



National Nuclear Security Administration

Overview

Introduction

- Giving structure to experimental data
 - PrIMe Data Warehouse
- New PrIMe application
 - •front-end application to the CCMSC coal database (filter, visualization, and export data)
- Bound-to-Bound Data Collaboration workflow for model validation

Summary

Introduction

- Predictive modeling starts with validation
- Experimental data stored in various file formats
 - CSV, Excel, tab delimited, ASCII, etc.
 - No standard
- Each record requires specialized knowledge of how the data was stored
 - Can be an incomplete record of experiment with missing information
- We would like automated access to data
 - Without structure, query requests are quickly intractable across a diverse collection of data
- Efficiently discover validation data to incorporate in the model validation process

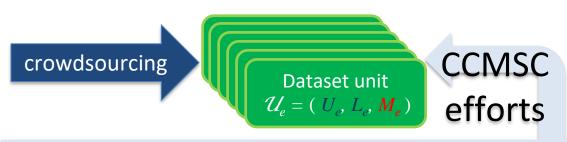
Providing Structure to Experimental Data



primekinetics.org

- What is PrIMe?
 - Data Warehouse repository of experimental records
 - Applications aid in development of predictive models
- Transformation of information into a usable form
- PrIMe's data models use XML schemas to provide structure
 - Contains complete information of an experiment
 - Experimental data is stored in XML or HDF5 files
- Storage of raw experimental data and derived properties
 - Ability for instrumentation modeling

CCMSC Coal Database for V/UQ



- International Flame Research Foundation, Livorno, Italy
- Sandia National Laboratory, Livermore, CA

269 Solid Fuels & Blends

Fossil, Biomass, Sludge, Waste, Char

2710 Data Groups collected from 1016 Records

Varying Conditions (Temperatures, %O₂, %H₂O, Gas Mixture) Experiment Types: *Devolatilization, Char oxidation*

In collaboration with Salvatore Iavaron and Alessandro Parente, Université Libre de Bruxelles

