A compact kinetic reaction mechanism for NH₃/H₂ flames

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1. INTRODUCTION

Ammonia's potential as a zero-carbon fuel and hydrogen (H₂) carrier has attracted scientific interest in its application as a fuel in combustion systems. However, harnessing ammonia as a fuel source for energy applications presents notable challenges due to its low flammability and the potential for high emissions. Blending NH₃ with H₂ presents a prospect for enhancing combustibility, albeit with a notable increase in NOx emissions, particularly in fuel-lean conditions. To address these challenges and underscore the underlying factors, a comprehensive understanding and analysis of the chemical kinetics of NH₃ combustion are necessary. The current study focuses on developing a compact and robust kinetic model for smooth Computational Fluid Dynamics (CFD) simulations of NH₃/H₂ flames by optimising the rate parameters of the San Diego 2018 mechanism against a large experimental data collection. This mechanism was selected due to its exceptionally small size (21 species and 64 reactions) and its fair overall performance [1]. The results of the present work are available in a journal publication [2].

2. METHODS

The Optima++ code [3] was used to tune the Arrhenius parameters (**P**) of rate constants within loose uncertainty limits (f=1), to minimize the experimental uncertainty ($\sigma_{fsd}^{\text{exp}}$)- normalized mean square deviation

Table 1: Optimisation targets

(*E*, Eq. (1)) of simulation results from a large collection of laminar burning velocity (LBV) data, concentration data from jet-stirred reactors (JSRs, Zhang et al., Osipova et al.), and concentration data from burner-stabilised stagnation flames (BSSFs, Hayakawa et al.) of 70/30 NH₃/H₂ mixtures (see Table 1).

$$E(\mathbf{P}) = \frac{1}{N} \sum_{f=1}^{N_f} \sum_{s=1}^{N_{fs}} \frac{w_{fsd}}{N_{fsd}} \sum_{d=1}^{N_{fsd}} \left(\frac{Y_{fsd}^{\text{sim}}(\mathbf{P}) - Y_{fsd}^{\text{exp}}}{\sigma_{fsd}^{\text{exp}}} \right)^2$$
 (1)

Tempera ture Method / No. Of Pressure Equivain fuel Measured Quantity data data lence mixture range series points (atm) ratio LBV 185 0-100 0.5-10 0.2-2.0 1311 295-584 JSR conc 0.15-1.5 47 538 10-70 800-1300 BSSF conc 119 30 298 0.57-1.4 Overall 0-100 0.5-10 0.15-2.0

If $\sqrt{E} \le 1$, $\sqrt{E} \sim 2$, $\sqrt{E} < 3$ the model is considered perfect, accurate and satisfactory, respectively.

3. RESULTS

The results in Table 2 show the greatly improved accuracy of the optimized mechanism (Present work, PW) vs. the San Diego 2018 (SD) model in predicting the LBVs (\sqrt{E} : 3.36 \rightarrow 1.97) and BSSF concentrations

(13.91 \rightarrow 3.24). Actually, PW is currently the most accurate model for LBV simulations while featuring the shortest computational time. In BSSF simulations of 70/30 vol% NH₃/H₂ mixtures, it predicts NO emission accurately, while improvement is needed for other species. The performance of the SD model in predicting JSR concentrations slightly deteriorated upon optimization (2.43 \rightarrow 2.72) but \sqrt{E} error value stayed below 3, and PW model can also predict NO and N₂O emissions accurately.

Table 2: Performance of mechanisms in predicting experimental data.

