



Application of Sequential Design of Experiments (SDoE) to Solvent-Based CO₂ Capture Systems at Multiple Scales

Joshua C. Morgan, Ph.D.
Advanced Process Modeling Engineer – KeyLogic Systems
National Energy Technology Laboratory – Pittsburgh PA

**2019 Carbon Capture, Utilization, and Storage, and Oil and Gas Technologies
Integrated Review Meeting**

Pittsburgh PA
August 27, 2019



Disclaimer

This report was prepared as an account of work sponsored by an agency of the United States Government. Neither the United States Government nor any agency thereof, nor any of their employees, makes any warranty, express or implied, or assumes any legal liability or responsibility for the accuracy, completeness, or usefulness of any information, apparatus, product, or process disclosed, or represents that its use would not infringe privately owned rights. Reference herein to any specific commercial product, process, or service by trade name, trademark, manufacturer, or otherwise does not necessarily constitute or imply its endorsement, recommendation, or favoring by the United States Government or any agency thereof. The views and opinions of authors expressed herein do not necessarily state or reflect those of the United States Government or any agency thereof.

KeyLogic Systems, Inc.'s contributions to this work were funded by the National Energy Technology Laboratory under the Mission Execution and Strategic Analysis contract (DE-FE0025912) for support services.



Outline

- Executive Summary
- Background and Motivation for SDoE
- Applications of SDoE
 - MEA campaigns for NCCC and TCM pilot test facilities
 - Bench scale SDoE for CO₂BOL solvent system
 - Future campaigns at TCM for novel technologies
- Conclusions

Executive Summary

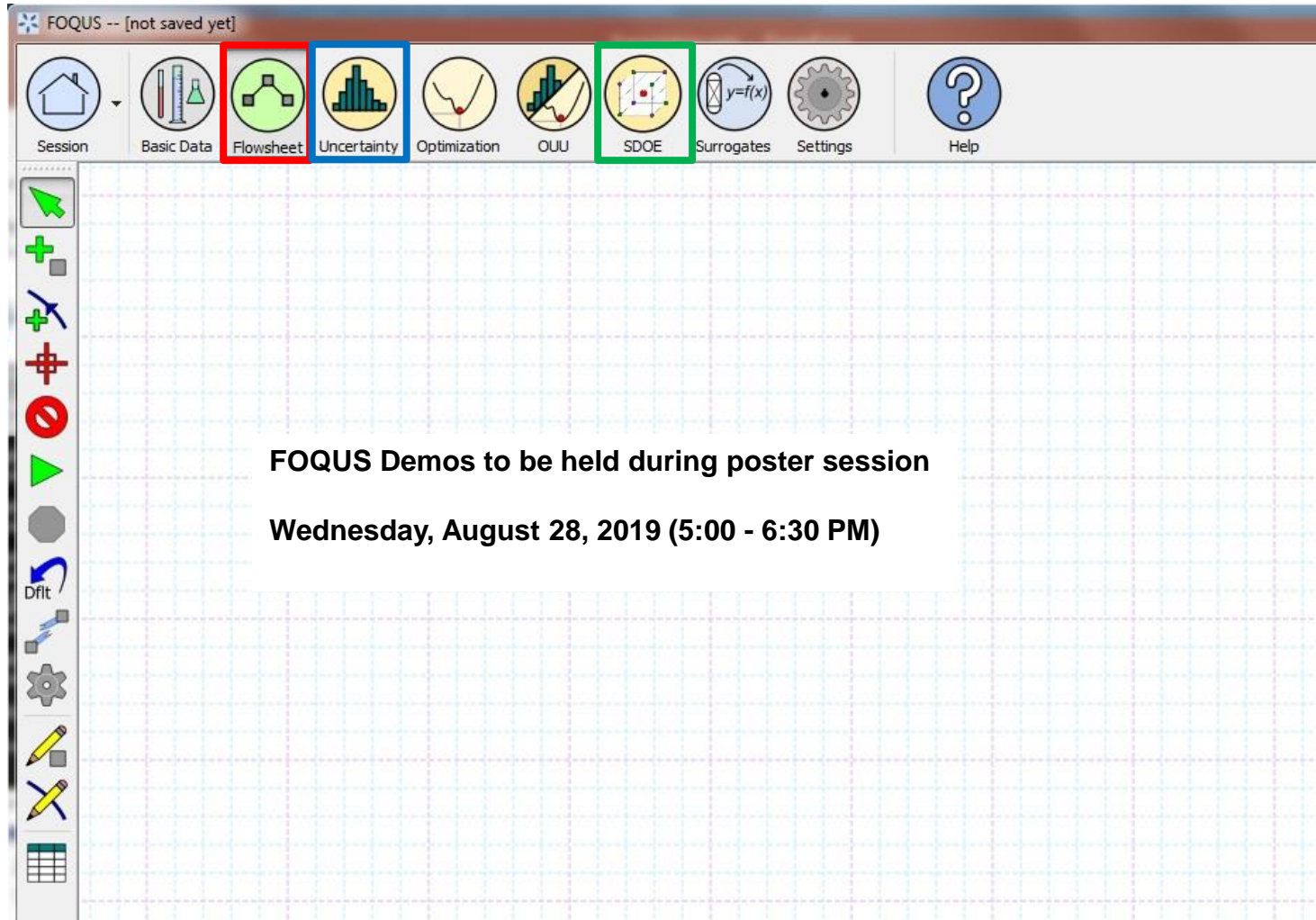
- CCSI² has developed and demonstrated methodology for sequential design of experiments (SDoE) to improve solvent-based CO₂ capture pilot testing
 - Applied to aqueous monoethanolamine (MEA) campaigns at National Carbon Capture Center (NCCC) [0.5 MWe] and Technology Centre Mongstad [12 MWe]
 - Reduced uncertainty of CO₂ capture predictions by approximately 60% for both campaigns
- SDoE work is ongoing for bench scale CO₂BOL process developed by Pacific Northwest National Laboratory
- Future work will focus on application of SDoE to novel technologies – including solvents, sorbents, and membranes

