



OPTIMIZATION OF CO₂ CAPTURE SIMULATIONS

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Nick Sahinidis¹

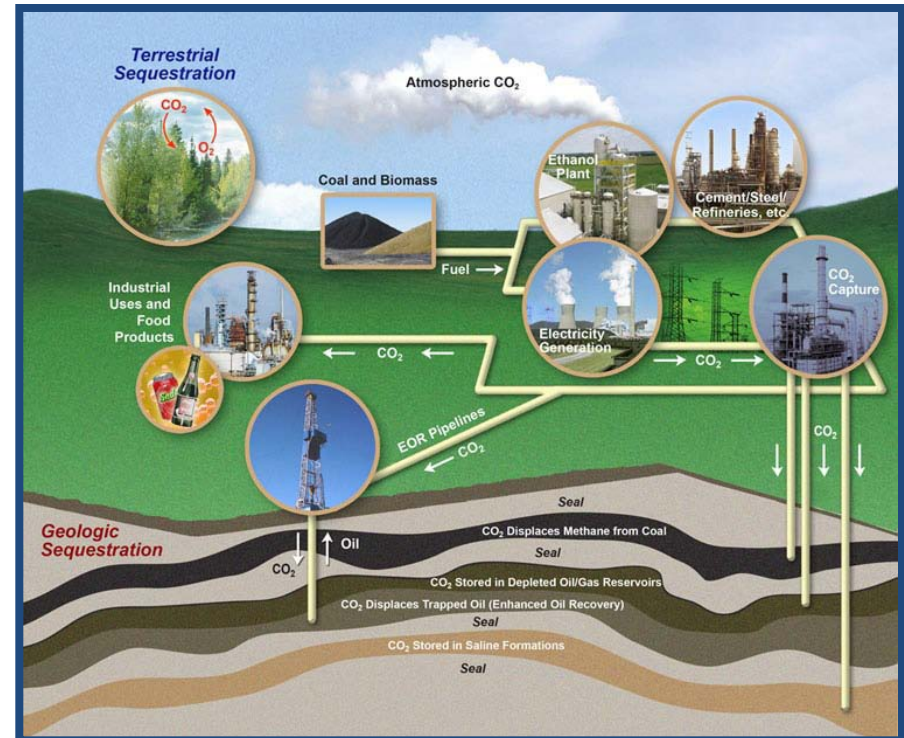
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MOTIVATION

- One-third of U.S. CO₂ emissions come from power plants and other point sources
- Available carbon capture technologies would increase electricity costs
 - Pulverized coal plants
 - *Currently: 75% increase*
 - *Goal: <30% increase*



http://www.netl.doe.gov/technologies/carbon_seq/index.html

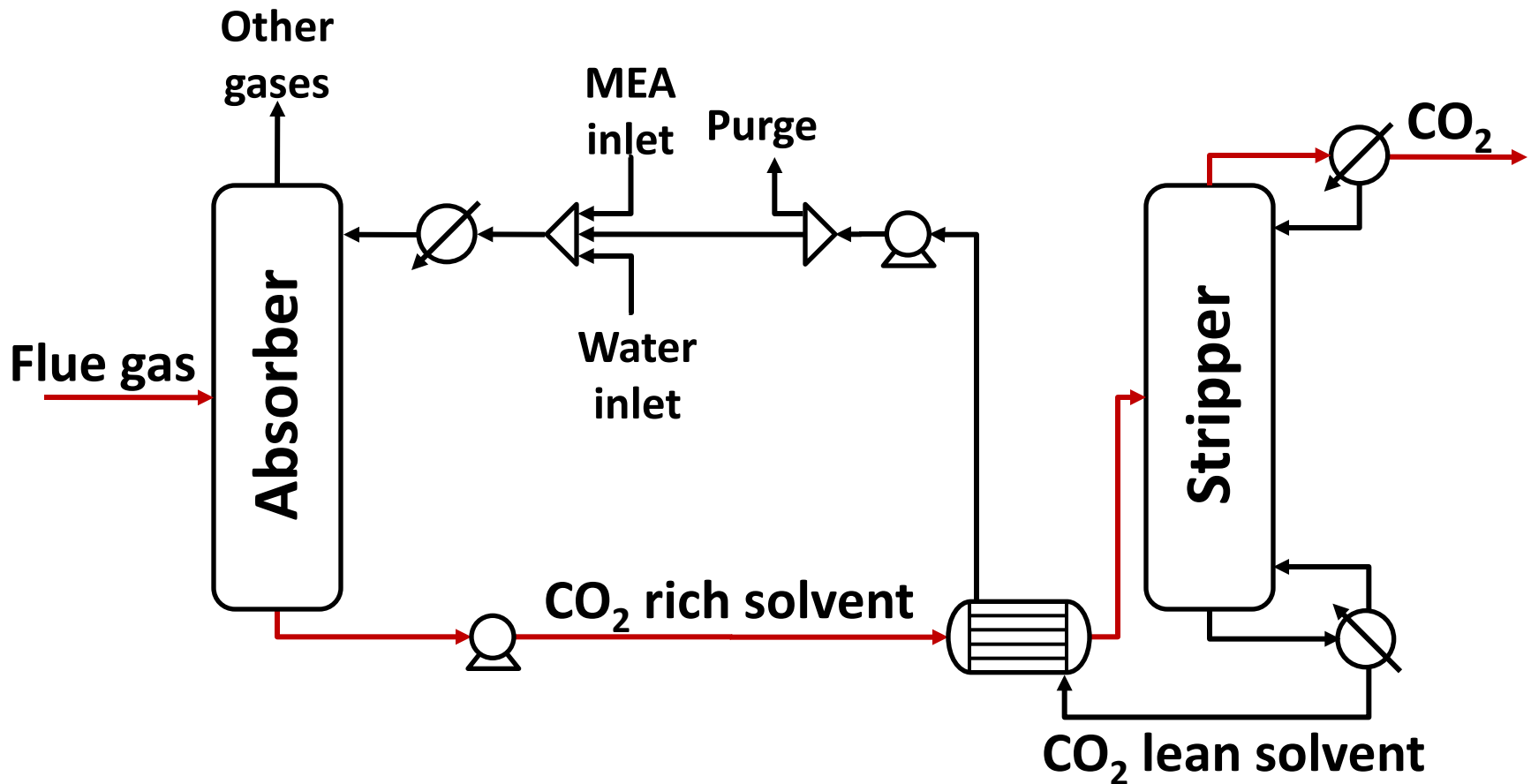
OVERVIEW

- 1. Carbon capture process**
- 2. Method overview**
- 3. Surrogate model generation**
- 4. Preliminary results**
- 5. Conclusions**

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MEA CARBON CAPTURE PROCESS



→ Path of CO₂ through the process

CURRENT METHODS

- **Simulation methods**

- **Aspen Plus**

- *(Amann and Bouallou, 2009)*
 - *(Fashami et al., 2007)*
 - *(Abu-Zahra et al., 2007)*
 - *(Alie et al., 2005)*
 - *(Chang and Shih, 2005)*
 - *(Corradetti and Desideri, 2005)*
 - *(Fisher et al., 2005)*
 - *(Alie, 2004)*
 - *(Freguia and Rochelle, 2003)*
 - *(Report DOE/NETL, 2002)*
 - *(Desideri and Paolucci, 1999)*
 - *(Desideri and Corbelli, 1998)*

- **HYSYS**

- *(Amann and Bouallou, 2009)*
 - *(Oi, 2007)*
 - *(Singh et al., 2003)*

- **MATLAB**

- *(Mofarahi et al., 2008)*

- **Fortran code**

- *(Tobiesen et al., 2007)*
 - *(Tobiesen and Svendsen, 2006)*

CURRENT METHODS

- **Simulation optimization methods**

- **Direct**

- *(Emun, Gadalla, Majozi, and Boer, 2010)*
 - *(Chen, Shao, and Qian, 2009)*
 - *(Leboreiro and Acevedo, 2004)*
 - *(De Simon, Parodi, Fermeglia, and R. Taccani, 2003)*
 - *(Ernst, Garro, Winkler, Venkataraman, Langer, Cooney, and Sasisekharan, 1997)*

- **Surrogate model based**

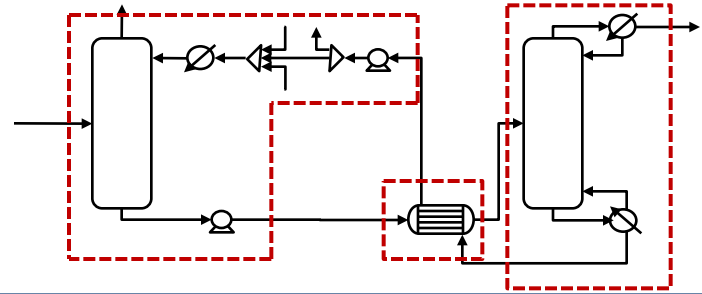
- *(Henao and Maravelias, 2010)*
 - *(Zhou, Xinping, Kefa, and Fan, 2004)*

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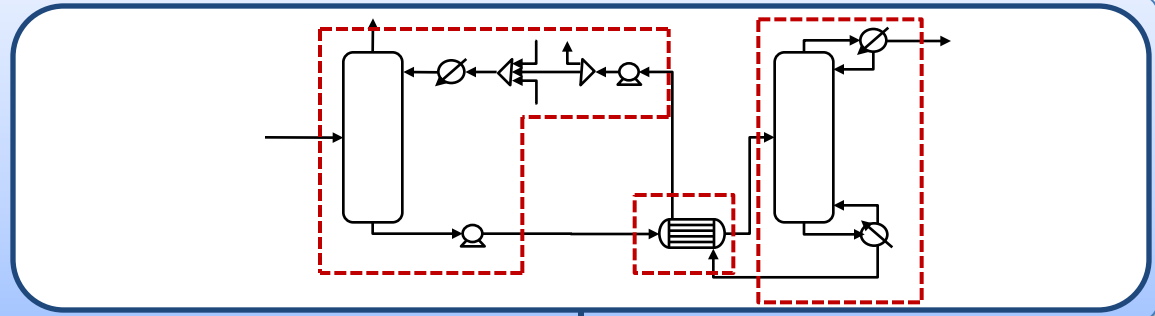
OVERVIEW OF THE METHOD

