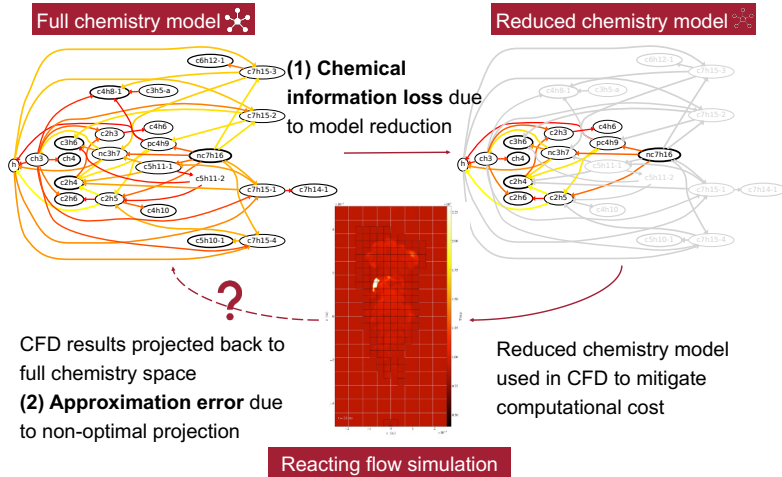


# Iterative Workflow for Quantification and Minimization of Reduced Chemistry-Induced Uncertainties in Reacting Flow Simulation

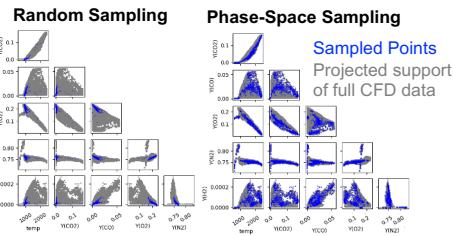
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## 1. INTRO | Reduced chemistry models mitigate computational cost but introduce two sources of uncertainties in reacting flow simulation

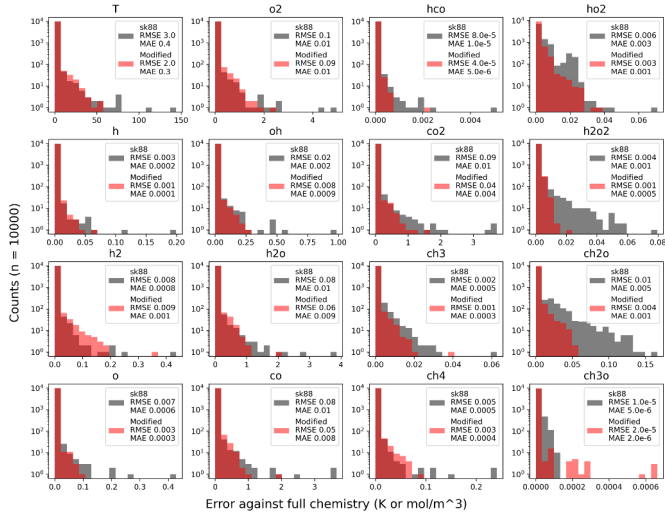


## 3. CASE STUDY

- Models used for case study
- Full chemistry model: n-heptane oxidation model w 561 species by Curran et al
- Reduced chemistry model: skeleton model with 88 species (sk88) by Yoo et al.
- Aiming to reduce uncertainties in predicted temperature compared to full chemistry
- Modified model contains 90 species
- Modified model outperforms sk88 for not only temperature but also many other key species

