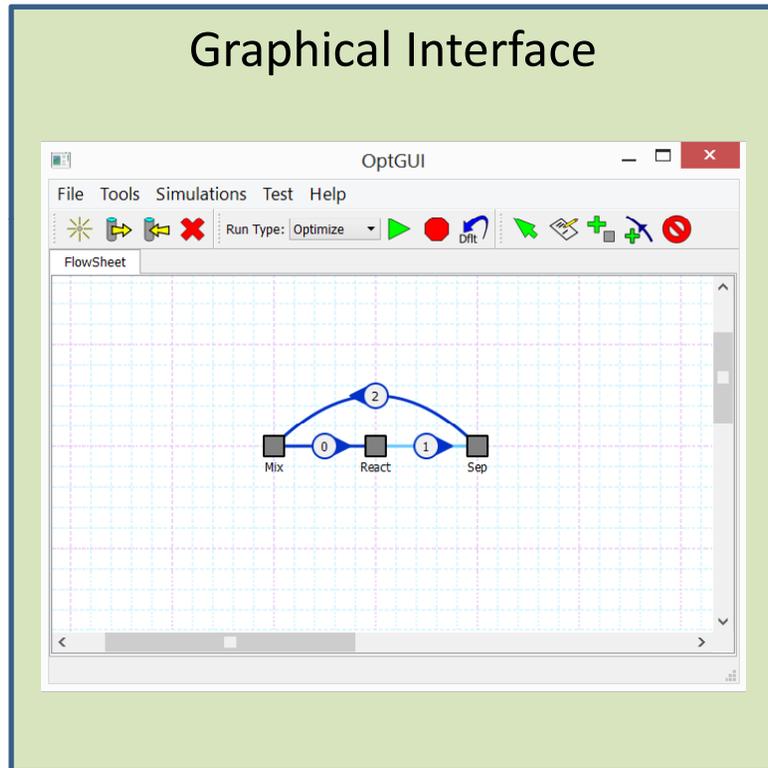
Insert Algebraic Surrogates into Superstructure



Initial Superstructure Solution



Simulation-Based Optimization Framework



Turbine Science Gateway

Turbine Science Gateway
Parallel Simulation Execution
SaaS On-Demand Provisioning
Amazon EC2 Cloud

PSE gPROMS is now supported. Deployed on EC2 with 5 gORun_xml licenses

ACM Hybrid Split Optimization

- Experiment ran 13000 simulations
- 100 simulations per iteration
- 130 iterations Over 2 days using 50 virtual machines

amazon web services™

gPROMS

aspentech

Data Management
Real-time Log
analysis of parallel
simulations

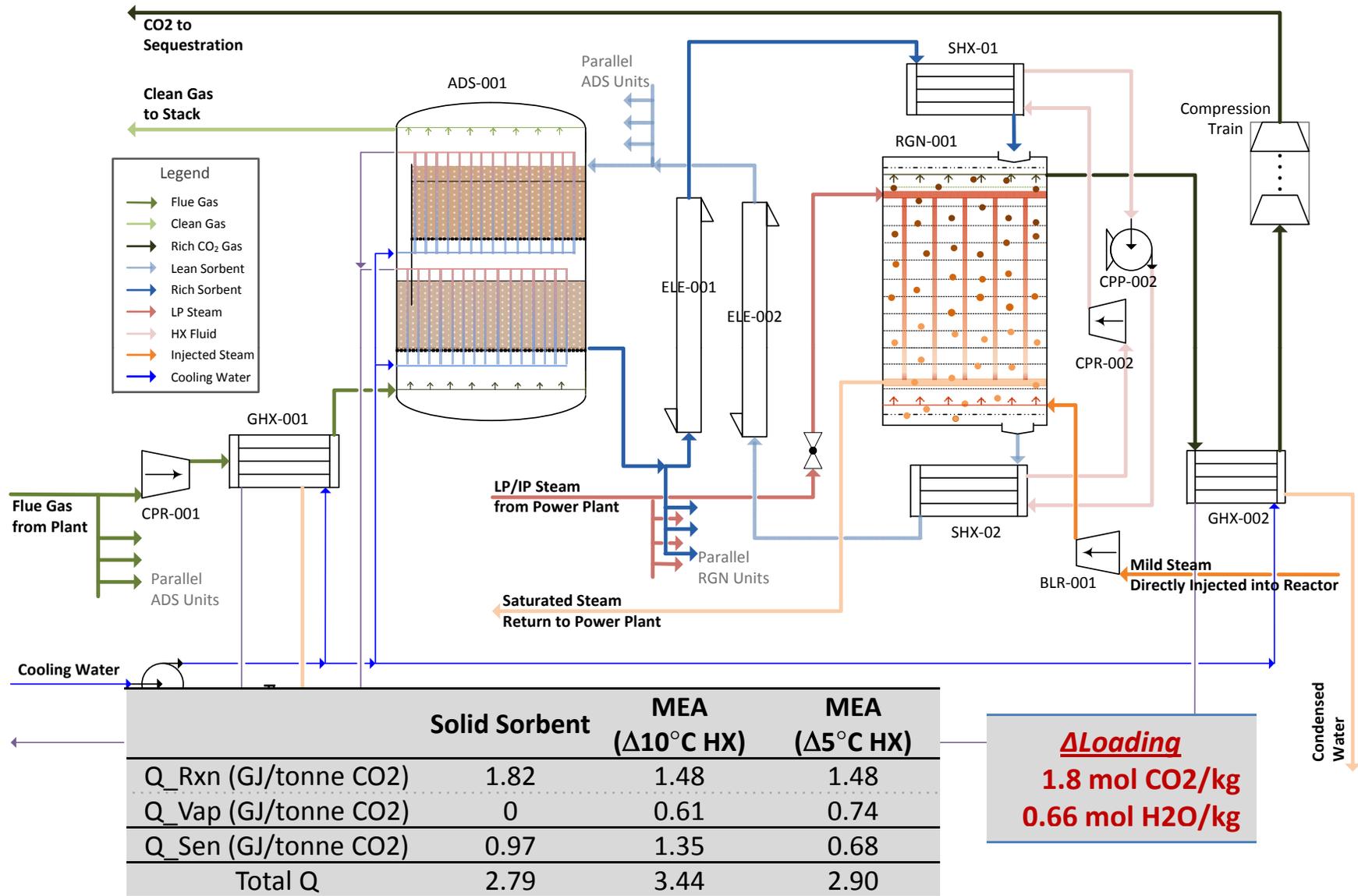
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Total Consumer Activity Over Last Hour