Motivation

- Develop systematic approach to conducting pilot plant testing, regardless of scale, process configuration, technology type, etc.
- Ensure right data is collected – improve understanding, refine models
- **Design of Experiments (DoE)** is a tool to accelerate learning by targeting maximally useful input combinations to match experiment goals
- **Sequential DoE (SDoE)** expands on DoE capabilities, allowing for incorporation of information from an experiment as it is being run, by updating input selection criteria based on new information

Ultimate Goal: Reduce technical risk associated with scale-up

Framework for Optimization, Quantification of Uncertainty, and Surrogates



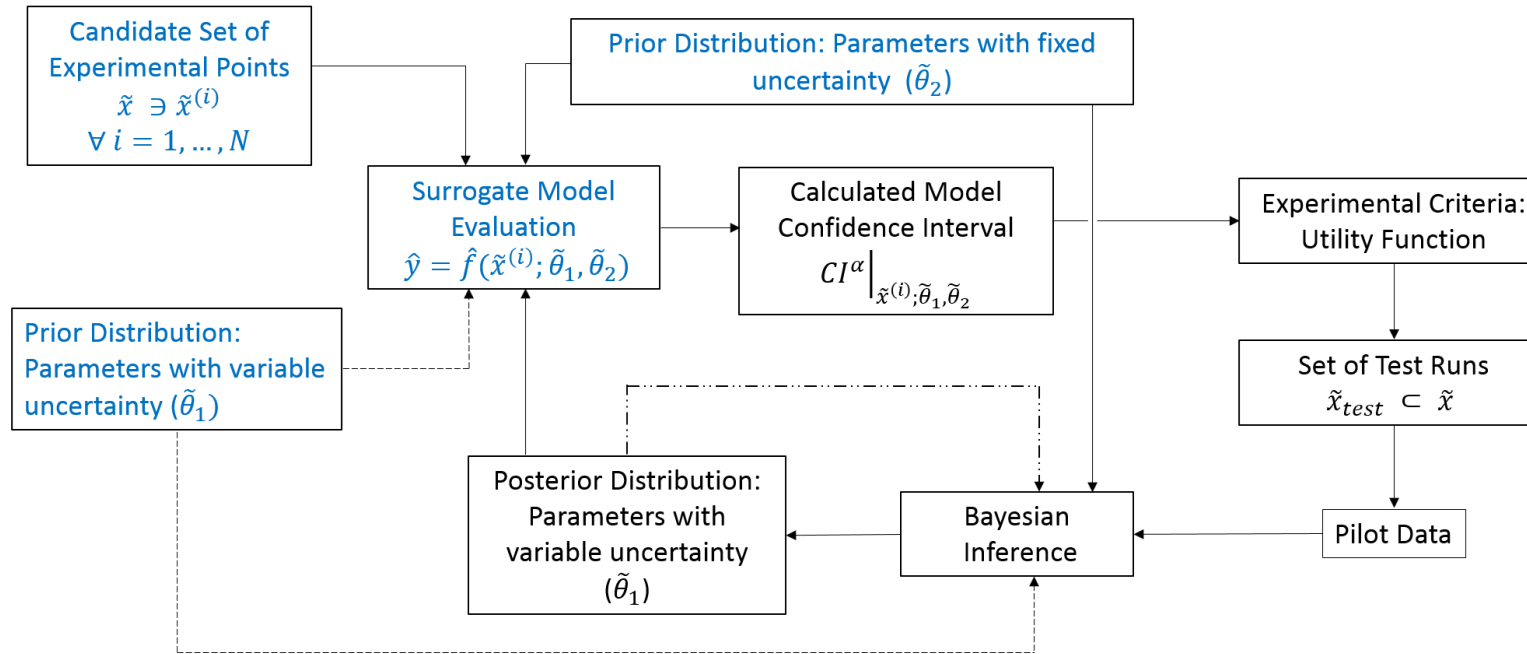
Flowsheet Tab – Used for propagating uncertainty through simulation model

Uncertainty Tab – PSUADE used for Bayesian inference and surrogate modeling

SDoE Tab – Currently being developed for streamlining process described in this work

Open-source software available at:
<https://github.com/CCSI-Toolset>

SDoE Process



Denotes input to SDoE algorithm

----- Denotes use of prior distribution of $\tilde{\theta}_1$ for first iteration only

..... Denotes use of posterior distribution of $\tilde{\theta}_1$ as prior distribution for next iteration

$$\tilde{\theta} = [\tilde{\theta}_1 \ \tilde{\theta}_2]$$

Full Set of Model Parameters

$$\Omega_i = \{\hat{y}(\tilde{x}^{(i)}; \tilde{\theta}^{(1)}), \dots, \hat{y}(\tilde{x}^{(i)}; \tilde{\theta}^{(M)})\}$$

Propagation of Parametric Uncertainty

$$CI^\alpha \Big|_{\tilde{x}^{(i)}; \tilde{\theta}_1, \tilde{\theta}_2} = F_{1-\alpha/2}(\Omega_i) - F_{\alpha/2}(\Omega_i)$$

