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Recent advances in ALAMO

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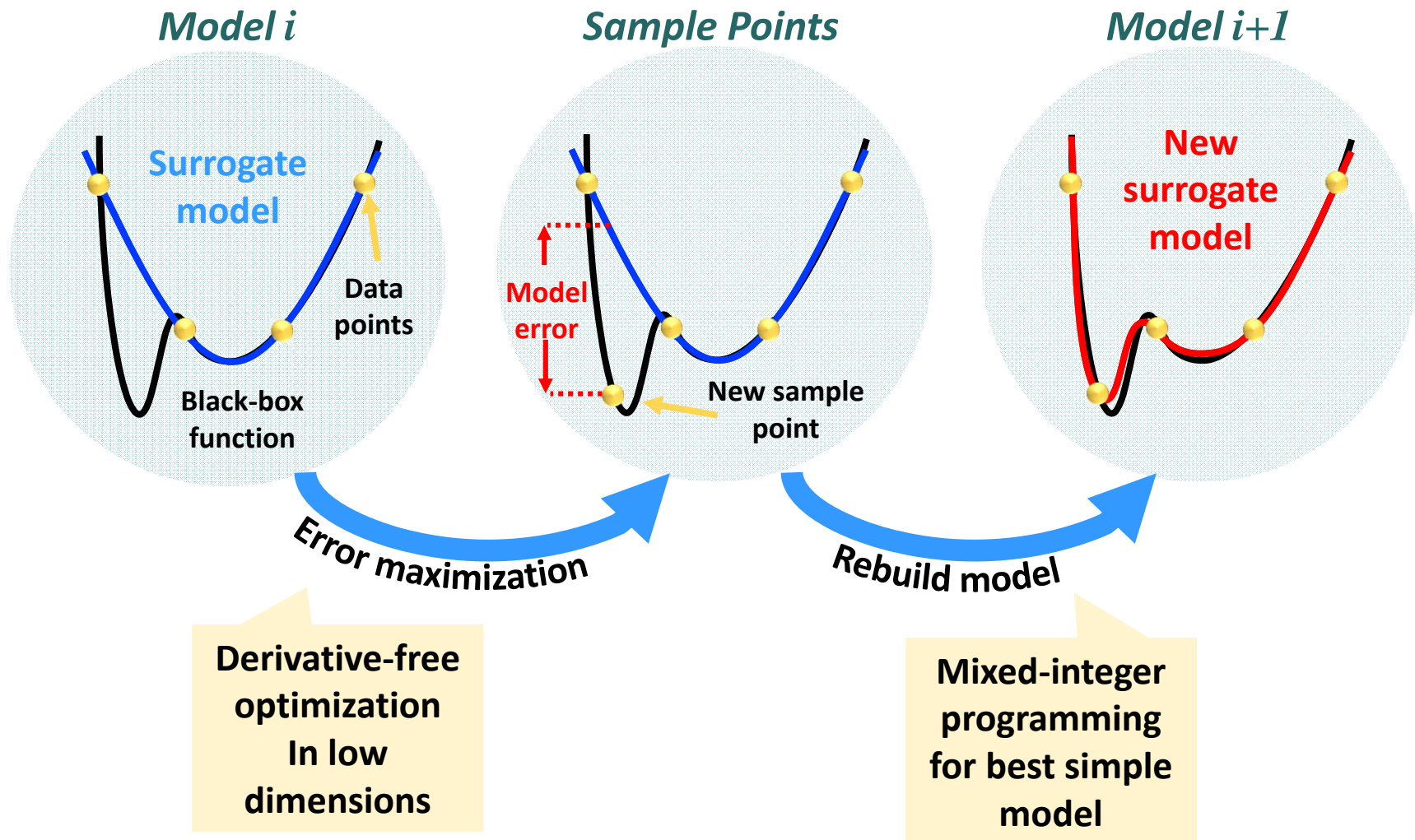
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ALAMO METHODOLOGY



ALAMO SOFTWARE

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 - 4000 lines of **Matlab code**
 - Written in 2011-2013
 - Used GAMS to solve algebraic optimization subproblems
 - Demonstrated advantages vs. least squares and lasso

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- **Current version**
 - **7000 lines of Fortran code**
 - Uses GAMS to solve algebraic optimization subproblems

