

Joint Computational/Experimental Study of Flashback in Hydrogen-rich Gas Turbines

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Challenges in Flashback Simulation

- **Unique physics**

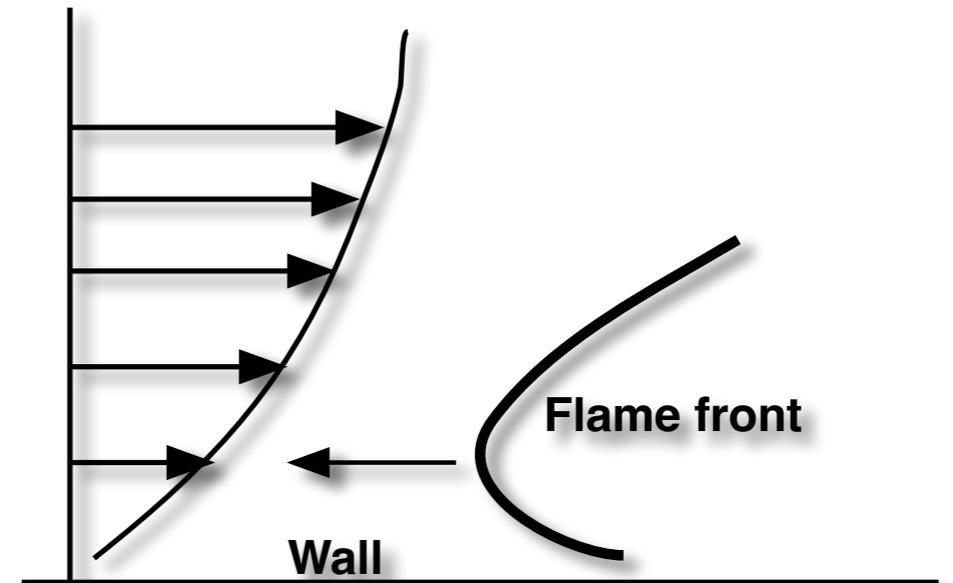
- ➔ Flame propagation in a turbulent boundary layer
- ➔ Non-uniform equivalence ratio

- **Coupled to upstream processes**

- ➔ Fuel-injection and mixing affects fuel stratification

- **Input uncertainties**

- ➔ Chemistry mechanism and boundary condition uncertainties could affect flashback speed + initiation

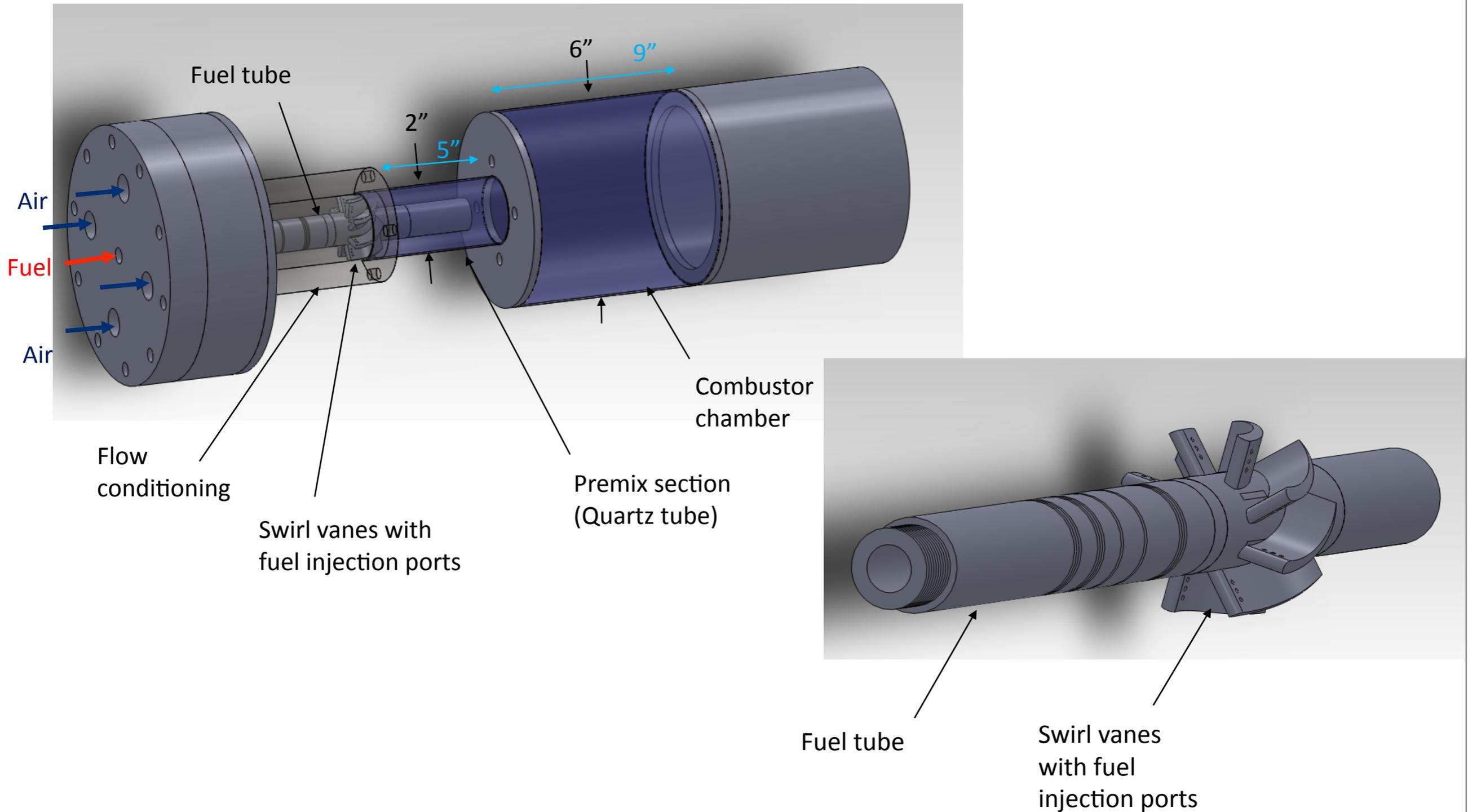


Operating Hypotheses

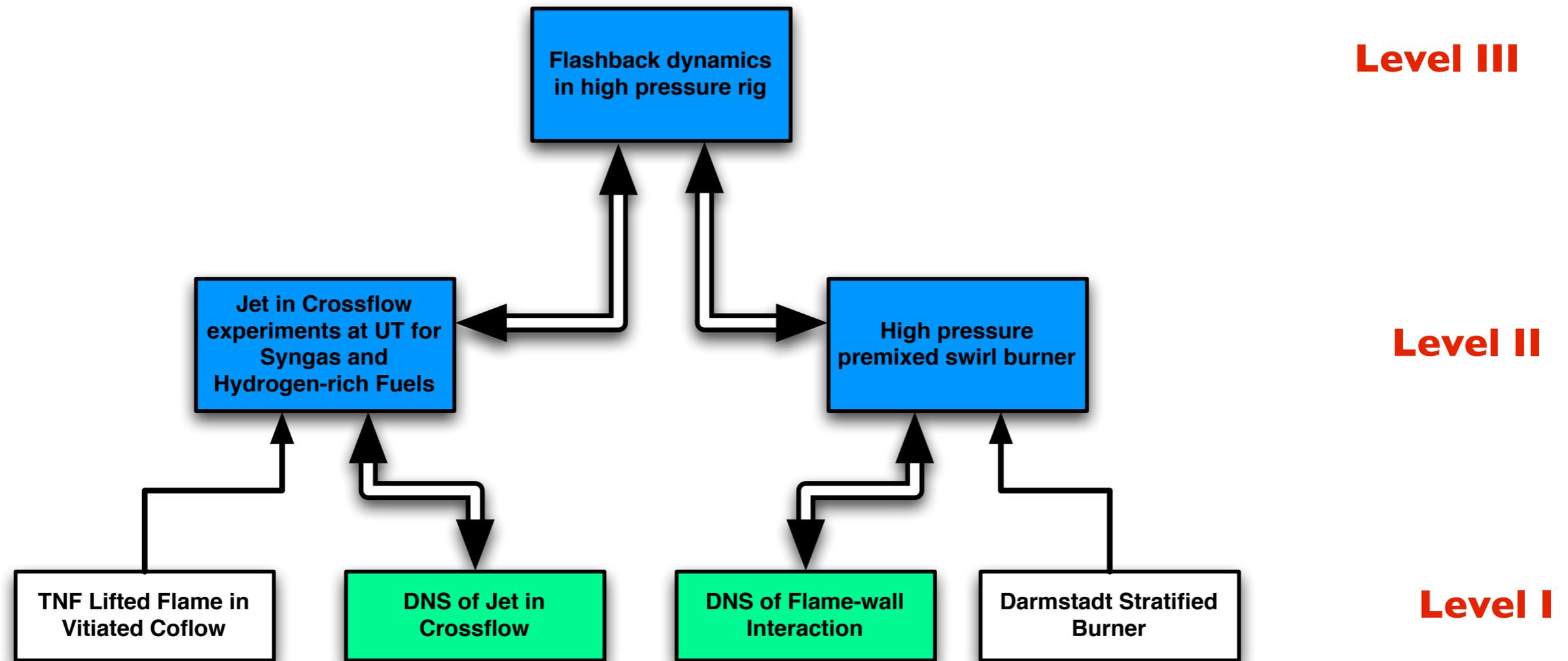
- **Both Large eddy simulation (LES) and Reynolds-averaged Navier-Stokes (RANS) approaches necessary**
- **Models developed should be applicable in complex geometries**
 - ➔ Restricts model formulation
- **Rigorous validation procedures necessary**
 - ➔ Individual processes (flame-wall interaction, jet mixing etc.) need to be validated
 - ➔ Interaction of processes also need to be tested
- **It is not sufficient to produce predictions, but also uncertainty in the predictions**
 - ➔ A statistical framework for uncertainty evaluation needed

Target-based Flashback Modeling

- UT high-pressure swirl combustor



Hierarchical Validation Pyramid



- **Level 1 - Fundamental data from legacy expts. and direct numerical simulations (DNS)**
- **Level 2 - UT re-configurable experiments designed for validation**
- **Level 3 - UT target system experiments**

Topics of Discussion

- **Quadrature approach (DQMOM) for modeling combustion of jets-in-crossflow**
 - ➔ Formulation of DQMOM approach
 - ➔ Implementation in OpenFOAM general purpose CFD solver
- **Analysis of flame-wall interaction using direct numerical simulation (DNS) data**
- **Bayesian approach for uncertainty quantification**
 - ➔ Formulation and UQ of syngas chemistry mechanisms
- **Preliminary studies of UT swirl burner**
 - ➔ Experimental and LES studies

**Direct Quadrature Method of Moments
(DQMOM)
Approach for LES/RANS of Turbulent
Combustion**

DQMOM Basics

- **Probability density function (PDF) approach**

→ Solve a high-dimensional transport equation for joint-PDF of gas phase scalars

- **In LES calculations, the filtered moments of the composition vector are required**

$$\tilde{\phi} = \int \mathcal{G}(\zeta) P_{\xi}(\zeta; \mathbf{x}, t) d\zeta$$

- **PDF transport equation**

$$\frac{\partial P}{\partial t} + \frac{\partial}{\partial x_j} \left[P \widetilde{u_j | \zeta} \right] = - \underbrace{\frac{\partial}{\partial \zeta_{\alpha}} \left[P \widetilde{\mathcal{M}_{\alpha} | \zeta} \right]}_{\text{Conditional Diffusion}} - \frac{\partial}{\partial \zeta_{\alpha}} \left[P S_{\alpha} \right]_{\text{Chemical Source}}$$