Confidence Interval Calculation

SDoE Applied at National Carbon Capture Center – Summer 2017



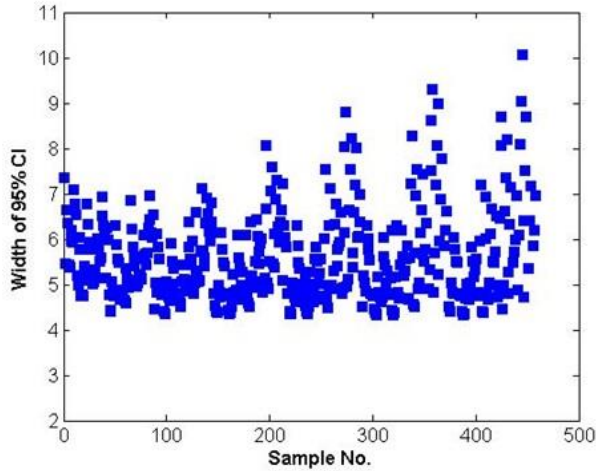
nationalcarboncapturecenter.com

- 0.5 MWe scale facility
- Variability in operating conditions for experimental design
 - Lean solvent flowrate
 - Flue gas flowrate
 - Lean solvent CO₂ loading
 - Flue gas CO₂ fraction
- Variability in absorber configuration also tested
 - Multiple solvent inlets allow operation with 1, 2, or 3 packing beds
 - Optional intercooling stages between beds
- Goal of pilot testing: Refine stochastic model prediction of CO₂ capture percentage

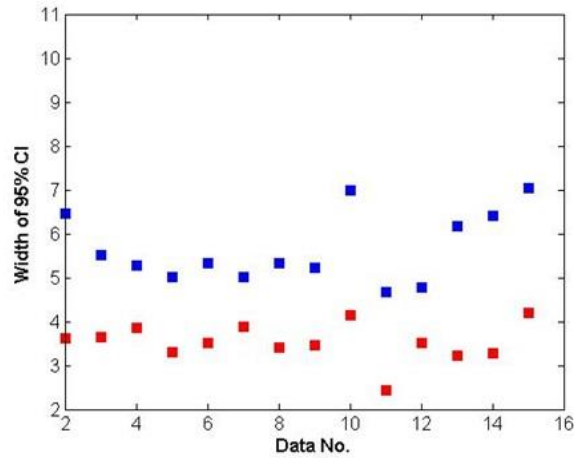
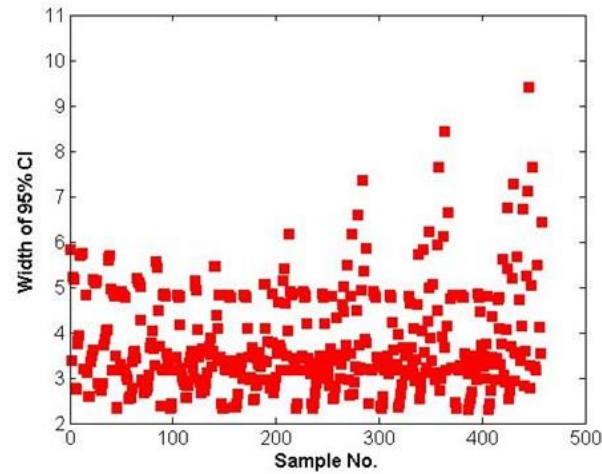
SDoE Results – Reduction in Prediction of CO₂ Capture Percentage

First Round

Prior



Posterior



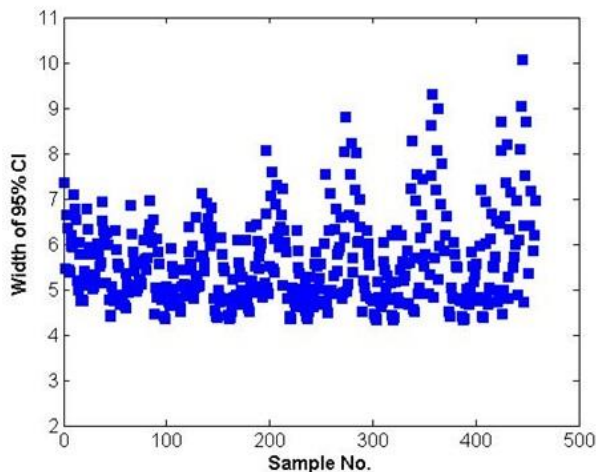
Candidate Set
(Discrete
Points in
Operating
Region of
Interest)

Experimental
Runs

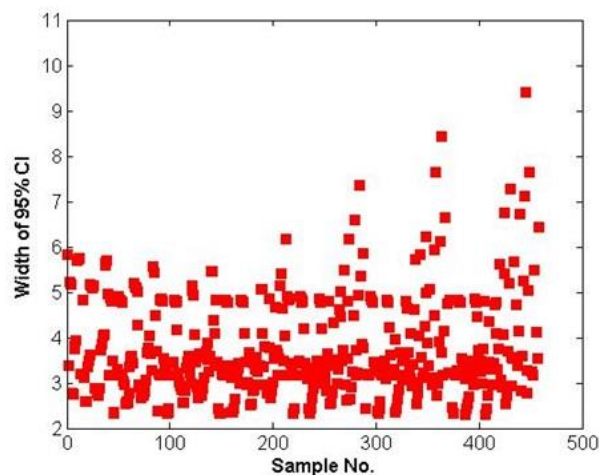
SDoE Results – Reduction in Prediction of CO₂ Capture Percentage

Second Round

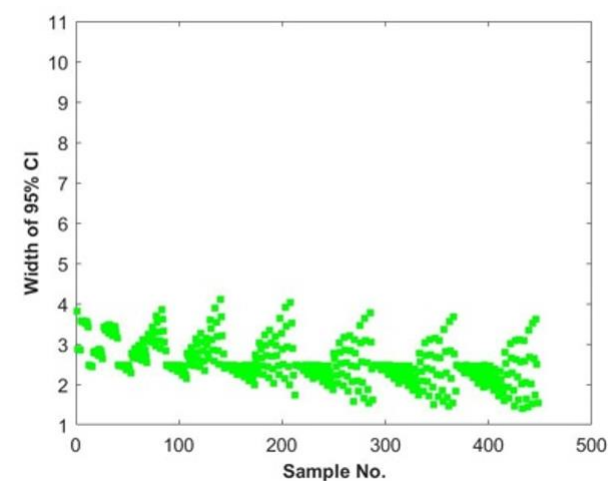
Candidate Set
(Discrete
Points in
Operating
Region of
Interest)