BASIC ALAMO FUNCTIONS

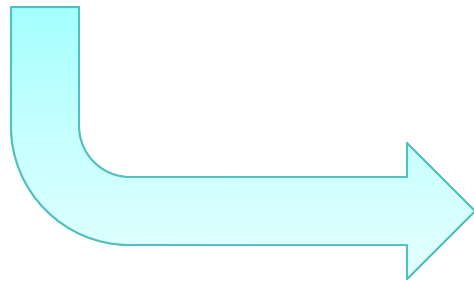
- **Curve fitting**
 - Given a collection of points in an input-output space, find a function that relates inputs to outputs
 - ALAMO considers potential functions and **automatically identifies a simple function that fits the data**

BASIC ALAMO FUNCTIONS

- **Curve fitting**
 - Given a collection of points in an input-output space, find a function that relates inputs to outputs
 - ALAMO considers potential functions and automatically identifies a simple function that fits the data
- **Design of experiments**
 - ALAMO can interface to a simulator (or experimental system) to **guide collection of data** with the aim to
 - *Build a model entirely from scratch*
 - *Validate quality of model built based on a pre-existing data set*
 - *Supplement existing experimental data*

EXAMPLE ALAMO INPUT FILE

Input-output data used to fit model



```
ninputs 1  
noutputs 1  
ndata 11
```

```
BEGIN_DATA  
-5      25  
-4      16  
-3       9  
-2       4  
-1       1  
0        0  
1         1  
2         4  
3         9  
4        16  
5        25  
END_DATA
```


ALAMO OUTPUT

Step 1: Build model

Successive models for variable Z1

Z1 = 0.00

Z1 = 1.0 * X1**2.0

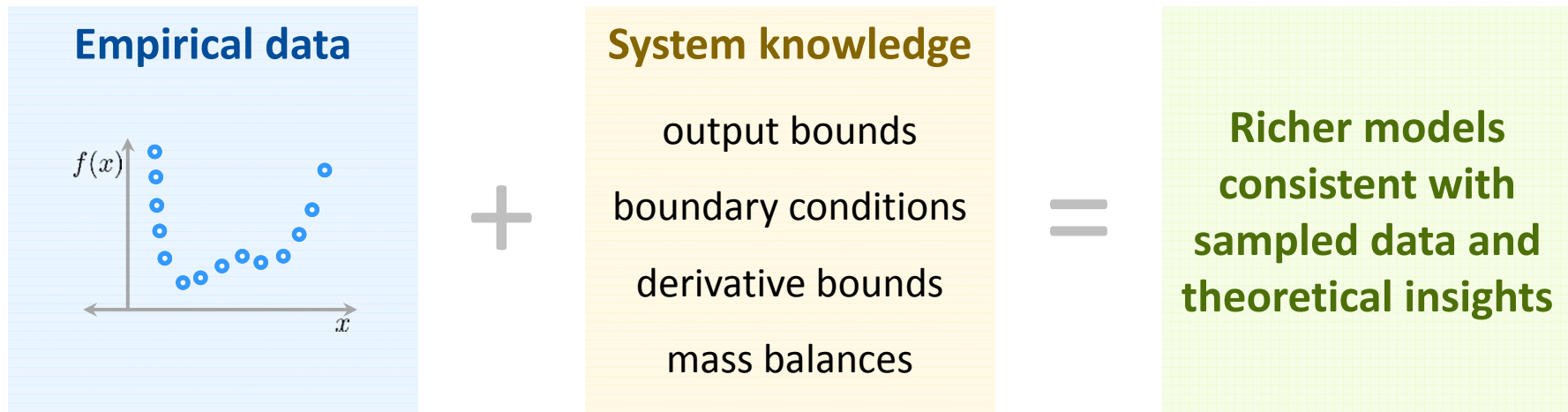
Calculating errors on observed data set

RMSE for observed data set:

