

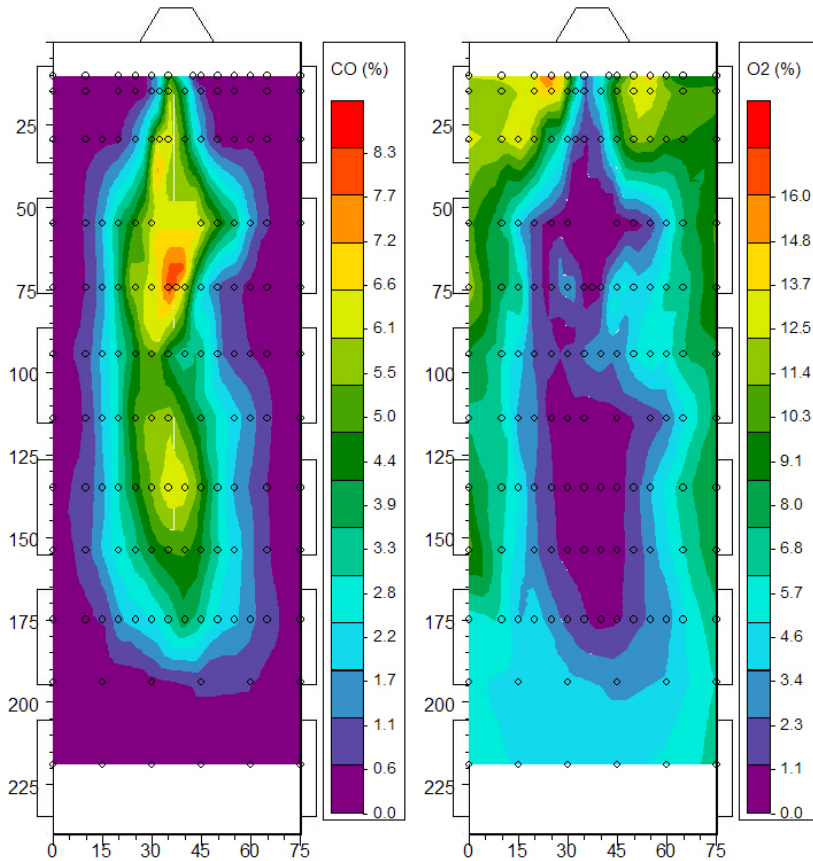
Two vertical images of flames are positioned on the left side of the slide. The top image shows a bright, turbulent flame within a dark, rectangular container. The bottom image shows a similar flame, but with a more distinct, swirling or vortex-like structure, also within a dark container.

Comparison of Two Mixing Models with Experimental, Solid-Fuel, Swirl-Stabilized Flame Data

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Introduction



- Significant overlap of CO and O₂
- How does this behavior occur?
- How to model this behavior?

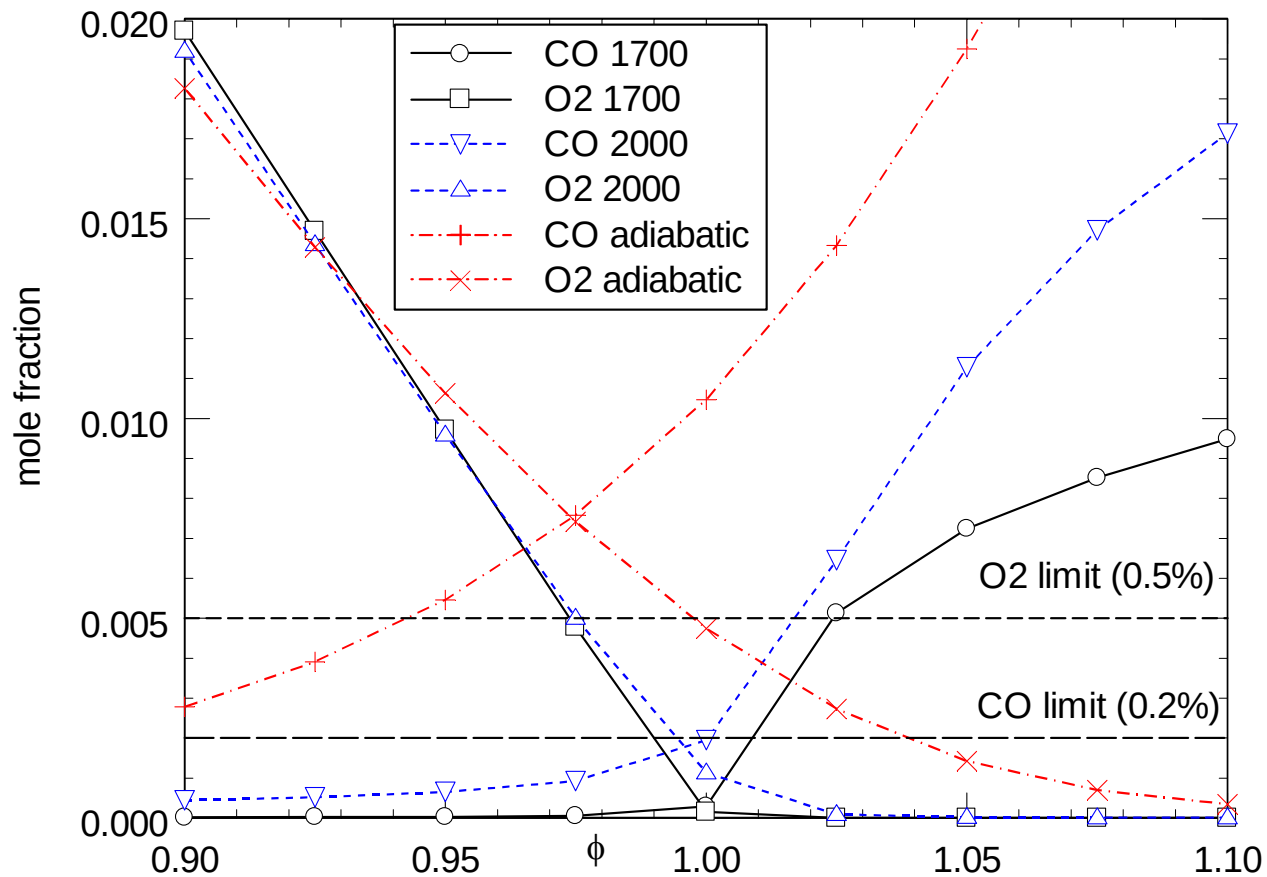
Objectives

- Show “Intermittency” in flame data
- Introduce new mixing model based on intermittency phenomenon
- Compare relative merits of new model and a current model often used in CFD predictions