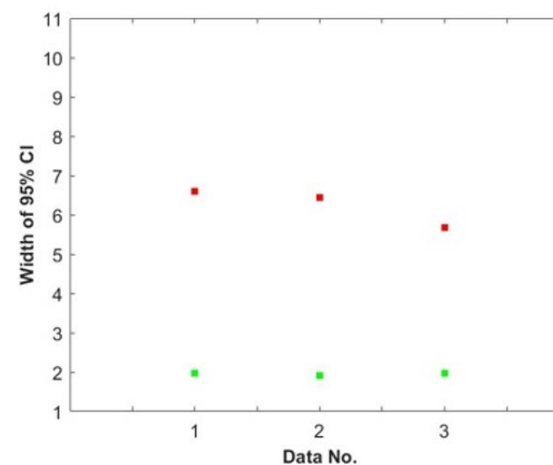
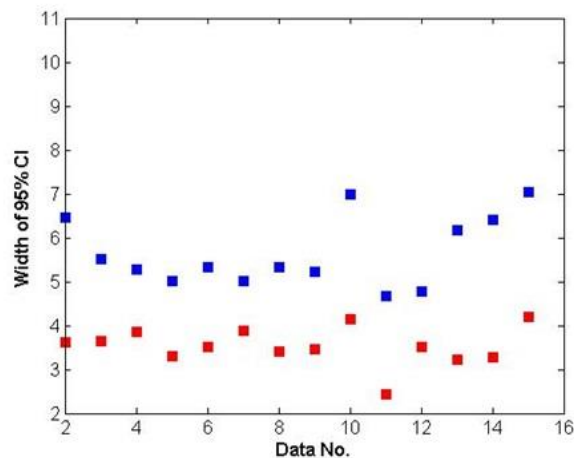
Prior



Posterior

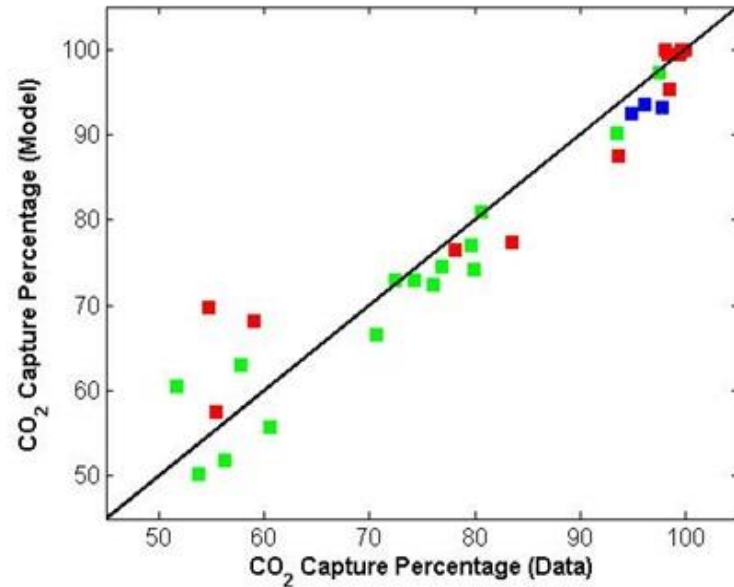


Experimental
Runs

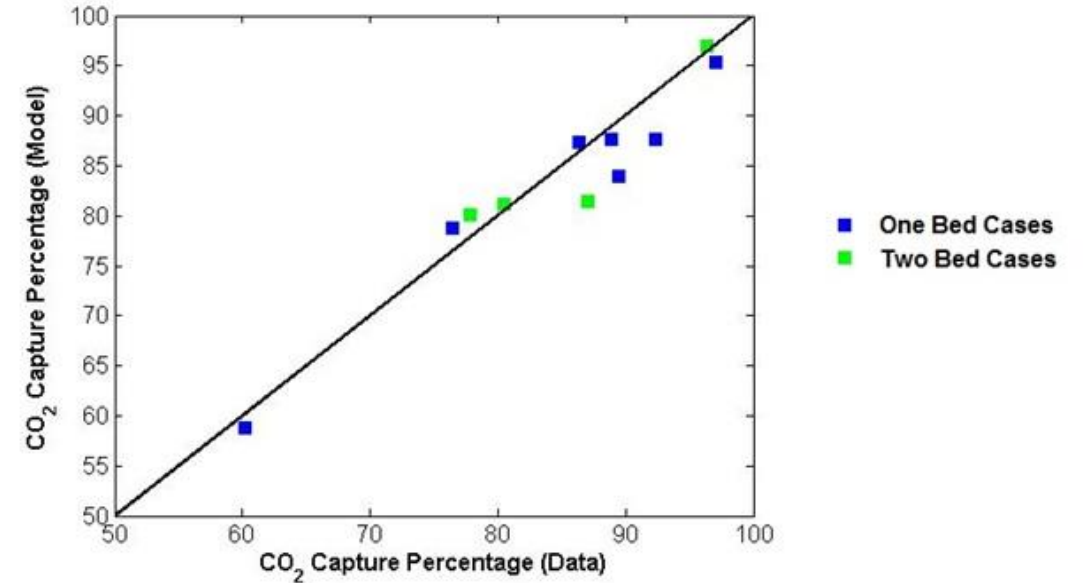


Fit of Model to NCCC Data

Three Beds with Intercooling Cases



One or Two Bed Cases



Note: These cases were not included in the sequential portion of the experimental design

SDoE Applied at Technology Centre Mongstad – Summer 2018



www.tcmda.com

- The world's largest facility for testing and improving CO₂ capture technologies (12 MWe scale)
- Located next to Equinor refinery in Mongstad, Norway
- Joint venture set up by Gassnova, Equinor, Total, and Shell
- Two flue gas sources
 - Combined Cycle Gas Turbine (CCGT)
 - Residual Fluidized Catalytic Cracker (RFCC)

