Proceedings of the Combustion Institute

Optimized chemistry for Large Eddy Simulations of wrinkled flames

--Manuscript Draft--

Manuscript Number:	
Article Type:	5. Turbulent Flames
Keywords:	Turbulent combustion modelling; Large Eddy Simulation; pollutants; flame wrinkling
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38th International Symposium on Combustion

Optimized chemistry for Large Eddy

Simulations of wrinkled flames

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Colloquium: TURBULENT FLAMES.

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WORD COUNT

Method of determination: Method 2

Words in text: 3215

Words in references: 540

Words in tables: 54

Words in equations: 276

Words in figures: 1715

Total length of paper: 5800 words

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Abstract

State-of-the-art LES models have made significant progress in the prediction of heat release and flame propagation for a wide range of combustion regimes. Predicting pollutants remains however a difficult issue, especially in situations where sub-grid scale wrinkling is high. A novel strategy, named COPLES (Chemistry OPTimized for LES), is presented in this paper to account for the effect of sub-grid scale wrinkling on pollutants formation in a transported chemistry context and under flamelet regime assumption. New chemical mechanisms are built through an optimization process to recover the properties of filtered flames. Two alternative formulations are proposed. The first one consists in optimizing the COPLES parameters to retrieve the structure of Filtered Planar Flame (FPF) and to include separately the impact of sub-grid scale wrinkling. The second formulation targets directly a collection of Filtered Wrinkled Flamelets (FWF), so that the influence

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of turbulence on the chemical flame structure is intrinsically accounted for. The two approaches are successfully validated on 1-D freely propagating premixed flames and then challenged against the standard Thickened Flame for LES model on a turbulent premixed swirled combustor, experimented at Cambridge University. Whereas all formulations retrieve the temperature field, the COPLES-FWF model improves significantly the prediction of CO formation.

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2010 MSC: 00-01, 99-00

1. Introduction

Prediction of pollutants formation in Large Eddy Simulation (LES) requires complex chemistry [1]. Issues relative to the computational cost induced by detailed chemistry are tackled by resorting either to chemistry reduction [2] or to tabulation techniques [3, 4]. By storing in a pre-processing step all thermo-chemical information in a look-up table, tabulated chemistry is extremely competitive in terms of CPU time. However, it lacks flexibility when the encountered flame structures differ from the tabulated archetypes [5]. Large discrepancies on the prediction of CO in a diluted combustion burner were for instance observed in the work of Lamouroux *et al.* [6]. Meanwhile, the progress in reduced mechanisms and the increase of available computational power enable a direct use of transported chemistry. Recently, analytic mechanisms have been for example introduced in LES [7, 8]. Also, virtual optimized schemes, developed to capture both heat release [9] and pollutant formation [10, 11] over a wide range of flame regimes have been successfully applied to LES of turbulent premixed flames.

Within both reduced and tabulated chemistry formalisms, the closure of filtered species balance equations is the key of turbulent combustion modeling. The LES combustion model must address two issues [12]: (i) it has to handle situations where the flame front is thinner than the LES grid. Commonly used solutions are either to filter [13] or thicken [14] the species equations at a scale larger than the mesh size; (ii) it must include a model for integrating sub-grid scale turbulent effects, for instance through a wrinkling factor [15] or a FDF [16]. We focus here on models based on a geometrical description of the flame front [12], which are efficient to compute flames in

complex geometries at low computational costs. While geometrical models
27 have been able to predict heat release and flame propagation in various con-
ditions [7, 13], intermediate species prediction remains challenging [17]. A
recent work by Abou-Taouk *et al.* [18] has focused on the inclusion of (i)
30 directly in the reaction mechanism by optimizing Arrhenius parameters on
filtered flames. The method recovers well the structure of filtered planar
laminar flames but is not able to predict intermediate species concentrations
33 if the flame is wrinkled.