Intermittency

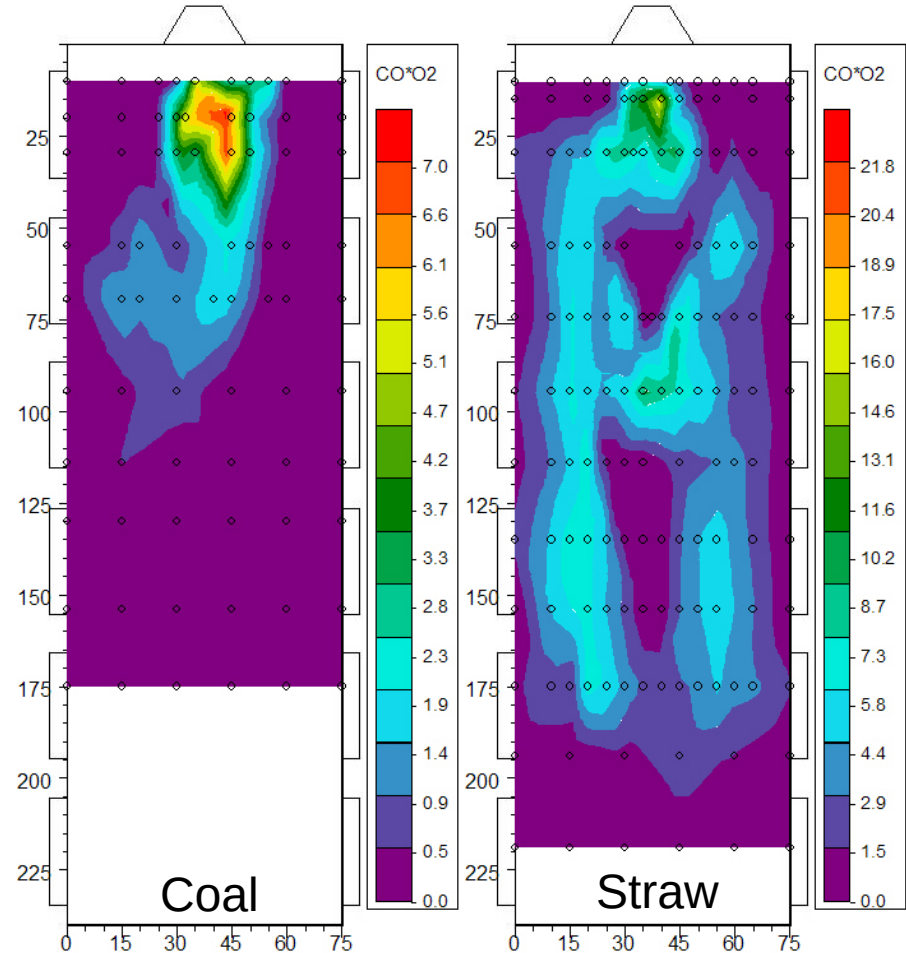
- Flame intermittency: The movement of a flamelet in and out of a region in space
- Distributed vs wrinkled vs candle flames
- In a sampling volume, both CO (from a flamelet) and O₂ (from the surrounding gas) will be recorded

Equilibrium O_2 and CO



Intermittency Maps

- Multiplying O_2 with CO gives an indication of where flame intermittency is occurring
- Even in coal, intermittent flame behavior is common



Intermittency Maps



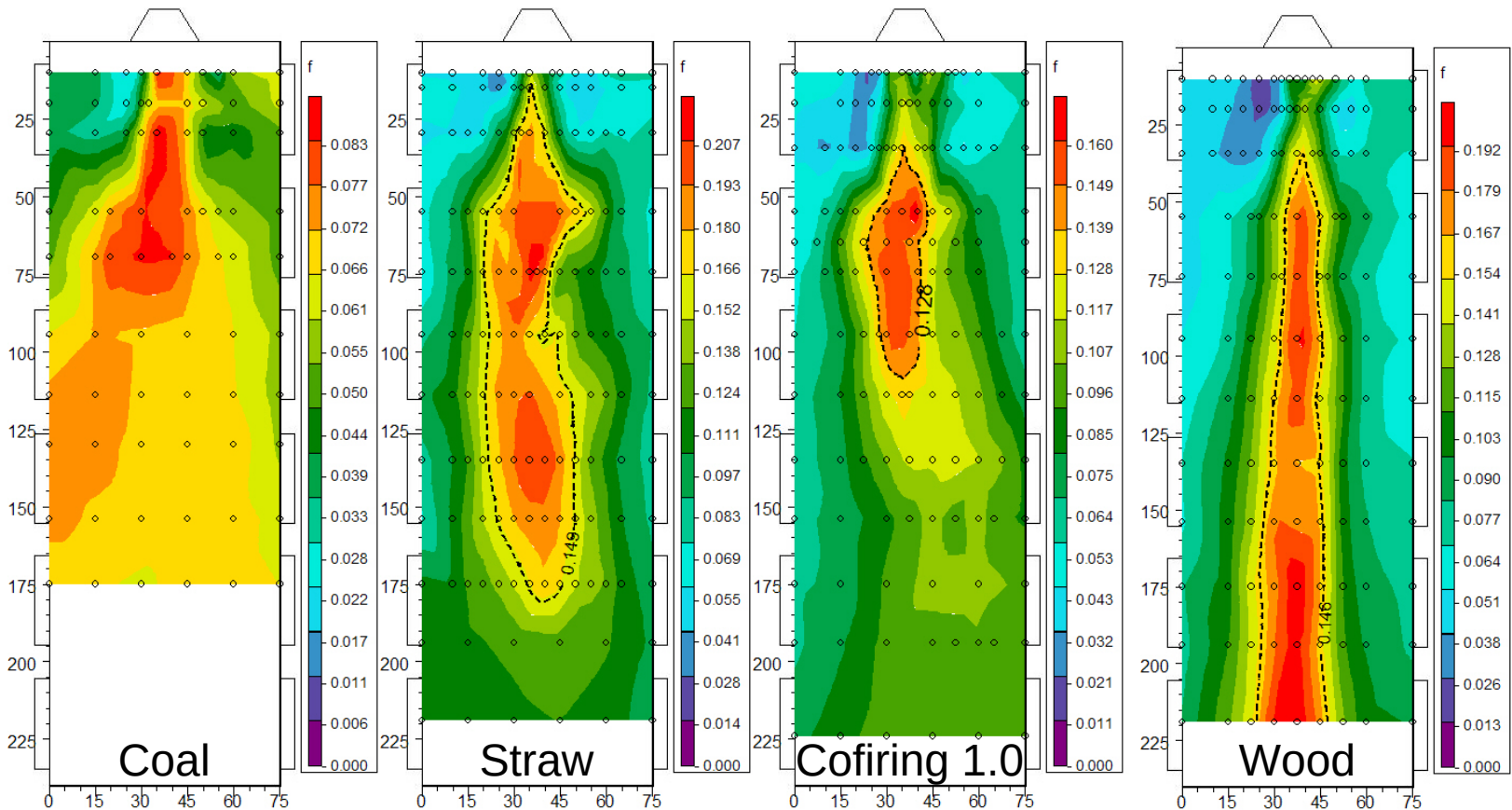
- Most intense near burner
- Holes in the straw, wood and cofiring flames
- Also show intermittent behavior:
- Non-penetration Coal Flames
 - Truelove 1988, Truelove 1990, Weber 1992, Godoy 1988, Godoy 1989, Costa 1991, Jensen 1994, Illerup 1994, Wu 2006
- Penetration Coal Flames
 - Smart 1988, Visser 1990, Abbas 1992, Abbas 1993, Smart 1996, Ballester 2005
- Cofiring Flames
 - Abbas 1994, Wu 2006, Damstedt 2006
- Biomass Flames
 - Ballester 2005, Wu 2006