Phases of TCM Test Campaign

Phase 1

Space-filling design for testing predictability of existing model

Phase 2

Selection of points for testing based on economic objective function

Phase 3

Sequential DoE

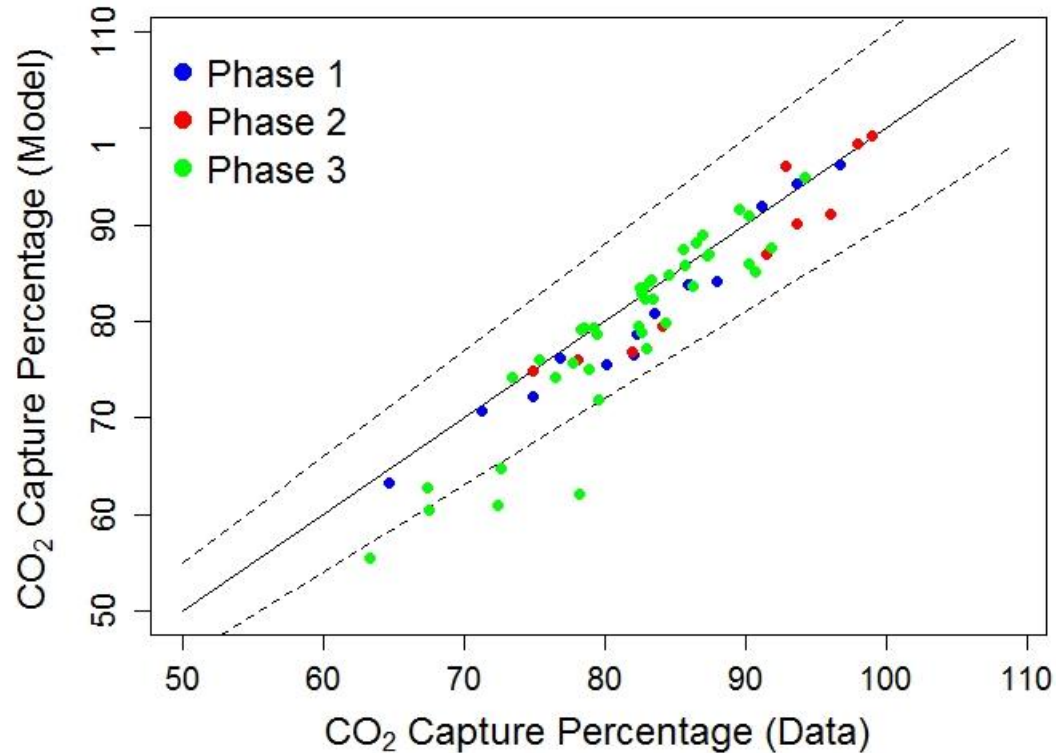
Selection of points based on G-optimality: minimize the maximum model prediction variance in the design space

Phase 4-5

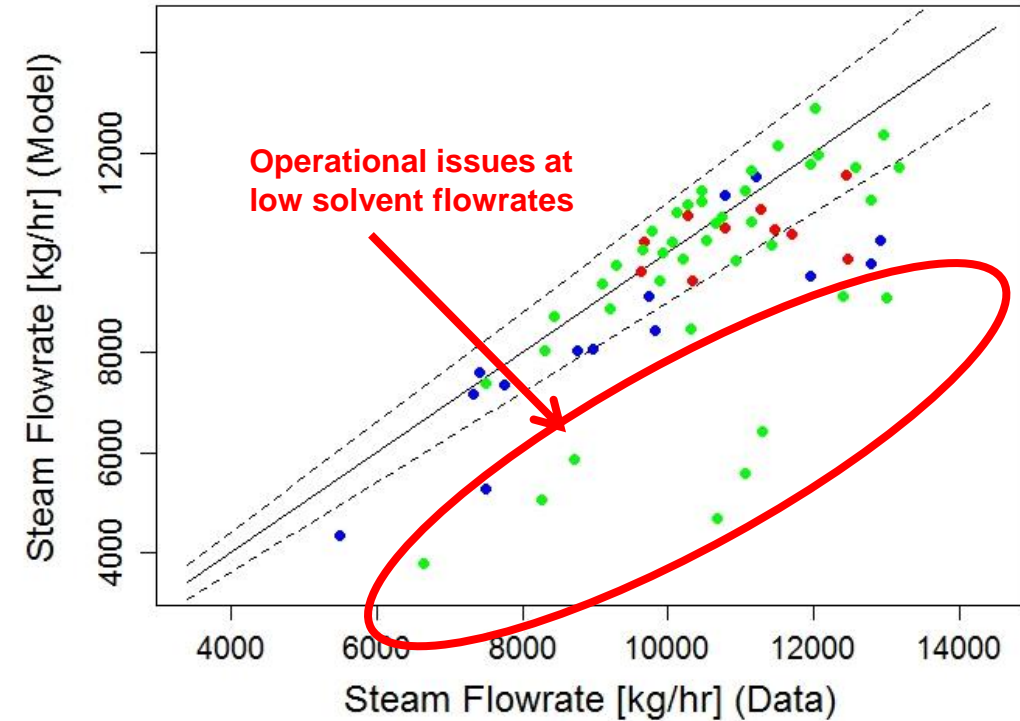
Minimization of reboiler duty
Variation in absorber packing height
Rich solvent bypass configuration

TCM Model Predictions (Deterministic)