Decision Variables

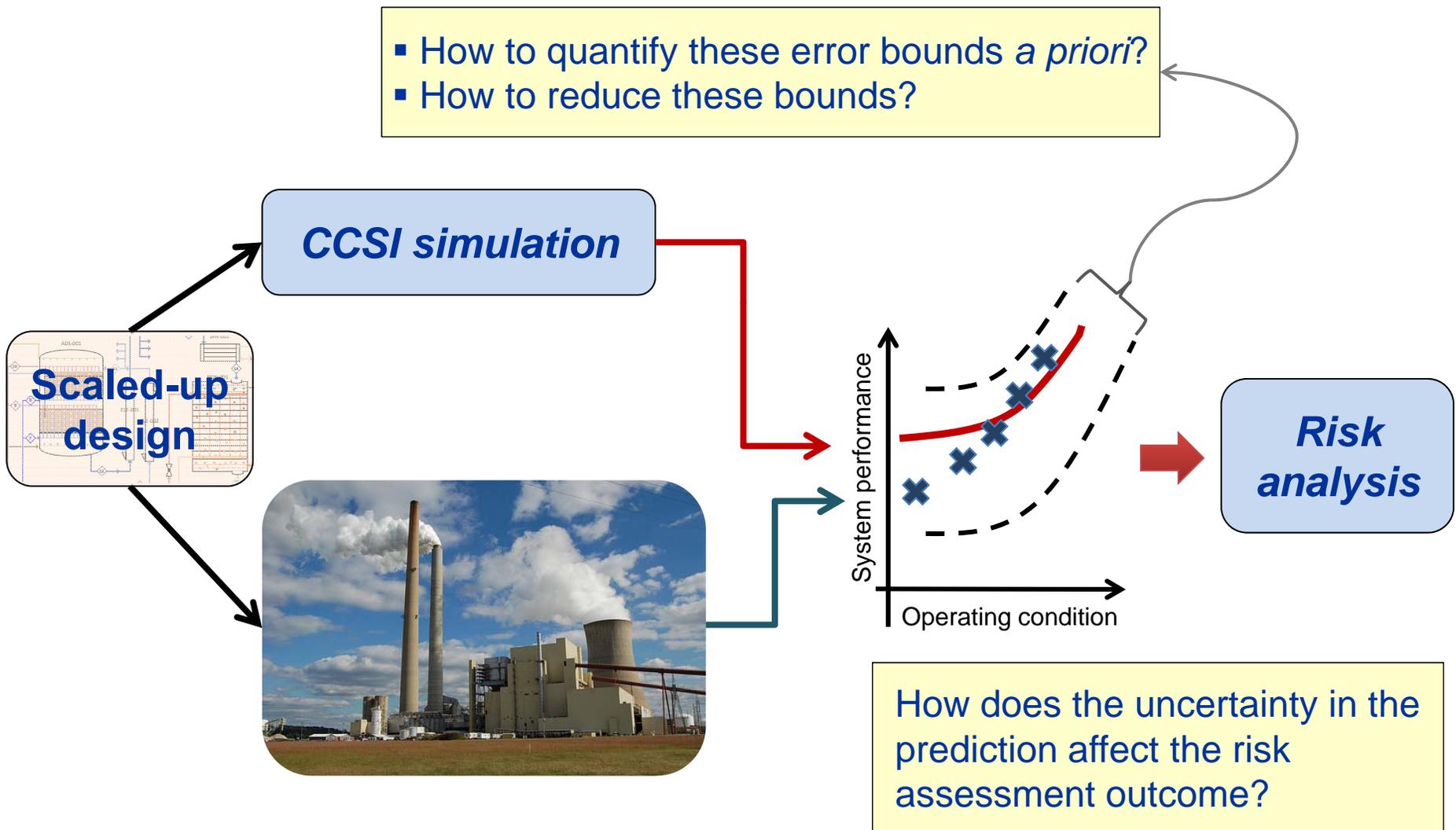
Input Variable	Lower Bound	Upper Bound
Adsorber Diameter (m)	7	10
Top & Bottom Adsorber Bed Depth (m)	4	10
Top & Bottom Adsorber Heat Exchanger Tube Diameter (m)	0.01	0.05
Top & Bottom Adsorber Heat Exchanger Tube Pitch (m)	0.1	0.2
Top & Bottom Adsorber Cooling Water Flowrate (kmol/hr)	30,000	60,000
Sorbent Flowrate per Adsorber (kg/hr)	350,000	600,000
Gas Pre-Cooler Temperature (°C)	40	60
Regenerator Height (m)	3	7
Regenerator Diameter (m)	6	10
Regenerator Heat Exchanger Tube Diameter (m)	0.01	0.05
Regenerator Direct Steam Injection Rate (kmol/hr)	900	1400
Regenerator Heat Exchanger Steam Flowrate (kmol/hr)	2,500	5,000

Optimized Capture Process Developed using CCSI Toolset

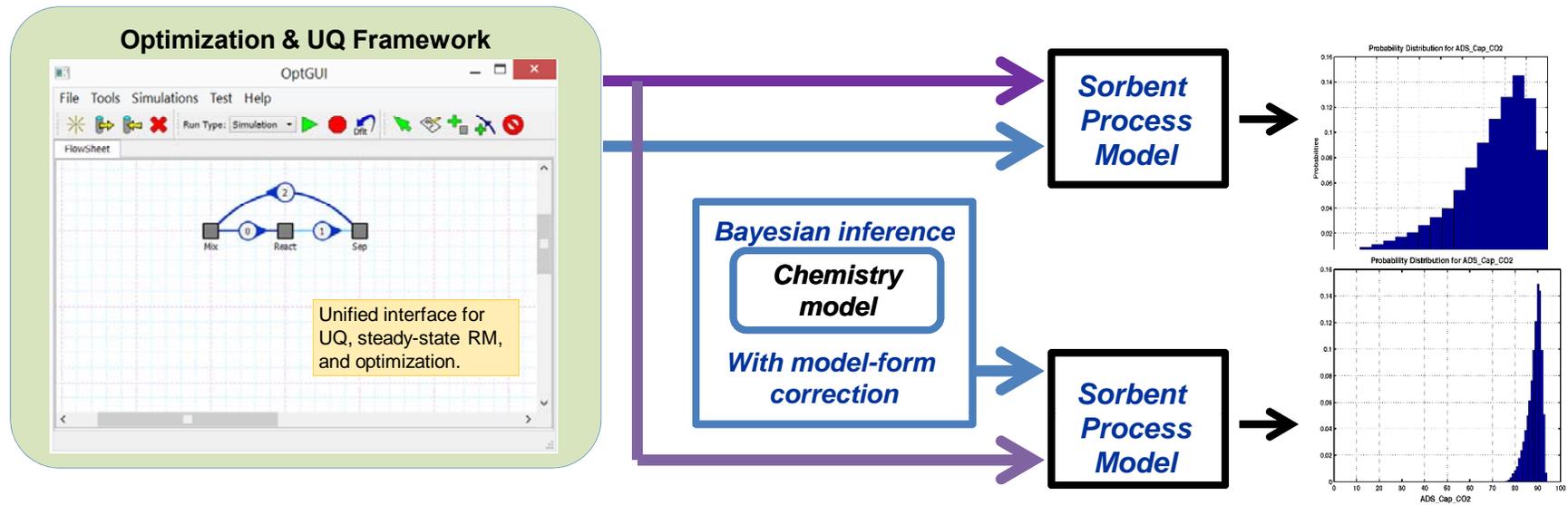


Uncertainty Quantification: How certain are we that our model can predict the system performance accurately?

- How to quantify these error bounds *a priori*?
- How to reduce these bounds?