Predictability

- Overlap of O_2 and CO not a new observation
- Predictive models in terms of the mixture fraction: f

Data Regression

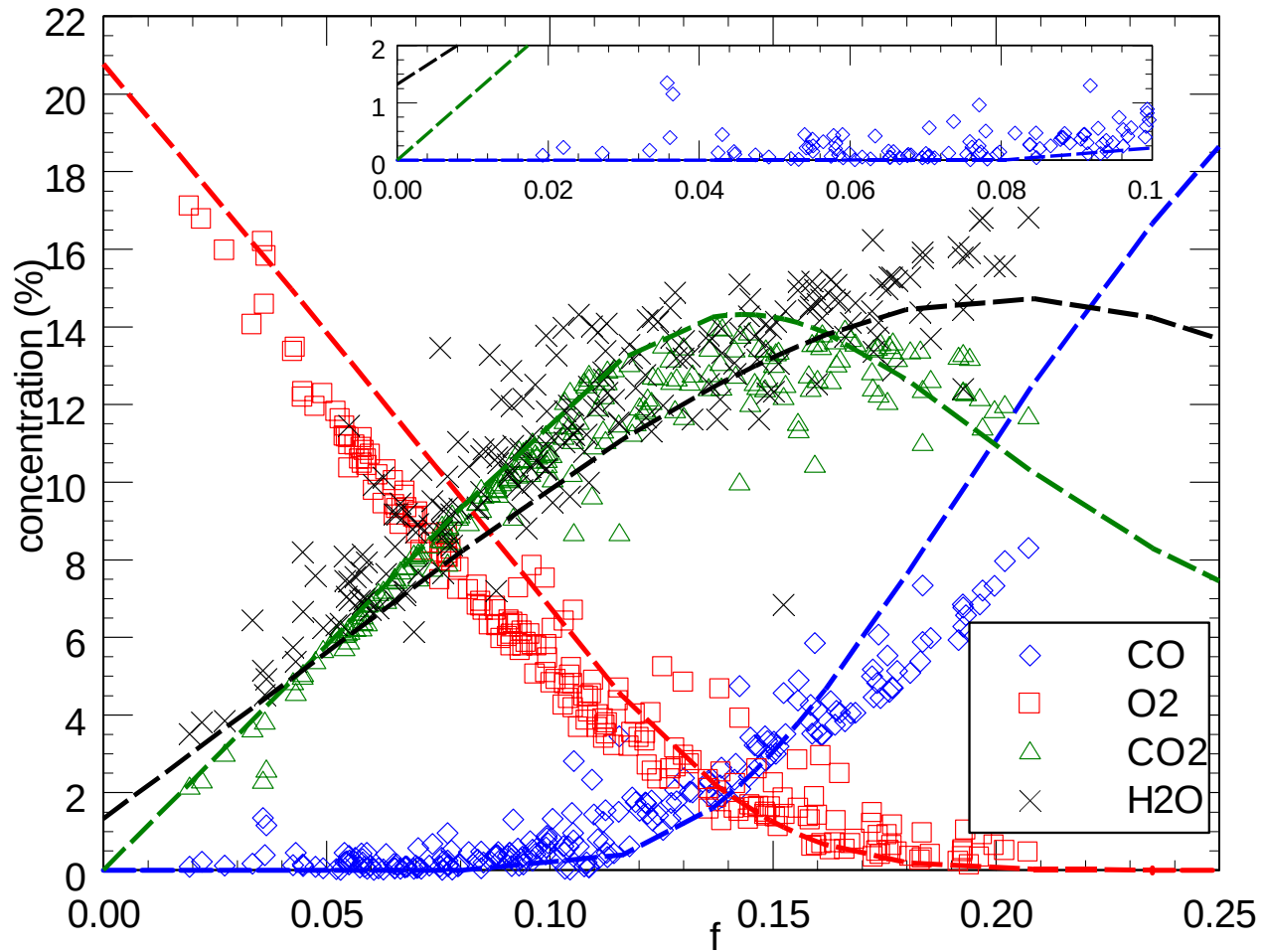
- Based on carbon species, estimate f



$$f = \frac{m_f}{m_{tot}} = \frac{MW_c \cdot \sum x_i \cdot \xi_i}{MW_{tot} \cdot y_c}$$