→ Condition diffusion requires a model for scalar dissipation rate

Modeling PDF Transport Equation

- PDF transport equation

$$\frac{\partial P}{\partial t} + \frac{\partial}{\partial x_j} \left[P \widetilde{u_j} | \zeta \right] = - \frac{\partial}{\partial \zeta_\alpha} \left[P \widetilde{\mathcal{M}_\alpha} | \zeta \right] - \frac{\partial}{\partial \zeta_\alpha} [P S_\alpha]$$

Conditional Diffusion

Chemical Source

- PDF equation is high-dimensional

→ If N species present in chemistry, N+5 dimensions

- Lagrangian Monte-Carlo approach typically used

→ Stochastic in nature

→ Numerical stability is highly flow dependent

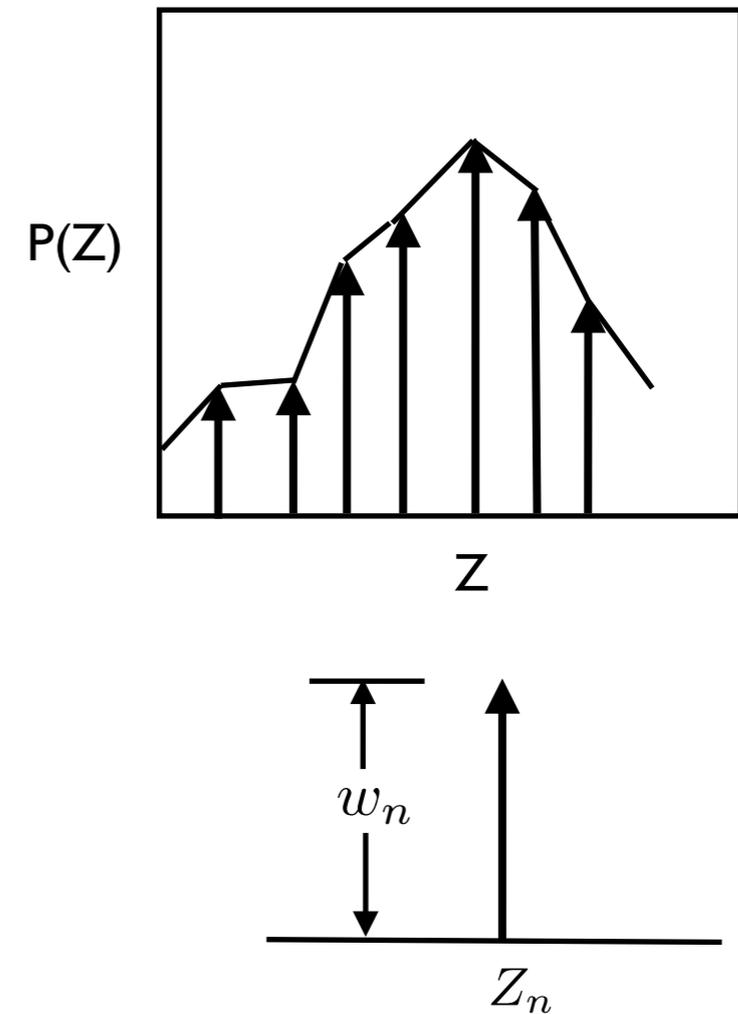
→ Difficult to maintain numerical accuracy in complex geometries

→ Highly expensive for realistic flow configurations

Eulerian DQMOM Approach

- **DQMOM uses dirac-delta functions to discretize the PDF**
- **Each delta-function characterized by a weight and abscissa**
 - ➔ Transport equations for these two variables can be formulated
- **Similar in structure to scalar transport equations**

$$\frac{\partial w_n}{\partial t} + \overline{U}_i \frac{\partial w_n}{\partial x_i} = \frac{\partial}{\partial x_i} \left(\Gamma \frac{\partial w_n}{\partial x_i} \right) + a_n$$



Implementation in OpenFOAM

- **Eulerian PDF approach**

- ➔ Easy transition to commercial and open source codes

- **OpenFOAM open source platform**

- ➔ C++ libraries for solving partial differential equations

- ➔ Arbitrary geometries handled using unstructured grids

- ➔ MPI-based parallelization

```
solve
(
    fvm::ddt(rho, G1[i])
    + mvConvection->fvmDiv(phi, G1[i])
    - fvm::laplacian(turbulence->muEff(), G1[i])
    ==
    rho*mixingSource[i]
    + chemistry1.RR(i)
);
```

Abscissas

```
solve
(
    fvm::ddt(rho, w1)
    + mvConvection->fvmDiv(phi, w1)
    - fvm::laplacian(turbulence->muEff(), w1)
);
```

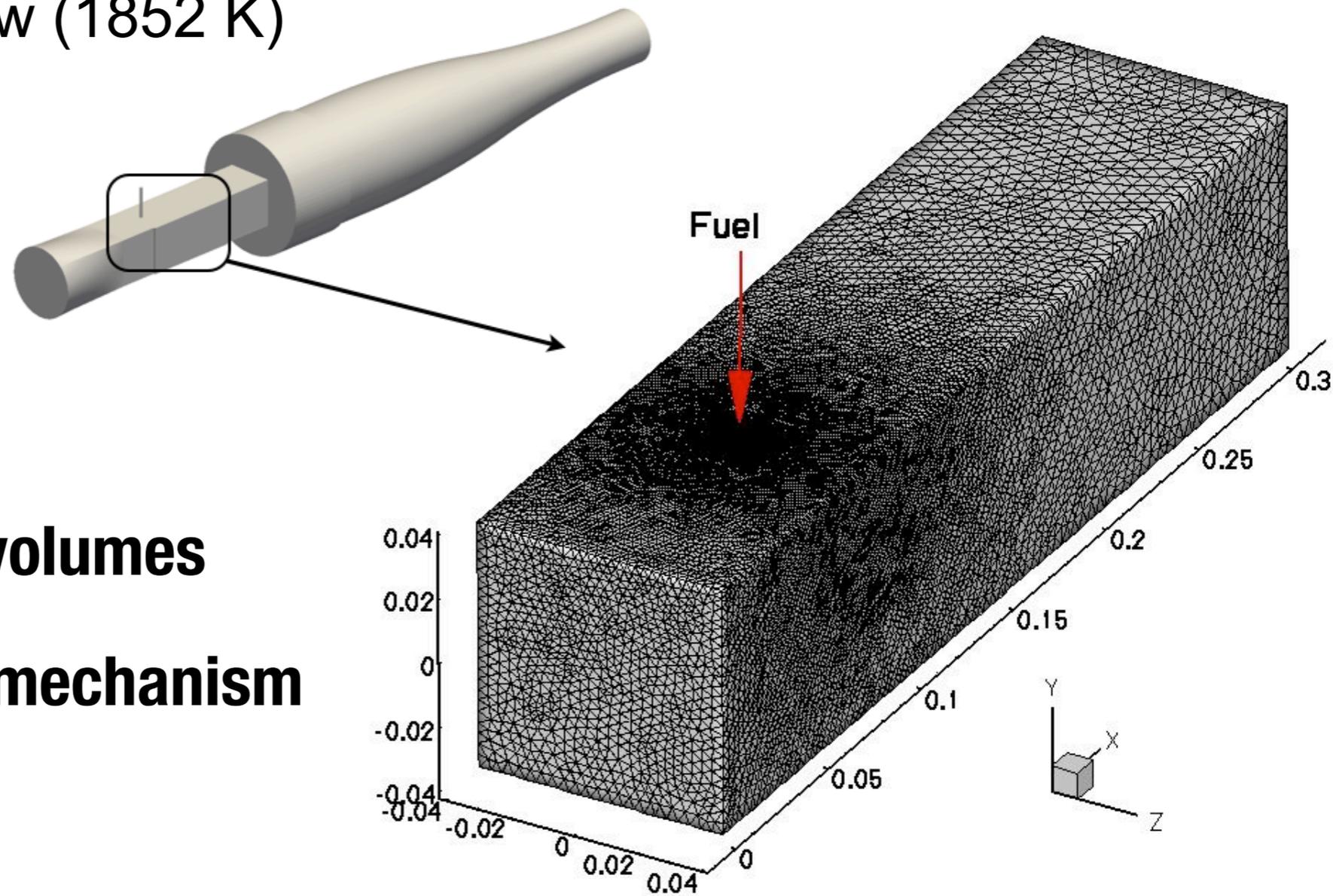
Weights

Simulation of Jet-in-Crossflow Configuration

- **Experiment from Lieuwen's group (GaTech Univ.)**

- ➔ Methane/hydrogen mixture (300K)

- ➔ Vitiated air crossflow (1852 K)