Absorber Performance



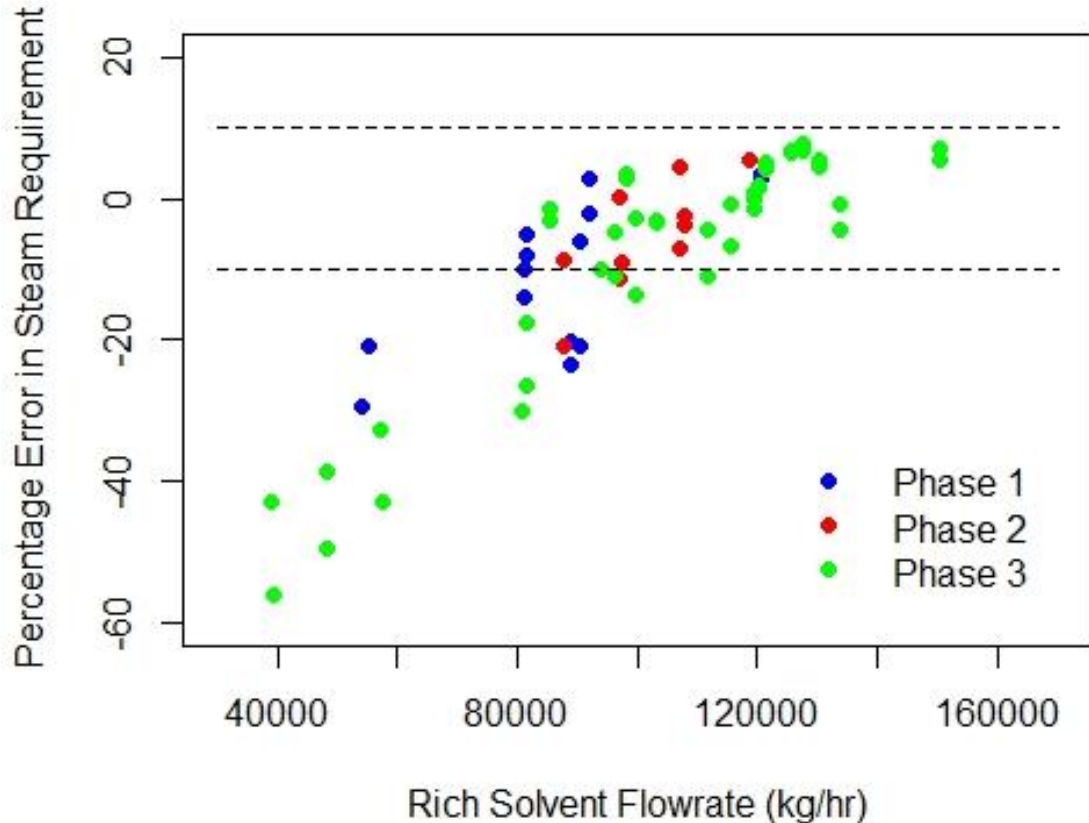
Stripper Performance



Dashed lines represent $\pm 10\%$

Data include variation in flowrates of solvent, flue gas, and steam as well as CO₂ composition in flow gas

TCM Stripper Performance



Two strippers available for use at TCM

- Stripper designed for CCGT flue gas (~3.5% CO₂) [Capacity: 80 tonne CO₂/day]
- Stripper designed for RFCC flue gas (~13-14% CO₂) [Capacity: 275 tonne CO₂/day]

CCSI² campaign used RFCC stripper and CCGT flue gas with recycle (8-10% CO₂), thus leading to over-designed stripper when running process with low flowrates

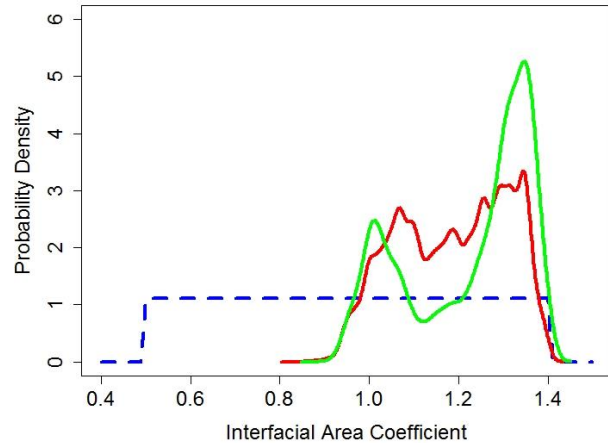
Potential maldistribution effect at low solvent flowrate not captured in Aspen Plus rate-based process model

Results – TCM SDoE (Phase 3)

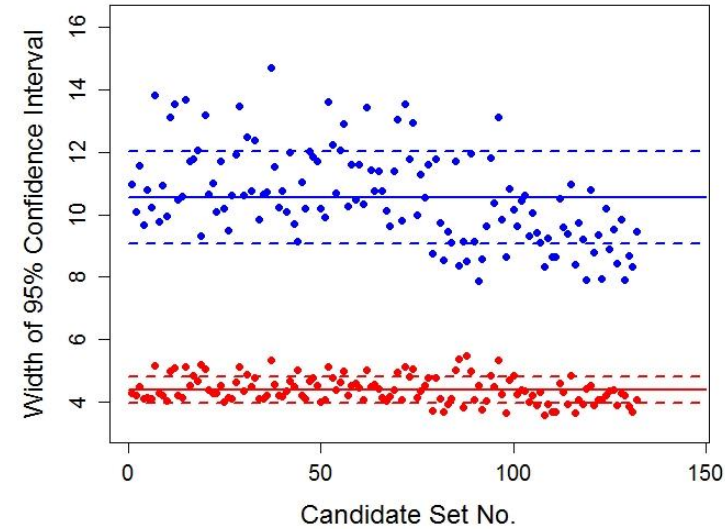
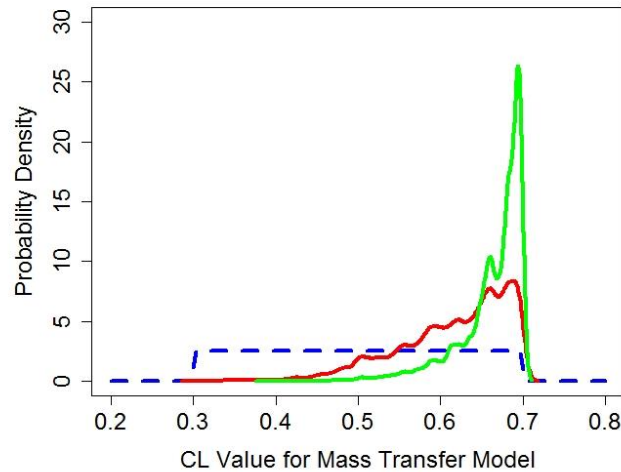
Update in Parameter Distributions
for Absorber Packing