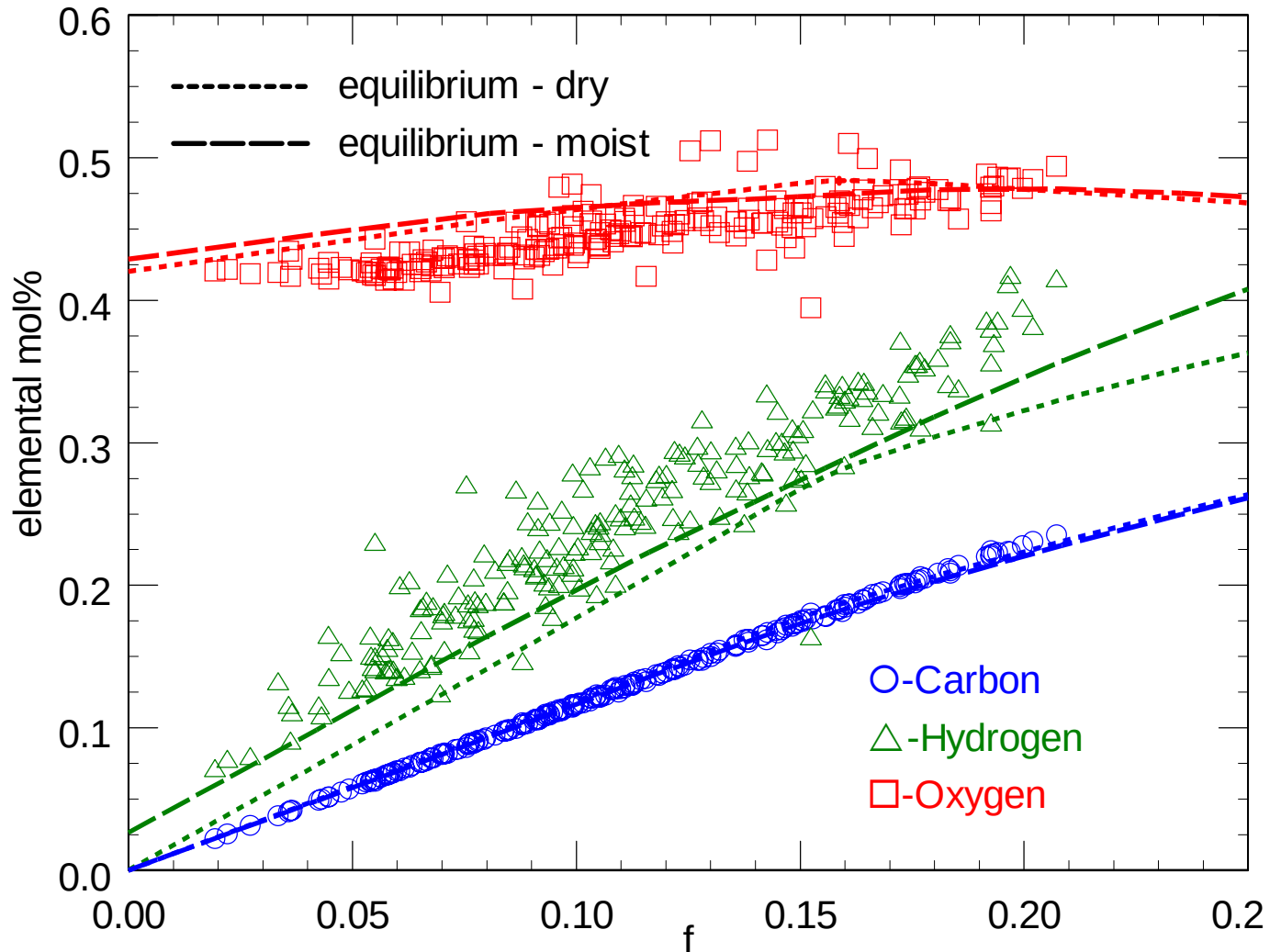
Data Regression

Data vs. Equilibrium

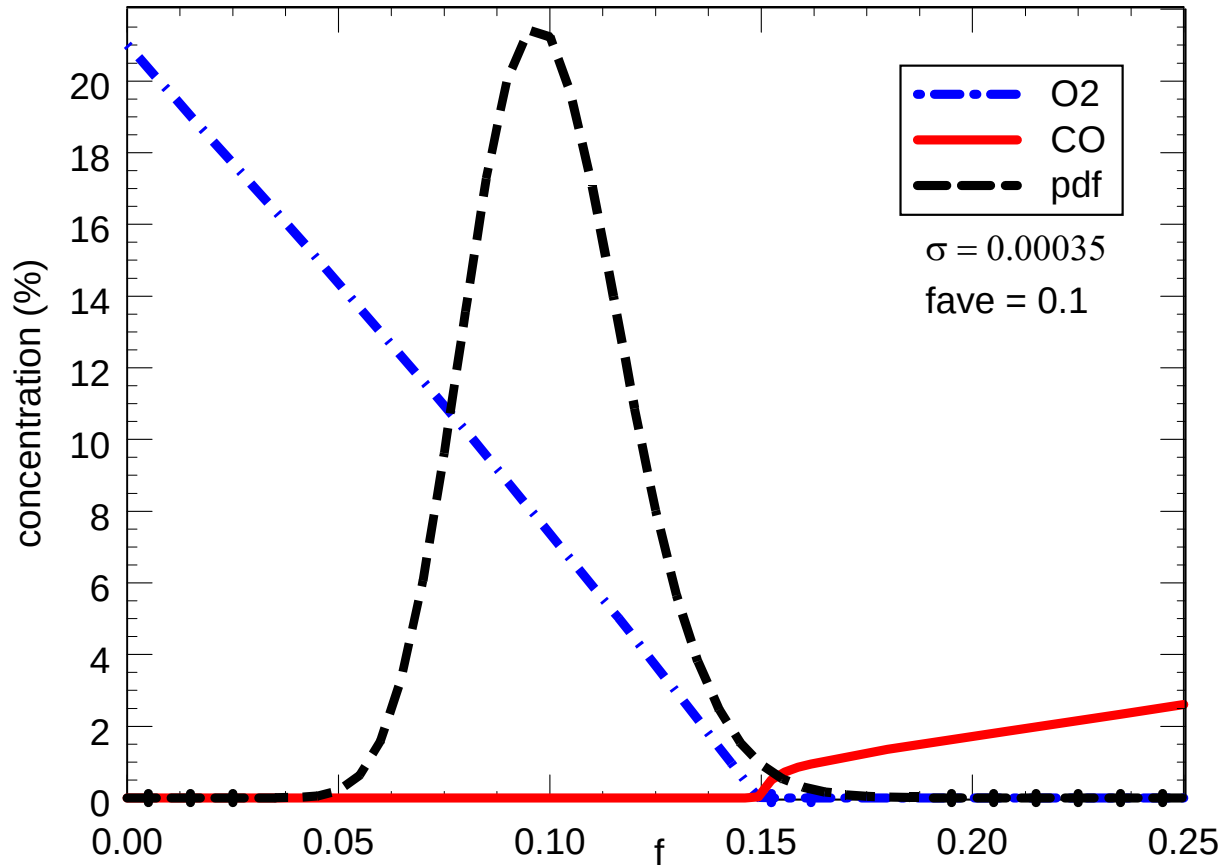


Data Regression

Element Balance

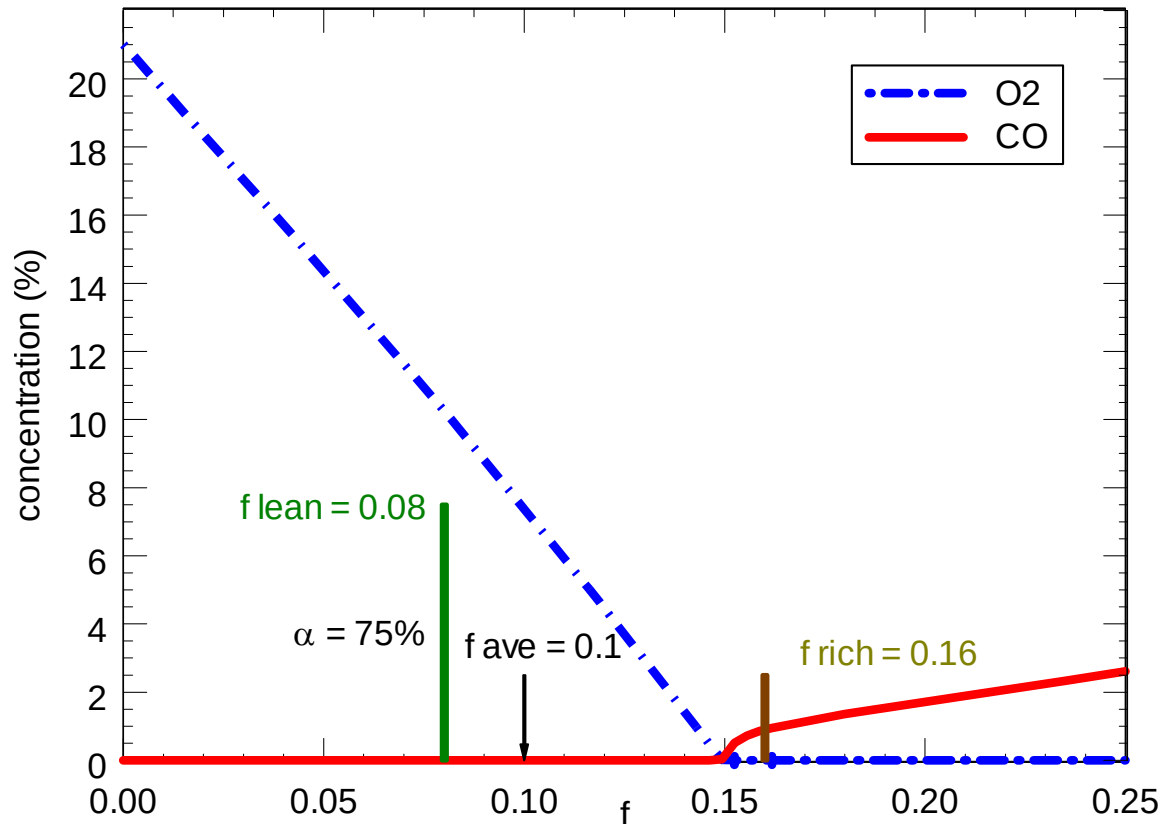


Gaussian pdf Model



$$\overline{x_i} = \int_0^1 pdf(f, \sigma) \cdot x_{i, equilibrium}(f) df$$

Dual Delta Function Model (DDF)