Z1: 0.96E-05

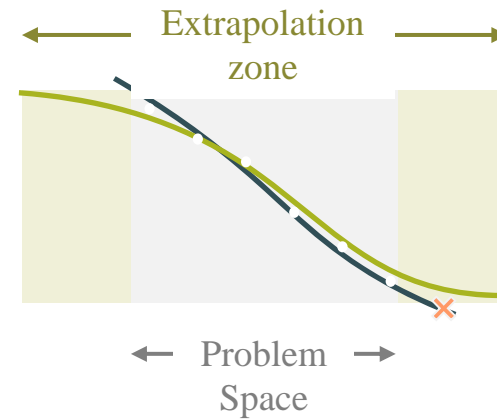
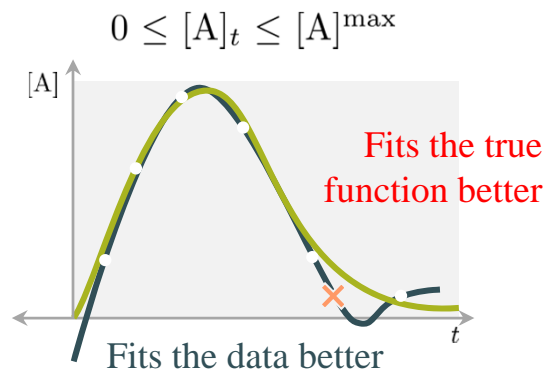
Total execution time 0.72E-01 s

A DATA- **AND** THEORY-DRIVEN APPROACH



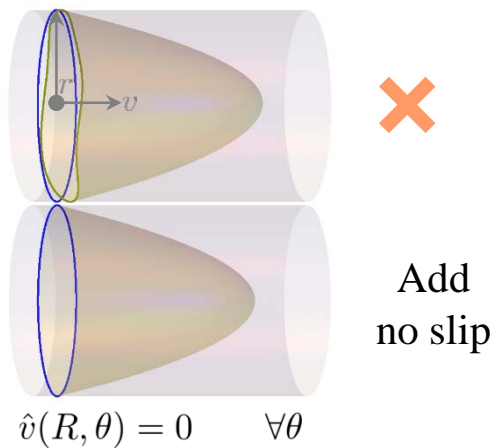
- **Apply system knowledge beyond sampled data to build regression models**
 - First principles derivation
 - Engineering insight
 - Output and derivative bounds
- **Insights that are generally available ‘free of charge’**

CONSTRAINED REGRESSION



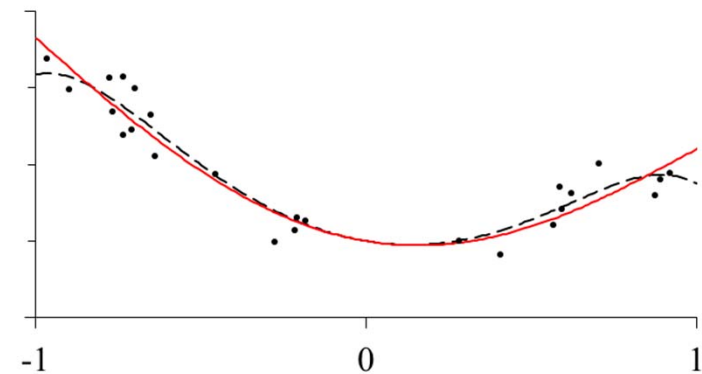
Safe Extrapolation

velocity profile model



Boundary values

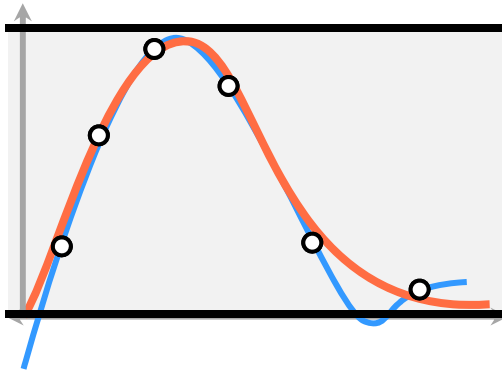
--- Unconstrained model — Forced convexity



Convexity

IMPLEMENTATION

Enforced everywhere



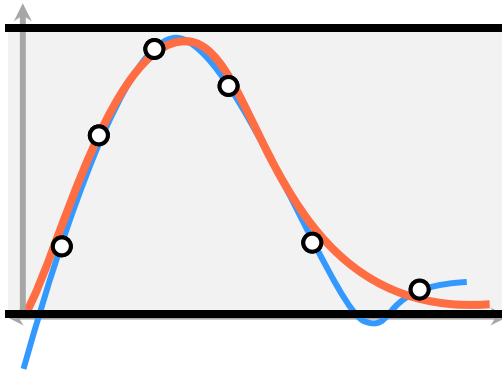
$$\min_{\beta} \sum_{i=1}^N (z_i - [\beta_0 + \beta_1 x + \beta_2 x^2 + \dots])^2$$

$$\text{s.t. } z^l \leq \beta_0 + \beta_1 x + \beta_2 x^2 + \dots \leq z^u \quad \forall x$$