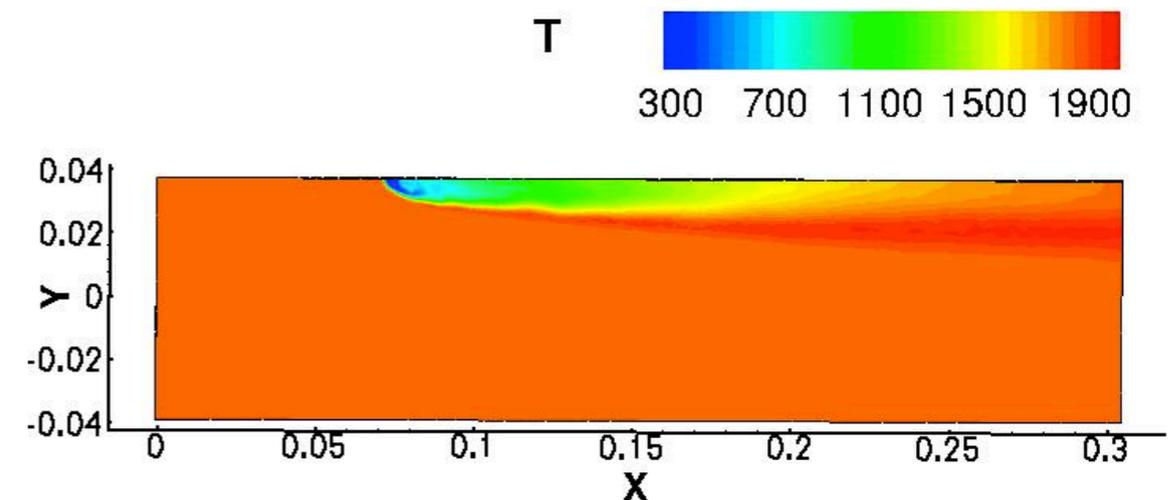
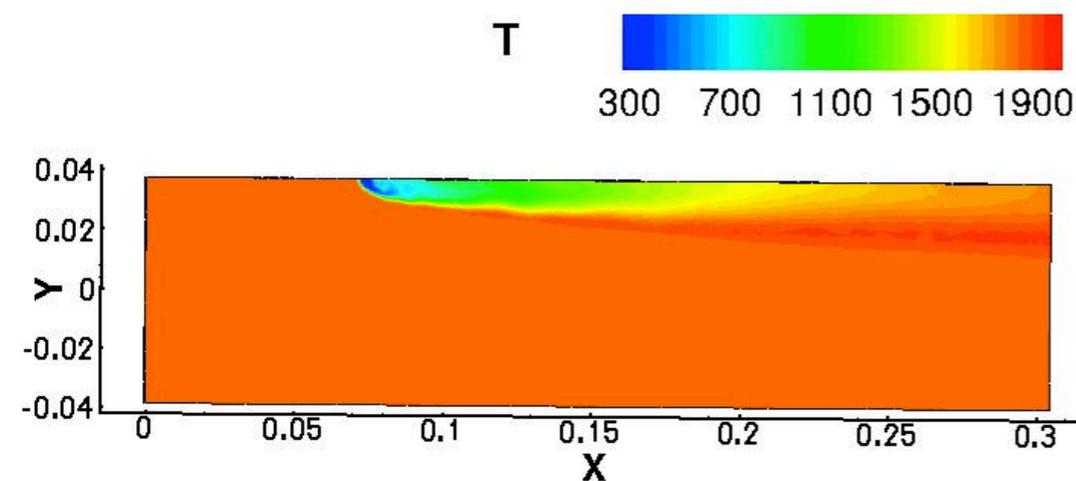
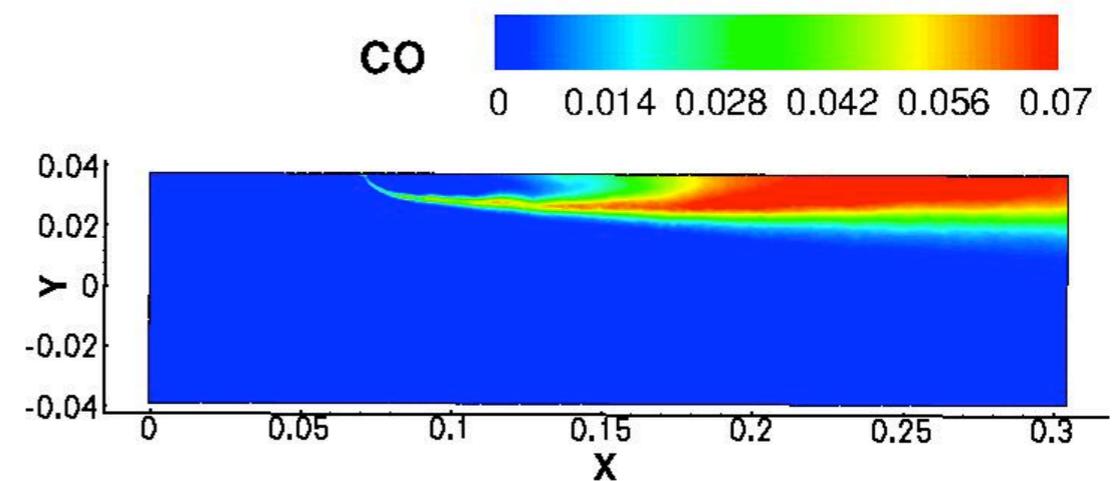
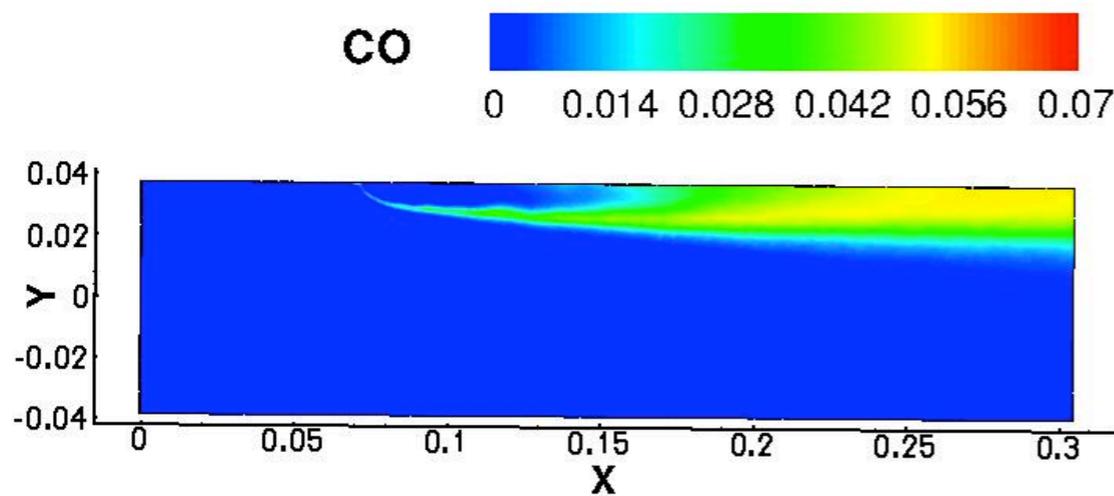
- **Simulation details**

- **1.7 million control volumes**

- **7-species Linstedt mechanism**

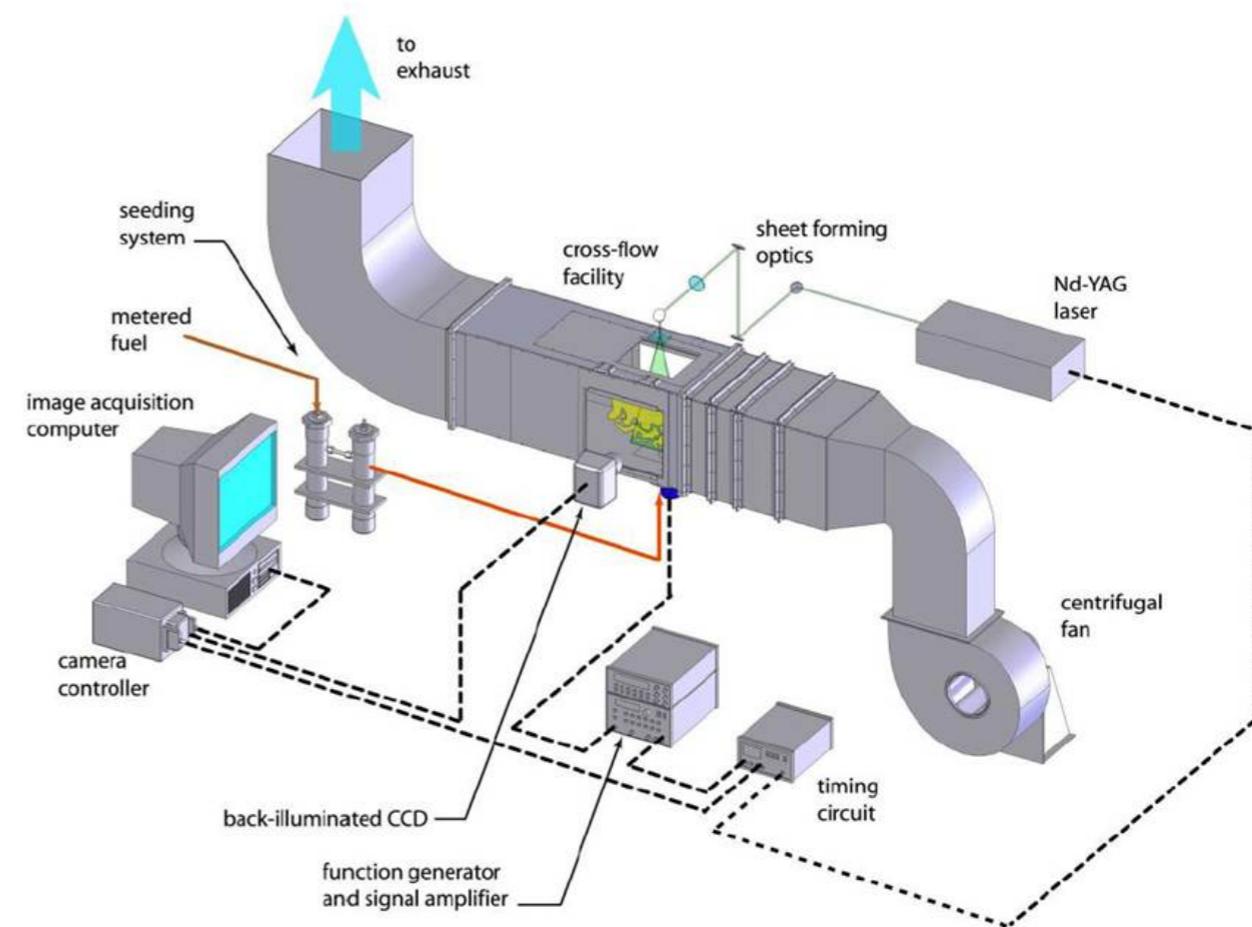
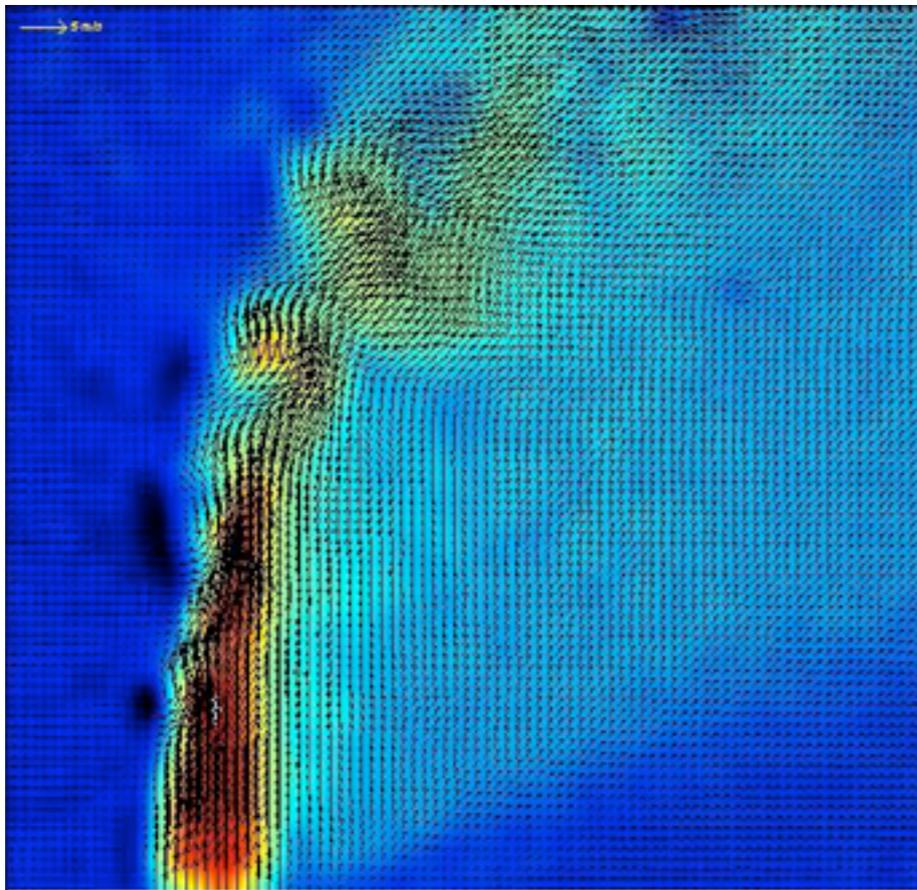
Species Distribution

- **DQMOM results compared with no-combustion model simulation**
 - ➔ DQMOM predicts lower combustion rates
 - ➔ Finite-rate mixing slows reactions