leveraging existing cloud infrastructure and data models

```
<?xml version="1.0"?>
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xmlns:xsi="http://www.w3.org/2001/XMLSchema-instance"
   <copyright>@primekinetics.org 2017</copyright>
   <bibliographyLink primeID="b00019060" preferredKey="Leis</pre>
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       <kind>drop-tube furnace</kind>
     - - cproperty name="length" label="Length" units="m">
          <value>4</value>
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     - - cproperty name="length" label="D_inner" units="mm">
          <value>150</value>
       </property>
   </apparatus>
 - <commonProperties>
     + + property name="initial composition">
     + property name="temperature" label="T_furnace" units
     + property name="pressure" label="P" units="atm">
   </commonProperties>

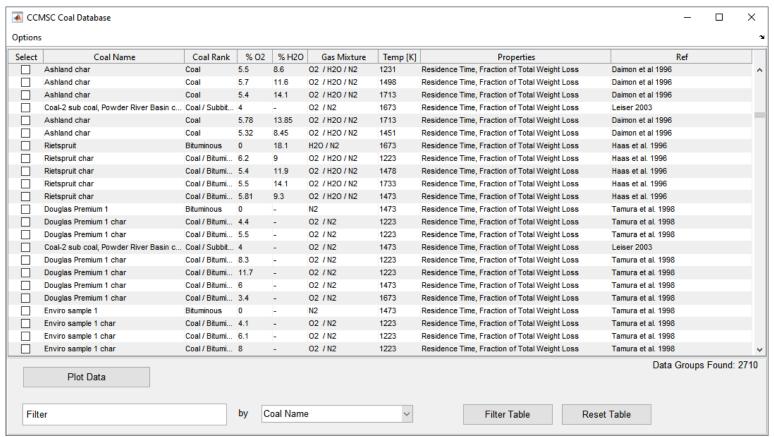
    - <dataGroup label="Residence time vs Weight Loss" id="dg</li>

       cproperty description="Residence Time" name="resider
       property description="Fraction of Total Weight Loss" r
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          < x1 > 0 < /x1 >
          <x2>0</x2>
       </dataPoint>
     - <dataPoint id="dp2">
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          <x2>0.144</x2>
       </dataPoint>
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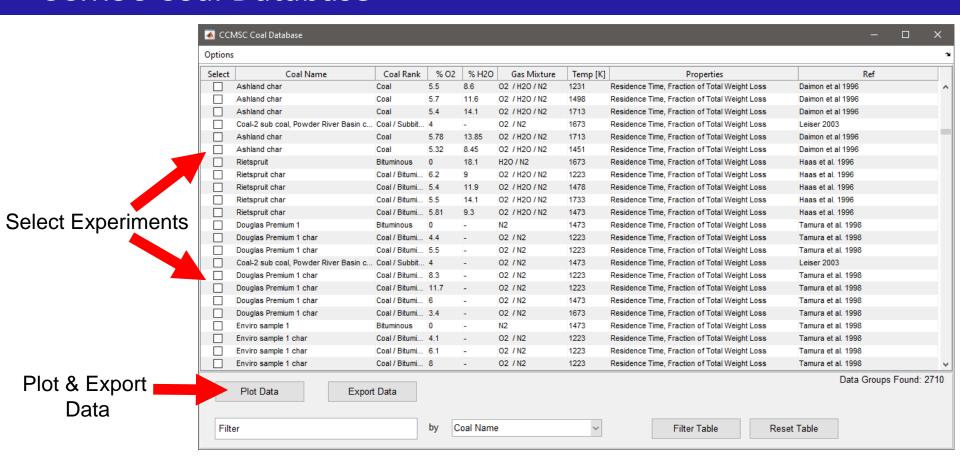