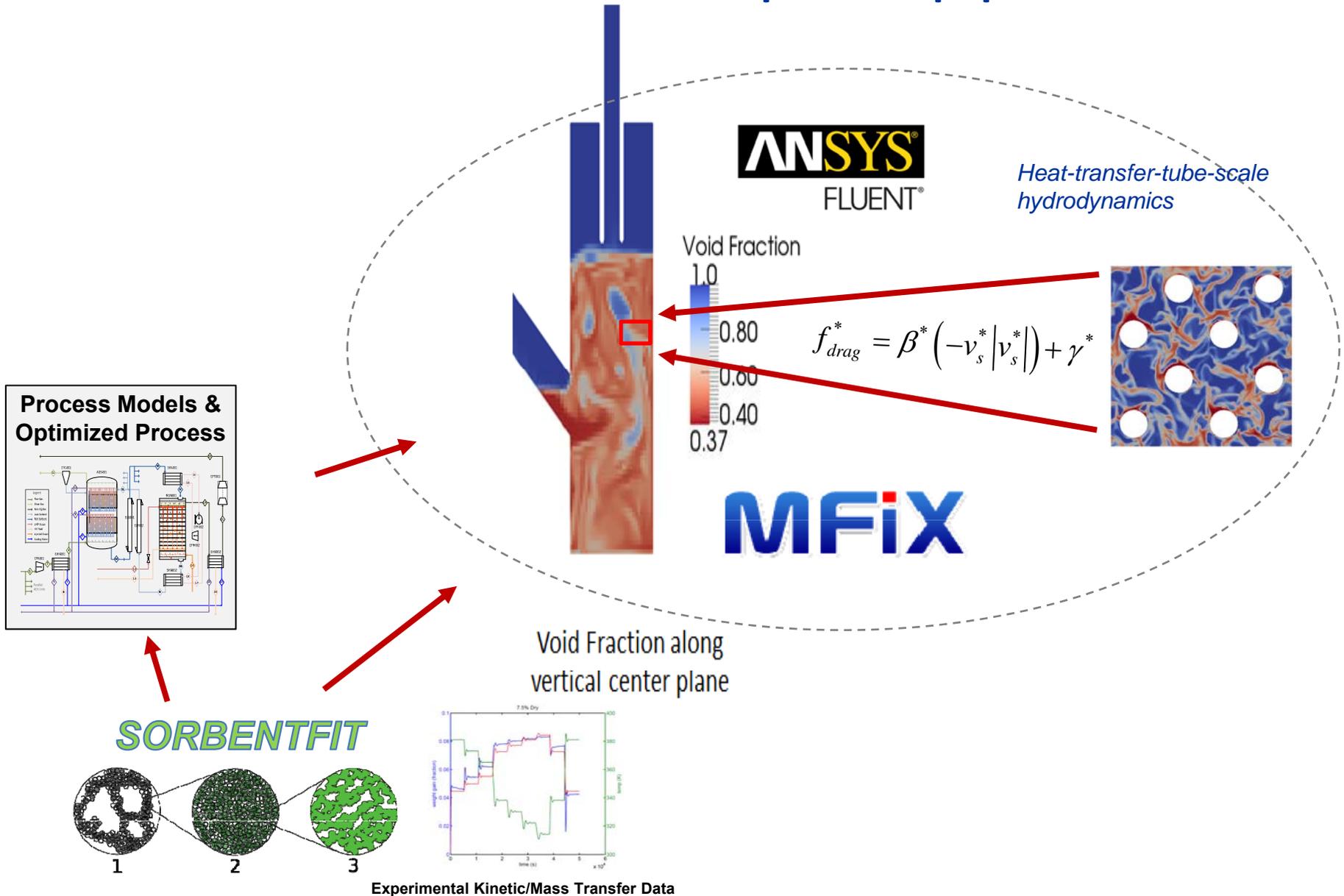


Multi-Scale Uncertainty Quantification Framework



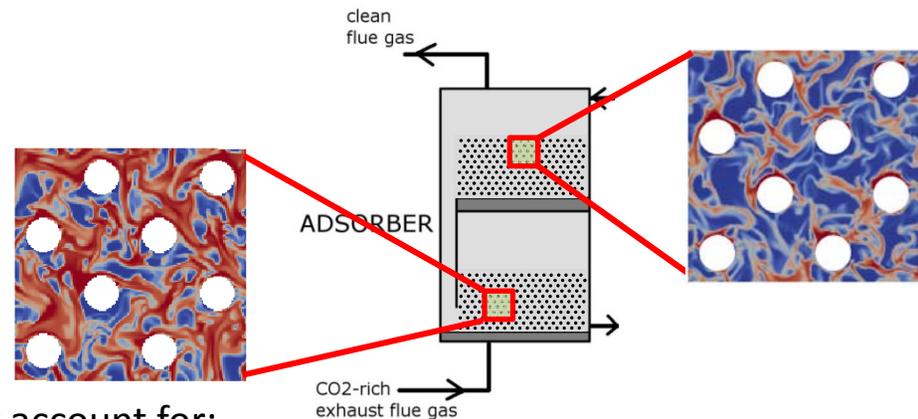
- **UQ for basic data models**
 - Bayesian UQ methodology
 - Integration of model form discrepancy into process & CFD models
- **UQ for CFD models**
 - Adaptive sampling capability for RM/UQ
 - Bayesian calibration capability
 - UQ of discrepancy between CFD/process models
- **UQ for process models**
 - Integration with optimization platform
 - Optimization under uncertainty

Detailed CFD Simulations of Specific Equipment

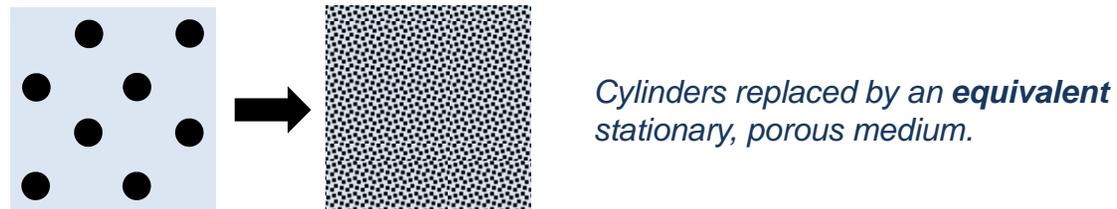


High Resolution Particle Models

- **Problem:** Explicit resolution of small-scale particle clusters and cooling tubes in large devices computationally infeasible!