UT JICF Configurations

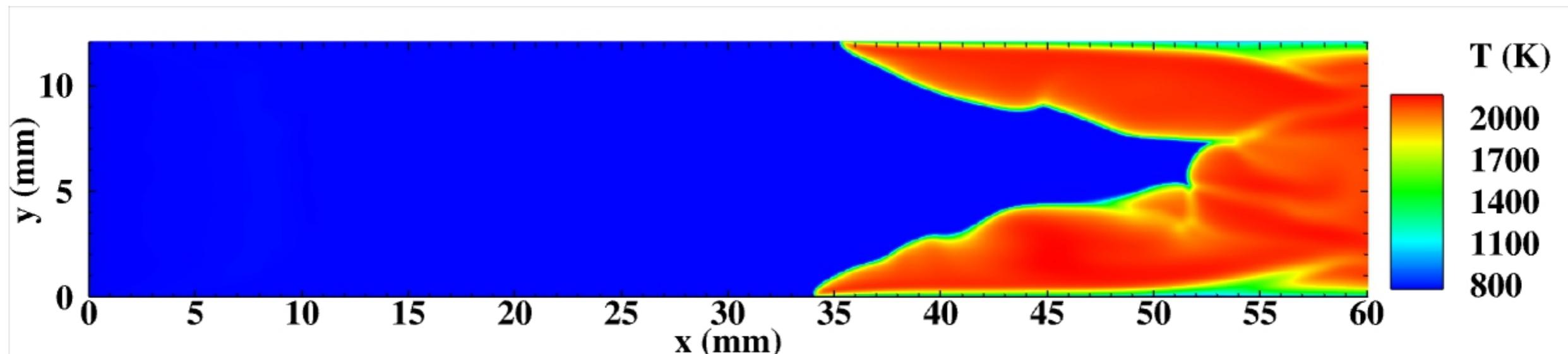
- Objective is to develop joint velocity/scalar statistics for methane/hydrogen fuel mixtures
- 3-component PIV measurements completed



DNS-based Analysis of Flame Propagation Through Turbulent Boundary Layers

Flame Propagation

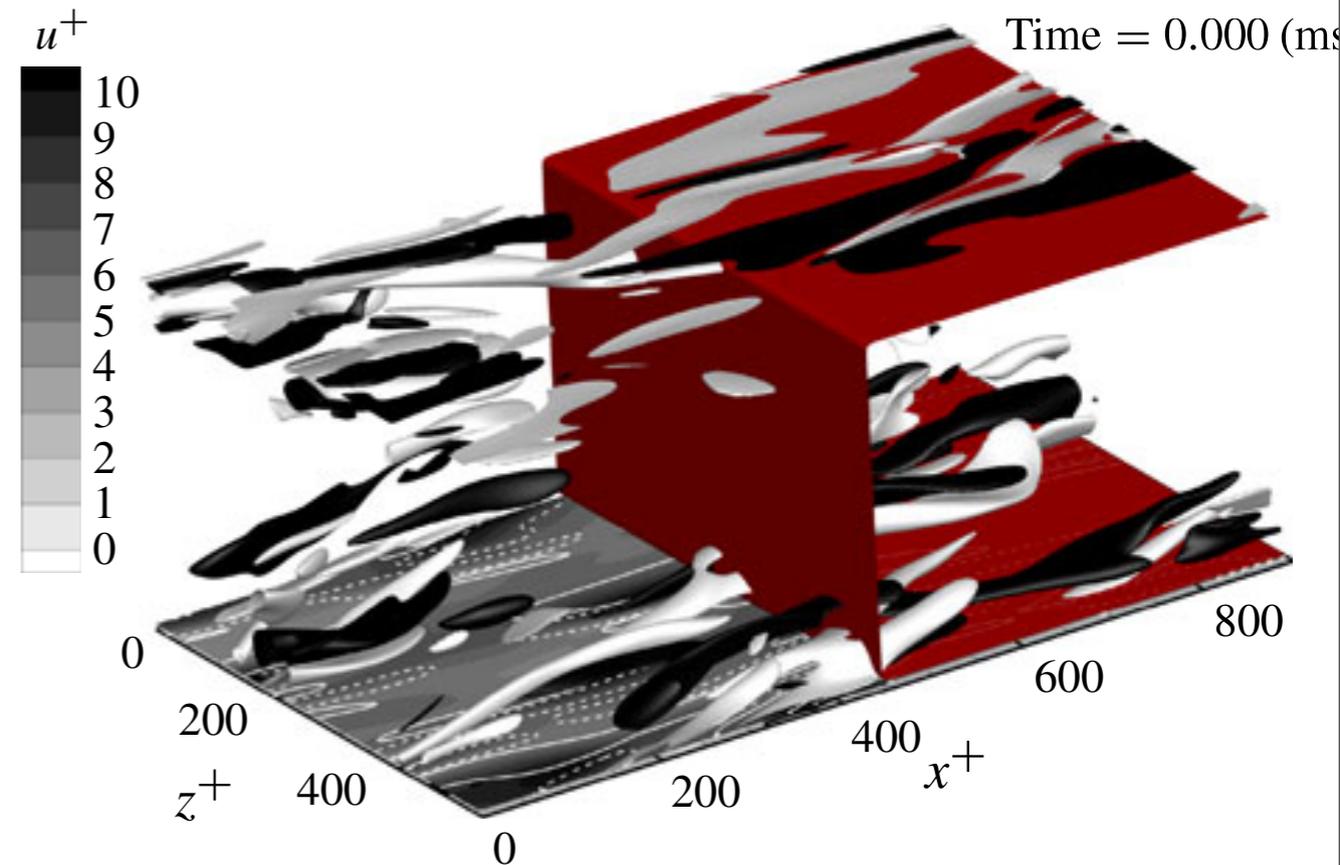
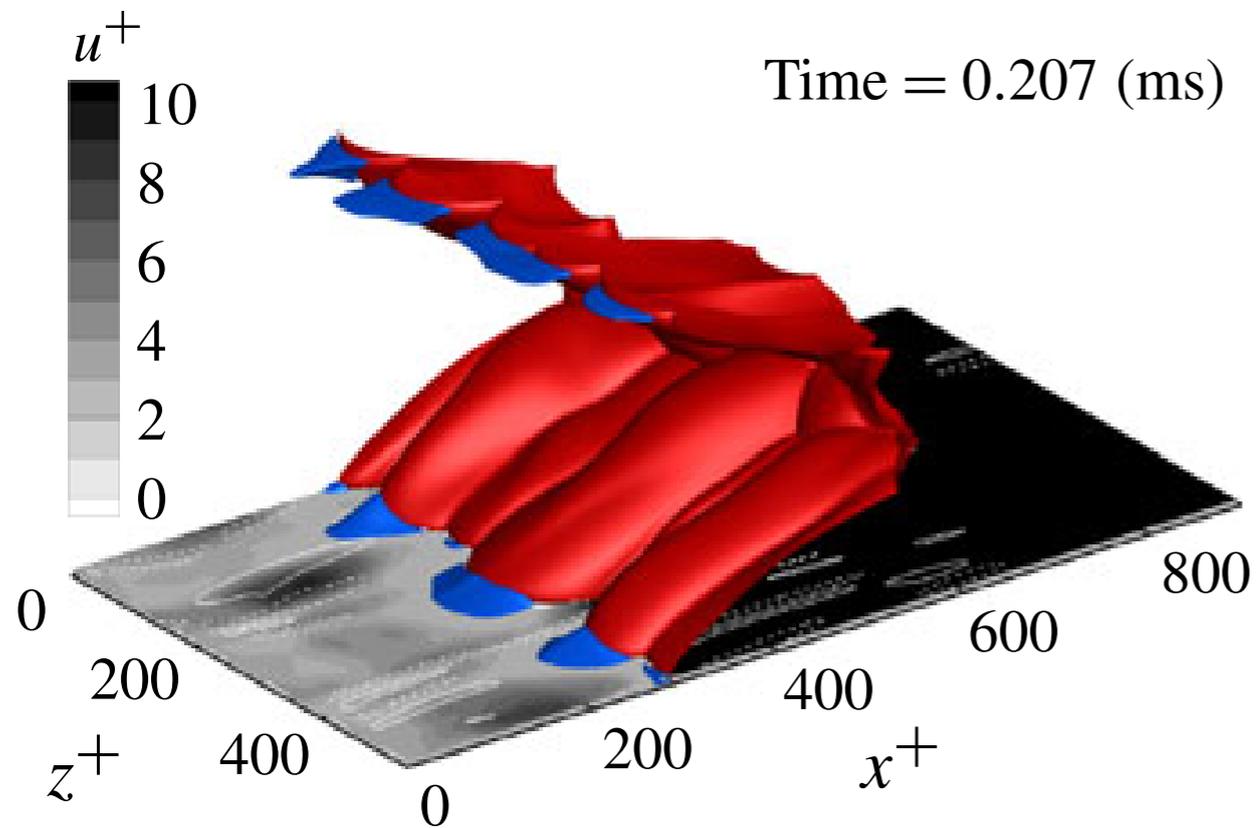
- Premixed flame propagation generally studied in free-stream or shear layer turbulence
- Wall-bounded flows
 - ➔ Exhibit significant flow anisotropy
 - ➔ Turbulence modified through flame propagation
 - Similar to shock-turbulence interaction



Sandia Flashback DNS

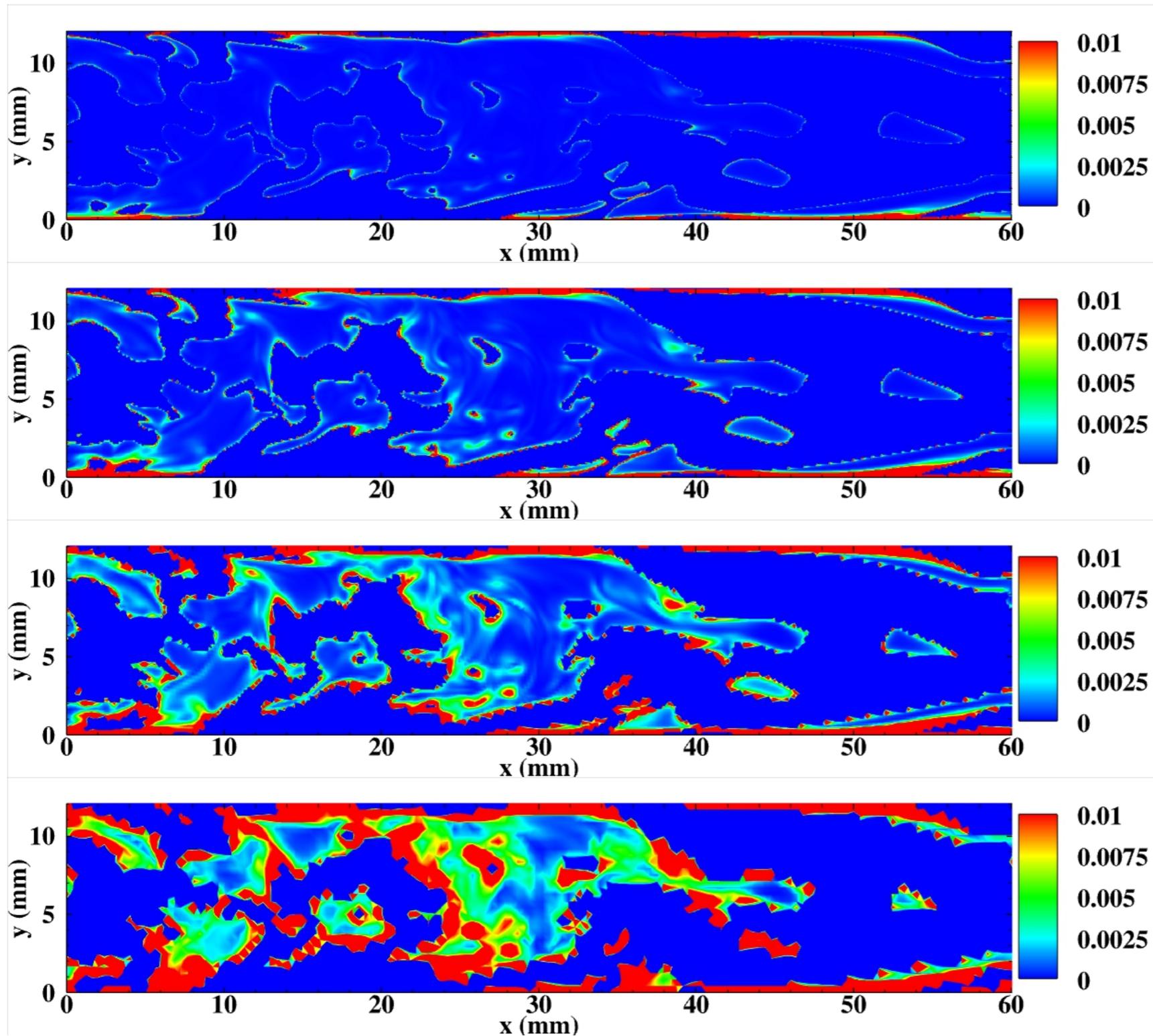
- **Petascale simulation of flame flashback in a turbulent channel flow**