Simulation

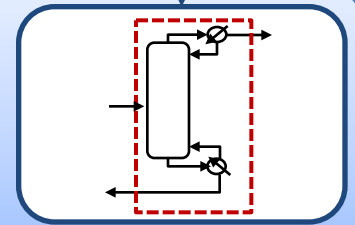
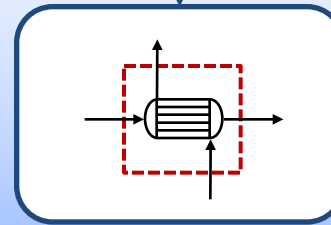
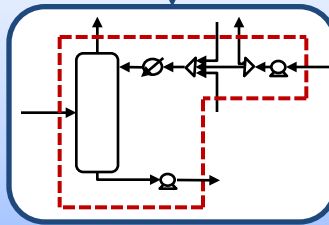


OVERVIEW OF THE METHOD

Simulation

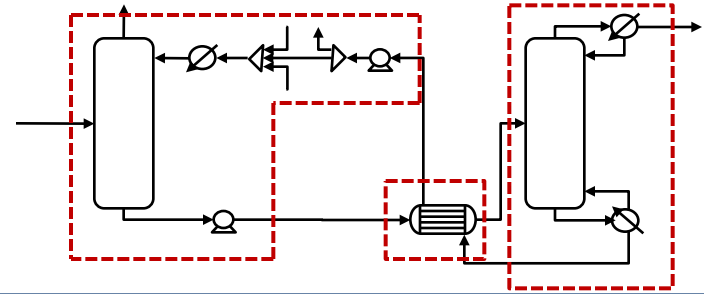


Disaggregated blocks of process unit(s)

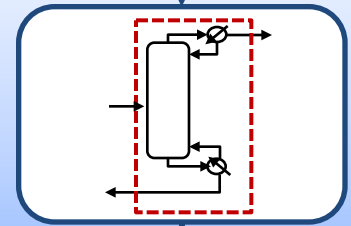
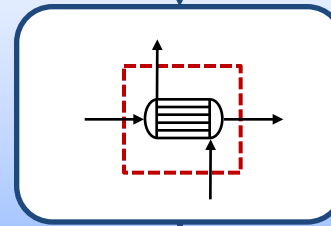
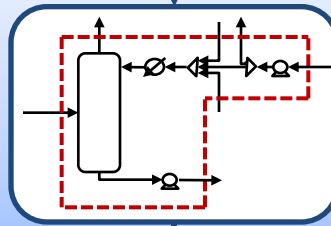


OVERVIEW OF THE METHOD

Simulation



Disaggregated blocks of process unit(s)



Surrogate models of blocks

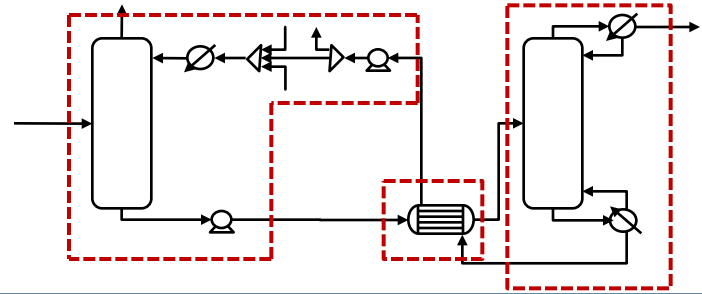
$$f_1(x)$$

$$f_2(x)$$

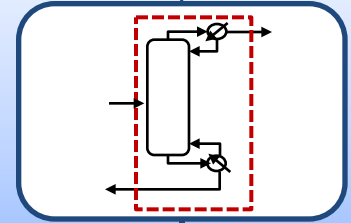
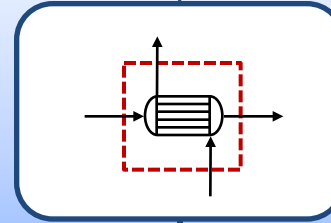
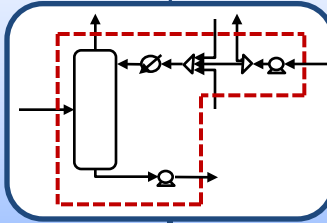
$$f_3(x)$$

OVERVIEW OF THE METHOD

Simulation



Disaggregated blocks of process unit(s)



Surrogate models of blocks

$$f_1(x)$$

$$f_2(x)$$

$$f_3(x)$$

Algebraic constraints

Mass balances

Design specs

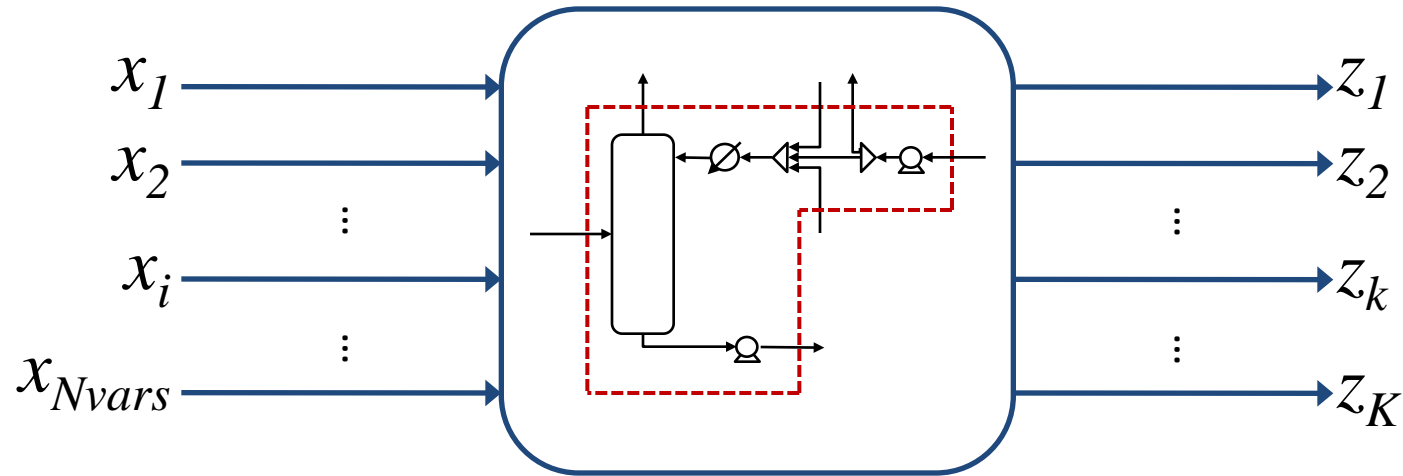
Nonlinear program

Algebraic model for optimization

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Disaggregated blocks



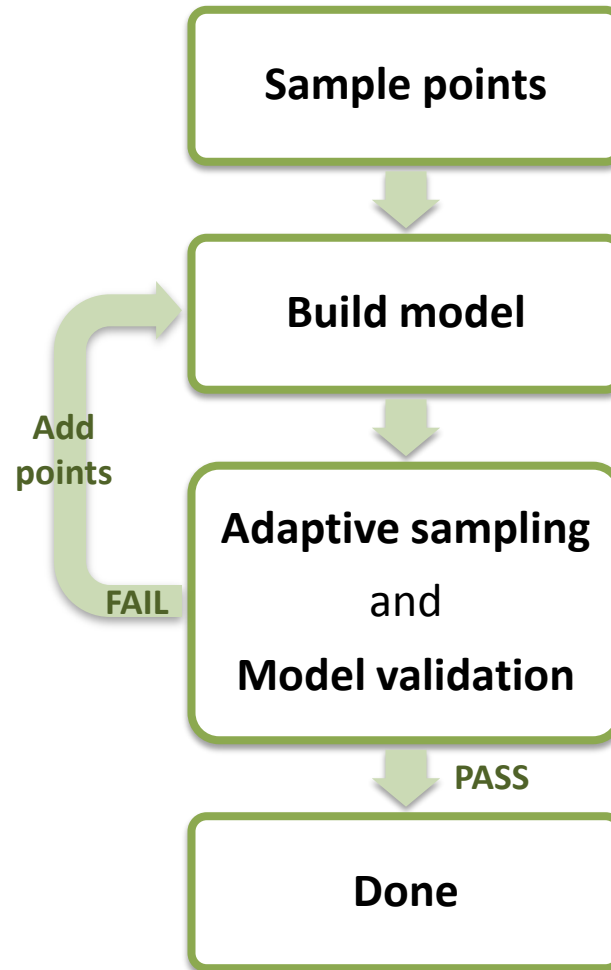
Independent variables, x

- Inlet conditions
 - Flow rates
 - Composition
 - Temperature
 - Pressure
- Operating conditions
- $x_i^l \leq x_i \leq x_i^u$

Dependent variables, z

- Outlet conditions
 - Flow rates
 - Composition
 - Temperature
 - Pressure
- Other relevant conditions

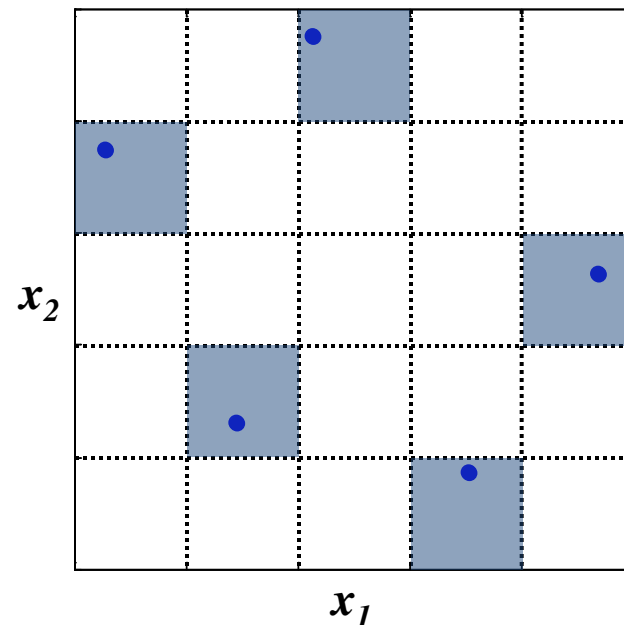
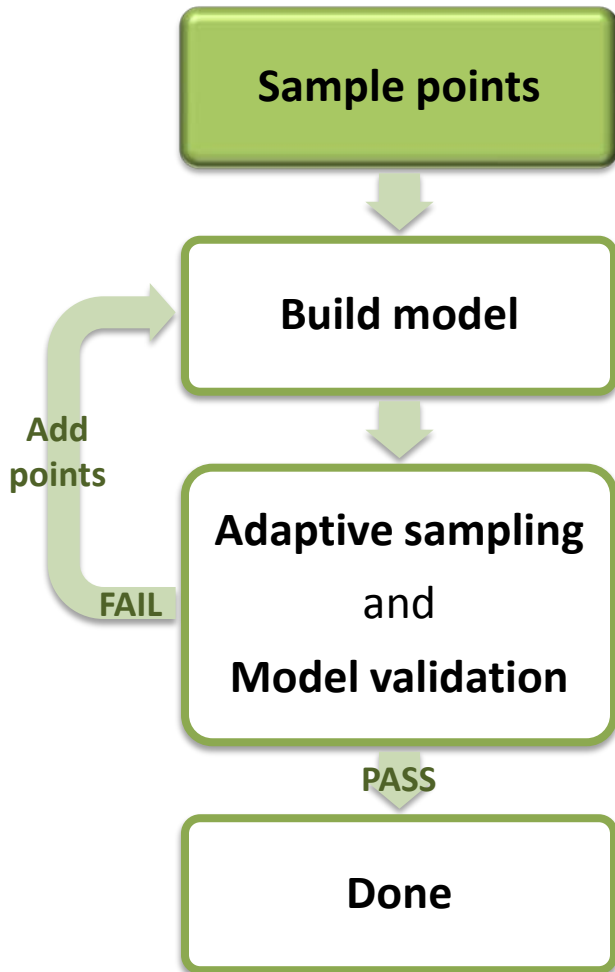
SURROGATE MODEL GENERATION