CCMSC Coal Database

primekinetics.org

github.com/oreluk/coalDB

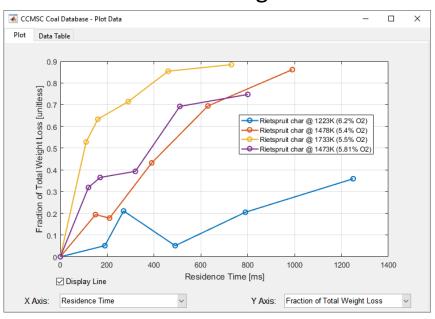


CCMSC Coal Database

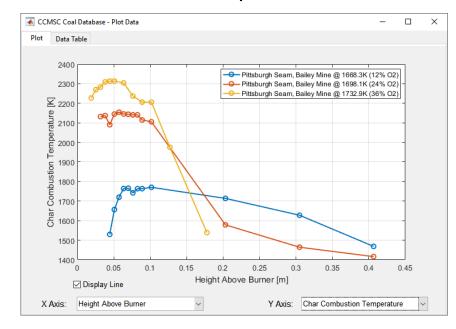


CCMSC Coal Database

Fraction of Weight Loss



Char Temperature



Char Oxidation Example

Experimental Data of Utah Skyline coal from Sandia's Laminar Entrained Flow Reactor

Features:

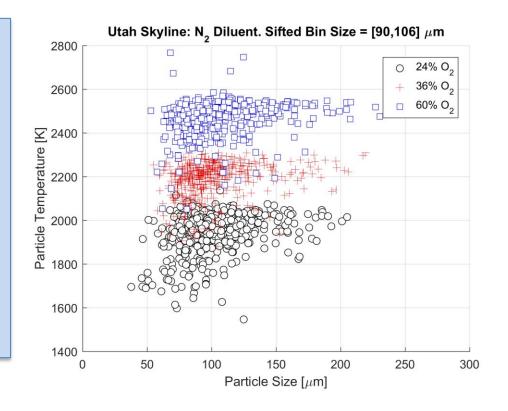
CO₂ or N₂ diluent

Initial Particle Diameter: $53 - 150 \mu m$

 O_2 : 24 - 60%

 $H_2O: 10-16\%$

Validation data at 399 different gas conditions & heights above burner



Uncertain parameters: $x \in \mathcal{H} \subseteq \mathbb{R}^n$

$$\mathcal{H} = \{ x \in \mathbb{R}^n : l_i \le x_i \le u_i, \ i = 1, \dots, n \}$$

Bounds on QOI model: $L_e \leq M_e(x) \leq U_e$, for e = 1, ..., N

Dataset:
$$x \in \mathcal{H} \subseteq \mathbb{R}^n$$

$$L_e \le M_e(x) \le U_e$$
, for $e = 1, ..., N$

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$$x \in \mathcal{H} \subseteq \mathbb{R}^n$$

 $L_e < M_e(x) < U_e$, for $e = 1, ..., N$

Feeley, Seiler, Packard, Frenklach. J. Phy. Chem. A, 2004, 108, 9573-9583.

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Dataset:
$$x \in \mathcal{H} \subseteq \mathbb{R}^n$$

$$L_e \leq M_e(x) \leq U_e$$
, for $e = 1, ..., N$

Feasible set:
$$\mathcal{F} = \bigcap_{e=1}^{N} \{x \in \mathcal{H} : L_e \leq M_e(x) \leq U_e\}$$

Scalar consistency measure

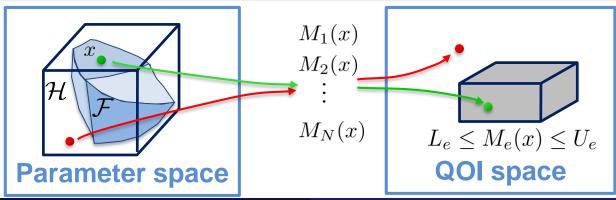