➔ Gruber et al., JFM, 2012



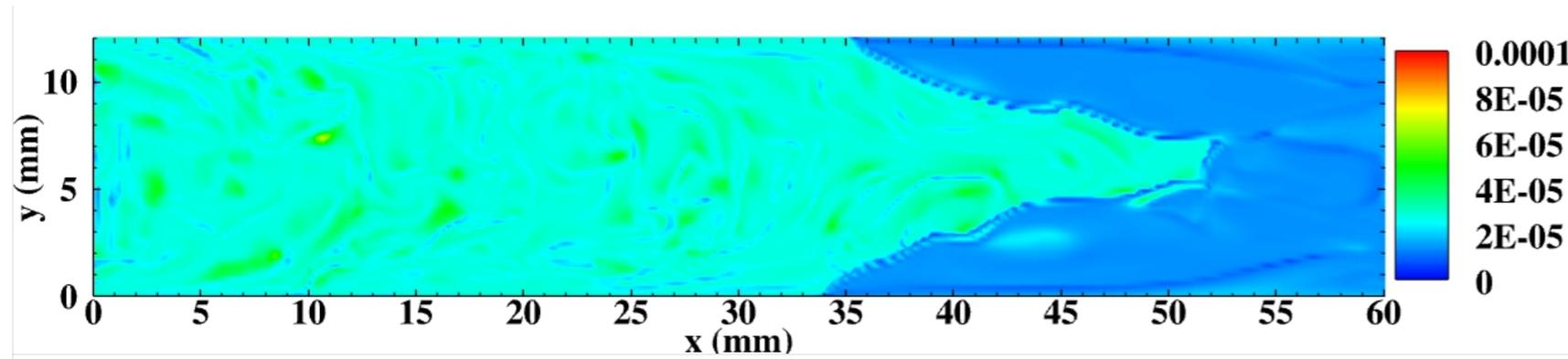
A Priori Results

- DNS-based analysis of unresolved kinetic energy



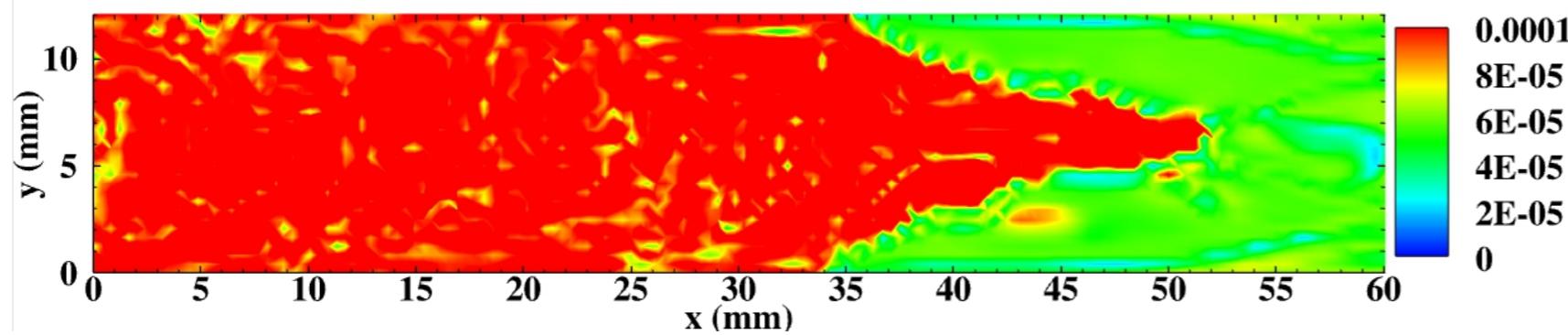
**Decreasing LES
resolution**

Mixing Timescale and Kinetic Energy Production



High

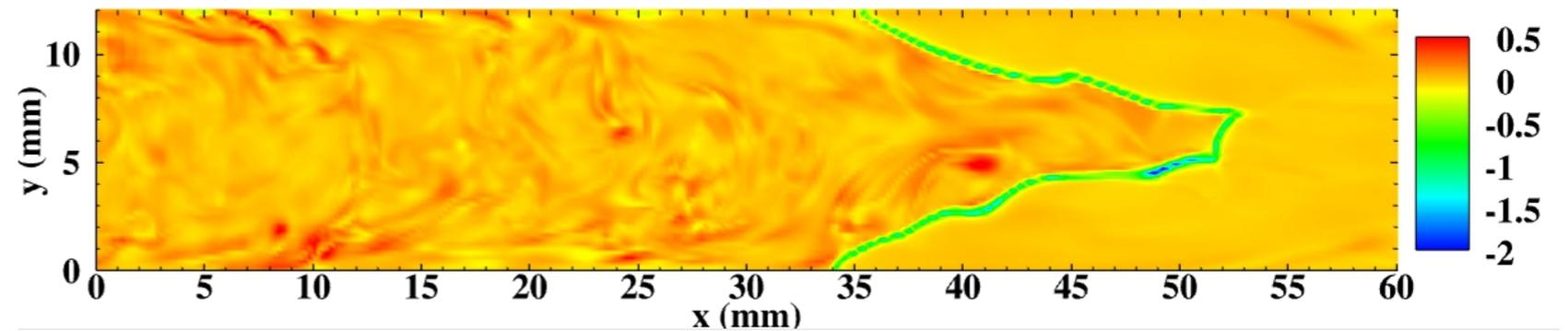
k_{sgs}/ϵ



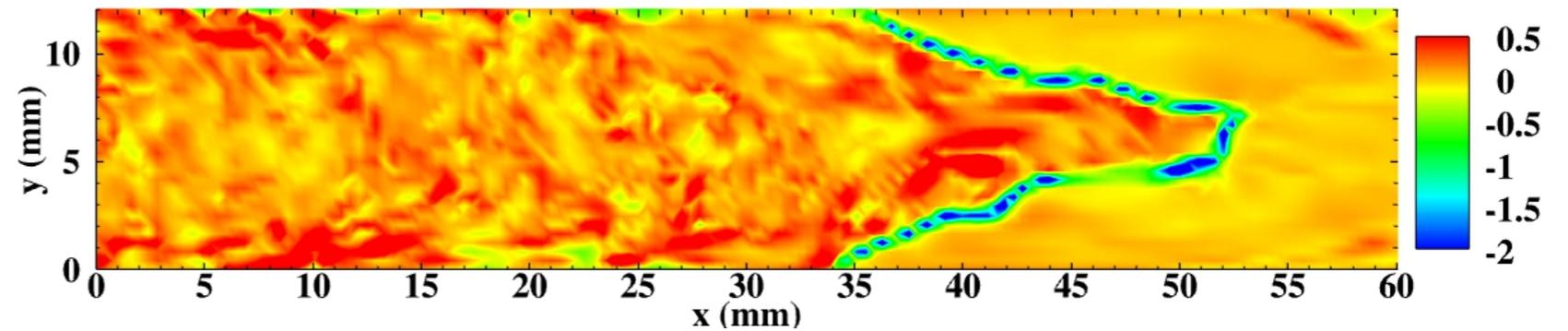
Low

High

P/ϵ



Low



Uncertainty Quantification in Gas Turbine Simulations

Application to Chemistry Model

CFD and Predictions

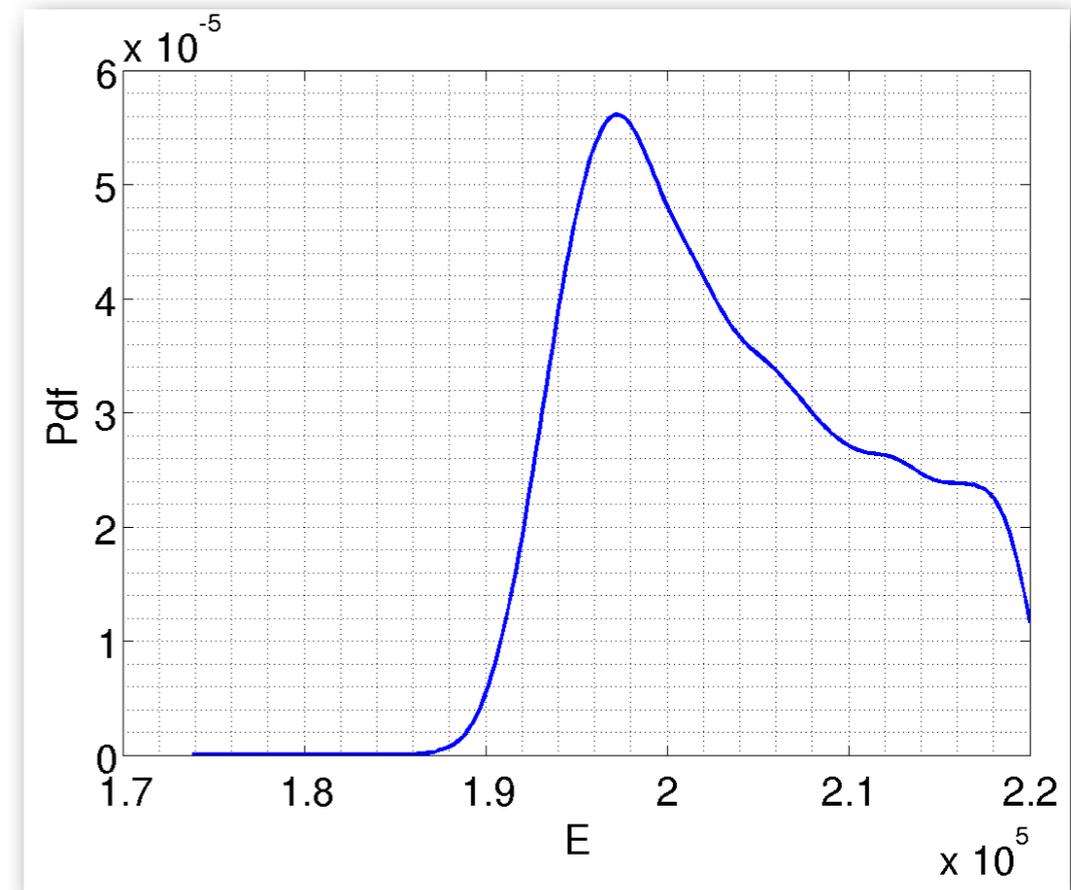
- **The goal of CFD is to issue predictions at future conditions**
 - ➔ No corroborating data exist
- **CFD models are highly imperfect**
 - ➔ Varying degree of error
- **How reliable are the predictions?**
 - ➔ Measure of prediction error is necessary
 - ➔ Termed as prediction uncertainty
 - ➔ Field of uncertainty quantification (UQ)