								$\sqrt{E_{ m JSR}}$							$\sqrt{E_{ m BSSF}}$						
#	Mechanism	$N_{ m spec}$	$N_{ m reac}$	$\sqrt{E_{\mathrm{LBV}}}$	$\sqrt{E_{ m JSR}}_{ m V}$	$\overline{E_{\mathrm{BSSF}}}$	$\sqrt{E_{\text{Overall}}}$	NH_3	H_2	O_2	H ₂ O	N_2	NO	N ₂ O	NH ₃	H_2	O ₂	H ₂ O	NO	NO_2	N_2O
1	Zhu 2024		312	2.97	1.11	2.27	2.25	2.1	0.8		1.4	0.6	0.4	0.5	0.9	0.9	1.3	1.9	0.7	1.5	5.2
2	Han 2023	32	171	2.24	1.63	3.70	2.67	2.5	0.8	0.6	1.3	0.9	0.7	2.8	1.1	5.4	0.7	3.9	6.3	3.0	1.0
3	Present work		64	1.97	2.72	3.24	2.70		2.3		3.3	2.7		1.5	0.7	5.6	0.8	3.9	1.5	3.6	3.2
4	Jian 2024		233	3.23	1.80	3.79	3.06	2.2	2.6		1.3		0.5	0.5	1.8	5.1	0.7	4.0	2.8	2.2	6.5
5	Otomo 2018		213	3.67	2.03	3.65	3.21	2.9		2.5	2.3	2.4	0.5	1.0	0.8	5.3	0.8	3.8	5.4	1.9	4.1
6	X. Zhang 2021		224	2.45	2.78	4.59	3.41	3.5		1.3		3.5		3.0	1.3	5.6	0.7	3.8	3.9	2.8	8.7
7	Stagni 2023		203	3.46	1.75	4.69	3.51	2.0	1.3	0.8	1.8	2.1	0.6	2.7	1.5	5.4	0.7	4.0	3.3	2.7	9.4
8	Gotama 2022		165	3.28	2.91	4.59	3.67	2.9		2.1	2.4	1.2	1.2	5.0	1.0	5.6	0.7	3.8	5.1	3.1	8.1
9	Nakamura 2019		229	3.75	2.87	4.71	3.85	3.9		0.9			0.6	1.4	2.7	5.3	0.7	4.0	2.7	2.7	9.4
10	Stagni 2020		203	3.32	3.31	4.90	3.91	2.0		0.5	1.6	1.5		8.1	1.1	5.7	0.7	3.8	7.2	2.8	7.6
11	Liu 2024		238	3.96	2.39	5.19	4.01	2.7		1.4	2.7	1.1	1.2	4.3	1.5	5.5	0.7	3.9	6.2	3.2	9.6
12	Glarborg 2022		227	6.42	2.55	4.45	4.74	3.1		1.2		3.2		2.9	2.1	5.3	0.8	4.1	1.8	2.2	9.0
13	Glarborg 2023		228	6.52	2.54	4.45	4.79	3.1		1.2			1.4	2.9	2.2	5.3	0.8	4.0	1.8	2.2	9.0
14	He 2023		221	7.37	2.46	4.45	5.17	3.2	2.7	1.1	2.7	3.1	1.0	2.4	2.9	4.1	0.8	3.9	2.5	2.3	9.3
15	Z. Zhang 2024		224	8.46	1.14	4.50	5.57	2.1	0.6	1.1	1.5	0.6		0.4	3.1	4.1	0.7	4.0	2.9	2.4	9.2
16	Mei 2021		239	4.02	1.65	9.84	6.21	2.3	2.5	2.7	1.6	2.0	0.6	0.8	17.1	6.1 8.2	9.4 9.3	10.4	1.8	3.6	11.6
17 18	Wang 2022 Tamaoki 2024		140 228	2.53	2.64	10.13 10.17	6.22	3.6		2.7	3.0 1.8	2.3	1.5	0.9	17.2			10.9	5.1	3.8	10.5
19	Meng 2023		269		2.14	4.62	6.29 6.68	2.9		1.6	4.1	2000			17.2	8.2	9.3		2.5	2.4	
	20 Klippenstein 2018		108	10.14	3.03	4.73	6.76	4.1	2.4		4.1	3.6	1.3	3.4 2.8	2.9	5.1 5.1	0.8	4.1 4.2	2.5	2.4	9.3 9.6
21	Glarborg 2018		211		3.03	4.73	6.77	4.1	2.4		4.1	3.6		2.8	2.9	5.1	0.8	4.2	2.9	2.2	9.6
22	San Diego 2018		64		2.43	13.94	8.40	2 8	2.4	1.6	2 1	1.0	0.7	3.4	24.9	10.1	12.3	15.8	0.7	5.5	10.7
44	San Diego 2018	21	04	5.56	2.43	13.94	6.40	4.8	5.1	1.1	5.1	1.0	0.7	5.4	24.9	10.1	12.3	13.8	7./	٥.٥	10.7

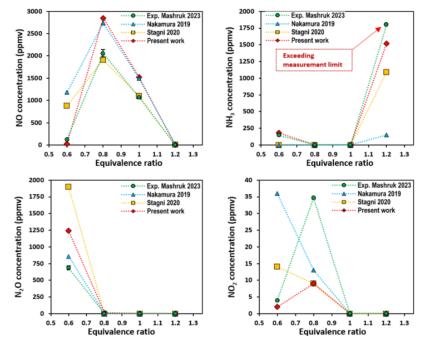


Figure 1: Validation in CFD RANS simulations of a swirl burner.

The PW model was validated in Reynolds Averaged Navier Stokes (RANS) CFD simulations of a swirl burner against experimental data of Mashruk et al. [4] (Fig. 1). It gave improved predictions for all four main emission species and twice faster simulations as the Nakamura 2019 and Stagni 2020 models.

4. CONCLUSIONS

A model with good qualitative and often good quantitative predictions for LBV and NOx emissions has been developed. However, its structural deficiencies limit its accuracy for the prediction of other combustion features.

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