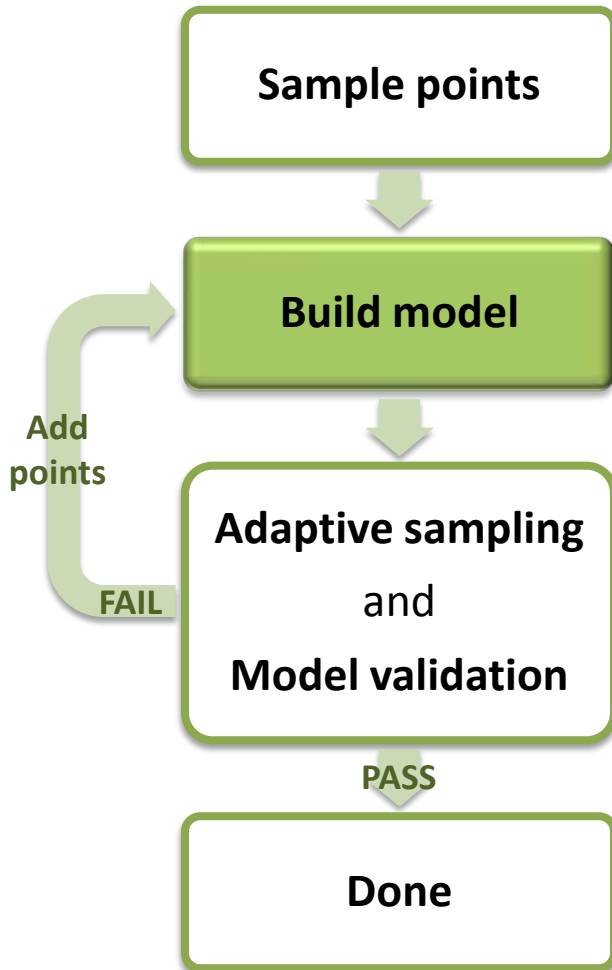
SAMPLE PROBLEM SPACE

Design of experiments

1. Random points sampling
2. Factorial design
3. **Latin hypercube design**
 - *Space-filling design*



MODEL BUILDING



- Build model for output z_k

$$\hat{z}_k = \sum_{j=1}^J C_{kj} B_j(z, x) + C_{k0}$$

- Basis functions $B_j(z, x)$

- Explicit $B_j(x)$

- *Polynomial*

$$x_i^\beta$$

- *Pairwise polynomial*

$$x_i^\beta x_{i'}^\beta$$

$$\frac{x_i}{x_{i'}}$$

- *Exponential and logarithmic*

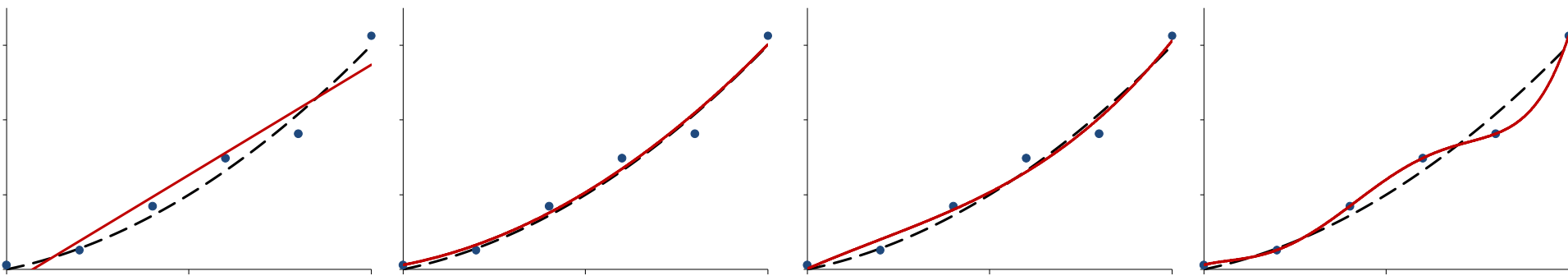
- *Expected basis functions*

- Implicit $B_j(z, x)$

- *Ex: log mean temperature difference*

AVOID OVERFITTING THE DATA

Increasing model complexity



Linear

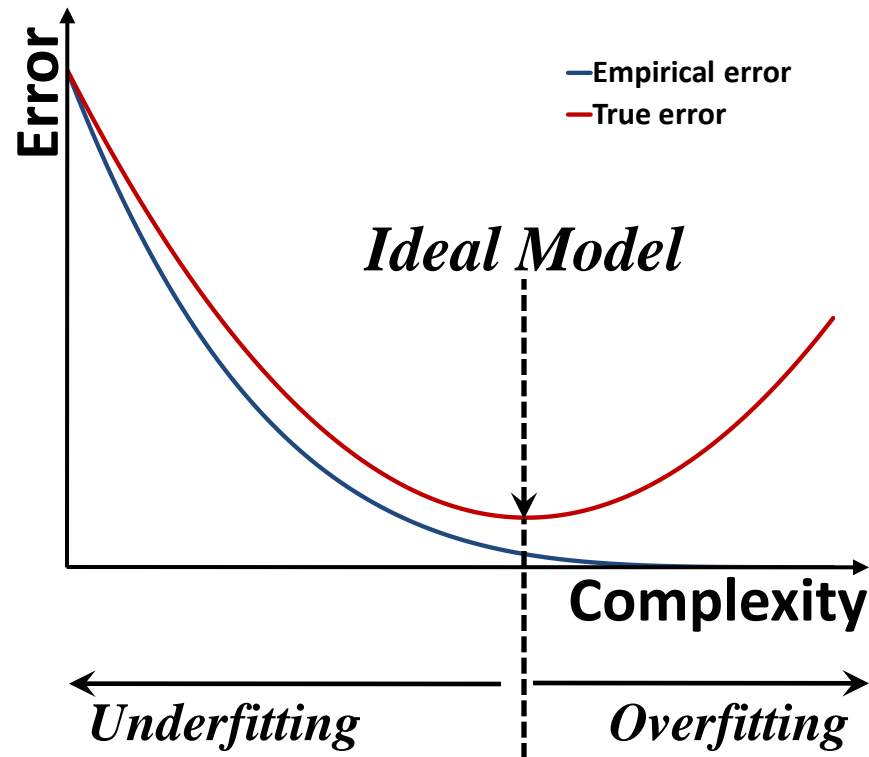
Quadratic

Cubic

5th Order

- True Function
- Data Points
- Model

EMPIRICAL VERSUS TRUE ERROR



- **Empirical error:**
Error between the model and the sampled data points
- **True error:**
Error between the model and the true function

BEST SUBSET METHOD

1. Start with a linear combination of basis functions

$d=4:$ $\hat{y} = \alpha + \beta_1x + \beta_2x^2 + \beta_3\sqrt{x}$

2. Generate all possible subsets of basis functions

One term	Two terms	Three terms	Four terms
$\hat{y} = \alpha$	$\hat{y} = \beta_1x + \alpha$	$\hat{y} = \beta_1x + \beta_2x^2 + \beta_3\sqrt{x}$	$\hat{y} = \alpha + \beta_1x + \beta_2x^2 + \beta_3\sqrt{x}$
$\hat{y} = \beta_1x$	$\hat{y} = \beta_2x^2 + \alpha$	$\hat{y} = \alpha + \beta_2x^2 + \beta_3\sqrt{x}$	
$\hat{y} = \beta_2x^2$	$\hat{y} = \beta_3\sqrt{x} + \alpha$	$\hat{y} = \alpha + \beta_1x + \beta_3\sqrt{x}$	
$\hat{y} = \beta_3\sqrt{x}$	$\hat{y} = \beta_2x^2 + \beta_1x$	$\hat{y} = \alpha + \beta_1x + \beta_2x^2$	
	$\hat{y} = \beta_3\sqrt{x} + \beta_1x$		
	$\hat{y} = \beta_3\sqrt{x} + \beta_2x^2$		