Indeed, as observed in a turbulent flame DNS [19], filtered quantities,
especially intermediate species such as CO, are influenced by the surface of
36 the flame contained at the sub-filter scale. The recent work developed by
Mercier *et al.* [20] includes these effects on pollutant formation by defining a
new flame archetype, which takes sub-grid scale flame wrinkling intrinsically
39 into account. The model relies on the Filtered Wrinkled Flamelets (FWF)
archetype, obtained by manufacturing two-dimensional wrinkled flames, which
are filtered at the LES filter size. FWF has been used to build-up a filtered
42 chemical look-up table in order to model the unclosed terms of the filtered
progress variable equation. Simulations of the swirling stabilized Cambridge
flame [21] showed that accounting for sub-filter flame wrinkling on the chem-
45 ical flame structure improves the prediction of intermediate species such as
CO [20]. Whereas the benefit of FWF has been demonstrated within a tab-
ulated chemistry framework, accounting for the impact of SGS flame wrin-
48 kling on the species formation in a reduced chemistry context remains an
open question.

The objective of the present work is to extend the work of Mercier *et*

51 *al.* [20] in a reduced chemistry context. Developments are made under a
flamelet regime assumption. Contrarily to classical LES models [14], the im-
pact of sub-filter turbulence on CO formation [22, 20] is accounted for. Other
54 physical phenomena which may impact the formation of CO, such as flame
generated turbulence or counter-gradient transport, are not tackled. The se-
lected methodology features the development of optimized filtered schemes to
57 retrieve the properties of the FWF flame archetype. The paper is organized
as follow: model equations are detailed in Sec. 2, the filtered mechanism
generation is presented in Sec. 3 whereas validations in both 1-D and 3-D
60 turbulent premixed flames configurations are shown in Sec. 4.

2. Chemistry Optimized for LES of turbulent flames

2.1. Model equations

63 In LES, the balance equation of the filtered species mass fraction $\tilde{Y}_k$ reads:

$$\frac{\partial \tilde{\rho} \tilde{Y}_k}{\partial t} + \nabla \cdot (\tilde{\rho} \tilde{u} \tilde{Y}_k) = \overline{\text{RHS}(\Phi)} \quad (1)$$

where u is the flow velocity and $\Phi = (P, T, Y_1, \dots, Y_N)$ is the thermo-
chemical state vector with P and T the pressure and temperature. $\overline{\varphi}$ and
66 $\tilde{\varphi}$ denote Reynolds and Favre averaging of quantity φ , respectively. The
unclosed filtered Right Hand Side $\overline{\text{RHS}(\Phi)}$ reads:

$$\overline{\text{RHS}(\Phi)} = \bar{\tau}_k + \nabla \cdot (\overline{\rho D_k \nabla Y_k}) + \overline{\rho \dot{\omega}_k^A(\Phi)} \quad (2)$$

where D_k is the molecular diffusivity of species k , $\bar{\tau}_k = -\nabla \cdot (\overline{\rho u \tilde{Y}_k} - \tilde{\rho} \tilde{u} \tilde{Y}_k)$
69 is the sub-grid scale convective term and $\dot{\omega}_k$ the chemical reaction rate of

species k parametrized by the set of kinetic rate parameters $\mathcal{A}$. Because of non-linearities of the terms that compose Eq.2, the RHS cannot be directly estimated from the filtered thermochemical variable [12]: $\overline{\text{RHS}(\Phi)} \neq \text{RHS}(\overline{\Phi})$.

The concept of Chemistry Optimized for LES (COPLES) is to identify a function RHS^* , which verifies the equality $\overline{\text{RHS}(\Phi)} = \text{RHS}^*(\overline{\Phi})$. The following formulation of RHS^* is retained for that purpose:

$$\text{RHS}^* = \nabla \cdot \left(\bar{\rho} \alpha^* \tilde{D} \nabla \tilde{Y}_k \right) + \bar{\rho} \tilde{\omega}_k^{\mathcal{A}^*} \quad (3)$$

with:

$$\bar{\rho} \tilde{\omega}_k^{\mathcal{A}^*} = W_i \sum_{j=1}^{N_R} A_j^* \left(\nu_{kj}^b - \nu_{kj}^f \right) \times \prod_{i \in \mathcal{S}_j} \left(\frac{\bar{\rho} \tilde{Y}_i}{W_i} \right)^{n_{ij}^*} \exp \left(\frac{-E_{a,j}^*}{\mathcal{R} \tilde{T}} \right) \quad (4)$$

where N_R is the number of reactions, $\mathcal{S}_j$ are the species involved in reaction j , and ν_{kj}^b , ν_{kj}^f are respectively the backward and forward stoichiometric coefficients of species k in reaction j . Variables with * superscripts are model parameters: α^* is constant in space and identical for each species and $\mathcal{A}^* = (A_j^*, E_{a,j}^*, n_{ij}^*)$ where A_j^* are the pre-exponential constants, $E_{a,j}^*$ the activation energies and n_{ij}^* the reaction orders of the modeled chemical mechanism.

The parameters are optimized through a genetic algorithm [9, 11] to recover the properties of an ensemble made of N canonical 1-D target flames. The optimization process consists in minimizing the following function:

$$\zeta(\mathcal{A}^*, \alpha^*) = \sum_{n=1}^N \left(\frac{\|\tilde{\varphi}_n^* - \tilde{\varphi}_n^{ref}\|_2}{\|\tilde{\varphi}_n^{ref}\|_2} + \beta \left| \frac{f_{\varphi,n}^* - f_{\varphi,n}^{ref}}{f_{\varphi,n}^{ref}} \right| \right) \quad (5)$$

where φ_n and $f_{\varphi,n}$ are respectively the thermo-chemical spatial profile and the specific global flame properties of the n -th 1-D flame targeted by the

optimized scheme. Superscript *ref* and * refer to the target flame and to the solution obtained with optimized parameters, respectively. β is a weighting
90 parameter set to 1 in the present work.

The choice of the reference canonical flames is critical. The retained filtered archetype must be representative of the sub-grid turbulent flame structure encountered for various level of sub-grid flame wrinkling. As discussed
93 in [20], the issue is to account for the influence of the subfilter flame front geometry on the pollutant formation at the resolved scale. Two archetypes
96 are discussed in the following section: the Filtered Planar Flames (FPF) and the Filtered Wrinkled Flamelets (FWF).

2.2. Reference filtered flame archetypes

99 2.2.1. Optimization on Filtered Planar Flames (FPF)

The first solution to build-up the filtered optimized scheme is to isolate the impact of sub-grid scale wrinkling from the filtering process. The identification of diffusion and kinetic rate parameters is split in two steps.
102

i) A collection of 1-D freely propagating laminar flames is computed for various equivalence ratios ϕ using a detailed chemical scheme and then
105 filtered in the spatial dimension with a Gaussian filter of size Δ . This collection of Filtered Planar Flames (FPF) provides the filtered reference thermo-chemical quantities $\tilde{\varphi}_{\text{FPF}} = \tilde{\varphi}_{\text{FPF}}(x, \phi, \Delta)$ on which is trained the optimized scheme. This step leads to the identification of parameters $\mathcal{A}_{\text{FPF}}^* = (A_{j,\text{FPF}}^*, E_{a,j,\text{FPF}}^*, n_{ij,\text{FPF}}^*)$ and α_{FPF}^* , function of ϕ and Δ as shown in Tab. 1.

ii) The introduction of sub-grid scale flame wrinkling Ξ_{Δ} is done *a posteriori* by multiplying pre-exponential constants and the diffusive correction coefficient by Ξ_{Δ} : $\mathcal{A}^* = (\Xi_{\Delta} A_{j,\text{FPF}}^*, E_{a,j,\text{FPF}}^*, n_{ij,\text{FPF}}^*)$ and $\alpha^* = \Xi_{\Delta} \alpha_{\text{FPF}}^*$.
111

This operation, which is common in many turbulent LES combustion model
 114 [12, 14] does not affect the filtered flame structure but ensures that the fil-
 tered flame propagates at a speed $S_\Delta = \Xi_\Delta S_l$, where S_l is the unstretched
 laminar flame speed. However, as pointed out in [20], this assumption in-
 117 cludes biases in the prediction of some pollutants, such as CO. To address
 this issue, wrinkling patterns can be included in the filtering process.