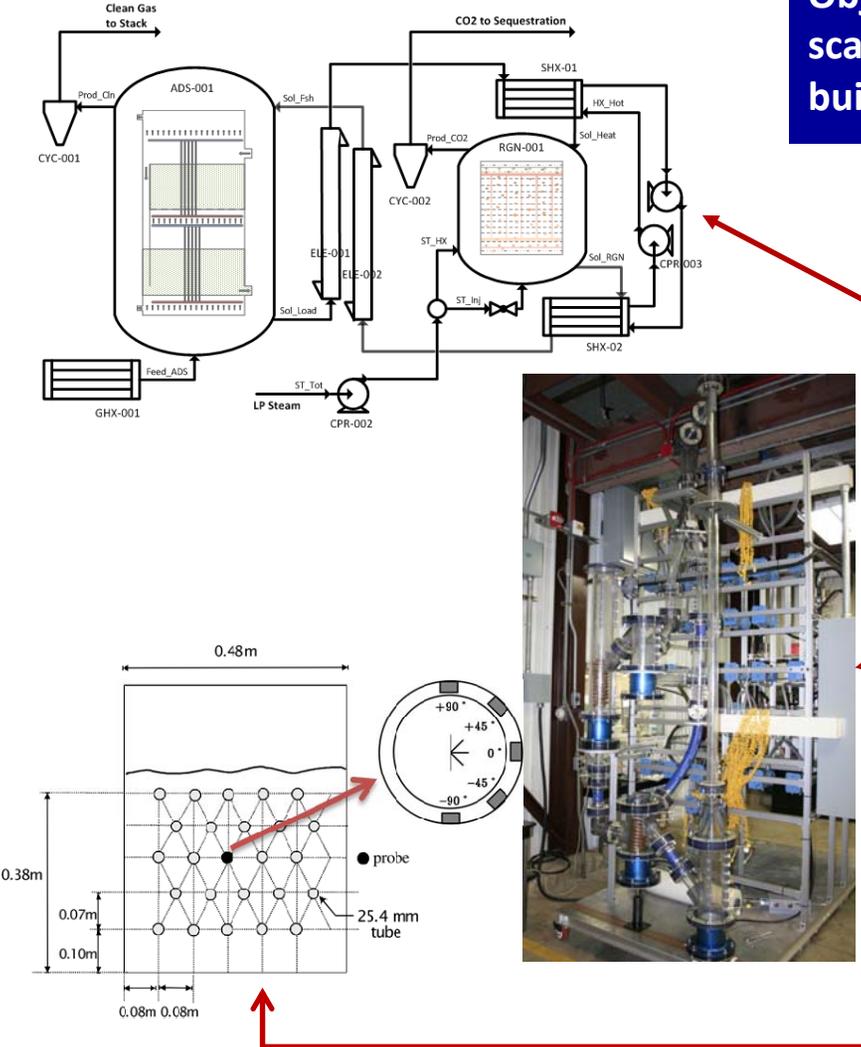
- **Solution** – Develop sub-grid ‘filtered’ that account for:
 - Unresolved particle clusters,
 - Drag exerted by cylinders on suspension. **Work well along the way for horizontal tubes**



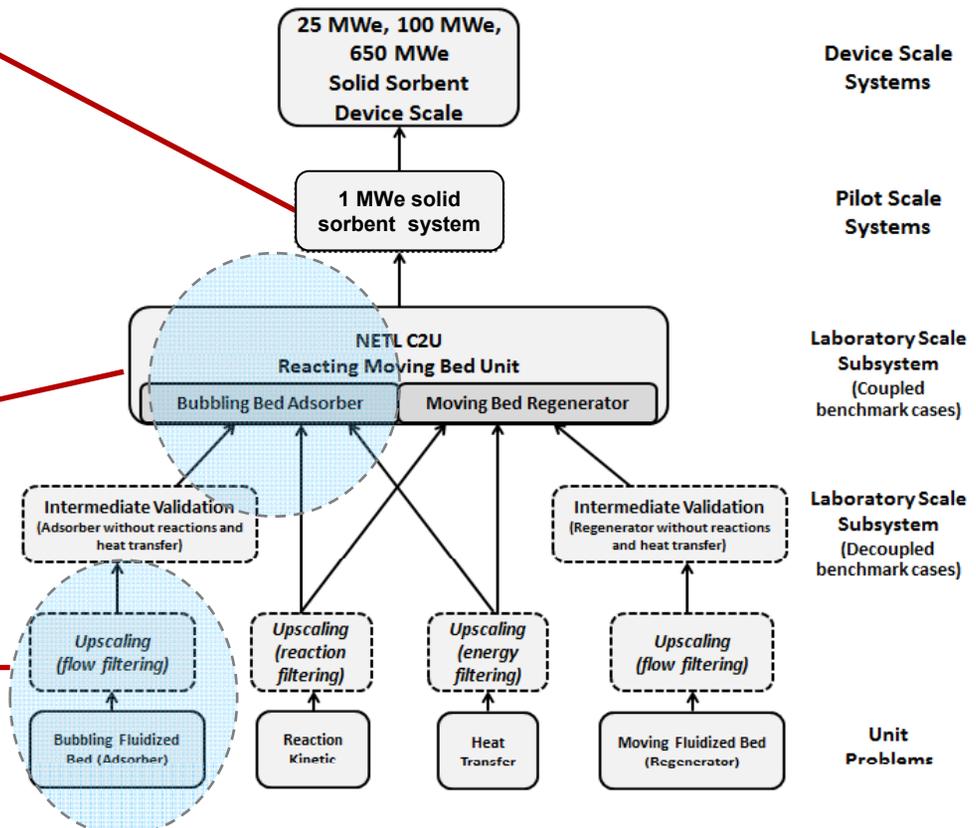
- **Benefit:** The sub-grid ‘filtered’ models can be implemented in faster, coarse-grid simulations.

Validated CFD Models at the Device Scale

Objective: To provide quantitative confidence on device-scale (CFD) model predictions for devices that are yet to be built.