Reduction in CO₂ Capture Percentage (First Iteration)



--- Prior
— Posterior 1
— Posterior 2



Prior CI Width: 10.5 ± 1.5

Posterior CI Width: 4.4 ± 0.4

Average reduction in uncertainty: $58.0 \pm 4.7\%$

Candidate set includes variation in:

- Solvent circulation rate
- Flue Gas flowrate and CO₂ concentration
- Reboiler steam flowrate

Test Phases 4-5

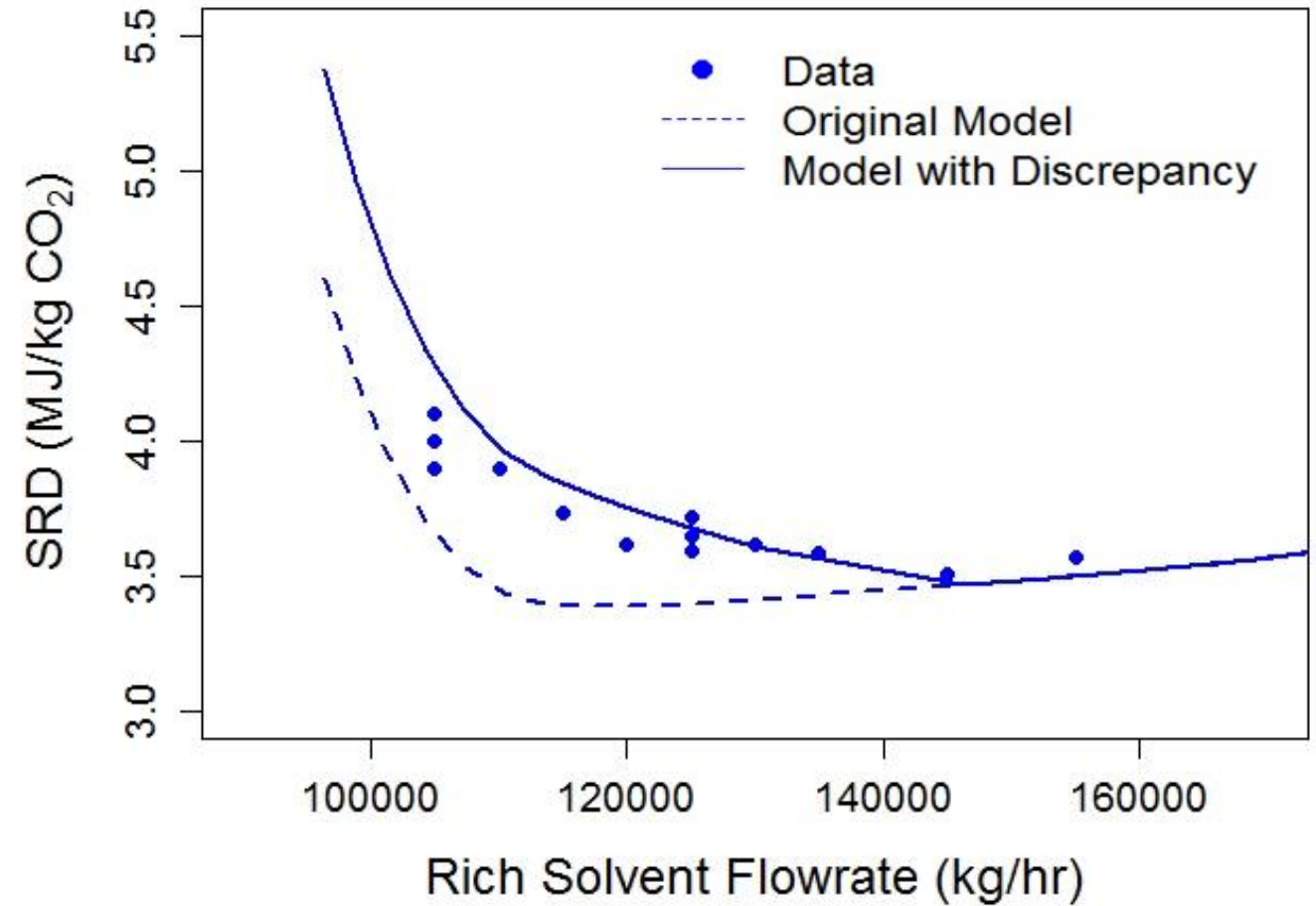
- Operated pilot plant with portion of rich solvent by-passing lean-rich heat exchanger routed to water wash bed of stripper column
- Reduced absorber packing height to 18 m (Phase 4) and 12 m (Phase 5)
- Space-filling design used to minimize specific reboiler duty (SRD) by varying solvent circulation rate
 - Fixed flowrate and composition of flue gas (50,000 sm³/hr; 8 mol% CO₂) and percentage of CO₂ capture (85%)