Uncertainty Quantification (UQ) Basics

- **Models are necessarily imperfect**
- **Measurements designed to calibrate models also contain errors**
- **Two forms of errors**
 - ➔ Model form error arising from specific model formulation
 - Example: Specific set of reactions to describe fuel combustion
 - ➔ Parametric uncertainty
 - Reaction rates cannot be determined to arbitrary precision
- **UQ theory**
 - ➔ Uses experiments (data) to develop a probabilistic estimate of the parameters and model form errors

Bayesian Theory of UQ

- **Based on assigning probabilistic values for parameters**
 - ➔ No single value but a likely range of values
- **Uncertainty in knowledge expressed through PDFs**
 - ➔ For instance, PDF of parameters
- **Use Bayes' theorem to utilize data for finding these PDFs**
 - ➔ As more data is available, PDFs change
 - ➔ Reflects a change in our knowledge of the system



PDF of activation energy obtained by using many different experiments

Bayesian Learning or Bayesian Inference

- The basis of Bayesian inference is the Bayes theorem
- Consider two events A and B
- Bayes' theorem relates the conditional PDFs of the two events to the marginal PDFs

$$P(A|B) = \frac{P(B|A) P(A)}{P(B)}$$

Likelihood (points to $P(B|A)$)

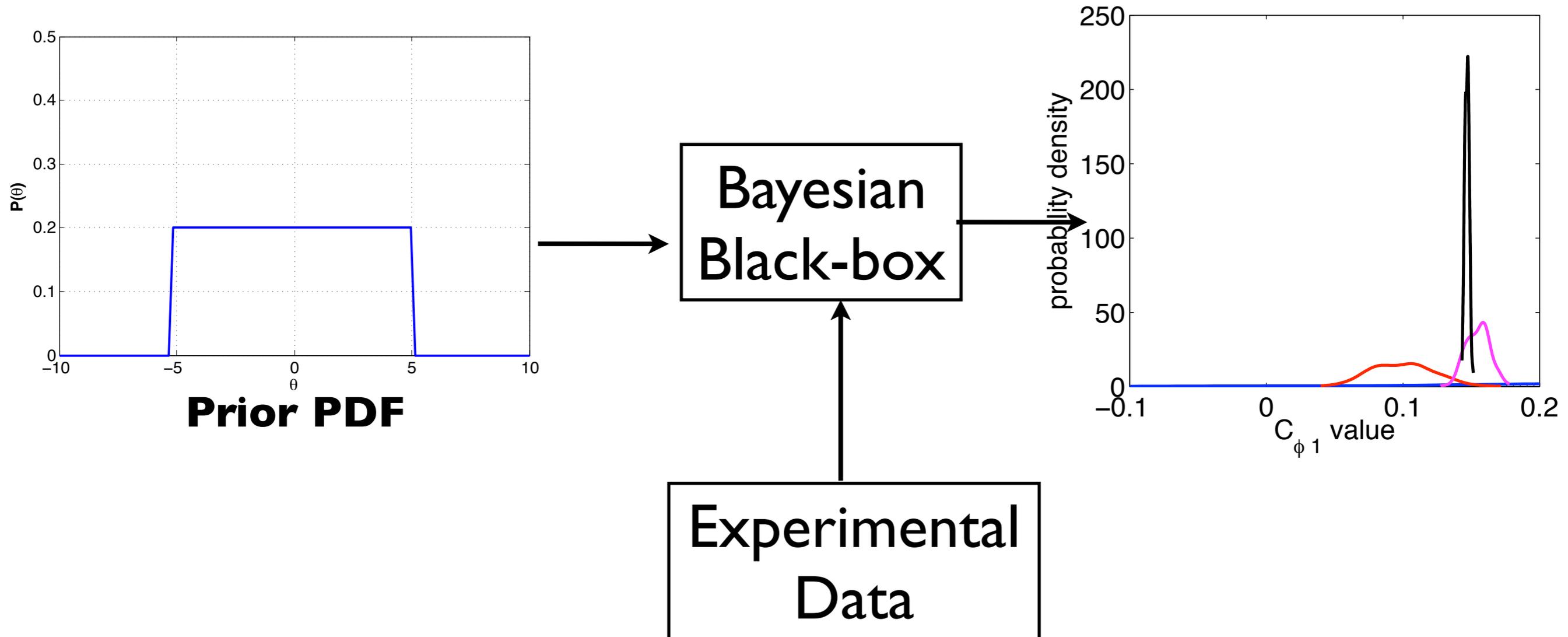
Prior (points to $P(A)$)

Posterior (points to $P(A|B)$)

Probability of Evidence (points to $P(B)$)

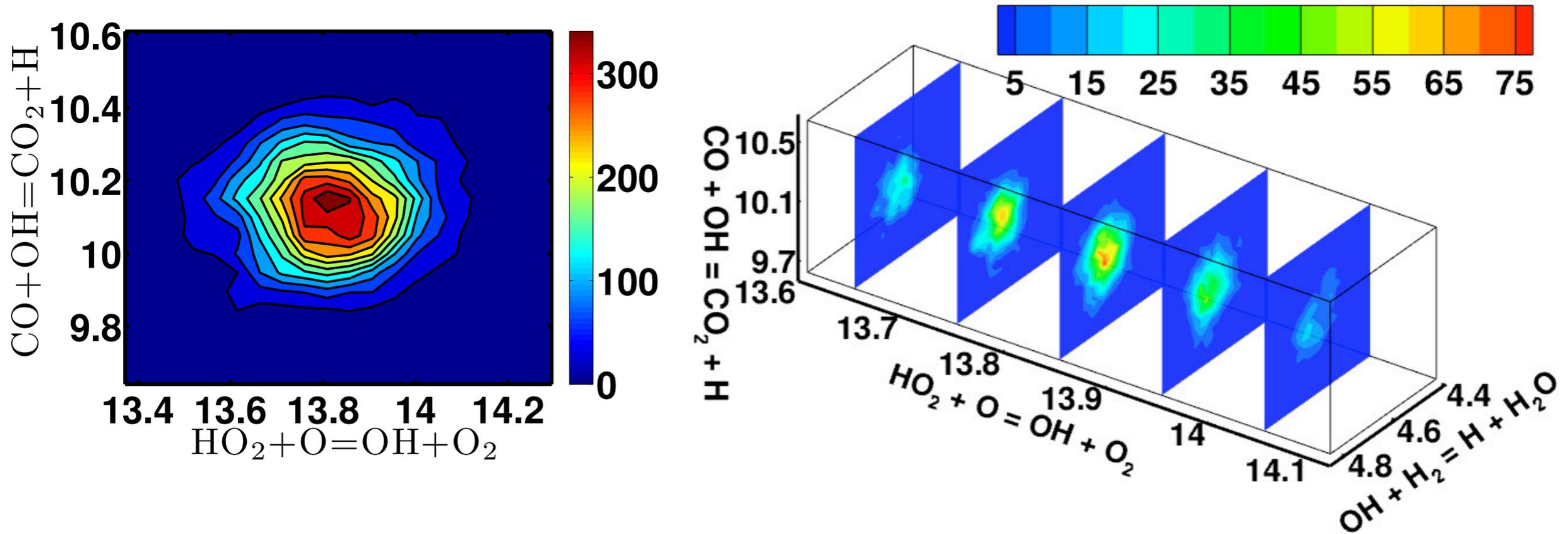
- ➔ For example, A could be the activation energy, and B will be experimental data
- ➔ The Bayes' theorem then updates the activation energy given the experimental data

Bayesian Learning Process



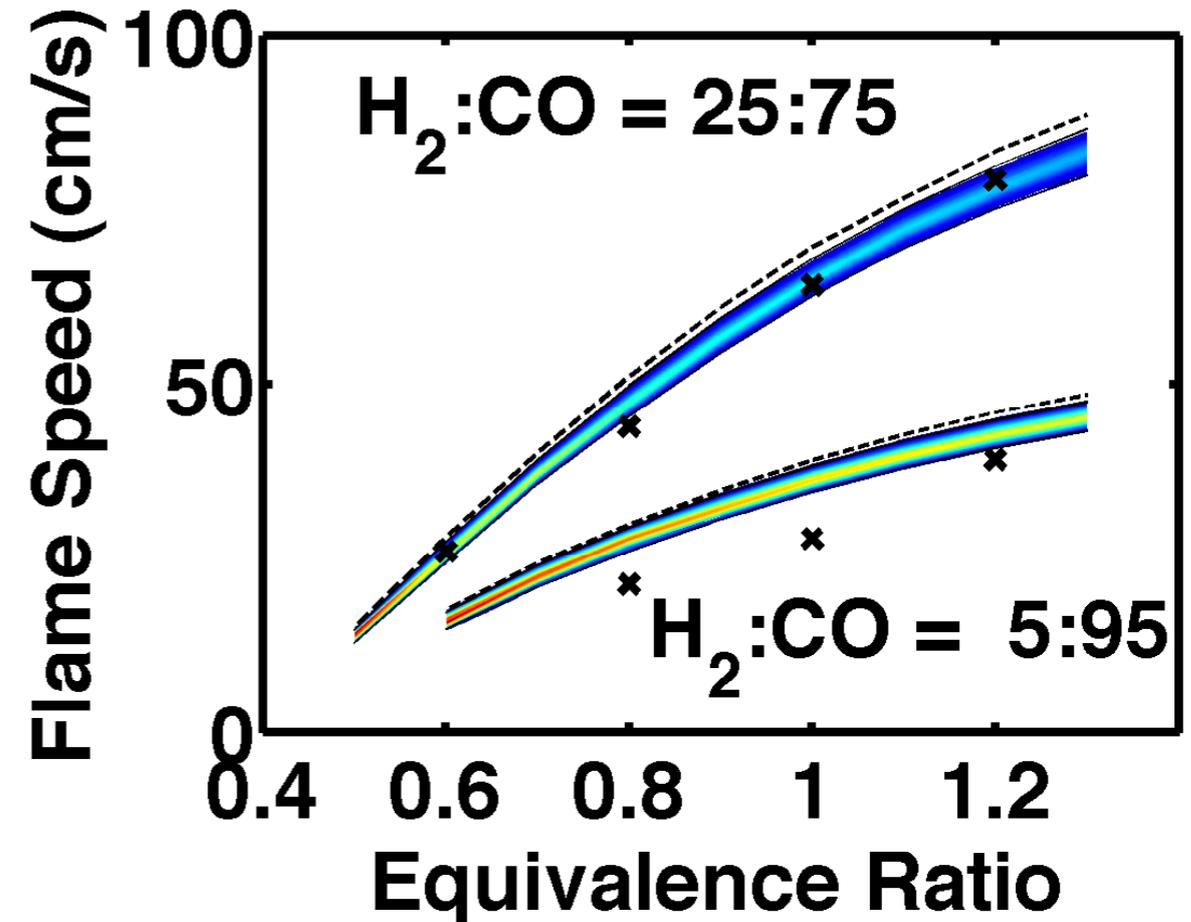
Application to Syngas Chemistry Mechanism