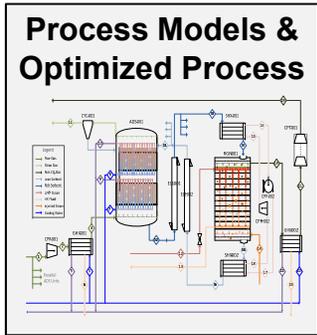


CCSI CFD Validation Hierarchy

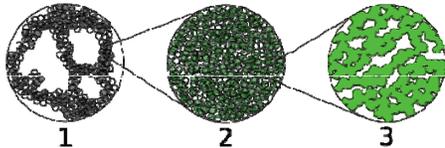


Rigorous Validation of Models

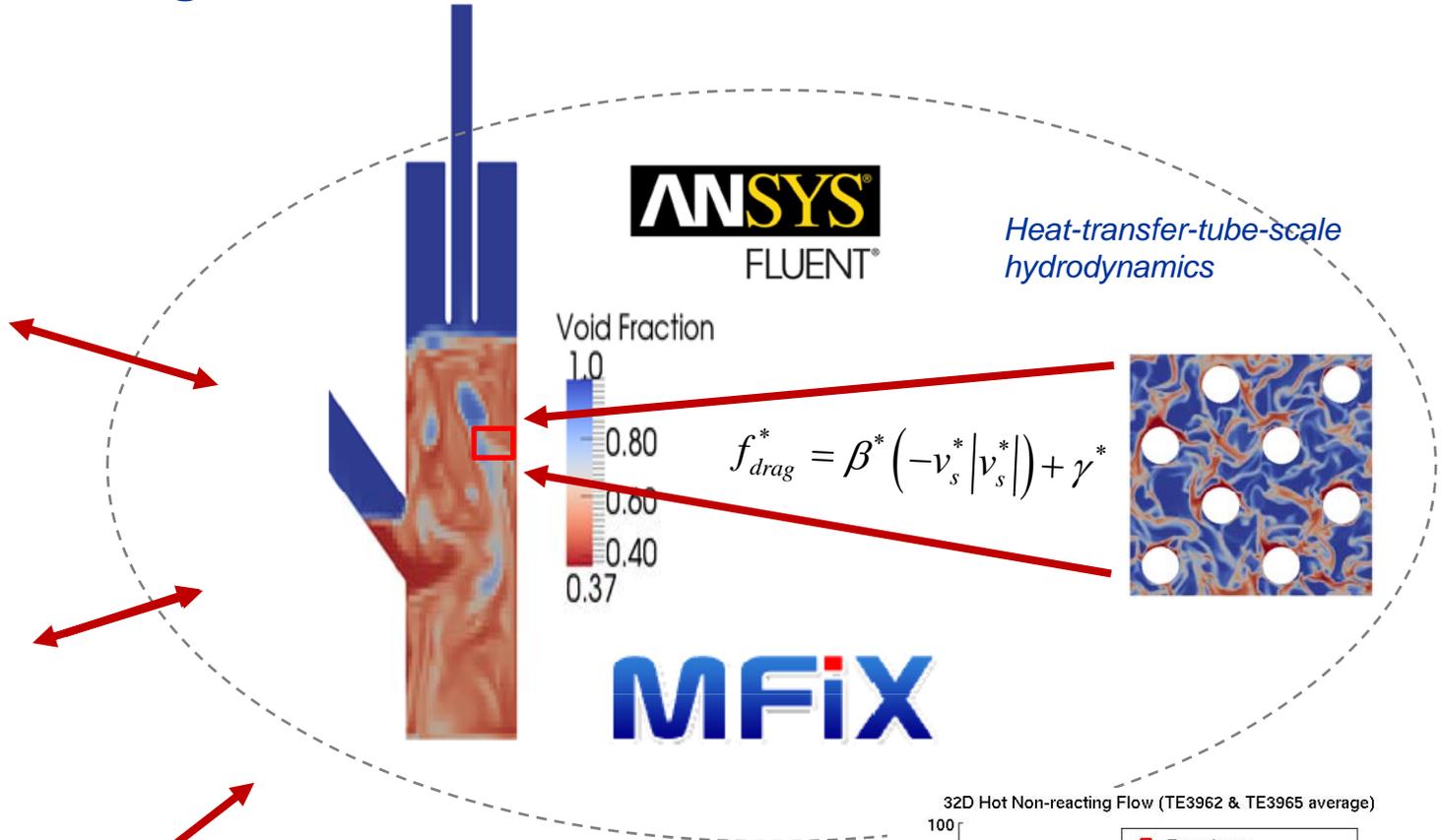
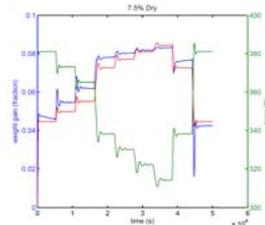
Experimental Validation