3. Pick the best

BEST SUBSET METHOD

1. Start with a linear combination of basis functions

$d=4:$ $\hat{y} = \alpha + \beta_1x + \beta_2x^2 + \beta_3\sqrt{x}$

2. Generate all possible subsets of basis functions

One term

$$\hat{y} = \alpha$$

$$\hat{y} = \beta_1x$$

$$\hat{y} = \beta_2x^2$$

$$\hat{y} = \beta_3\sqrt{x}$$

Two terms

$$\hat{y} = \beta_1x + \alpha$$

$$\hat{y} = \beta_2x^2 + \alpha$$

$$\hat{y} = \beta_3\sqrt{x} + \alpha$$

$$\hat{y} = \beta_2x^2 + \beta_1x$$

$$\hat{y} = \beta_3\sqrt{x} + \beta_1x$$

$$\hat{y} = \beta_3\sqrt{x} + \beta_2x^2$$

Three terms

$$\hat{y} = \beta_1x + \beta_2x^2 + \beta_3\sqrt{x}$$

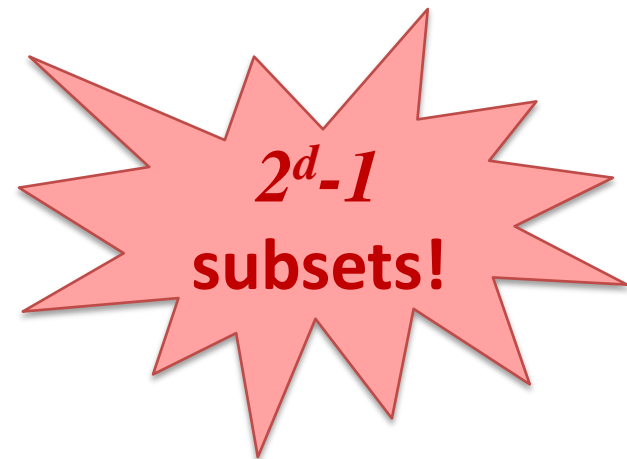
$$\hat{y} = \alpha + \beta_2x^2 + \beta_3\sqrt{x}$$

$$\hat{y} = \alpha + \beta_1x + \beta_3\sqrt{x}$$

$$\hat{y} = \alpha + \beta_1x + \beta_2x^2$$

Four terms

$$\hat{y} = \alpha + \beta_1x + \beta_2x^2 + \beta_3\sqrt{x}$$

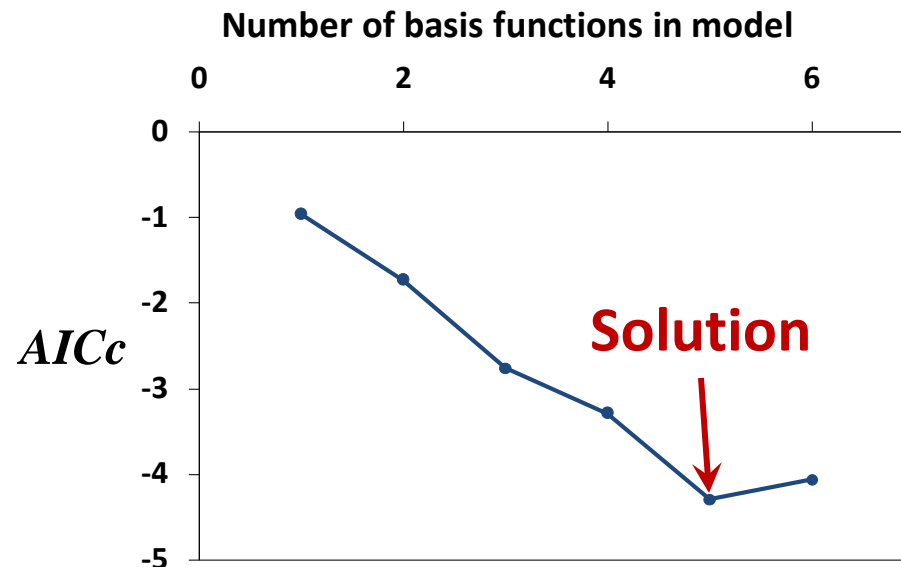


3. Pick the best

HOW TO PICK THE BEST SUBSET

- **Corrected Akaike Information Criterion (AICc)**
 - Gives an estimate of the difference between a model and the true function

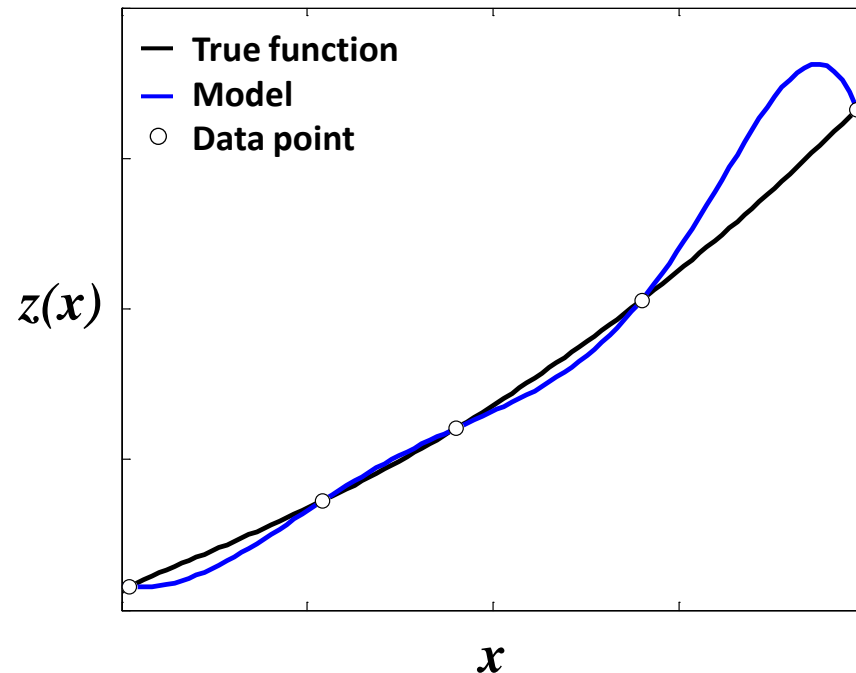
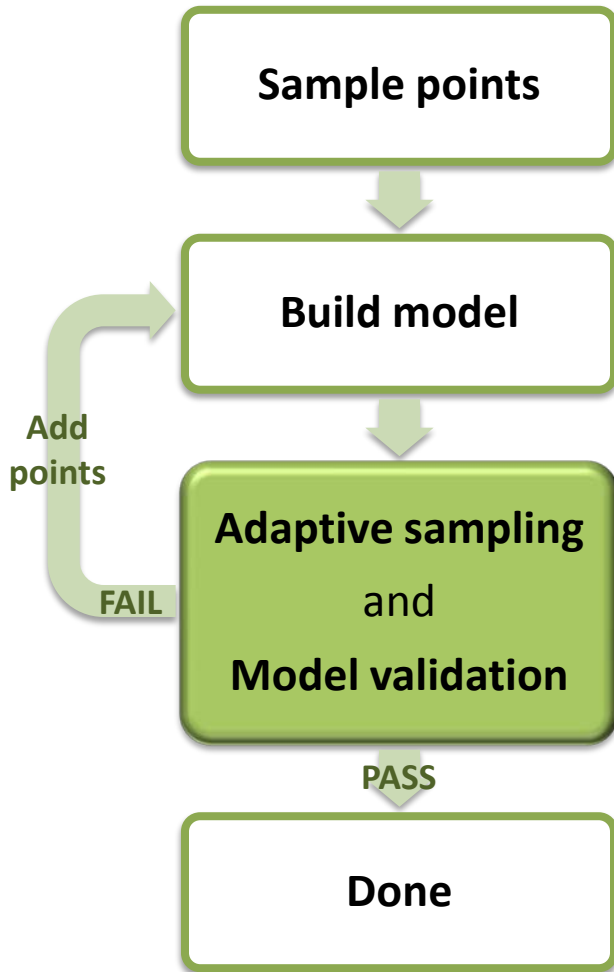
$$AICc = \underbrace{n \log \left(\frac{SSE}{n} \right)}_{\text{Accuracy}} + \underbrace{2p + \frac{2p(p+1)}{n-p-1}}_{\text{Complexity}}$$



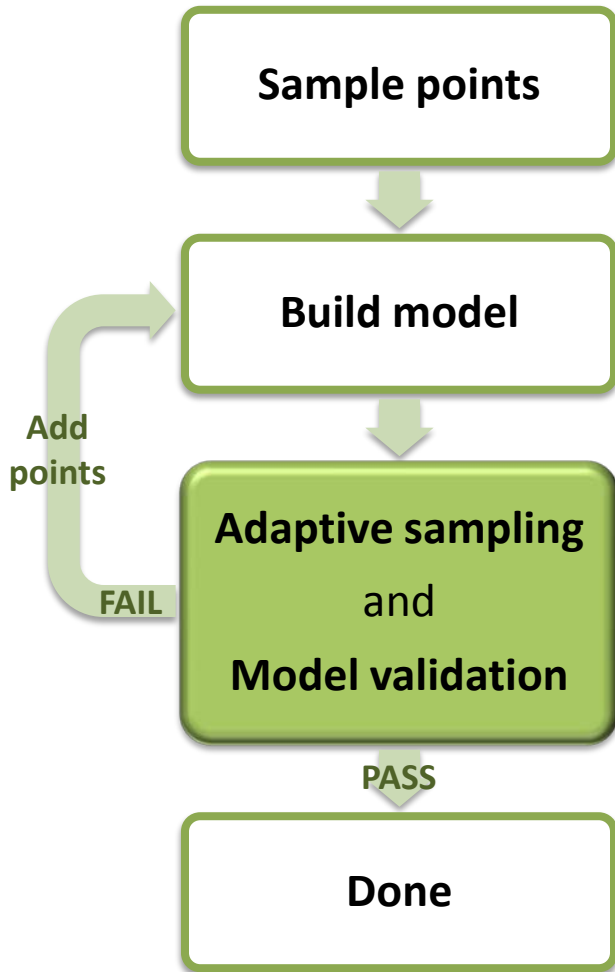
ADAPTIVE SAMPLING

- Find values of x where the model is inconsistent
 - Maximize the relative model error using a black-box solver:

