

# Tabulation of Combustion Chemistry via Artificial Neural Networks

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# Introduction

In order to develop more efficient and environmental friendly combustion engines comprehensive CFD simulations have to be performed. Such simulations are extremely time demanding since they involve:

- ▶ Turbulence-chemistry interaction models (i.e. PDF).
- ▶ Chemical mechanisms with hundreds of species and thousands of reactions.
- ▶ Wide range of time and length scales.

**Therefore, chemical step CPU time consumption must be reduced!!**

## Replace the production/destruction rate term

Our goal is to replace the source term by a non-linear regression process.

$$\dot{\omega} = \frac{\partial \phi}{\partial t} = f(x_1, \dots, x_k; w_1, \dots, w_n) \quad (1)$$

where

- ▶  $\phi$  Reactive scalar.
- ▶  $x_k$  The  $k_{th}$  input.
- ▶  $w_n$  The  $n_{th}$  polynomial constant.

# ANN Tabulation framework

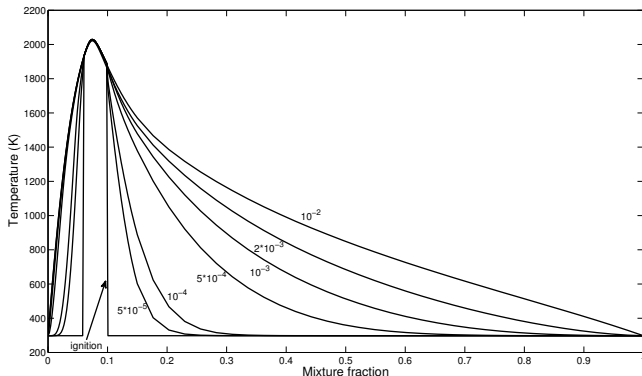
- ▶ Mechanism Reduction.
- ▶ Sample generating.
- ▶ ANN training.
- ▶ Validation.
- ▶ Application.

# 1 - Mechanism Reduction via Rate-Controlled Constrained Equilibrium (RCCE)

- ▶ Reduced mechanism framework.
- ▶ Kinetically constrained: Solved directly from the kinetics of the detailed mechanism.
- ▶ Equilibrated: Assumed to be in a constrained equilibrium.

## 2 - Sample generating data via ignited flamelet

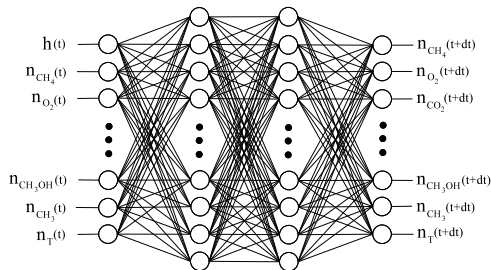
- ▶ Varying strain rate from  $1s^{-1}$  to  $1100s^{-1}$ .
- ▶ Randomly selected mixture fraction points.
- ▶ Store information of enthalpy alongside with kinetically constrained species before and after the reaction step.



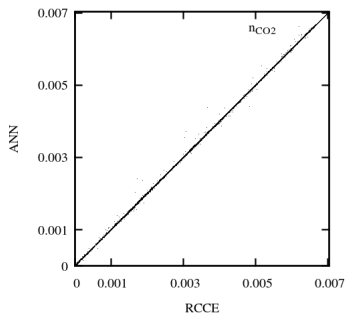
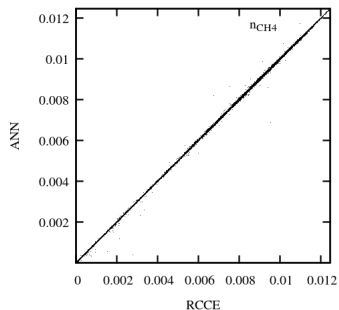
## 3 - ANN training

- ▶ The ANN is a non-linear mapping between two sets of variables.
- ▶ It involves a number of unknown coefficients, called weights ( $w_{ij}$ ), to be determined by a training step following the pairs of input-output data.
- ▶ Training is performed by conjugate gradient optimization, minimizing the mean square error between  $ANN_{outputs}$  and  $RCCE_{outputs}$ .

$$y_k = f\left(\sum w_{jk}y_j\right) \quad (2)$$



## 4 - Validation with unseen data



**Figure:** Comparison of unseen data points ANN prediction against the RCCE target values

$$\text{Mean}_{RMS} = 0.836\text{E-}6$$



## 5 - Application to turbulent non-premixed flame

We consider a combined LES-PDF model:

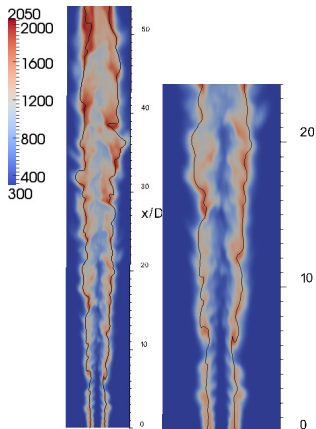
- ▶ LES of a variable density flow advecting a number of reactive scalars.
- ▶ The turbulence-chemistry interaction is closed by the PDF method.
- ▶ The Eulerian stochastic fields method comprising 8 fields is applied as the solution scheme via the Monte Carlo estimator.

$$\rho \partial \xi_{\alpha}^n = -\rho \tilde{u} \frac{\partial \xi_{\alpha}^n}{\partial x} dt + \rho \dot{\omega}(\phi) \xi_{\alpha}^n dt + \frac{\partial}{\partial x} \left[ \left( \frac{\mu}{\sigma} + \frac{\mu_{sgs}}{\sigma_{sgs}} \right) \frac{\partial \xi_{\alpha}^n}{\partial x} \right] dt - \frac{\rho}{2 \zeta_{sgs}} [\xi_{\alpha}^n - \tilde{\phi}] dt + \sqrt{2 \rho \Gamma} \frac{d \xi_{\alpha}^n}{dx} dW \quad (3)$$

$$\tilde{\phi}_{\alpha} = \frac{1}{N} \sum_{n=1}^N \xi_{\alpha}^n \quad (4)$$

- ▶ Simulations are performed on the ARCHER UK National Supercomputing Service, employing 80 processors.

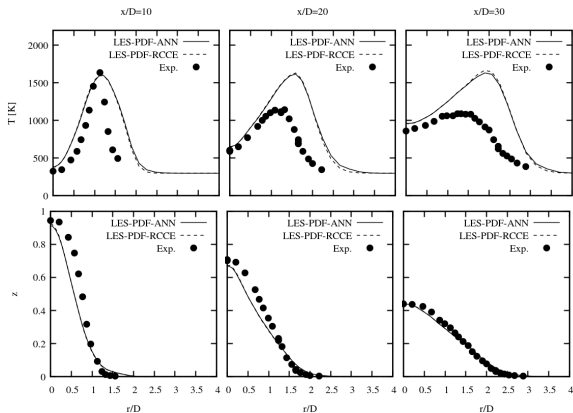
# Sydney Flame L



**Figure:** Contour plots of Sydney flame L computed with LES-PDF-ANN: a) full domain  
b) zoom on the middle section of the flame.

Results presented in the Combustion and Flame Journal Paper: Tabulation of Combustion Chemistry via Artificial Neural Networks (ANNs): Methodology and Application to LES-PDF Simulation of Sydney Flame L

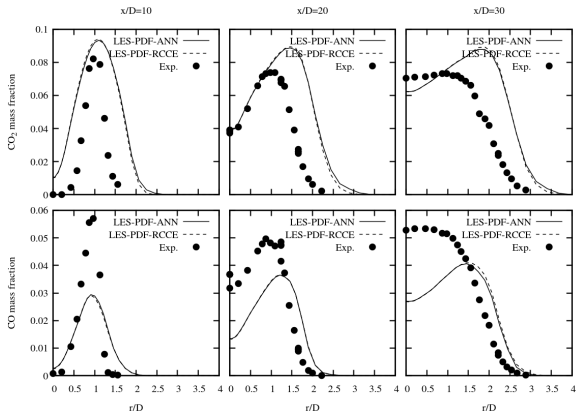
# Mean Fields - I



**Figure:** LES-PDF-ANN and LES-PDF-RCCE mean temperature and mixture fraction radial profiles

Results presented in the Combustion and Flame Journal Paper: Tabulation of Combustion Chemistry via Artificial Neural Networks (ANNs): Methodology and Application to LES-PDF Simulation of Sydney Flame L

# Mean Fields - II



**Figure:** LES-PDF-ANN and LES-PDF-RCCE mean  $CO_2$  and  $CO$  mass fraction radial profiles

Results presented in the Combustion and Flame Journal Paper: Tabulation of Combustion Chemistry via Artificial Neural Networks (ANNs): Methodology and Application to LES-PDF Simulation of Sydney Flame L

# Time measurement and storage space

Method	Convection-Diffusion (s)	Reaction (s)
RCCE	7	125
ANN	7	1.5

Table: Average CPU time for ANN and RCCE real time integrator

Finally, the disk storage space required for the ANNs (i.e. the weights) is only 36 MB!

## Concluding remarks

Advantages of the RCCE-ANN tabulation framework:

- ▶ Easy to implement (or to combine with existing software).
- ▶ Accurately reproduces the mean and rms profiles given by the reduced mechanism (RCCE).
- ▶ Reduced the computational cost by two orders of magnitude compared to the reduced mechanism (RCCE).
- ▶ Very small disk storage space requirement.

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Thank you very much