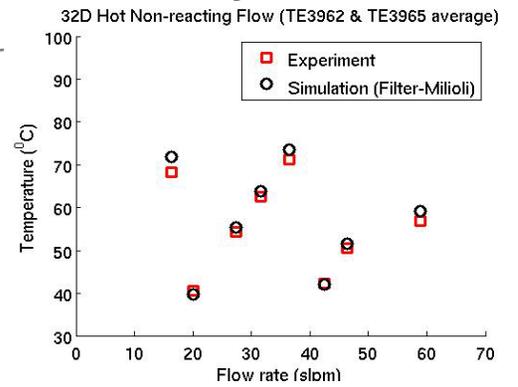
SORBENTFIT



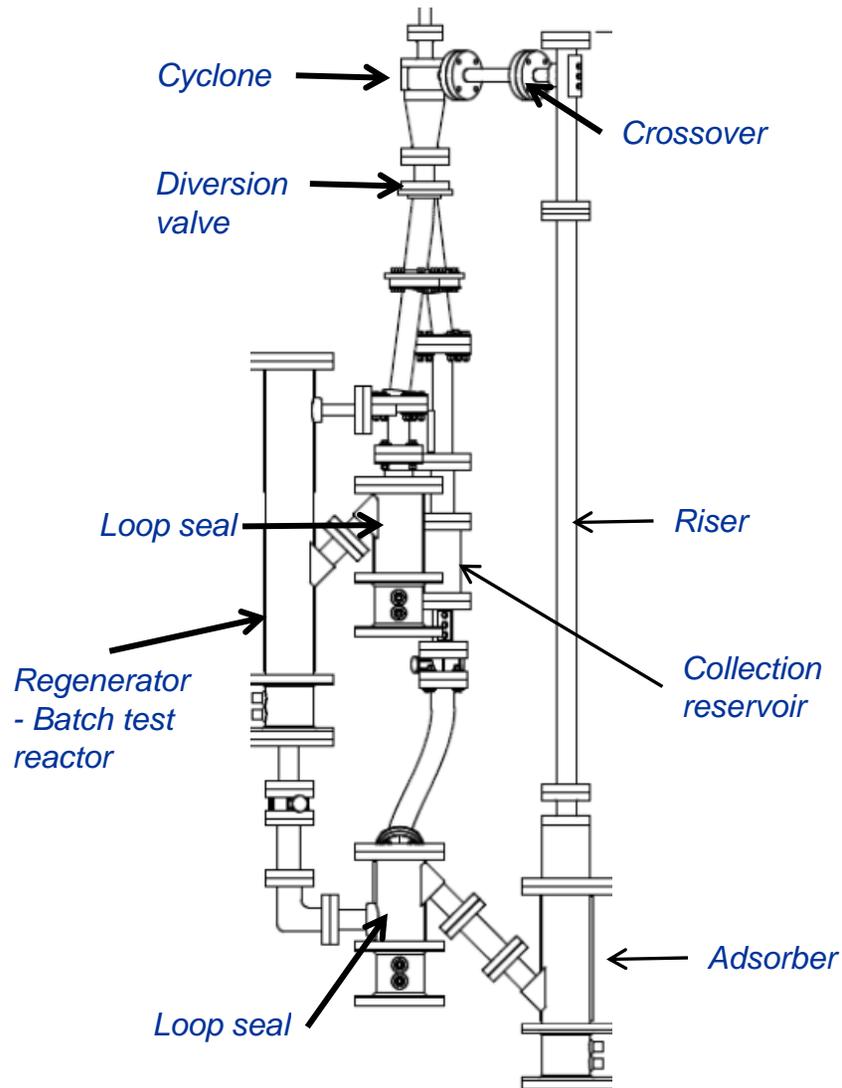
Experimental Kinetic/Mass Transfer Data



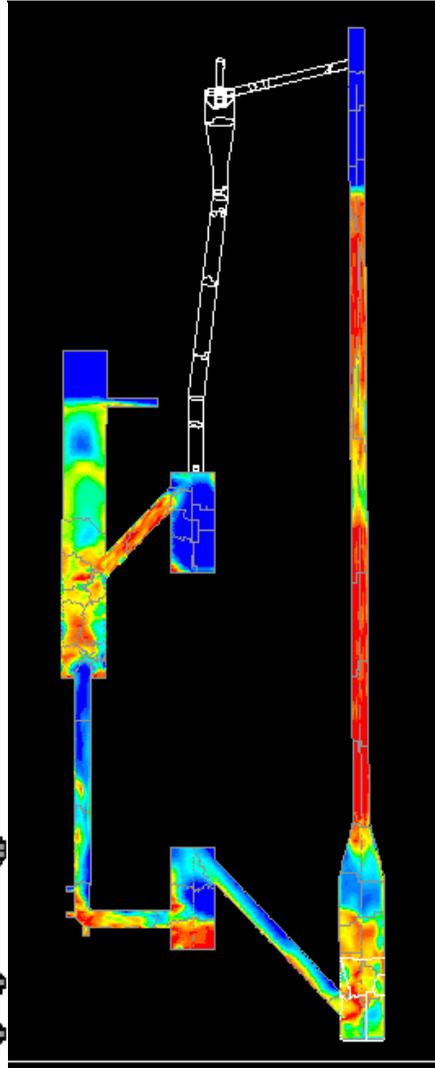
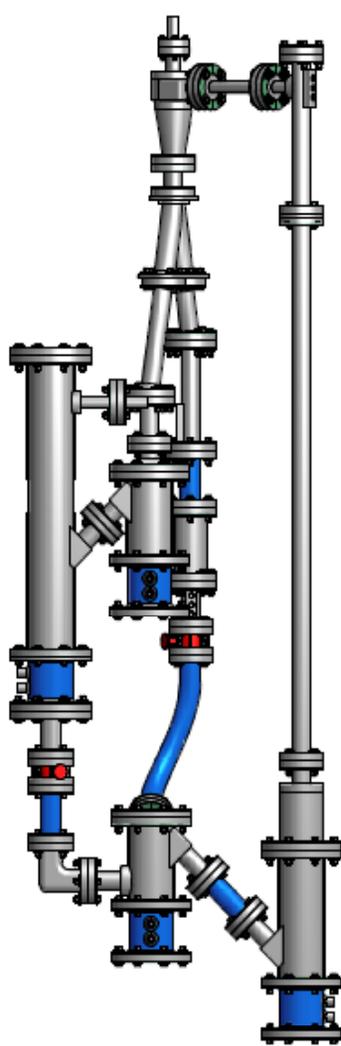
Void Fraction along vertical center plane



C2U and validation data



C2U and validation data



C2U Progression

1. Design and construction
2. Shakedown and modification as needed
3. Revise design drawing
4. Create system models from revised drawings.

C2U and validation data

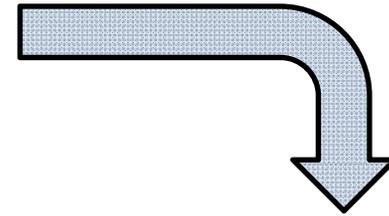
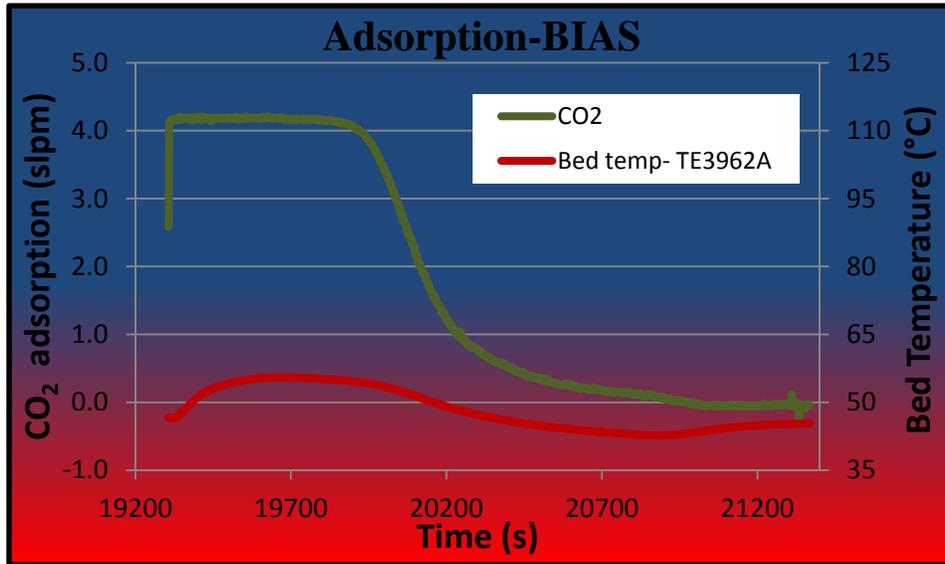
To validate individual models, three sets of tests were statistically devised and randomized

1. Cold Flow testing – hydrodynamics
2. Hot Flow testing – heat transfer, hydrodynamics
3. Reaction testing – reaction kinetics, heat transfer, hydrodynamics