2.2.2. Optimization on Filtered Wrinkled Flamelets (FWF)

Figure 1: Schematic view of sub-filter sinusoidal flame pattern.

A 2-D wrinkled flame pattern, shown in Fig. 1, is manufactured by ap-
 proximating the flame front position by a sinusoidal function of amplitude
 γ and wavelength P . Embedded in a 2-D square of size Δ , this flame
 123 pattern of wrinkling Ξ_Δ contains a number of wavelengths $n_\Delta = \Delta/P$.
 Assuming that the flame is in the flamelet regime, the 2-D flame struc-
 ture is manufactured from a 1-D laminar premixed flame computed with
 126 detailed chemistry. The reconstruction algorithm as well as the possible
 physical range covered by wrinkling parameters γ and n_Δ are detailed in
 [20]. The manufactured wrinkled flame is then 2-D filtered and, due to
 129 the periodicity of the sine function, filtered thermo-chemical variables re-

duce to a single dimension x , perpendicular to the flame front propagation:
 $\tilde{\varphi}_{\text{FWF}} = \tilde{\varphi}_{\text{FWF}}(x, \phi, \Delta, n_{\Delta}, \Xi_{\Delta})$. The filtered database serves as an ensemble
132 of targets for the filtered optimized scheme. This step leads to the identification of parameters $\mathcal{A}^* = (\mathcal{A}_{j,\text{FWF}}^*, E_{a,j,\text{FWF}}^*, n_{ij,\text{FWF}}^*)$ and $\alpha^* = \alpha_{\text{FWF}}^*$. The
parameter dependencies of the new filtered optimized scheme are listed in
135 Tab. 1. While the simplified approach to represent wrinkled flames neglects several physical phenomena, such as local variations in temperature due to
turbulence, it has been shown to recover with reasonable accuracy the effect
138 of sub-filter wrinkling on pollutants formation [20].

Table 1: Summary of Filtered Optimized Model with FPF and FWF flame libraries.

	FPF	FWF
$\mathbf{A}_j^*$	$\Xi_{\Delta} A_{j,\text{FPF}}^*(\phi, \Delta)$	$A_{j,\text{FWF}}^*(\phi, \Delta, \Xi_{\Delta}, n_{\Delta})$
$\mathbf{E}_{a,j}^*$	$E_{a,j,\text{FPF}}^*(\phi, \Delta)$	$E_{a,j,\text{FWF}}^*(\phi, \Delta)$
$\mathbf{n}_{ij}^*$	$n_{ij,\text{FPF}}^*(\phi, \Delta)$	$n_{ij,\text{FWF}}^*(\phi, \Delta)$
$\boldsymbol{\alpha}^*$	$\Xi_{\Delta} \alpha_{\text{FPF}}^*(\phi, \Delta)$	$\alpha_{\text{FWF}}^*(\phi, \Delta, \Xi_{\Delta}, n_{\Delta})$

2.2.3. Influence of FPF and FWF archetypes on the filtered chemical flame structure

141 FPF and FWF archetypes are both manufactured from a 1-D laminar premixed methane/air detailed chemistry solution obtained using the GRI3.0 mechanism [23] at an equivalence ratio $\phi = 0.75$. Filtered temperature and
144 CO mass fractions are plotted in Fig. 2. FWF solutions (lines) are shown for one flame pattern ($n_{\Delta} = 1$) but for different values of sub-grid scale flame wrinkling. For $\Xi = 1$, the FWF profiles degenerate as expected towards the

147 FPF solution (symbols). When the sub-grid scale flame wrinkling increases,
the filtered thermal layer is thickened. Fig. 2 (right) shows that the CO
peak is also enhanced by the sub-grid scale wrinkling. The total mass of CO
150 contained in the filter box of surface Δ^2 is: $\mathcal{M}_{CO}^\Delta(\Xi_\Delta) = \int_{\Delta^2} \rho Y_{CO} ds$.