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

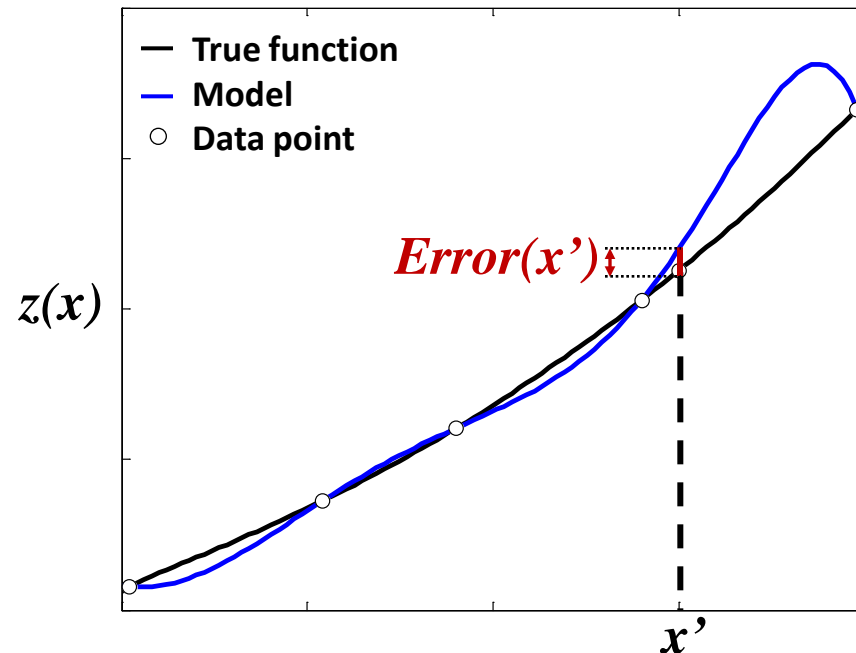


ADAPTIVE SAMPLING



- Find values of x where the model is inconsistent
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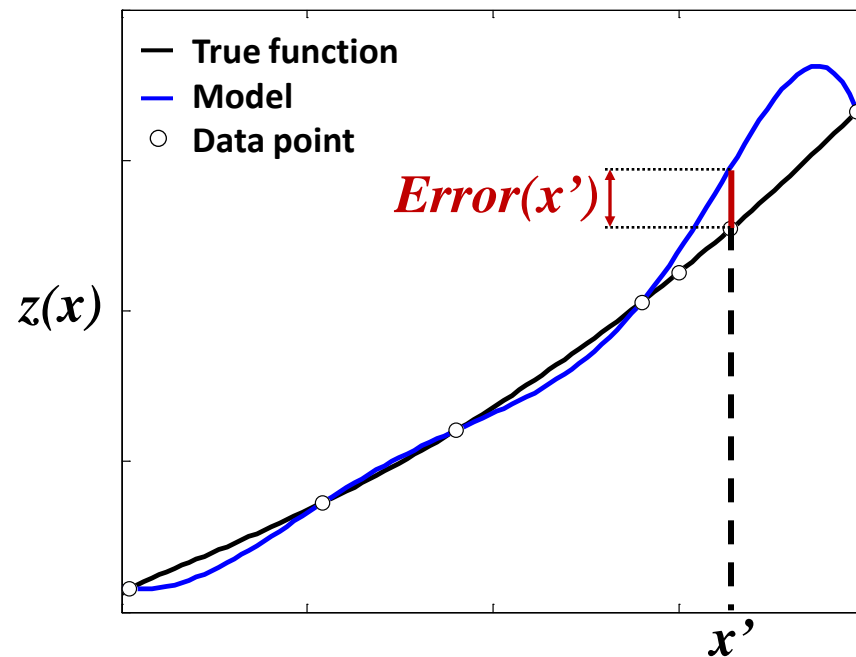
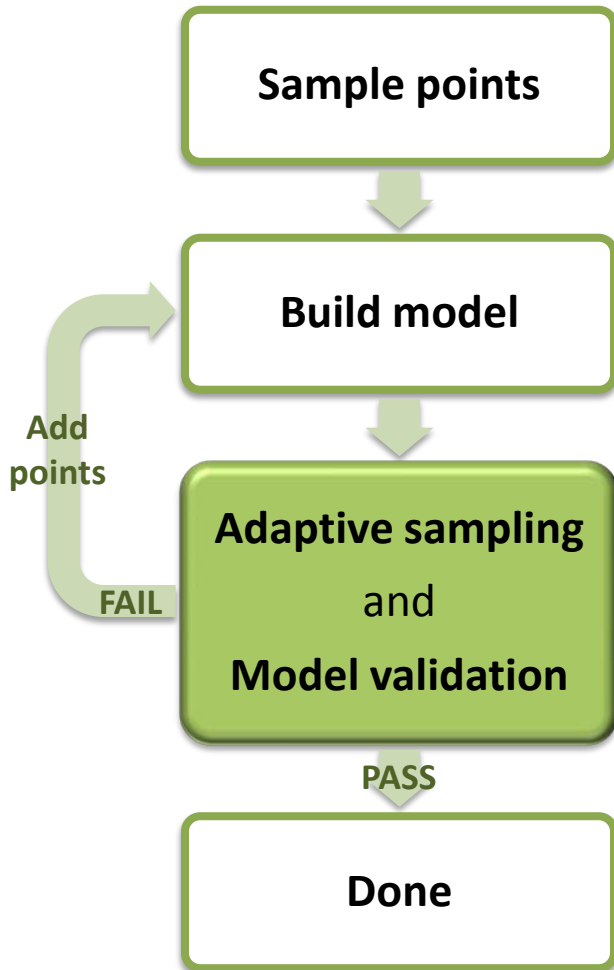
$$\max_x \left(\frac{z(x) - \hat{z}(x)}{z(x)} \right)^2$$



ADAPTIVE SAMPLING

- Find values of x where the model is inconsistent
 - Maximize the relative model error using a black-box solver:

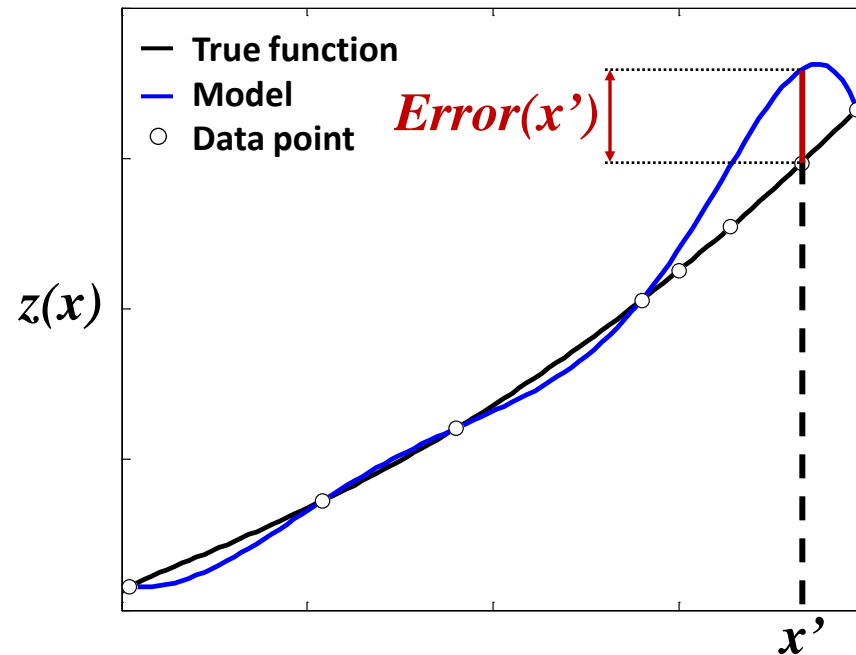
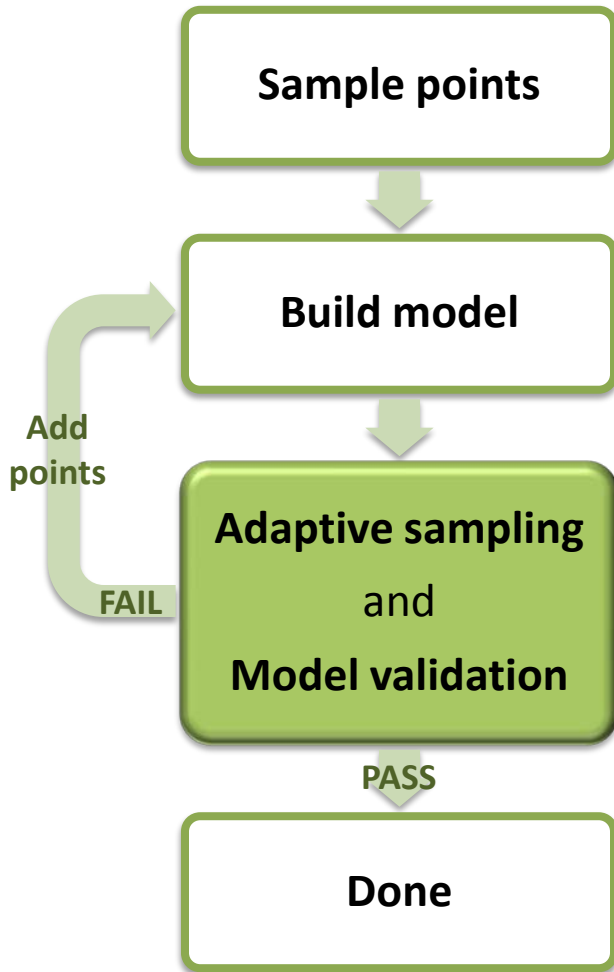
$$\max_x \left(\frac{z(x) - \hat{z}(x)}{z(x)} \right)^2$$