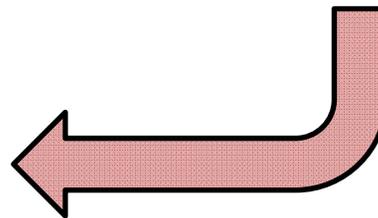
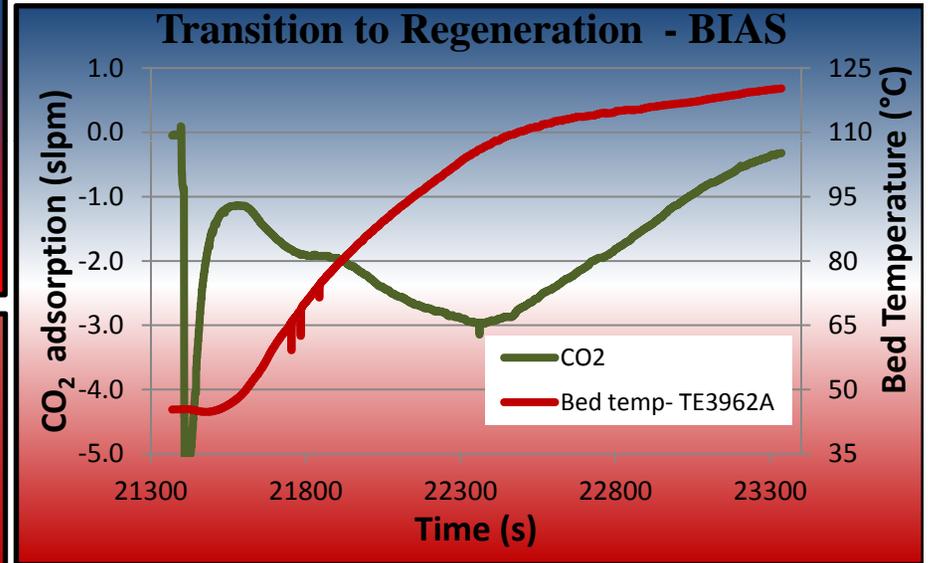
Sorbent AX			Sorbent 32D					
Cold Flow	Hot Flow		Cold Flow	Hot Flow		Reacting Flow		
Flow (SLPM)	Flow (SLPM)	Temp(°C)	Flow (SLPM)	Flow (SLPM)	Temp(°C)	Flow (SLPM)	Temp(°C)	CO ₂ Conc.
49.8	16.9	70.3	19.2	21.9	60.4	51.3	62.9	18.9
15	48.8	60.2	23.6	39.5	66.7	40	62.3	10.8
58.9	35.4	56.3	50.3	43.8	45.3	37	68.2	14.6
43.7	38.6	67.4	51.7	46.3	40.3	27.3	72.8	15.7
35.7	57.1	43.5	37	36.7	65	51.3	57.5	14.2
29.8	20.8	58.4	31.9	31.9	52.1	59	40.7	17.4
25.1	27.1	52.4	51.9	16.4	71.1	33.7	60.8	16.3
20.4	30	42.7	51.6	26.7	47.7	35.6	41.6	10.4
54.6	48.6	49	35.9	22.7	53.4	23.1	64.3	16.9
40.3	26	77.2	25.4	52.7	78.1	32.8	59	18.1
38.4	54.7	65.6	59.6	58.1	57.7	41.3	70.4	20
23.2	41.9	74.2	33.3	51	73.4	20.5	44.8	14.3
31.6	40.6	61.8	16.1	30.6	79.1	56.4	63.6	13.7
28	52	73	27.4	46	68.8	17.9	48	10.5
47.8	23	47.4	21.6	35.2	55.2	43	71.7	18.2
18	15.8	45.3	43.3	19.3	65.7	56.9	57.9	10.2
33.4	29.5	69.7	54.9	39.3	43.5	32	79.1	15.3
52.6	44.4	40.5	41.4	49.4	63	49.1	61.6	17.1
46	36	50	29.6	29.7	76	44.8	65.4	19.7
56.9	19.2	76.3	47.7	56.1	42.6	50.2	46.2	16.5
15	32.2	63.8	16.1	35.2	58.5	59.7	44.3	13
31.6	46.1	54.4	45.9	17.1	48.5	57.5	59.3	15.9
49.8	53.3	56.8	57.5	42.3	51.6	38.2	76.9	14.9
35.7	58.7	79.4	41.4	25.8	71.9	44.5	67.8	12
20.4	15.8	45.3	25.4	17.1	48.5	4.3	52.7	19.5
46	38.6	67.4	33.3	30.6	79.1	52	50.2	11.4
25.1	36	50	21.6	58.1	57.7	45.9	60.1	12.6
43.7	58.7	79.4	29.6	51	73.4	15.9	67	12.2
56.9	53.3	56.8	51.7	19.3	65.7	35.8	63.1	13.9
54.6	19.2	76.3	50.3	43.8	45.3	55	59	14.6
29.8	37	52.4	31.9	48.2	54.9	54.9	54.9	13.1
38.4	43.3	51.7	47.8	54.4	75.7	11.9	11.9	11.9
47.8	19.2	76.3	43.3	15.1	56.4	15.1	56.4	15.1
23.2	27.4	52.4	47.8	34.6	47	47	47	12.7
58.9	54.9	54.9	54.9	30.2	77.6	30.2	77.6	30.2
33.4	39.7	51.7	51.7	26.7	48.3	10.7	10.7	10.7
18	31.9	51.7	51.7	19.9	44	17.5	17.5	17.5
40.3	47.7	51.7	51.7	28.6	50.7	13.3	13.3	13.3
28	59.6	51.7	51.7	25.1	74.5	11.6	11.6	11.6
52.6	23.6	51.7	51.7	22.2	76.4	16.1	16.1	16.1
				46.8	40.9	19.2	19.2	19.2
				39.2	66.1	18.4	18.4	18.4
				30.8	42.7	13.4	13.4	13.4
				44.5	69.1	15.6	15.6	15.6
				42.3	56	11.1	11.1	11.1
				53.4	72.5	18.8	18.8	18.8
				19	79.3	17.7	17.7	17.7
				27.9	73.7	19.3	19.3	19.3
				17.6	51.8	17.5	17.5	17.5
				40.5	49.4	11.6	11.6	11.6
				19	79.3	17.7	17.7	17.7
				32.8	59	18.3	18.3	18.3
				54.4	75.7	11.9	11.9	11.9
				46.8	40.9	19.2	19.2	19.2
				38.2	76.9	14.9	14.9	14.9
				24.5	67.8	12	12	12
				20.5	44.8	14.3	14.3	14.3
				57.5	59.3	15.9	15.9	15.9
				56.9	57.9	10.2	10.2	10.2
				35.6	41.6	10.4	10.4	10.4

Sorbent 32D					
Cold Flow	Hot Flow		Reacting Flow		
Flow (SLPM)	Flow (SLPM)	Temp(°C)	Flow (SLPM)	Temp(°C)	CO ₂ Conc.
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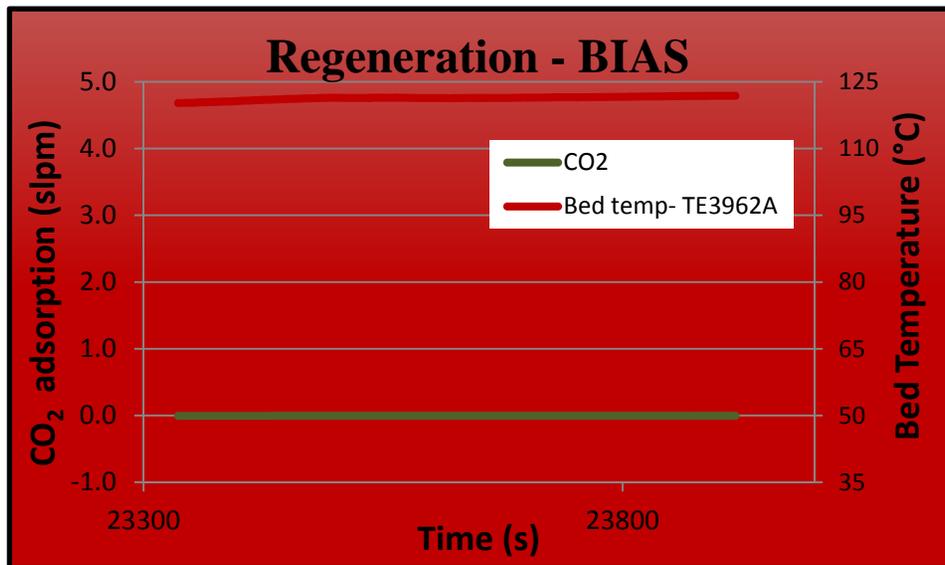
C2U Initial testing data