- Several mechanisms used as starting point
- Calibrations carried out using experimental data
 - ➔ >10 parameters jointly calibrated
 - ➔ PDFs are joint distribution of all variables



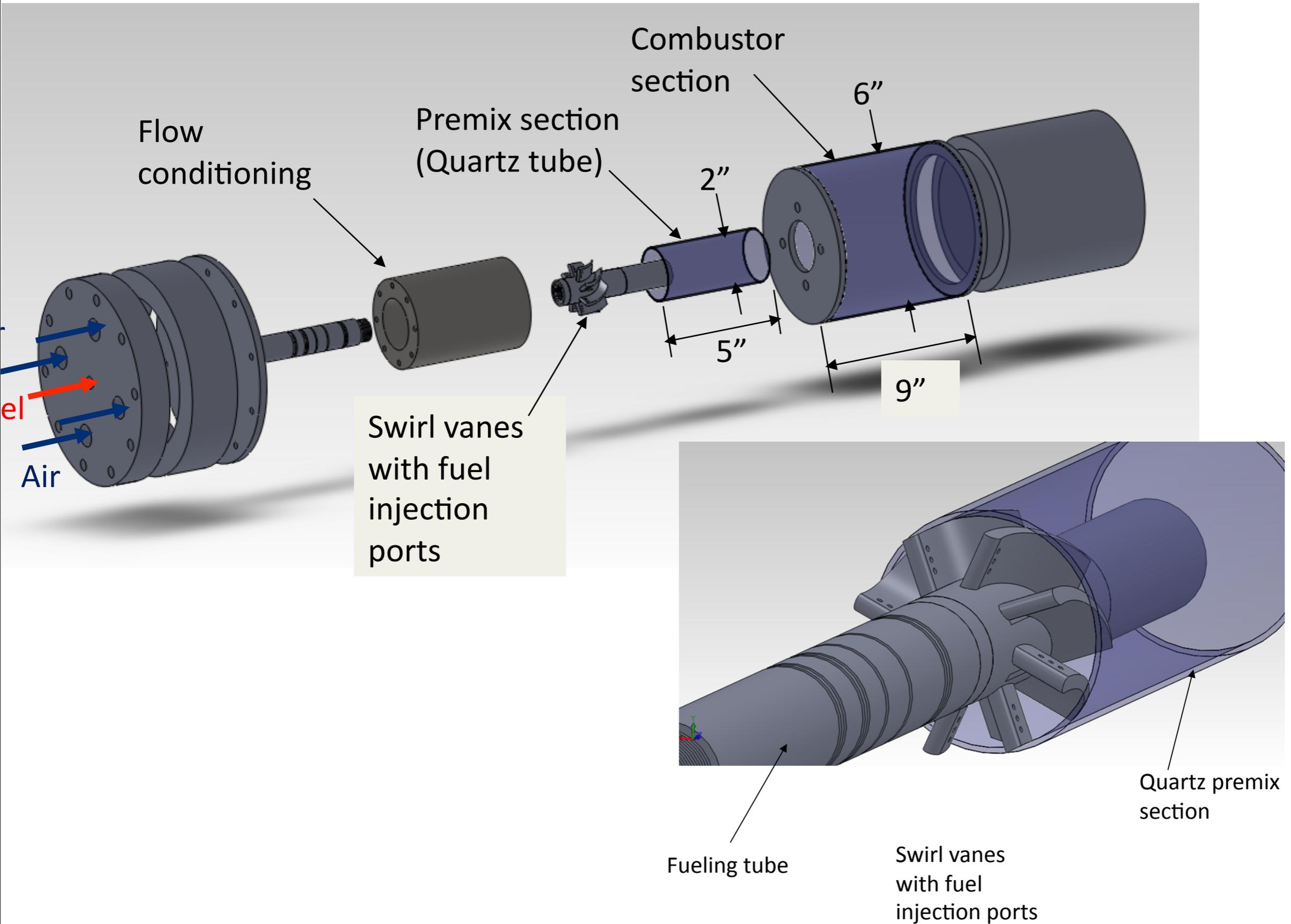
Prediction Uncertainty

- **Calibration using 10 atm. data**
 - ➔ Prediction at 20 atm.
- **PDFs used to develop prediction uncertainty**
- **Note that experiments also contain errors**
- **Inability to match experiments**
 - ➔ Points to model failure
 - ➔ Lack of appropriate experiments needed for calibration
 - ➔ Information used to design future experiments

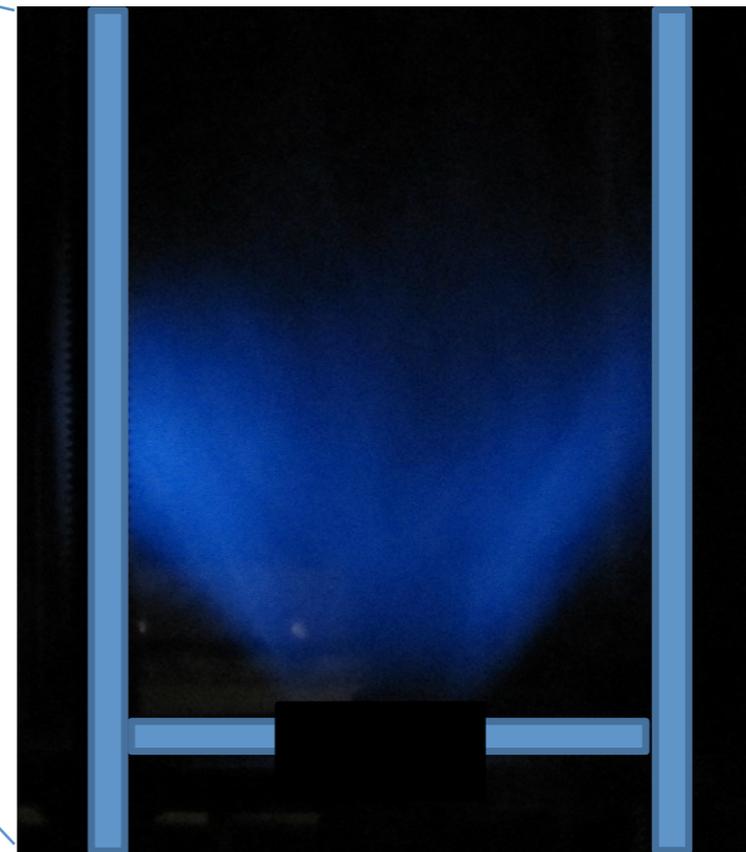
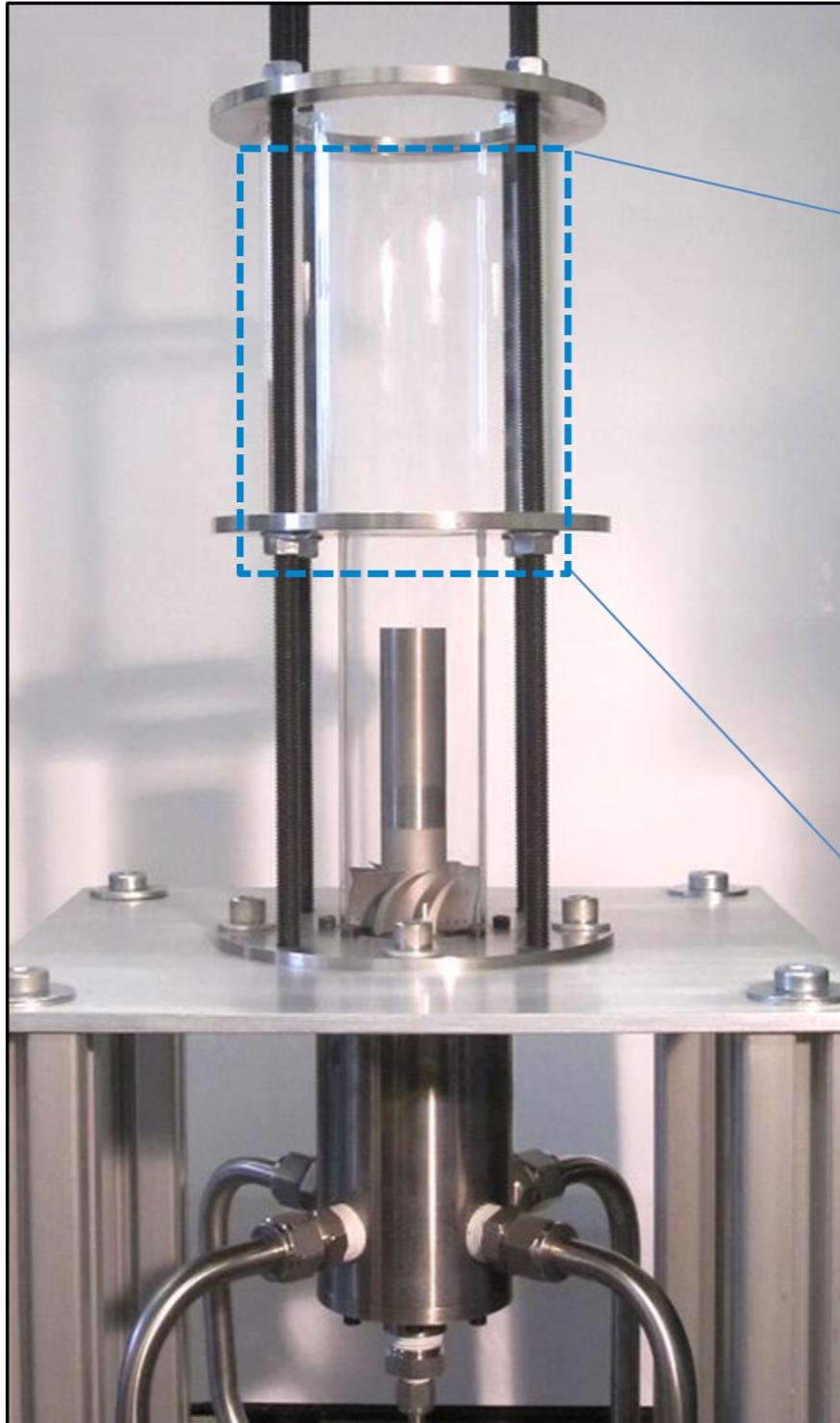