ADAPTIVE SAMPLING

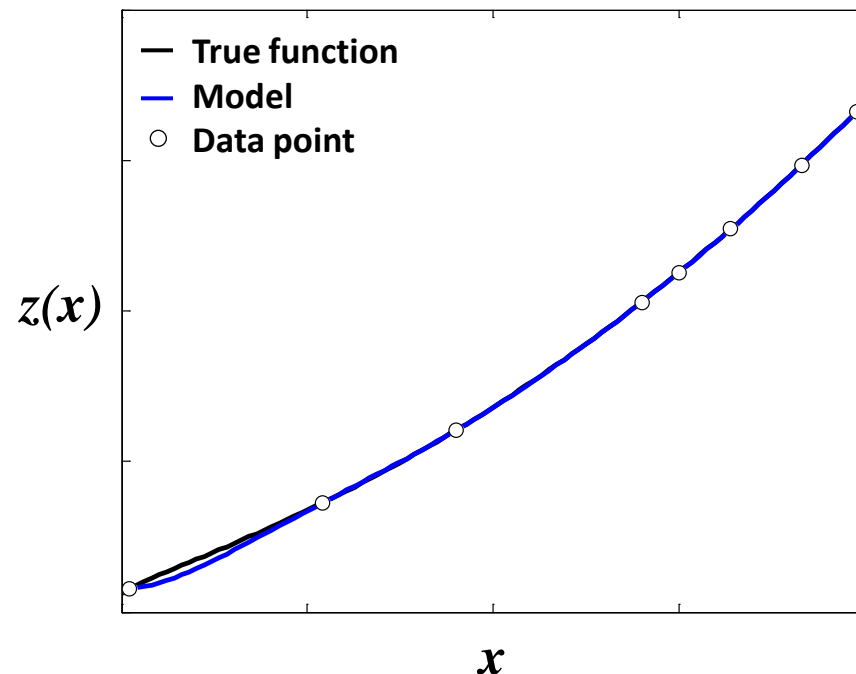
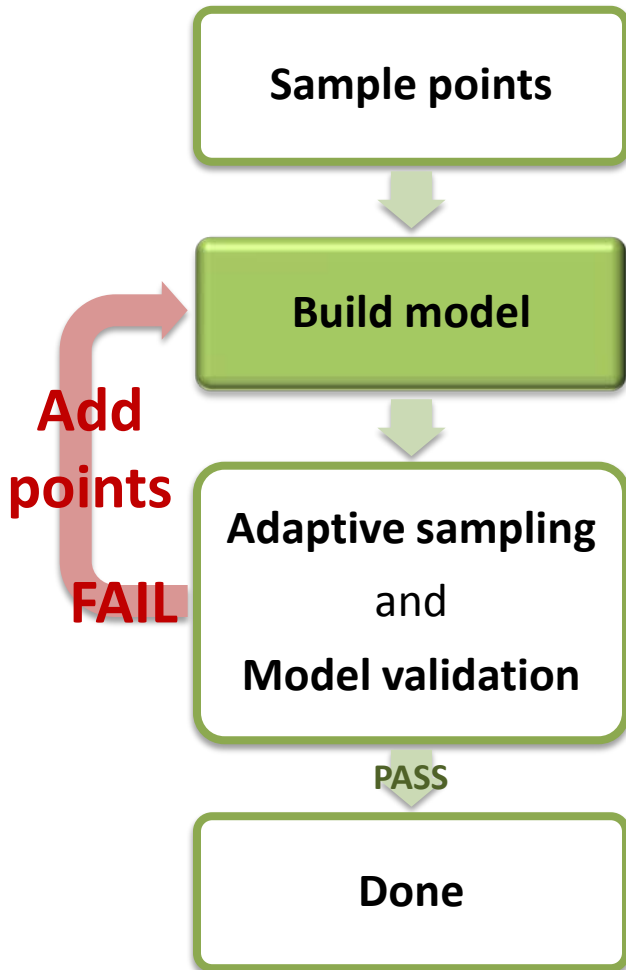
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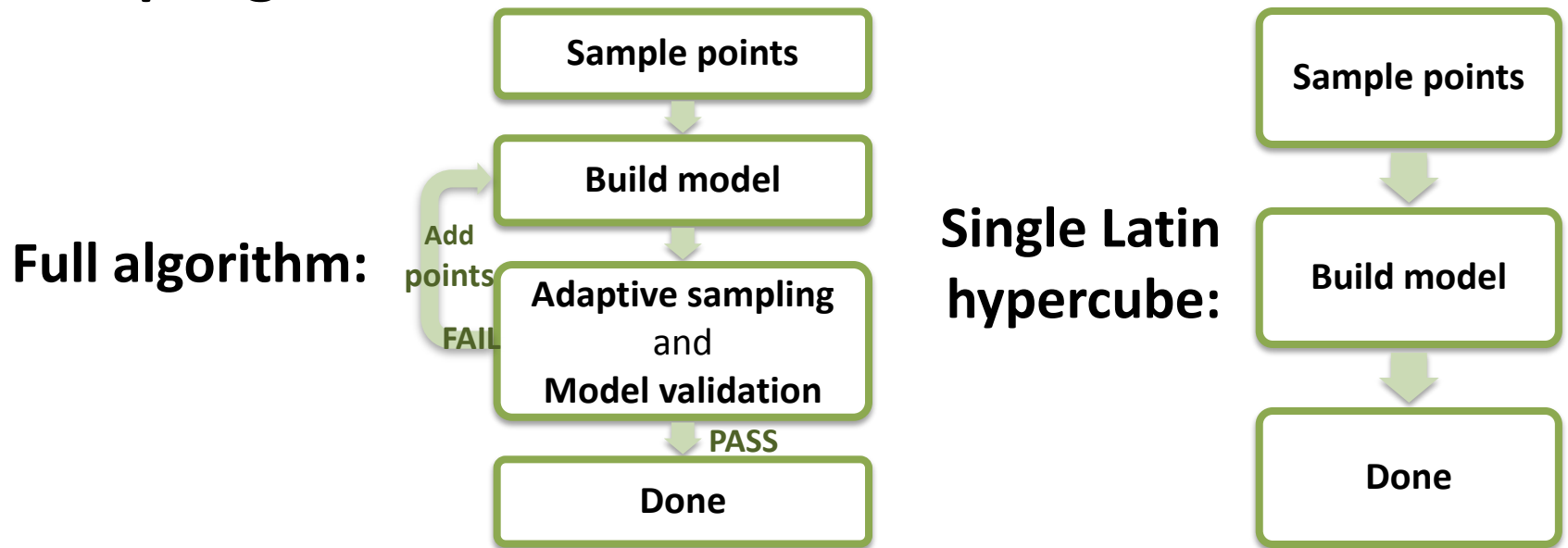
REBUILD THE MODEL

- Add the data points found during adaptive sampling to the training set
- Rebuild the model



TESTING MODEL GENERATION

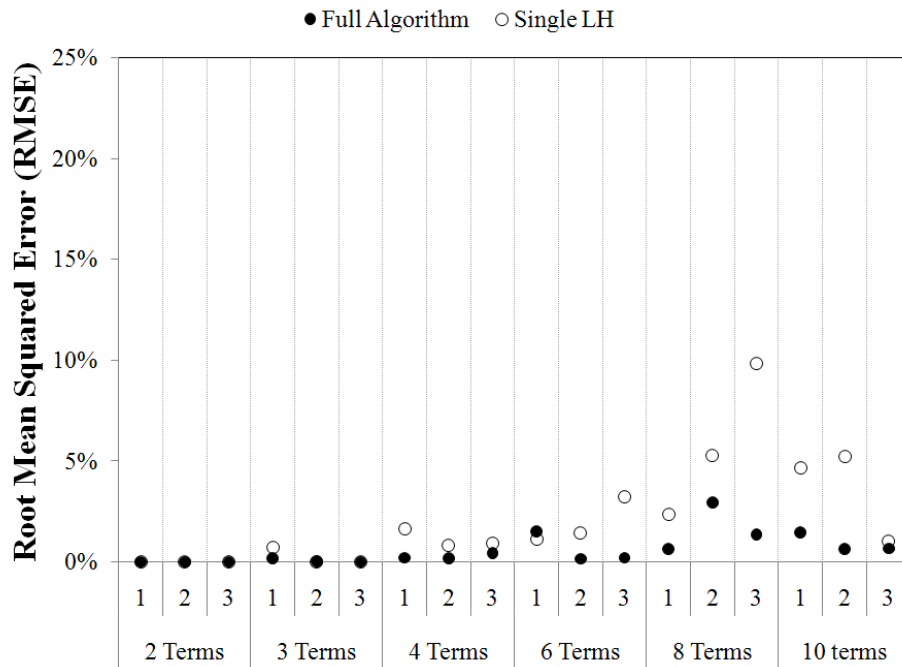
- Compare the method with and without adaptive sampling:



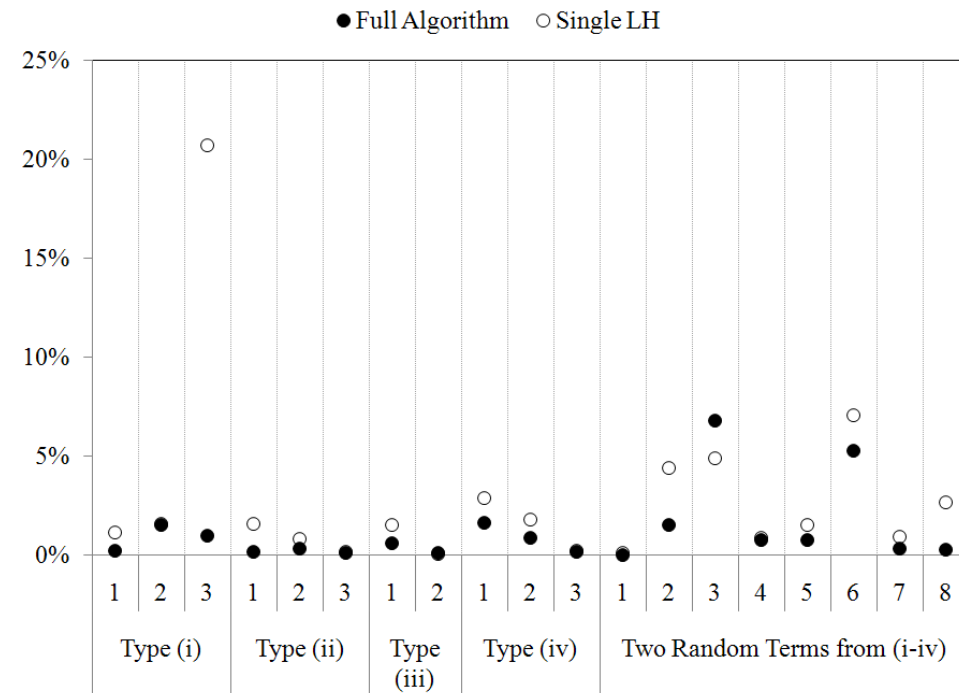
- Two sets of known equations made up of functions
 - Present in the algorithm's basis set
 - Not present in the algorithm's basis set

TESTING MODEL GENERATION

Functions present in the basis set



Functions NOT present in the basis set



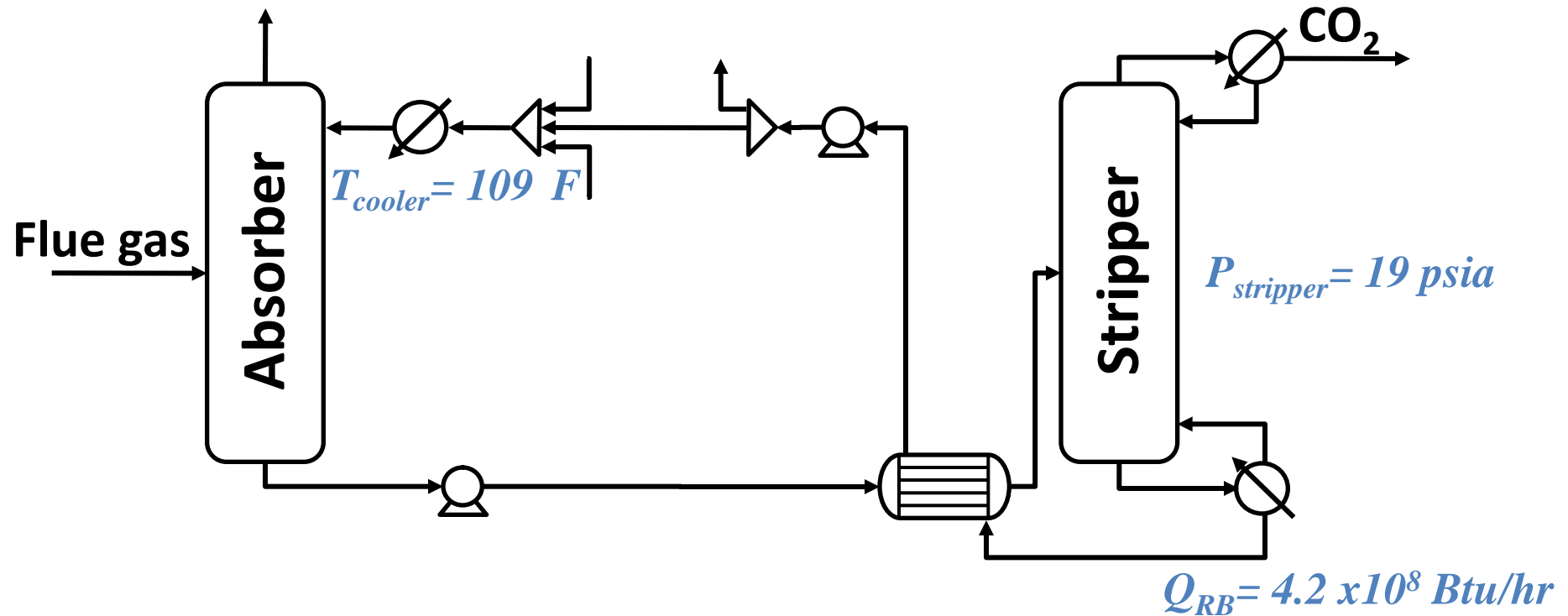
Averaged over three runs for every equation