Scalar consistency measure

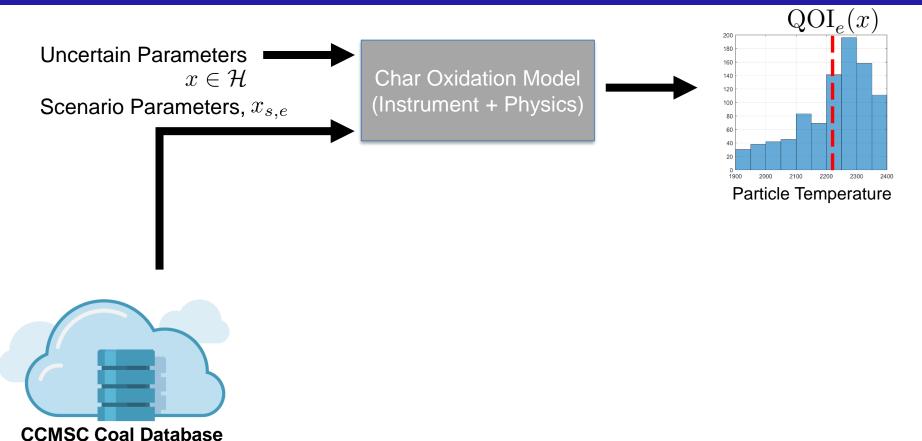
$$\gamma \geq 0$$
: Consistent Dataset

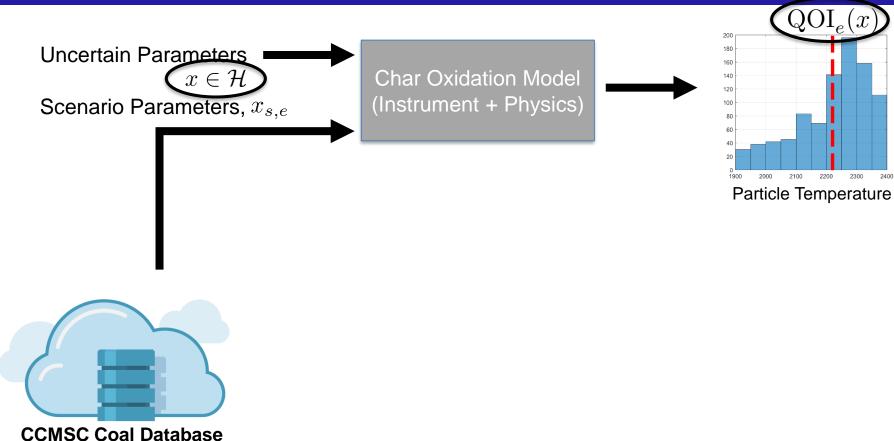
$$\gamma < 0$$
: Inconsistent Dataset

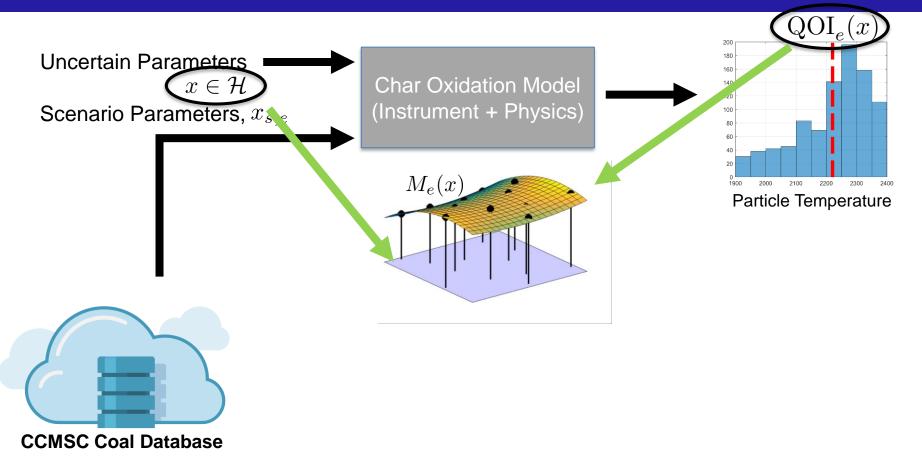
$$C_{\text{Dataset}} = \max_{\gamma, x \in \mathcal{H}} \quad \gamma$$

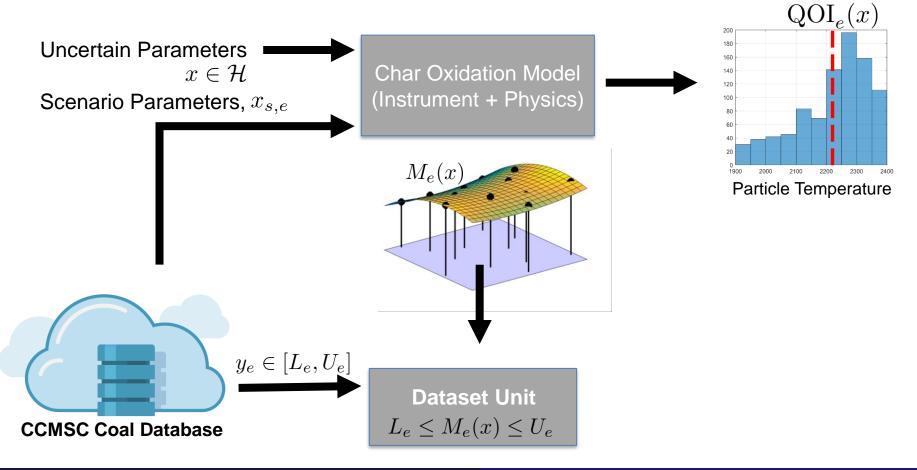
s.t. $L_e + \frac{(U_e - L_e)}{2} \gamma \leq M_e(x) \leq U_e - \frac{(U_e - L_e)}{2} \gamma$
for $e = 1, \dots, N$.

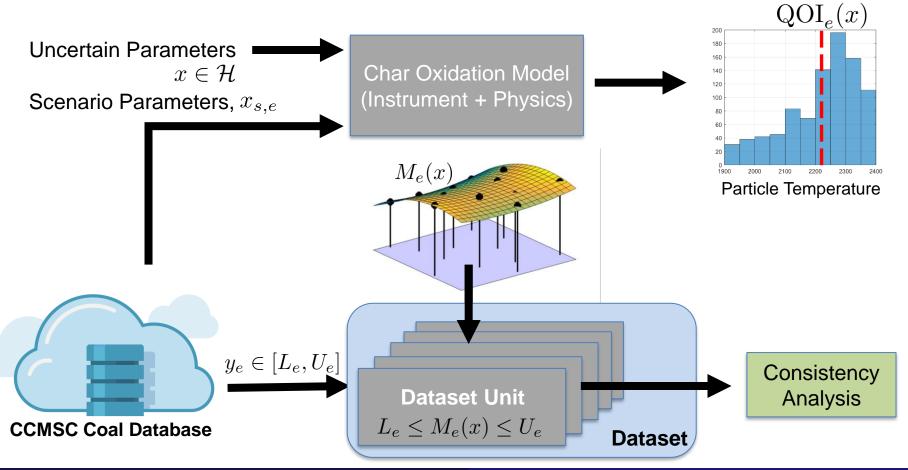












Model Form

$$\begin{array}{ccc} \mathbf{C_{(s)}} + \mathbf{O_2} & \longrightarrow & \mathbf{CO_2} \\ 2 \, \mathbf{C_{(s)}} + \mathbf{O_2} & \longrightarrow & 2 \, \mathbf{CO} \end{array} \} \ r = \mathbf{A} \exp \left(\frac{-\mathbf{Ea}}{RT_p} \right) C_{\mathrm{oxid}}$$

Transport

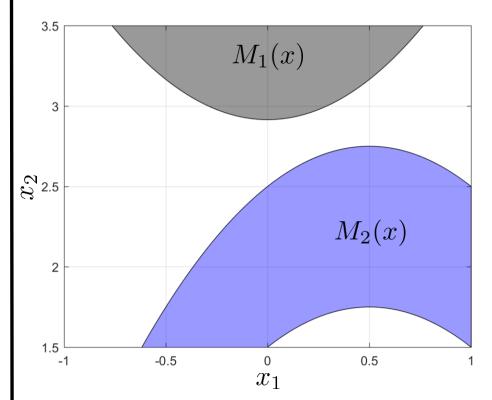
- Diffusion of oxidizer to particle surface
- Diffusion of products from particle surface

Scalar consistency measure:

$$C_{Dataset} = [-0.26, -0.19]$$

If **all** constraints are expanded by at least 26% the inconsistency can be resolved.

If **all** constraints are expanded by no more than 19% the inconsistency **cannot** be resolved.



Iavarone et al. *Energy Procedia*, 120 (2017): 500-507.

Model Form

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Transport

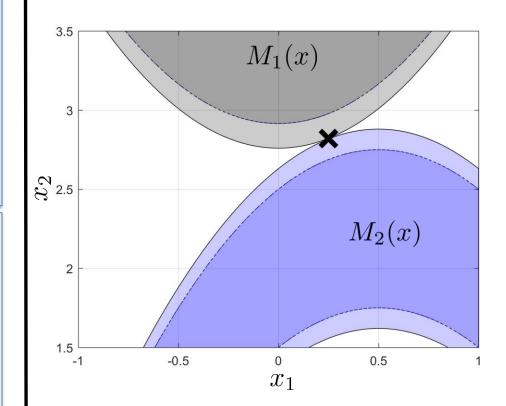
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Model Form

Uncertain Kinetic Parameters