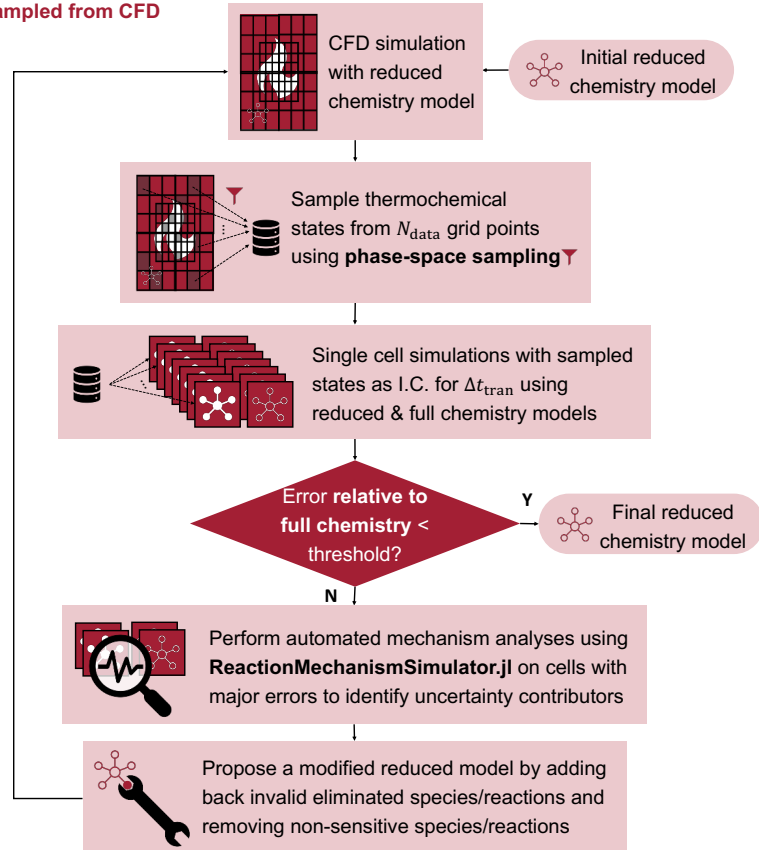


### Error in original & modified reduced chemistry models against full chemistry model



## 2. WORKFLOW

### 2.1. Decreases (1) chemical information loss by modifying reduced chemistry model based on error analyses against full chemistry model at thermochemical states sampled from CFD



### 2.2. Minimizes (2) approximation error by solving for optimal projection using constrained optimization with flux snapshots

$$\underline{Y} = \underline{L}\underline{Z}$$

$$\frac{d\underline{Z}}{dt} = \underline{\Gamma}(\underline{Z}, T, P; \underline{k}_f, \underline{K}_{eq}) + \underline{F} \quad \underline{\dot{Z}} = \underline{U}\underline{Y} = \underline{U}(\underline{L}\underline{Z}) \quad \underline{U}^* = ? \quad \frac{d\underline{Y}}{dt} = \underline{\gamma}(\underline{Y}, T, P; \underline{k}_f, \underline{K}_{eq}) + \underline{f}$$

Integrated over  $\Delta t_{tran}$  to smooth out effects from short-lived reactive intermediates

Collect snapshots of fluxes for both reduced & full chemistry models

$$\underline{W} = \left( \int_{t_0}^{t_0+\Delta t_{tran}} \underline{\gamma}(\underline{L}\underline{Z}_1, T_1, P_1; \underline{k}_f, \underline{K}_{eq}) + \underline{f} dt \quad \dots \quad \int_{t_0}^{t_0+\Delta t_{tran}} \underline{\gamma}(\underline{L}\underline{Z}_{N_{data}}, T_{N_{data}}, P_{N_{data}}; \underline{k}_f, \underline{K}_{eq}) + \underline{f} dt \right)$$

$$\underline{V} = \left( \int_{t_0}^{t_0+\Delta t_{tran}} \underline{\Gamma}(\underline{Z}_1, T_1, P_1; \underline{k}_f, \underline{K}_{eq}) + \underline{F} dt \quad \dots \quad \int_{t_0}^{t_0+\Delta t_{tran}} \underline{\Gamma}(\underline{Z}_{N_{data}}, T_{N_{data}}, P_{N_{data}}; \underline{k}_f, \underline{K}_{eq}) + \underline{F} dt \right)$$

Minimize differences between projected reduced and full snapshots

Ensure projection back to full chemistry space

$$\underline{U}^* = \underset{\underline{A}, \underline{U}}{\operatorname{argmin}} \frac{1}{2} \|\underline{W}^T \underline{U}^T - \underline{V}^T\|_2^2 + \lambda \underline{U}^T (\underline{L}^T \underline{U}^T - \underline{I})$$

combining derivative  $\frac{\partial \underline{L}}{\partial \underline{U}} = \underline{W}(\underline{W}^T \underline{U}^T - \underline{V}^T) + \underline{\Lambda}$

$\underline{\Lambda} = \underline{0}$  & constraint,  $\underline{U}^*$  can be solved with  $\begin{pmatrix} \underline{W}\underline{W}^T & \underline{L} \\ \underline{L}^T & \underline{0} \end{pmatrix} \begin{pmatrix} \underline{U}^T \\ \underline{\Lambda} \end{pmatrix} = \begin{pmatrix} \underline{W}\underline{V}^T \\ \underline{I} \end{pmatrix}$