The ratio $\mathcal{M}_{CO}^\Delta(\Xi_\Delta)/\mathcal{M}_{CO}^\Delta(\Xi_\Delta = 1)$ is plotted in terms of Ξ_Δ in Fig. 3
for two LES filter size values: $\Delta = 4\delta_l^0$ and $8\delta_l^0$, where δ_l^0 is the laminar flame
153 thickness. To avoid superposition of sine branches, the possible number of
flame patterns within the filter box is limited to $n_\Delta^{max} = E[\Delta/2\delta_l^0]$, where $E[.]$
refers to the integer part of a number. The normalized mass of CO is therefore
156 shown in Fig. 3 for all possible number of flame patterns ($n_\Delta = (1, 2)$ for
 $\Delta = 4\delta_l^0$ and $n_\Delta = (1, 2, 3, 4)$ for $\Delta = 8\delta_l^0$). A significant increase of CO
mass is first observed when the SGS wrinkling grows. The mass of CO
159 remains however lower than the value (dashed line) obtained under infinitely
thin flame front assumption: $\mathcal{M}_{CO}^\Delta(\Xi_\Delta) = \Xi_\Delta \mathcal{M}_{CO}^\Delta(\Xi_\Delta = 1)$. This is an
effect due to the non-zero Y_{CO} equilibrium value [20]. Moreover, n_Δ has
162 little impact on the mass of CO.

Figure 2: Illustration of the FPF and FWF flamelet libraries for $\phi = 0.75$, $\Delta = 4\delta_l^0$ and $n_\Delta = 1$. Legend: $\blacksquare$ FPF flame. - - - FWF flamelet with $\Xi_\Delta = 1$. - . - FWF flamelet with $\Xi_\Delta = 1.5$. — FWF flamelet with $\Xi_\Delta = 2.4$.

Figure 3: Normalized mass of CO in a subfilter box as a function of Ξ_Δ for $\phi = 0.75$ and two different filter sizes $4\delta_l^0$ and $8\delta_l^0$. Squares: $n_\Delta = 1$. Crosses: $n_\Delta = 2$. Triangles: $n_\Delta = 3$. Circles: $n_\Delta = 4$. Dashed line: unitary line $\mathcal{M}_{CO}^\Delta(\Xi_\Delta) = \Xi_\Delta \mathcal{M}_{CO}^\Delta(\Xi_\Delta = 1)$.

3. Filtered mechanism generation

In the present work, the COPLES method is applied to predict both the
 165 thermal flame structure, including heat release and flame propagation speed,
 and the CO production. The chemical scheme retained as a basis for the
 optimization procedure is based on the virtual chemistry concept [9, 10, 11].
 168 This strategy relies on the optimization of a virtual mechanism composed
 of virtual reactions and virtual species whose properties are also optimized.
 The scheme is first made of the following two-step main virtual mechanism
 171 which converts CH_4 and O_2 into an intermediate species (I) which is then
 transformed through into a set of N_p virtual products P_k .

The number of virtual products (here $N_p=4$) as well as the stoichiometric
 174 coefficients α_{P_k} are optimized to retrieve real burnt gases temperature and
 enthalpy over the whole range of equivalence ratio conditions, as detailed in
 [11]. α_F , α_{Ox} , α_I are given by fuel composition.

177 Complementary independent virtual sub-mechanisms are devoted to the
 prediction of real pollutant species such as CO, NOx or soot precursors
 (PAH). By dedicating each sub-mechanism to a unique pollutant prediction,
 180 this approach enables an efficient separation between slow and rapid chem-
 ical species. The following complementary virtual sub-mechanism is added
 here to predict the CO mass fraction:

183 where CO is the targeted pollutant species and V_1 is a virtual species.
 Free Gibbs energy of reaction (R4) is identified to retrieve the CO mass
 fraction at equilibrium conditions [11].