- i. Pairwise polynomial with unequal exponents*
- ii. Complex, unavailable, fractional forms*
- iii. $x_i^\beta e^{x_i}$*
- iv. $x_i^\beta \log x_i$*

OVERVIEW

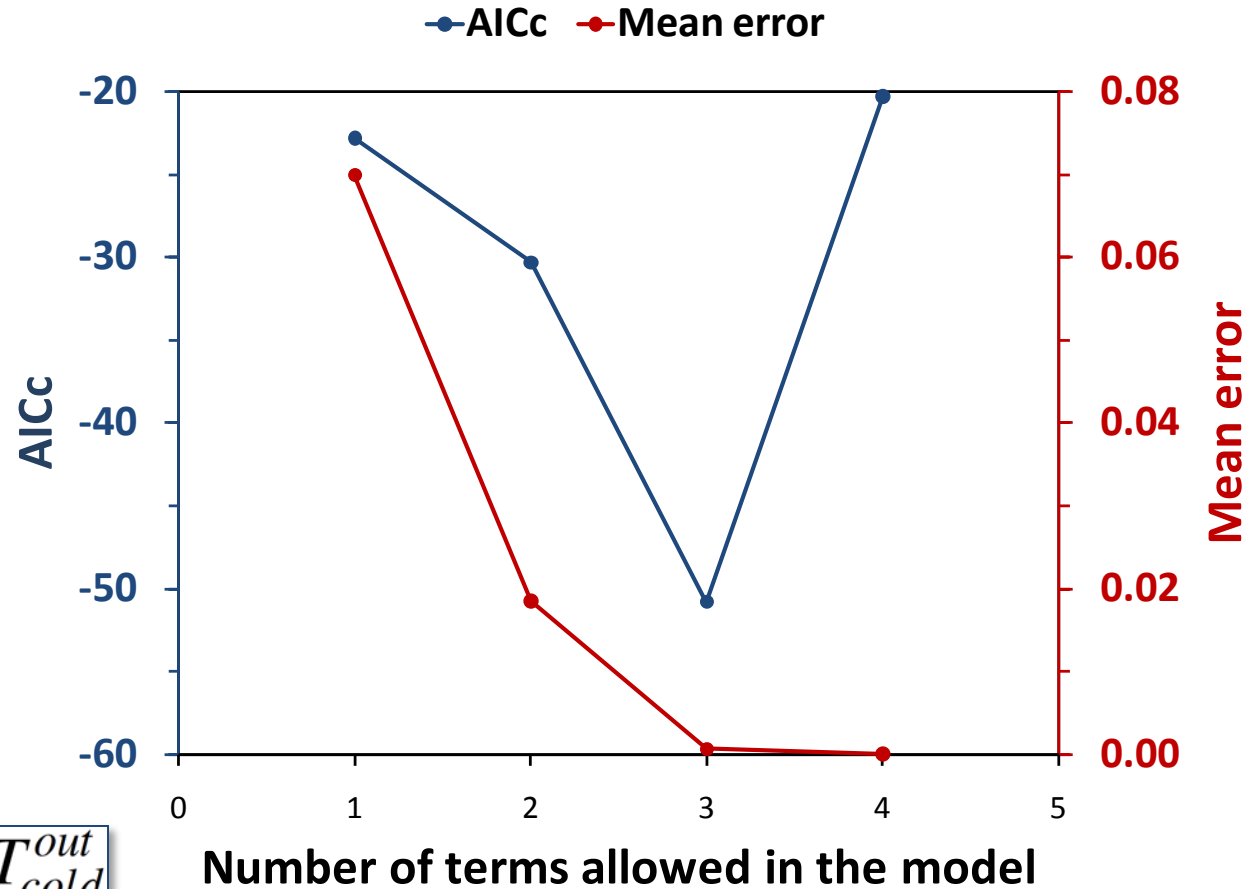
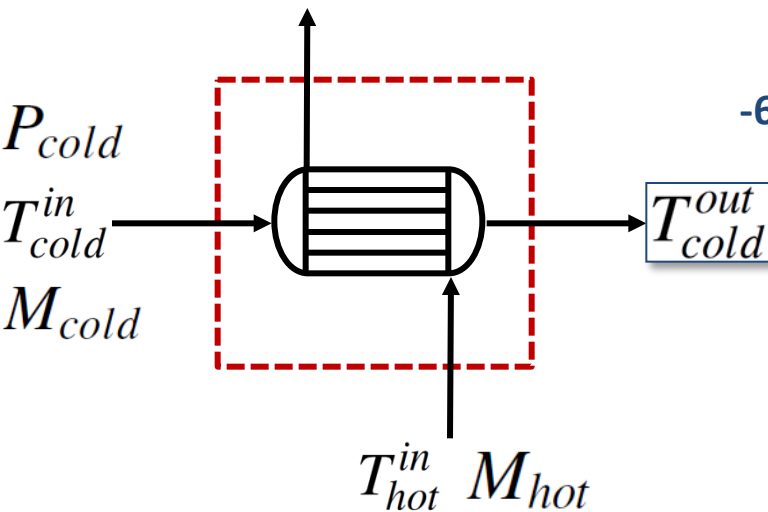
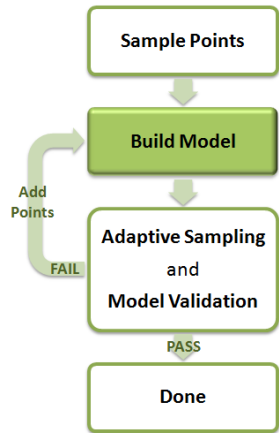
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PRELIMINARY RESULTS

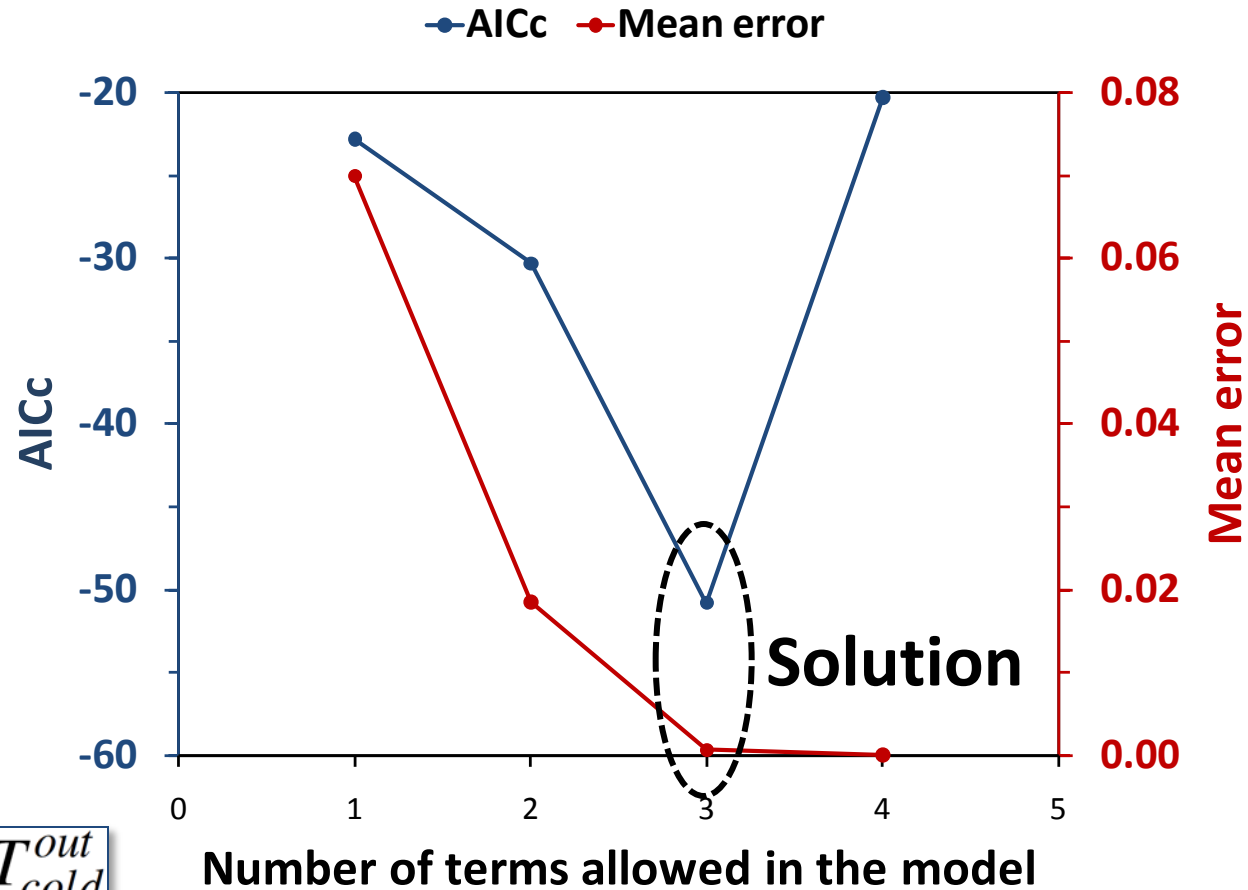
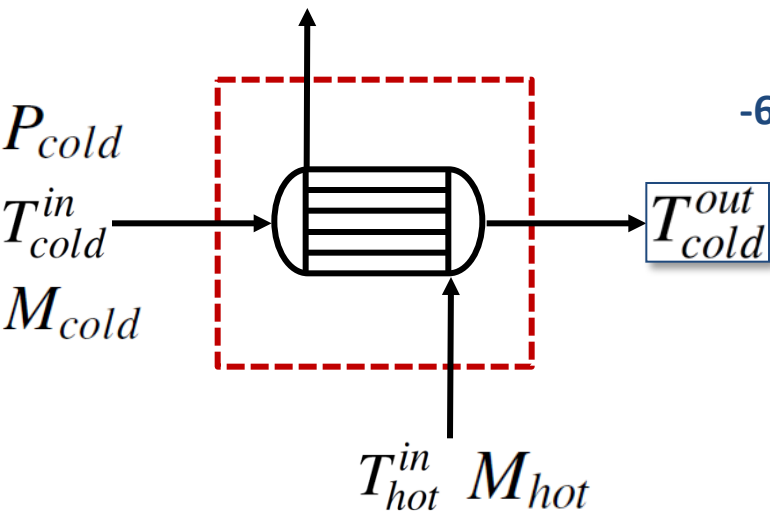
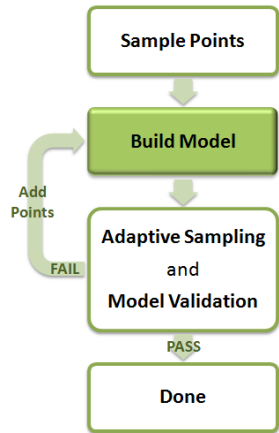