Ordinary least squares regression objective

IMPLEMENTATION

Enforced everywhere



Ordinary least squares regression objective

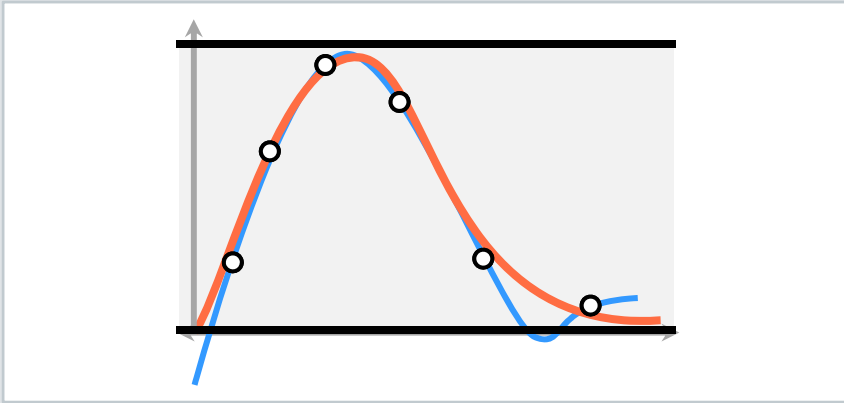
$$\min_{\beta} \sum_{i=1}^N (z_i - [\beta_0 + \beta_1 x + \beta_2 x^2 + \dots])^2$$

$$\text{s.t. } z^l \leq \beta_0 + \beta_1 x + \beta_2 x^2 + \dots \leq z^u \quad \forall x$$

Semi-infinite program

IMPLEMENTATION

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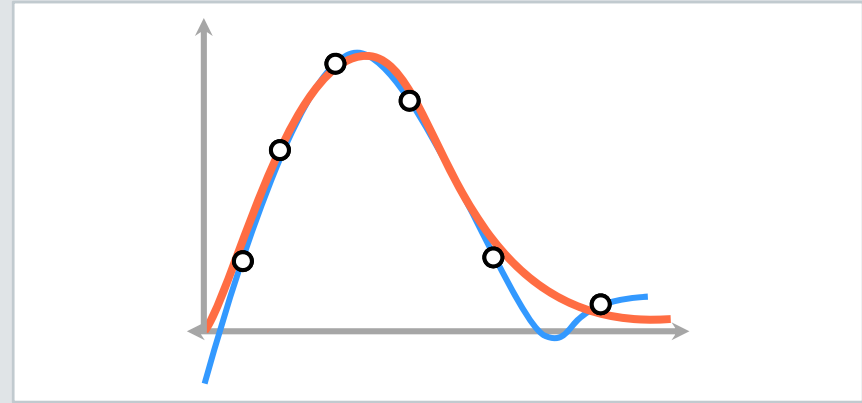


$$\min_{\beta} \sum_{i=1}^N (z_i - [\beta_0 + \beta_1 x + \beta_2 x^2 + \dots])^2$$

$$\text{s.t. } z^l \leq \beta_0 + \beta_1 x + \beta_2 x^2 + \dots \leq z^u \quad \forall x$$

Semi-infinite
program

Bounding set



$$\min_{\beta} \sum_{i=1}^N (z_i - [\beta_0 + \beta_1 x + \beta_2 x^2 + \dots])^2$$

$$\text{s.t. } z^l \leq \beta_0 + \beta_1 x_j + \beta_2 x_j^2 + \dots \leq z^u$$

$\forall j \in \text{Bounding set}$

Simple linear
constraints

CONCLUSIONS

- **Expanding the scope of MINLP algorithms**
 - Using low-complexity surrogate models to strike a balance between optimal decision-making and model fidelity
- **Enforcing theoretical insights to data-driven experimental science**
 - Iteratively use implicit regression constraints to ensure limits on models
- **ALAMO site:**
archimedes.cheme.cmu.edu/?q=alamo

$$\begin{aligned} \min & f(x, y) \quad \blacksquare \\ \text{s.t.} & g(x, y) \leq 0 \\ & h(x, y) = 0 \quad \blacksquare \end{aligned}$$