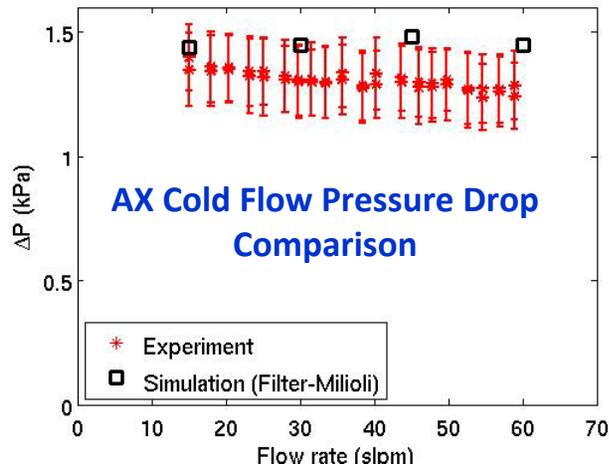
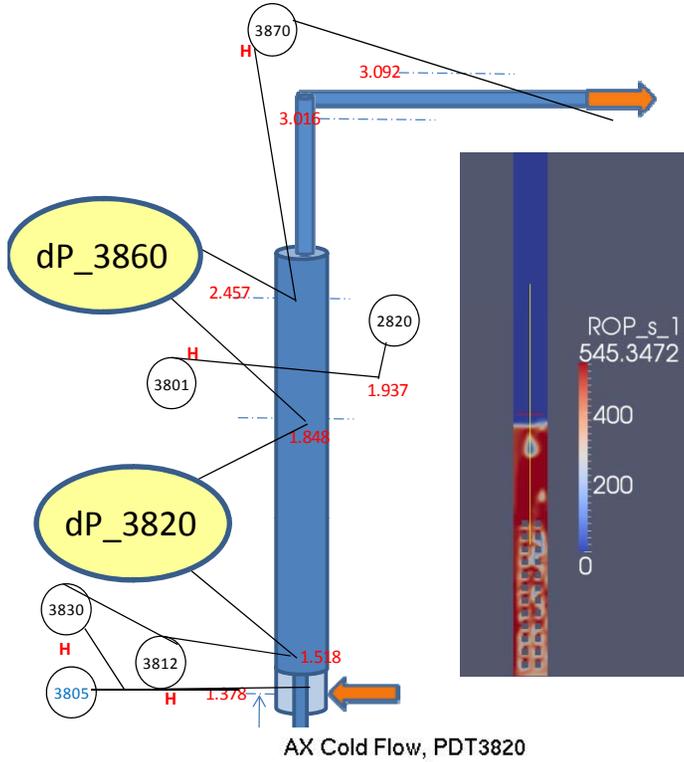
Batch test conducted in the regenerator with the BIAS sorbent



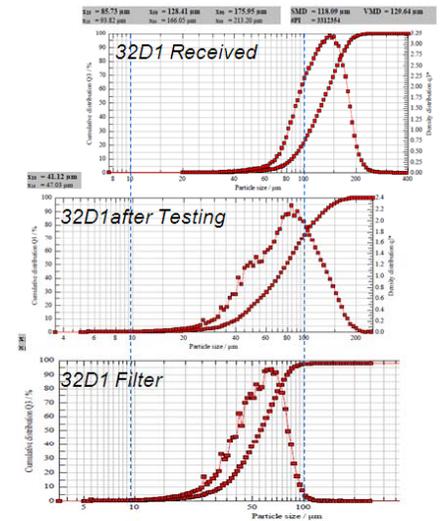
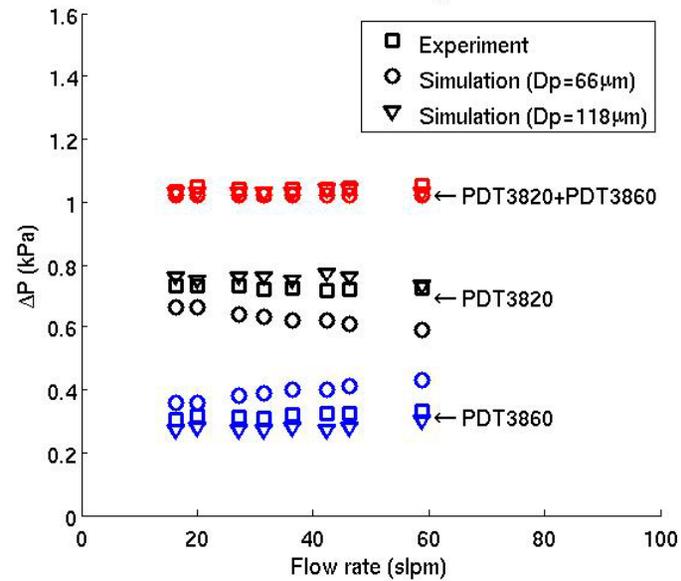
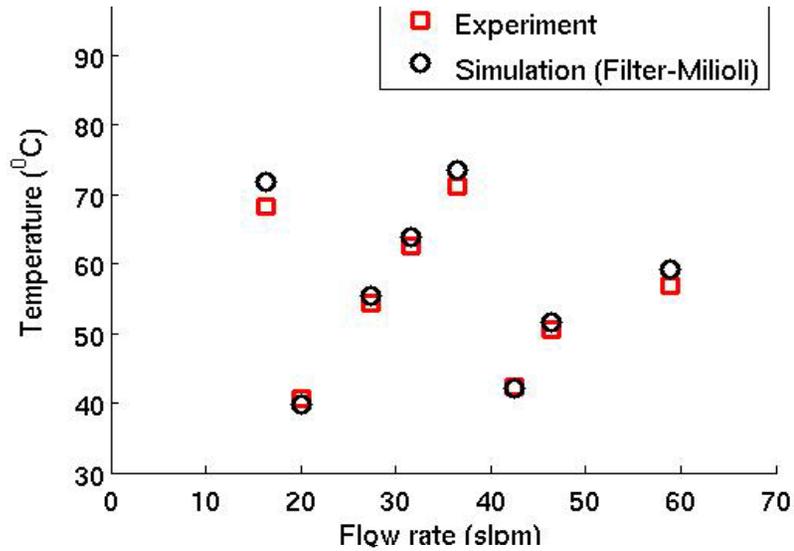
Most CO₂ is released before achieving the regeneration temperature



CFD Validation with C2U non-circulating Experiments

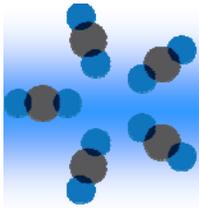


32D Hot Non-reactive Flow



Conclusion

- **Coupling experimental development with simulation enables**
 - New approaches to be screened more quickly
 - Focuses development on most promising material & process conditions
- **Simulation with uncertainty quantification**
 - Focuses experimental efforts on elements with the most impact to the process/technology
- **Focused experiments for model validation enable**
 - Lower risk for scale up through quantitative confidence of model predictions



CCSI

Carbon Capture Simulation Initiative

Computational Toolset Demonstration

- **5-7 PM today in Woodlawn I**
 - Demonstration & detailed discussion of capabilities
- **Sorbents, Solvents, Membranes, Oxycombustion**
- Initial toolset released Oct. 2012
 - 4 companies already have already licensed
 - Additional releases planned for Fall 2013, 2014, 2015.
 - Final release planned for Jan. 2016