$$\begin{array}{c} C_{(s)} + O_2 \longrightarrow & CO_2 \\ 2 \, C_{(s)} + O_2 \longrightarrow & 2 \, CO \end{array} \right\} \qquad \left\{ \begin{aligned} A_{O_2}, E_{O_2} \right\} \\ C_{(s)} + CO_2 \longrightarrow & 2 \, CO \end{array} \qquad \left\{ \begin{aligned} A_{CO_2}, E_{CO_2} \right\} \\ C_{(s)} + H_2O \longrightarrow & H_2 + CO \end{aligned} \qquad \left\{ \begin{aligned} A_{H_2O}, E_{H_2O} \right\} \end{aligned}$$

Transport

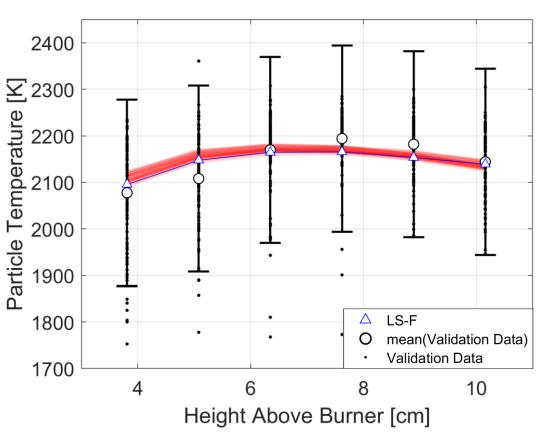
- Diffusion of oxidizer to particle surface
- Diffusion of products from particle surface
- Diffusion of oxidizer through coal particle
 - coal particle is a porous medium with internal surface area

Scalar consistency measure:

$$C_{\text{Dataset}} = [0.06, 0.32]$$

$$\mathcal{F} = \bigcap_{e=1}^{N} \{ x \in \mathcal{H} : L_e \leq M_e(x) \leq U_e \}$$

$$N_2:54\%, H_2O:10\%, O_2:36\%, d_{bin}=82.5\mu m$$



$$x_{\text{LS-F}} = \underset{x \in \mathcal{F}}{\operatorname{argmin}} ||M_e(x) - y_e||_2$$

 y_e : mean of the validation data

Summary

- Developed new data models for coal data
- Easy filtering through a diverse collection of experimental data
- B2BDC based test-bed for exploring parameter and model form uncertainty
 - With a consistent dataset we can do prediction of posterior QOI or parameter bounds, and sample the feasible set for correlations between parameters and QOIs

Acknowledgements

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