186 The COPLES method is then applied to optimize the molecular diffusion
 constant α^* and the kinetics rate parameter $\mathcal{A}^*$ required to close Eq. 3. As the
 main virtual mechanism is independent from the virtual CO sub-mechanism,
 189 the optimization process will be performed sequentially.

i) The diffusion constant α_{Main}^* and kinetics parameters $\mathcal{A}_{Main}^*$ of reactions
 (R1)-(R2) are first optimized to retrieve the temperature profile and the flame
 192 propagation speed over a wide range of sub-grid scale wrinkling conditions.
 The objective function introduced in Eq. 5 is therefore defined with $\tilde{\varphi} = \tilde{T}(x)$

and $f_\varphi = S_\Delta$, where S_Δ is the filtered flame consumption speed. Reference
 195 database is made of the filtered flame archetypes presented in section 2.2.

ii) Parameters α_{CO}^* and $\mathcal{A}_{CO}^*$ relative to reactions (R3)-(R4) are then
 optimized to retrieve the CO production and consumption. The same refer-
 198 ence database is used with a different objective function: $\tilde{\varphi} = \tilde{Y}_{CO}(x)$ and
 $f_\varphi = \text{Max}(Y_{CO})$.

The optimized pre-exponential constant of (R3), written A_3^* , and the
 201 optimized parameter α_{CO}^* obtained when using FPF and FWF flames as
 optimization targets are illustrated in Fig. 4 for a filter size $\Delta = 4\delta_l^0$. Both
 pre-exponential constant and diffusion factor differ rapidly as the sub-grid
 204 scale flame wrinkling increases.

Figure 4: Evolution of A_3^* and α_{CO}^* with the wrinkling factor for $\Delta = 4\delta_l^0$ and $n_\Delta = 2$.
 Legend: - - - COPLES-FPF model. — COPLES-FWF model.

4. Results

4.1. 1-D flame computations

207 To verify the realizability of COPLES modeling, filtered optimized schemes
 trained on FWF and FPF archetypes are used to compute 1-D planar fil-

Figure 5: Evaluation of the performance of the COPLES modeling strategy on 1-D flames for an equivalence ratio $\phi = 0.75$. n_Δ is set to 2 and Δ to $4\delta_l^0$. Legend: — Reference flames. - . - COPLES-FPF model. - - - COPLES-FWF model.

tered premixed flames, wrinkled at the sub-grid scale, at an equivalence ratio $\phi = 0.75$ and with $n_\Delta = 2$. Figure 5 shows that COPLES solutions match as desired the FWF archetypes for different values of wrinkling factor. Note that the case $\Xi_\Delta = 1$ corresponds to a filtered laminar premixed flame. While the flame consumption speed added in Fig. 5 remains well predicted for all values of Ξ_Δ , the COPLES-FPF does not capture the influence of the sub-grid scale flame wrinkling on the filtered flame structure.

216 *4.2. LES of a turbulent premixed burner: the Cambridge configuration.*

The Cambridge SwB burner jointly studied at Cambridge University and Sandia laboratories [21] is selected to challenge the proposed LES combustion
219 model. The burner is composed of a central bluff-body and two concentric annular tubes in which CH_4/air mixtures are flowing. In the retained operating conditions, a mixture with mean velocity $U_i = 8.31$ m/s and equivalence ratio $\phi_i = 0.75$ flows in the inner tube while a mixture of mean velocity $U_o = 18.7$ m/s and equivalence ratio $\phi_o = 0.75$ flows through the outer one. The burner is isolated by a large co-flow with mean velocity $U_{cf} = 0.4$ m/s.
222
225 Details about the burner can be found in [21].

The YALES2 low-Mach number solver [24] is retained to solve the conservation equations. It features a fourth-order finite volume discretization
228 and a fourth-order time integration using an explicit temporal scheme. An unstructured mesh containing 6 Million nodes and with a mesh size $\Delta_x = 0.5$ mm within the flame front is used. The flow field is transported using
231 a standard low Mach formulation for the equations along with the Sigma model [25] for Reynolds sub-grid stresses.