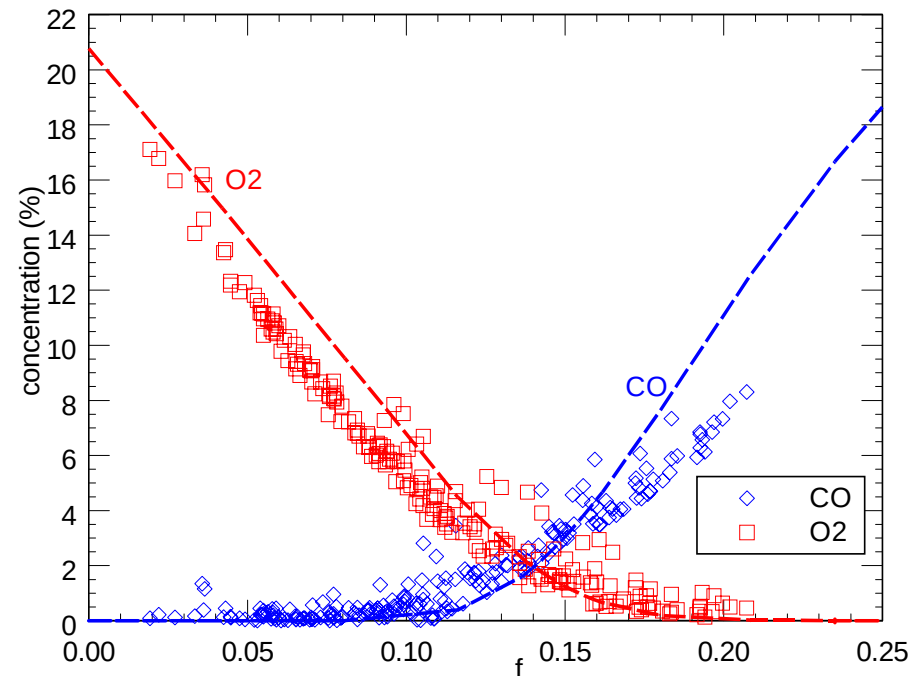


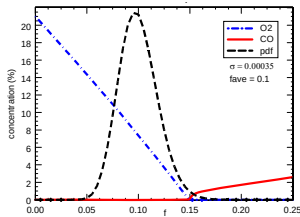
$$\bar{x}_i = \alpha \cdot x_{i,equl}(f_{lean}) + (1 - \alpha) \cdot x_{i,equl}(f_{rich})$$

$$\bar{f} = \alpha \cdot f_{lean} + (1 - \alpha) \cdot f_{rich}$$

Comparing Models

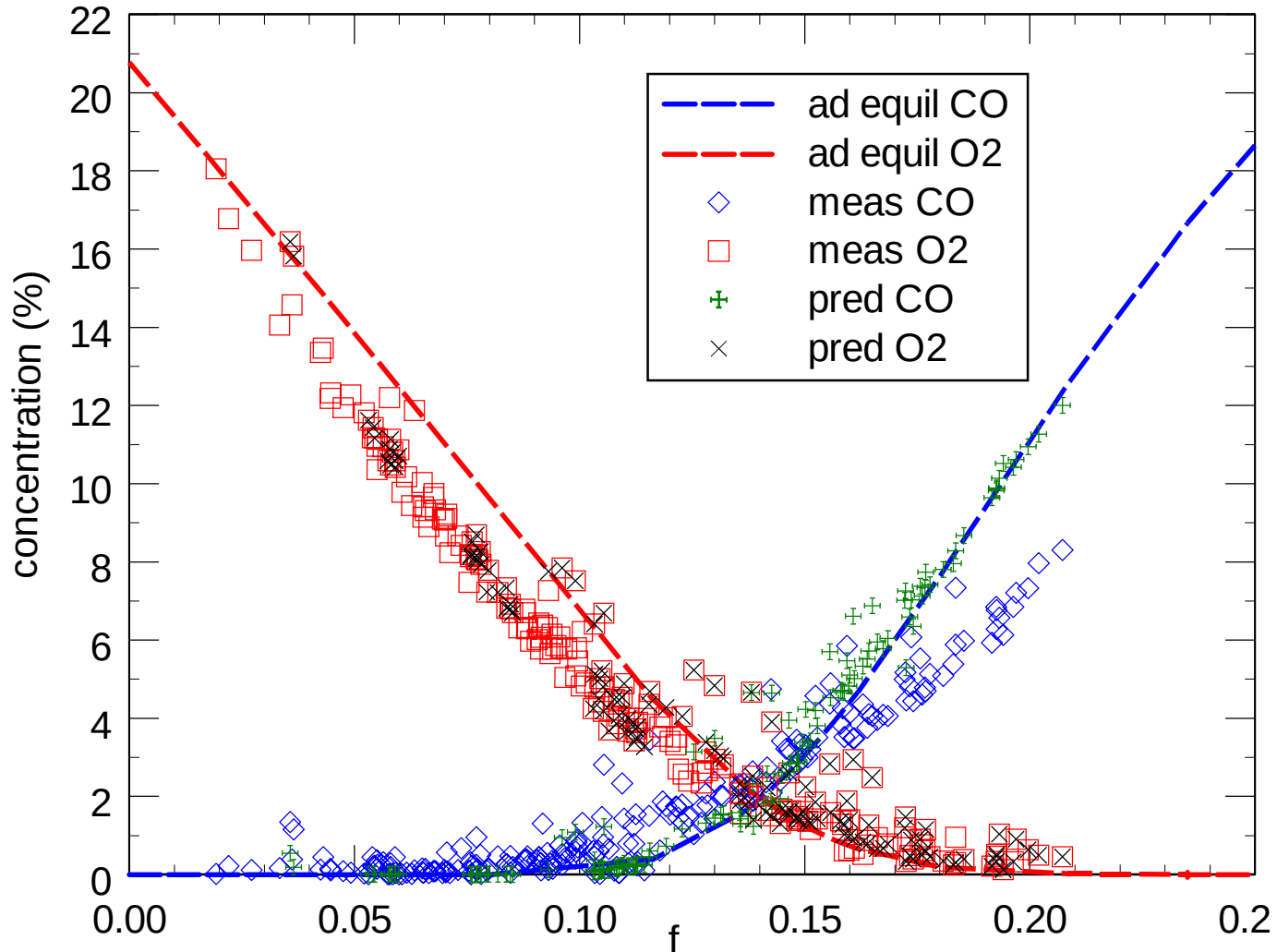
- Use both models to fit flame data by varying model parameters
 - Gaussian pdf: vary σ (get σ_{O_2} and σ_{CO})
 - DDF: vary f_{lean} and f_{rich}