Results – Phase 4

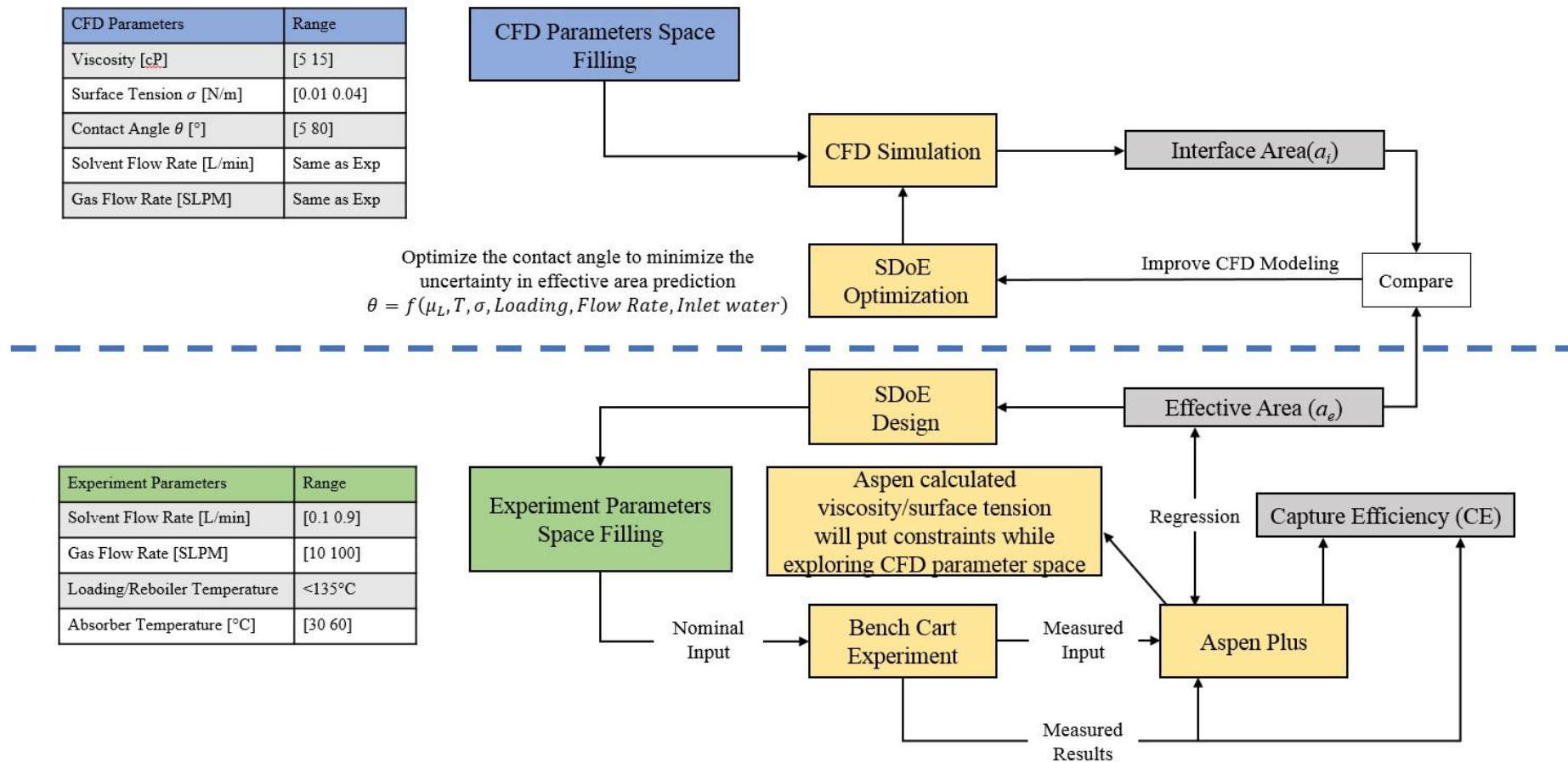
Statistical discrepancy model developed for reboiler steam requirement in order to account for mismatch between data and model prediction of SRD

$$\dot{m}_{steam} = \beta_0 + \beta_1 * L_{rich} + \beta_2 * bypass\ percentage$$

$$\dot{m}_{steam} = S_{calc} + \max(0, \Delta \dot{m}_{steam})$$



Ongoing Work: SDoE Application to CO₂BOL Bench-Cart System



Work will be presented in detail during:

“Low Aqueous Solvent System Optimization” – Zhijie Xu, Pacific Northwest National Laboratory

Capture and Utilization Session, Wednesday, August 28: 9:00 AM

Future Work

Upcoming SDoE projects at TCM

Industry Partner	Technology
Research Triangle Institute (RTI)	Non Aqueous Solvent
SRI International	Mixed Salt Solvent
Membrane Technology Research (MTR)	Membrane
TDA Research + MTR	Sorbent/Membrane Hybrid System

Summary and Conclusions

- Stochastic modeling framework enables quantification of model input uncertainty and propagation through model for risk assessment and economic analysis
- SDoE methodology has been shown to effectively inform design pilot test campaigns and reduce model uncertainty
 - SDoE demonstrates promise for accelerating development of novel CO₂ capture technologies
- Future work will focus on application of SDoE for novel CO₂ capture technologies, specifically for upcoming projects at TCM

Acknowledgements

Carbon Capture Simulation for Industry Impact (CCSI²)

Benjamin Omell, Michael Matuszewski (NETL-Pittsburgh)
Anderson Soares Chinen (NETL-Morgantown)
Christine Anderson-Cook, Towfiq Ahmed, Sham Bhat, John Baca (LANL)
Charles Tong, Brenda Ng, Pedro Sotorrio (LLNL)
Debangsu Bhattacharyya (WVU)

DOCCSS Collaboration – CO₂BOL System

Jay Xu, Charlie Freeman, Richard Zheng (PNNL)
Paul Mathias (Fluor Corp.)

National Carbon Capture Center



John Carroll
Chiranjib Saha
Justin Anthony

Technology Centre Mongstad



Thomas de Cazenove
Christophe Benquet
Muhammad Ismail Shah
Anette Beate Nesse Knarvik



For more information

<https://www.acceleratecarboncapture.org/>

joshua.morgan@netl.doe.gov