The tested turbulent combustion models tested are the dynamic TFLES
234 [14] and the COPLES models. The virtual chemical schemes described by reactions (R1)-(R2) and (R3)-(R4) are selected for chemistry modeling. COPLES schemes are optimized on either FPF or FWF archetypes, accordingly to section 2.2. A simple model has been derived for n_Δ in [20] by
237 assuming infinitely thin flame front. Recent simulations of the Cambridge configuration performed using tabulated chemistry based on FWF archetypes
240 however showed a small influence of n_Δ on the LES solutions [20]. n_Δ is hence

set constant to 2 in the simulations. The filter size retained to generate the reference database is $\Delta = 4\delta_l^0$, enabling a resolution of 5 nodes across the
243 flame reaction zone. The corresponding TFLES thickening factor is approximately 5 throughout the flame front. The wrinkling factor is computed using the modified Charlette model [26]. In accordance with previous studies of
246 the Cambridge burner [27], the fractal dimension is set to $\beta = 0.5$.

Instantaneous 2-D views of CO mass fraction predicted by both COPLES-FPF and COPLES-FWF simulations are qualitatively compared in Fig. 6
249 (left). The amount of CO mass fractions predicted by COPLES-FPF in the reactive layer remains identical for all axial positions. A similar quantity of CO produced is found in COPLES-FWF simulation in the vicinity of the
252 bluff body. However, the amount of CO predicted by the COPLES-FWF simulation significantly increases in the downstream direction, from 2 cm above the burner exit. This enhancement of CO production is due to the
255 sub-grid scale flame wrinkling, as shown in Fig. 6 (right).

To obtain statistics, filtered temperature and species mass fraction are averaged over 6 flow through times. Means of filtered functions can be compared to means of non-filtered experimental data when the mean flame brush
258 is much higher than the filter size [28]. However, as the mean flame brush is here close to the LES filter size for $z < 30mm$, the mean experimental quantities cannot be directly compared to simulated means. The experi-
261 mental profiles are then first filtered with the LES filter at the filter size $\Delta = 4\delta_l^0$. As thermo-chemical quantities gradients are dominated by radial contributions, filtering is done in 1-D on the radial axis. This approach is
264 justified by the fact that experimental probes averaging volumes are much

smaller than Δ (temperature and major species measurements are limited
 267 by the sampling resolution of $103\ \mu\text{m}$ and the laser beam diameters of $0.22\ \text{mm}$ [21]) and the LES filtering hence predominates. The impact of filtering
 on both mean experimental profiles of temperature and CO mass fraction is
 270 significant as shown in Fig. 7 for axial position $z=10\text{mm}$.

Fig. 7 shows the mean temperature and the mean CO mass fractions at
 different axial positions z in the burner. Agreement on the mean temperature
 273 is fair for any of the turbulent combustion models, with only slight discrepan-
 cies on the flame opening angle. Pronounced differences between models
 are observed for mean $\tilde{Y}_{CO}$. TFLES modeling significantly overestimates the
 276 CO mass fraction at all the axial positions. This overestimation of interme-
 diate species profiles by the thickening operator is inherent to the TFLES
 approach, as discussed in [9, 20]. Filtered Optimized Chemistry improves
 279 significantly the peak of CO prediction. COPLES-FPF and COPLES-FWF
 give similar results below $z=30\ \text{mm}$, where the flame wrinkling is moder-
 ate. However, for $z=40\ \text{mm}$ and $z=50\ \text{mm}$, the introduction of wrinkling in
 282 filtered flame structure improves significantly predictions.

Predicted profiles are however thinner than experimental data. These
 discrepancies may be attributed to differential diffusion effects, significant
 285 in this configuration [29], but which are not captured in the present virtual
 mechanism formulation. Measurements show that differential diffusion im-
 plies an increase of the atomic equivalence ratio from 0.75 to 0.8 across the
 288 flame front [29]. As more CO is produced when $\phi = 0.8$, experimental profiles
 are expected to be higher than simulations, as observed in Fig. 7.