- Minimize stripper reboiler duty (Q_{RB}) with 90% CO₂ capture
- Base case (blue)

BUILDING THE CC MODEL



BUILDING THE CC MODEL



BUILDING THE CC MODEL

Terms Allowed Model for T_{cold}^{out} , °F

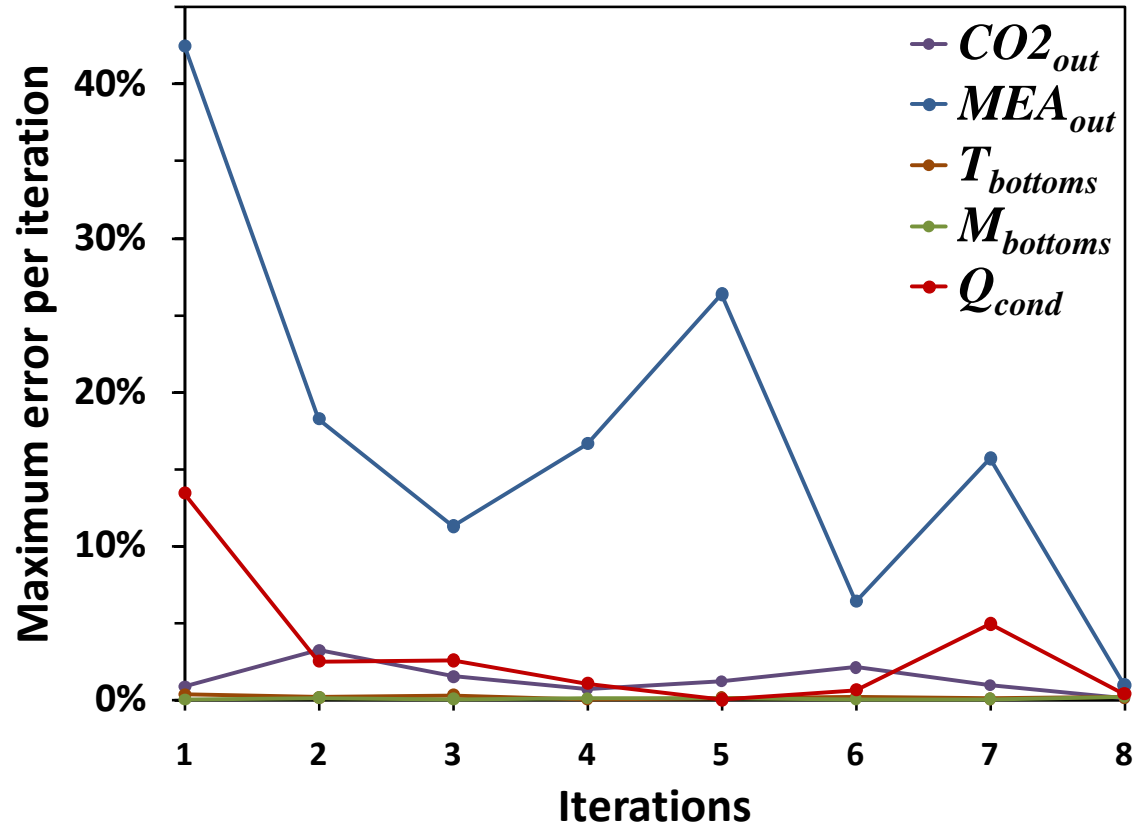
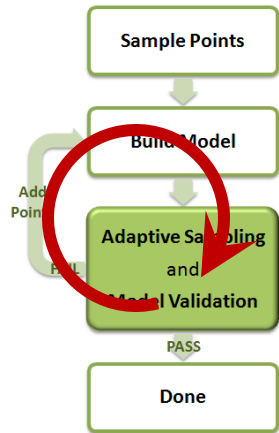
1 $0.017 \left(\frac{T_{hot}^{in}}{242} \right) \left(\frac{M_{cold}}{2.3 \cdot 10^6} \right)^{-1}$

2 $0.020 \left(\frac{T_{hot}^{in}}{242} \right) \left(\frac{M_{hot}}{2.0 \cdot 10^6} \right)^{-1} - 0.0021 \left(\frac{P_{cold}}{22} \right) \left(\frac{M_{hot}}{2.0 \cdot 10^6} \right)^{-1}$

3 $-0.0077 \left(\frac{T_{hot}^{in}}{242} \right)^2 + 0.013 \left(\frac{T_{cold}^{in}}{131} \right) \left(\frac{T_{hot}^{in}}{242} \right)^{-1} + 0.018 \Delta T_{LM}$

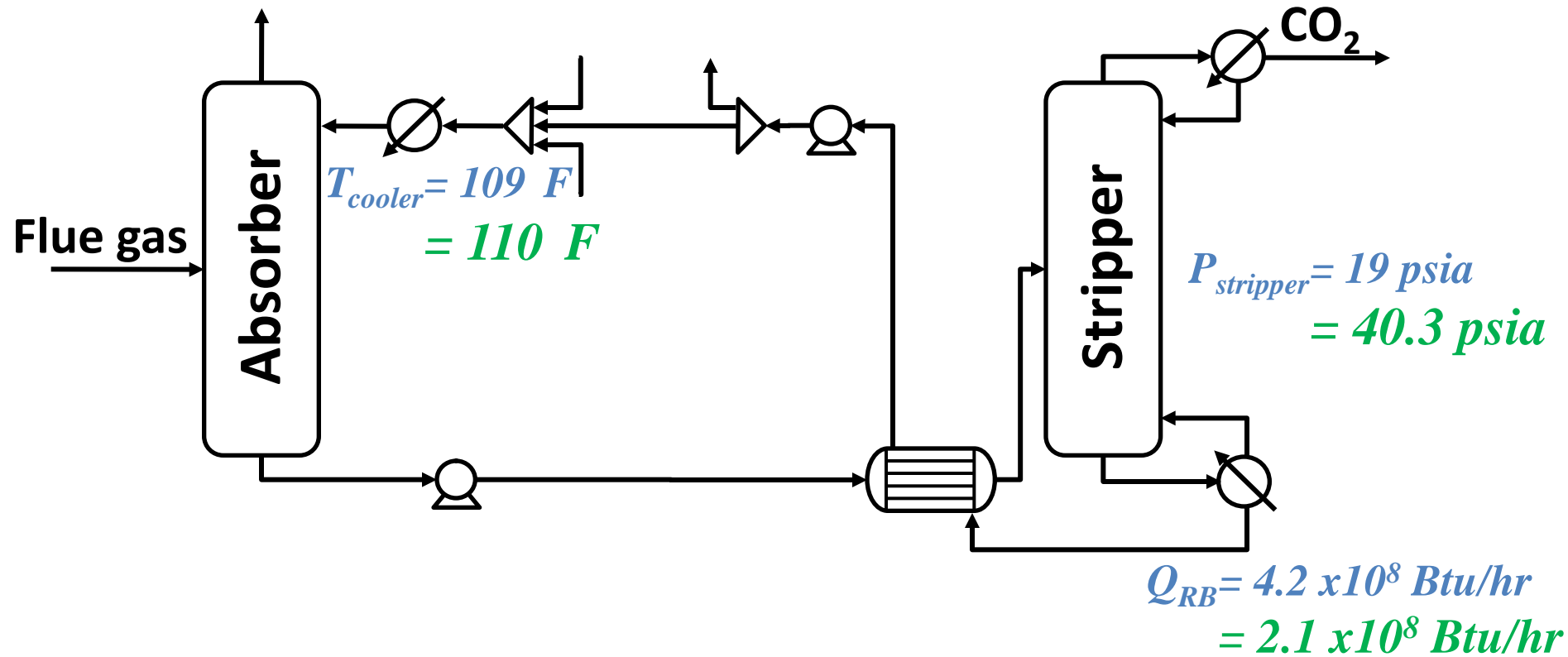
4 $0.011 \left(\frac{T_{hot}^{in}}{242} \right)^2 - 0.052 \left(\frac{T_{cold}^{in}}{131} \right) \left(\frac{T_{hot}^{in}}{242} \right)^{-1} + 0.018 \Delta T_{LM} + 0.046 \left(\frac{T_{cold}^{in}}{131} \right)^{-1}$

ADAPTIVE SAMPLING



- Progression through algorithm for Block 3
- Maximum error found can increase because the derivative-free solver “gets smarter” after each iteration

PRELIMINARY RESULTS



- Minimize stripper reboiler duty (Q_{RB}) with 90% CO₂ capture
- Base case (blue)
- Optimized results (green)

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CONCLUSIONS

- **We developed a method to generate surrogate models of process simulations that**
 - **Are highly accurate**
 - **Have low complexity**
 - **Use minimal function evaluations**
- **We optimized a carbon capture process with two primary decision variables that showed a 50% reduction in reboiler heat over the base case**
- **Future work includes optimizing a carbon capture process integrated with a pulverized coal plant**