UT Swirl Burner Studies

Swirl Burner Design



Burner Operation



Lean-premixed ($\phi = 0.6$)
methane-air swirl flame

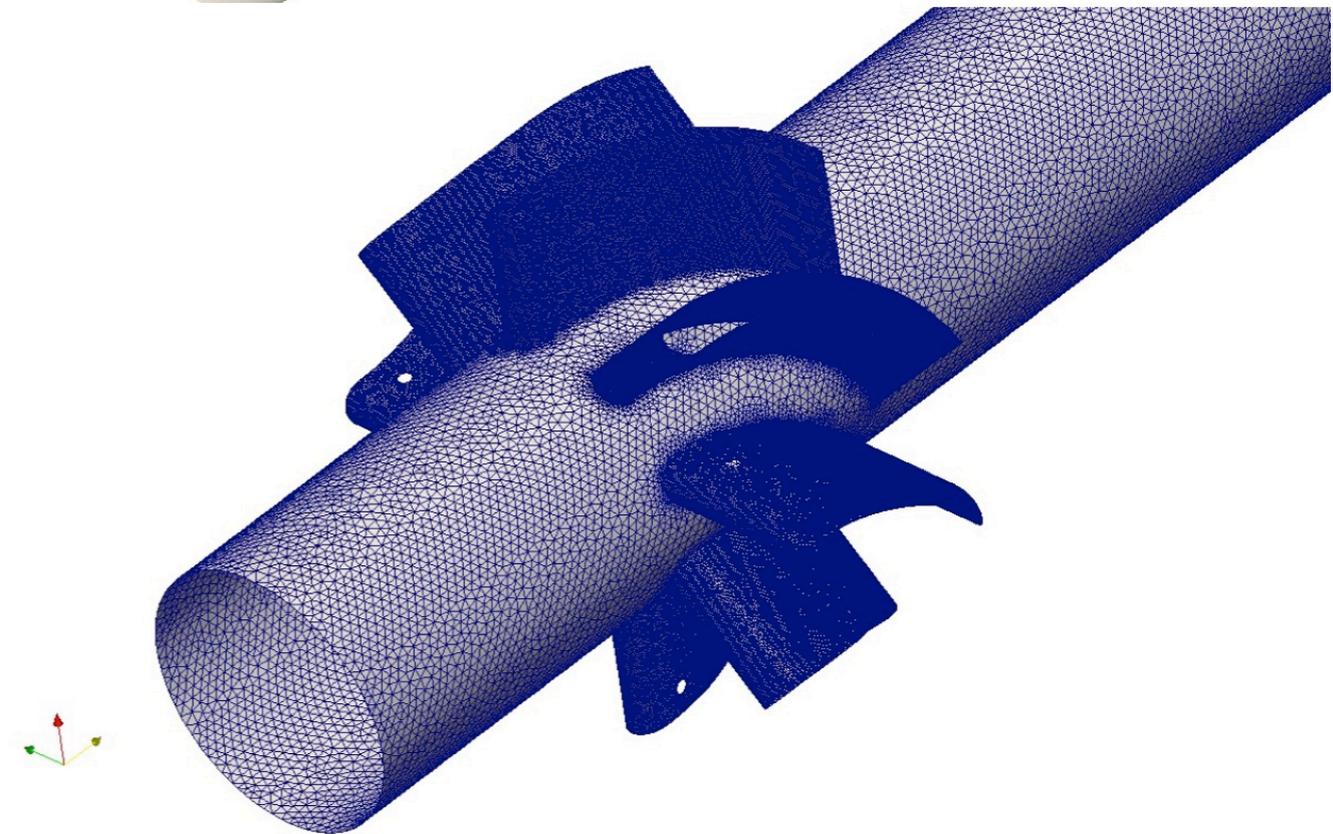
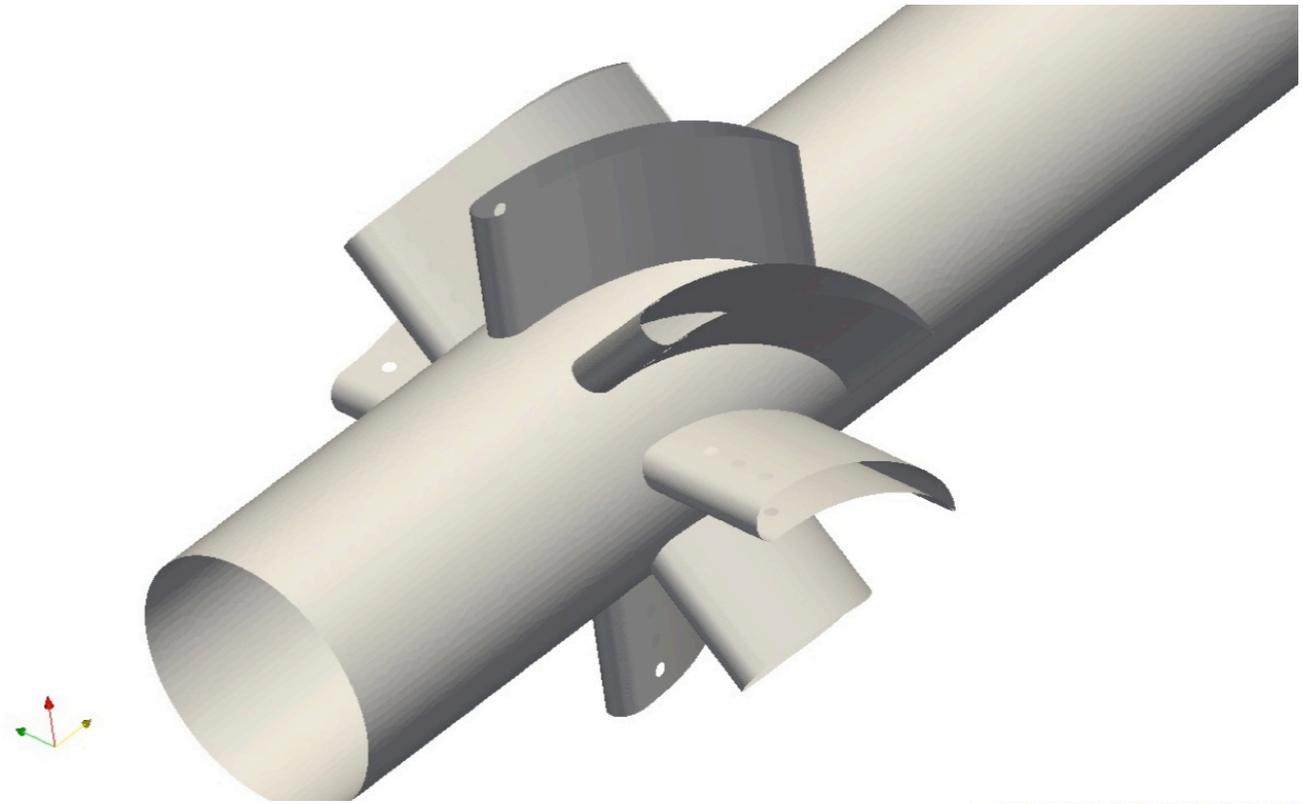
CFD for Burner Design and Operation

- **Use high-fidelity simulations to determine relevant experimental conditions**

- ➔ Minimizes experimental testing
- ➔ Better suited for model validation

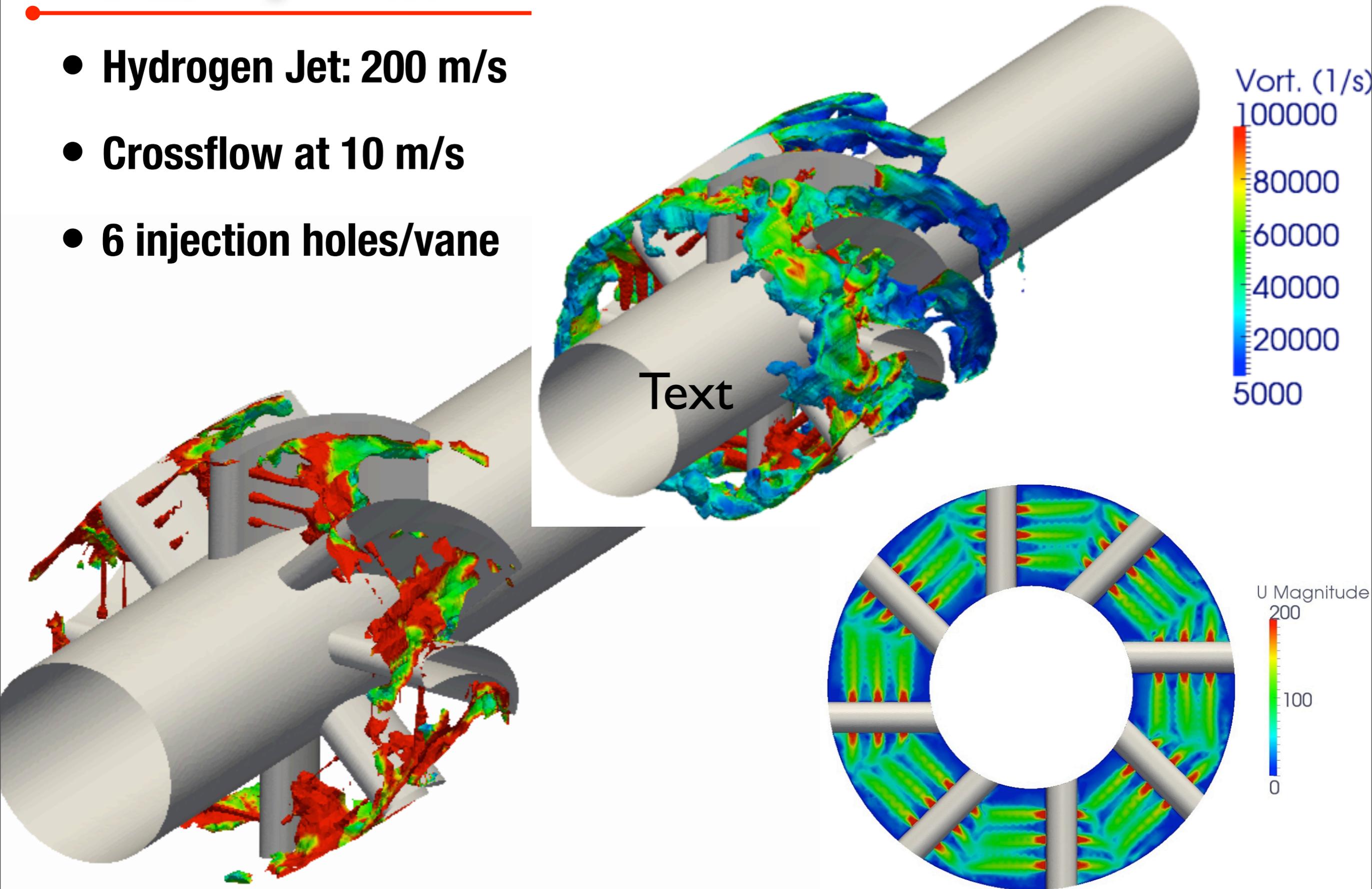
- **OpenFOAM based LES modeling of mixing**

- ➔ To understand the role of jet-jet interaction in vane-based injection
- ➔ Effect of hydrogen addition in methane jet



Preliminary Results

- **Hydrogen Jet: 200 m/s**
- **Crossflow at 10 m/s**
- **6 injection holes/vane**



Status

- **JICF studies**

- ➔ UT experiments ramping up; Initial data being analyzed
- ➔ UT simulations being performed
- ➔ OpenFOAM implementation being transferred to Siemens
 - (Open to other industrial partners)

- **UQ Computations**

- ➔ Chemistry UQ completed
- ➔ Transition to full CFD computations

- **UT swirl burner**

- ➔ Design, fabrication, and initial runs complete
- ➔ Experimental conditions being optimized using LES computations