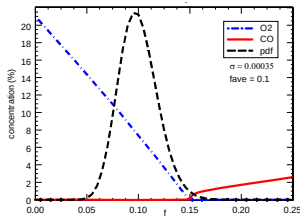




Gaussian pdf – O₂ fit

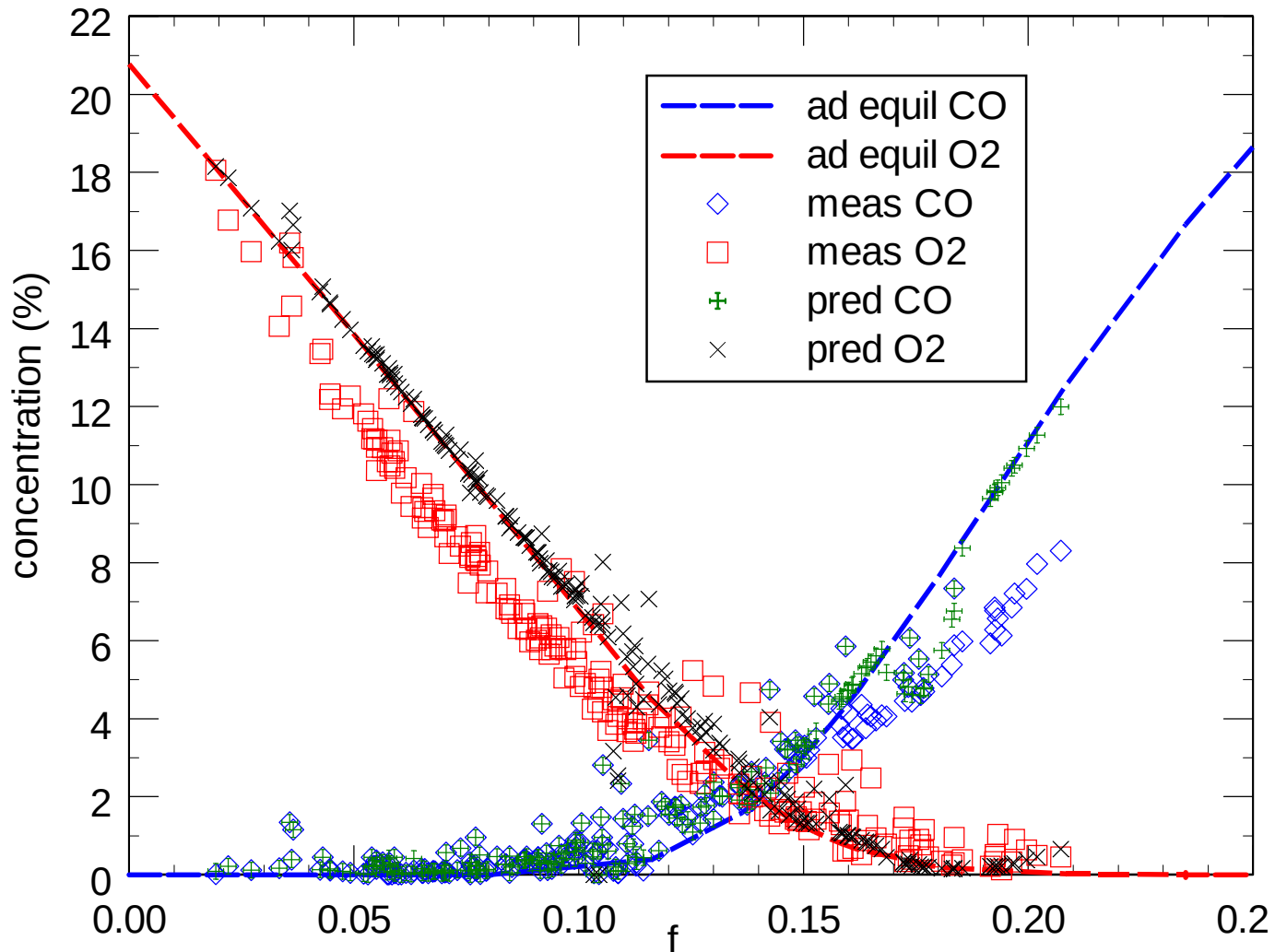
Gaussian Fit using O2

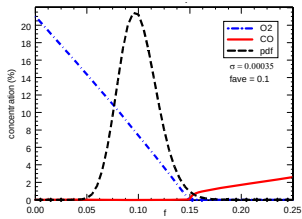




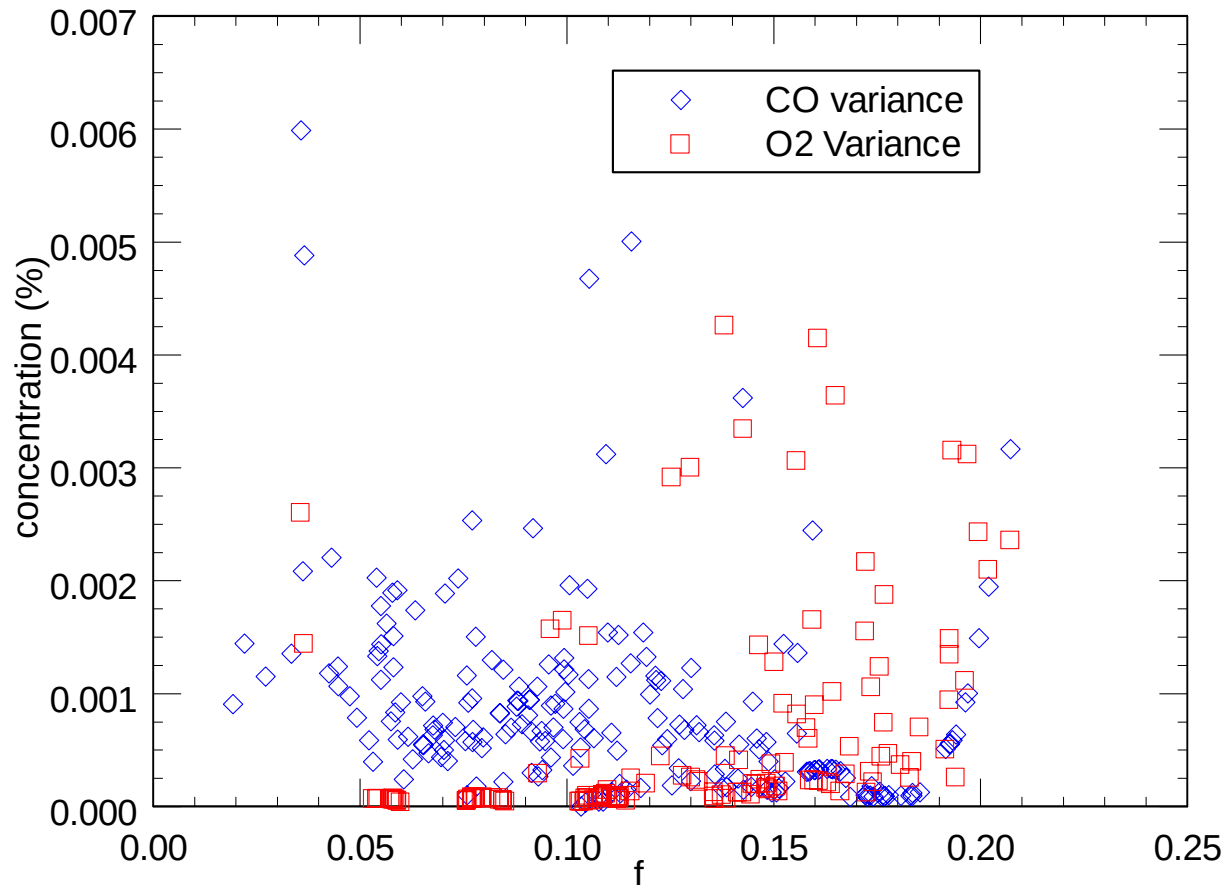
Gaussian pdf – CO fit

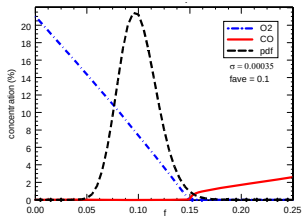
Gaussian Fit using CO



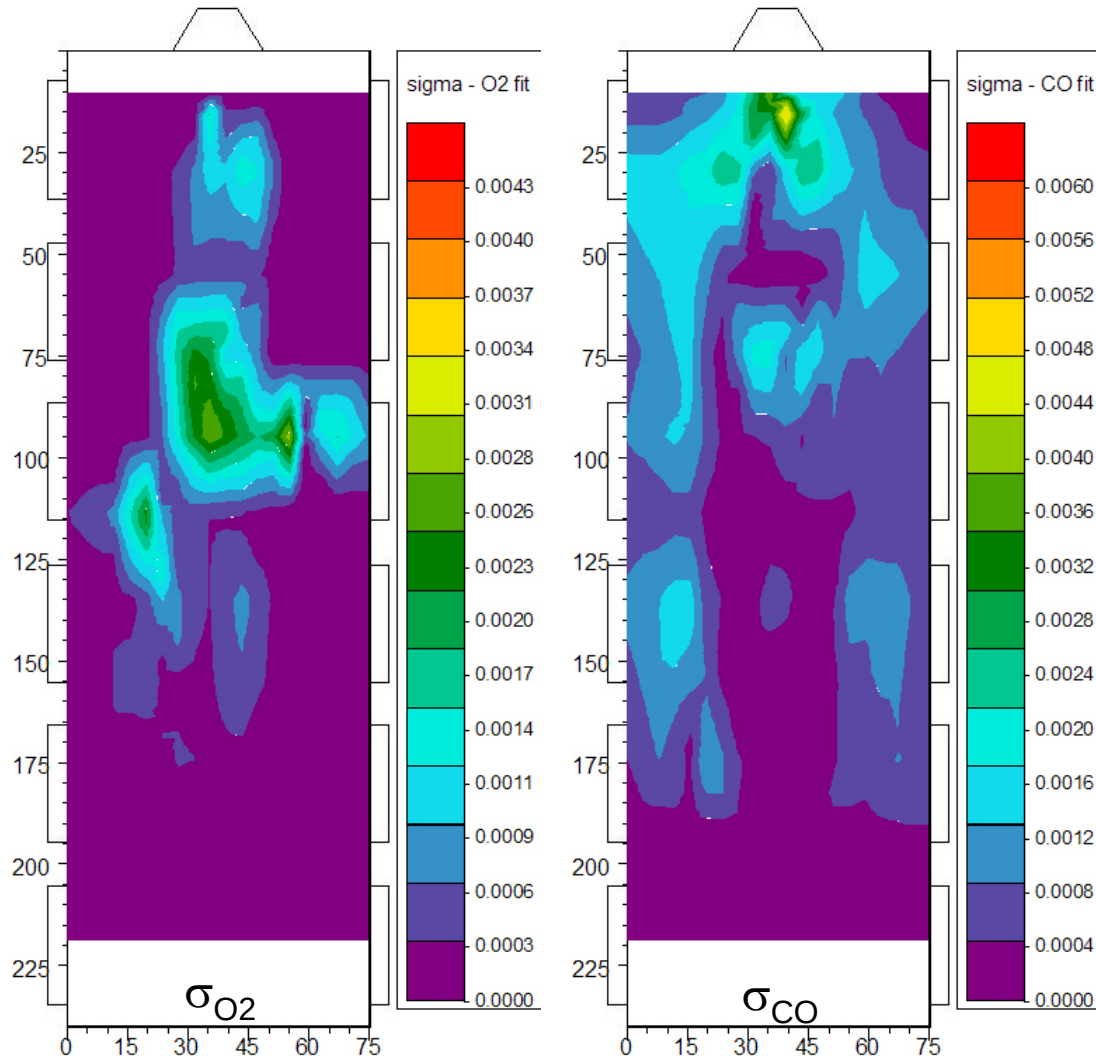