The following post-processing aims at removing the error induced by the

Figure 6: 2-D instantaneous fields on the plane $\theta \equiv 0 [\pi]$. Left: comparison between $\tilde{Y}_{CO}$ field obtained with COPLES-FPF (left) and COPLES-FWF (right) models. Right: comparison between Ξ_{Δ} (left) and $\tilde{Y}_{CO}$ (right) from COPLES-FWF simulations.

omission of differential diffusion effects in the present simulations. The mass of CO per unit of axial length is computed for that purpose in cylindrical coordinates as follows:

$$\mathcal{I}_{CO}^{z=z_0} = \int_{\theta=0}^{2\pi} \left(\int_{r=0}^R r \{ \tilde{Y}_{CO} \} (r, \theta) dr \right) d\theta \quad (6)$$

where $\{.\}$ refers to Favre averaging in time and R is a radius covering the entire reaction zone up to the co-flow where $\tilde{Y}_{CO}$ vanishes. Experiments show that the equivalence ratio variation along r induced by differential diffusion, remains roughly identical at any axial locations [29]. By selecting as a reference the plane located at $z = 10mm$, where the sub-grid scale flame wrinkling is resolved ($\Xi_{\Delta} \approx 1$), the relative mass of CO per unit of axial length is therefore defined as $\delta \mathcal{I}_{CO}^z = \mathcal{I}_{CO}^z - \mathcal{I}_{CO}^{z=10mm}$. $\delta \mathcal{I}_{CO}^z$ is plotted against z for experimental data and each of the simulations in Fig. 8. While TFLES over-estimates the CO integral evolution, the model featuring lami-

Figure 7: Radial profiles of mean $\tilde{Y}_{CO}$. Black circles: filtered experimental data. Red triangles: non filtered experimental data (shown for $z=10\text{mm}$). Dotted lines: TFLES model. Dashed lines: COPLES-FPF. Plain lines: COPLES-FWF.

303 nar premixed flames as optimization targets under-estimates it because the
influence of growing sub-grid scale flame wrinkling is not considered. Adding
the wrinkling influence in the target flame archetype leads to an excellent
306 agreement with experimental data.

5. Conclusion

A new methodology developed to model the influence of sub-grid scale
309 wrinkling on flame structure initially developed in a tabulated chemistry con-
text [20] has been extended to transported chemistry. Conservation equations
are re-written with optimized Arrhenius coefficients and molecular diffusion

Figure 8: Integral of CO mass fractions $\mathcal{I}_{CO}$ in planes (r, θ) as a function of the axial positions z . Legend: $\bullet$ — $\bullet$ Experimental data. $\blacktriangle$ - - $\blacktriangle$ COPLES-FPF model. $\blacklozenge$ - . - $\blacklozenge$ COPLES-FWF model. $\blacksquare$. . . $\blacksquare$ TFLES model.

factor to retrieve a set of reference flames. Two approaches have been studied. In a first case, the optimization is done by targeting laminar premixed flames and including wrinkling with the standard technique. A second possibility is to optimize coefficients on the newly developed 2-D manufactured wrinkled flames proposed in [20]. The simulation of a 3-D turbulent premixed burner has shown a significant advantage of the models based on optimization over a classical TFLES model which tends to over-estimate CO quantities. Moreover, adding the influence of wrinkling thanks to the FWF archetype improves significantly the results quality. Discrepancies observed between LES and experiments are attributed to the omission of preferential species transport in the model, which has a significant influence on the chemical structure of the turbulent flame configuration investigated here [29]. Future work will aim at improving the optimization of molecular diffusion in order to include these complex transport phenomena in the Filtered Optimized

Chemistry model.

327 **Acknowledgments**

This work was performed using HPC resources from GENCI-IDRIS (Grant 2017-A0012B00164).

330 **References**

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