σ_{O_2} and σ_{CO}

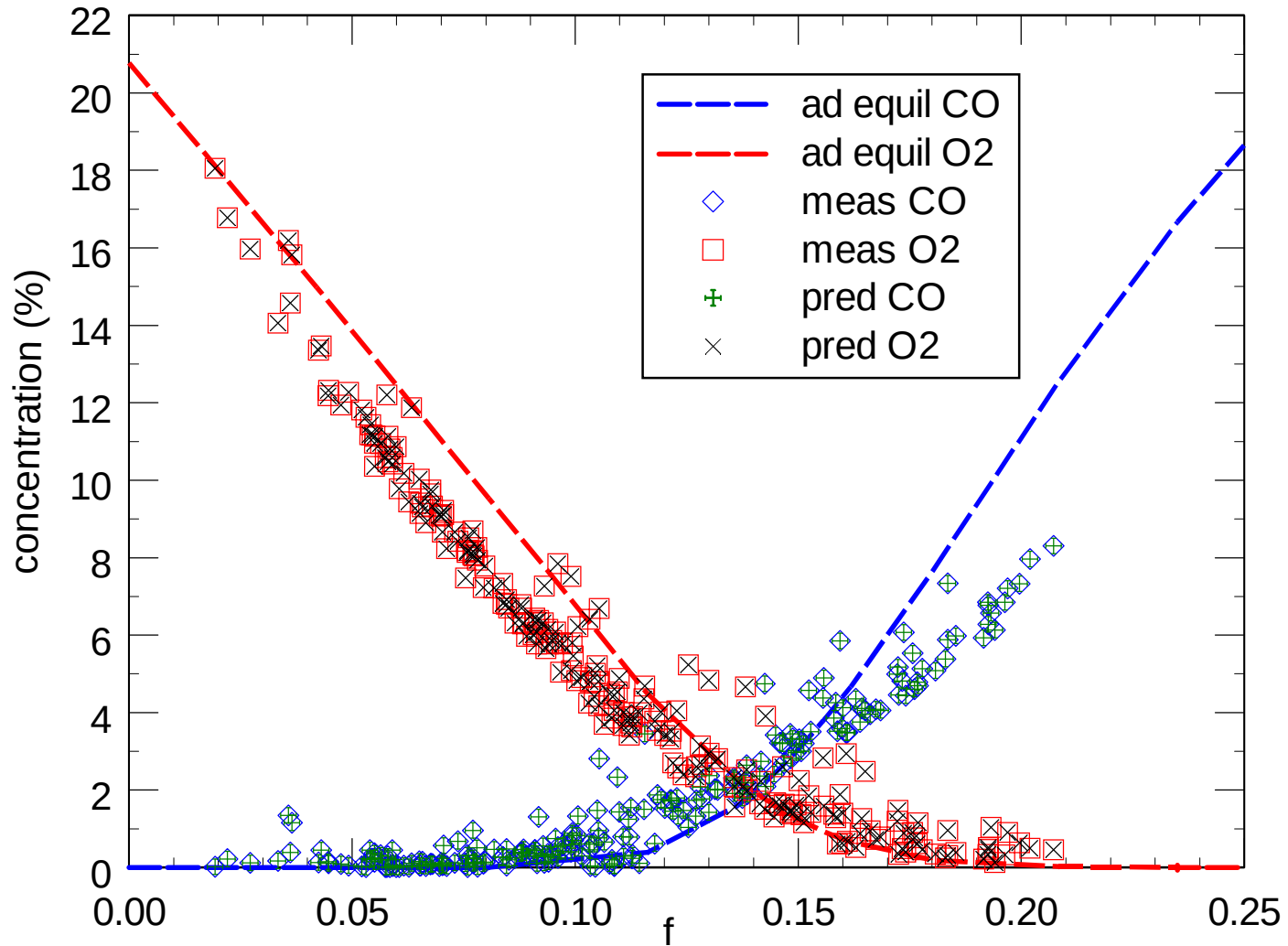
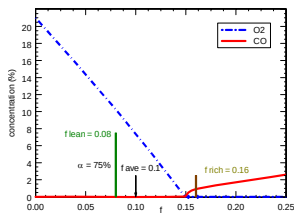


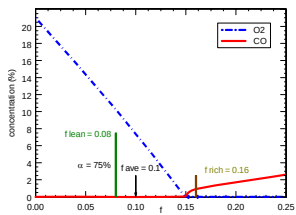


σ_{O_2} and σ_{CO}



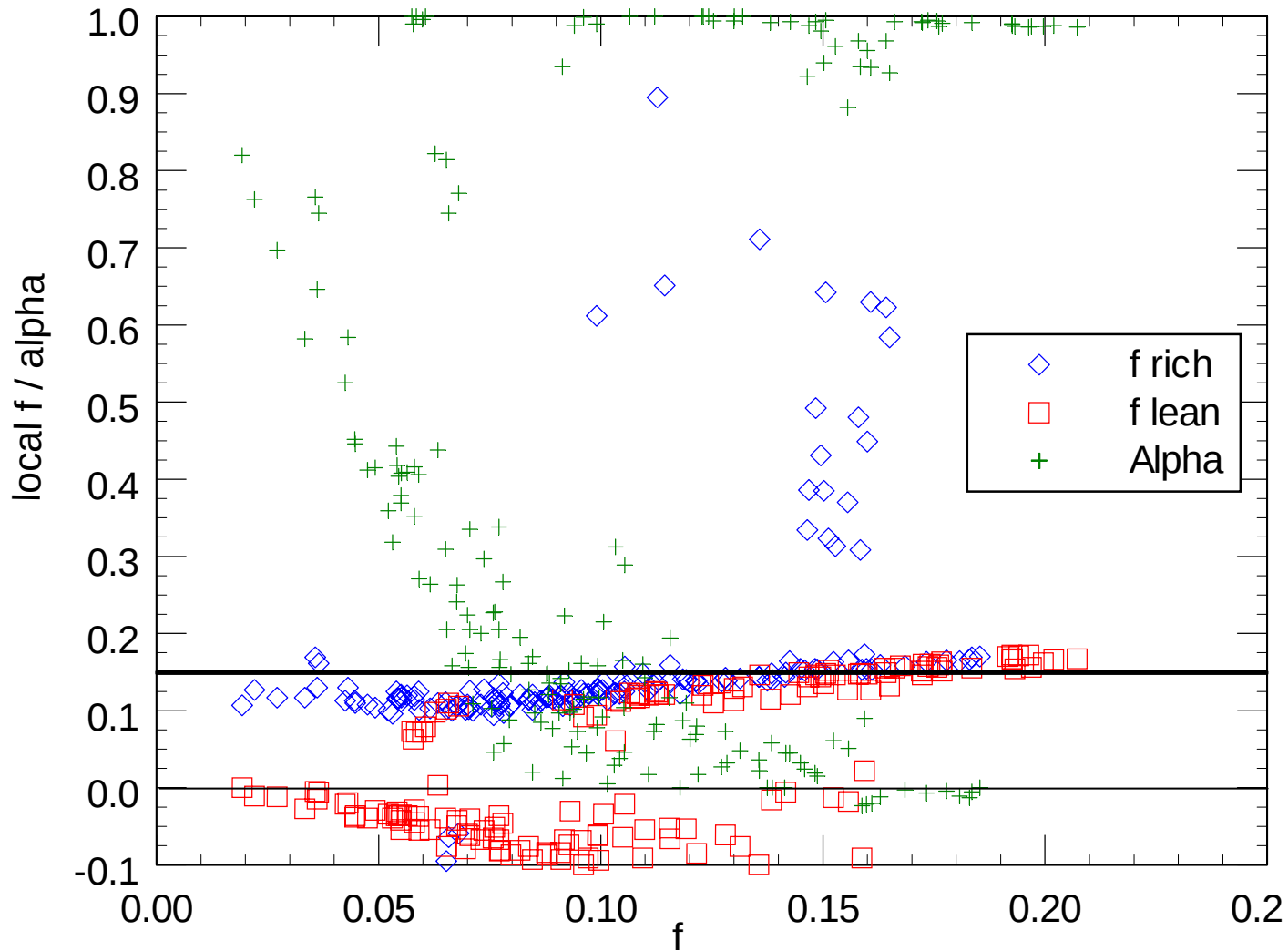
DDF fit





f_{lean} and f_{rich}

Parameters for DDF fit



Conclusions

- Intermittent flame behavior is prevalent in pulverized fuel flames
- The current modeling approach, Gaussian pdf, is unable to satisfactorily capture intermittent behavior
- An approach using dual delta functions captures intermittent phenomenon
 - Produces physically unrealistic f_{lean}
- Neither model shows great advantages over the other

Acknowledgements

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