



Supporting Sustainable Aviation Fuels (SAFs) Design and Certification via the *BayeSAF*-based Multi-Fidelity Modeling Framework

J. Liberatori¹, D. Cavalieri¹, M. Blandino^{1*}, M. Valorani¹, P.P. Ciottoli¹

*Lead presenter: matteo.blandino@uniroma1.it

¹ Department of Mechanical and Aerospace Engineering, Sapienza University of Rome
Via Eudossiana 18, 00184, Rome, Italy

Institutions and regulatory agencies have set ambitious targets for achieving mid-century carbon neutrality. As the demand for air transport is expected to double by 2040, Sustainable Aviation Fuels (SAFs) emerge as the near-term solution for decarbonizing commercial aviation. In this context, our research group developed a numerical framework based on the MATLAB[®]-implemented BayeSAF algorithm [1], designed to formulate surrogate mixtures that replicate the physicochemical properties of candidate SAFs leveraging Bayesian inference and Global Sensitivity Analysis (GSA), and meant to support fuel design and pre-screening by facilitating robust numerical analyses of SAF vaporization and combustion. By way of example, the statistical description of SAF surrogates' composition might drive uncertainty quantification (UQ) in reduced-order models (ROMs) for droplet vaporization [2]. Furthermore, the formulation of ad-hoc oxidation mechanisms through our recent development of the MC-HyChem methodology [3] fosters the multi-fidelity investigation of SAF properties' effect on combustor figures of merit (FOM), from low-dimensional reactors to computational fluid dynamics (CFD). In this contribution, we illustrate a comprehensive characterization of the alcohol-to-jet C1 POSF-11498 to demonstrate the potential of the BayeSAF-based framework. Relying on a database of 992 chemical species based on experimental measurements and descriptors-based machine learning (DB-ML), a surrogate fuel composed of two highly-branched C₁₂ and C₁₆ isoparaffins is formulated via the BayeSAF algorithm. Furthermore, experimental measurements of pyrolysis product evolution and ignition delay times support the development of a 130-species MC-HyChem reaction mechanism capturing high- and low-temperature chemistry. Finally, the impact of C1's distinctive physicochemical traits — such as its bimodal distillation curve — is assessed in: (i) a 0D vaporization model accounting for surrogate composition uncertainty via deep long short-term memory (LSTM) recurrent neural networks; (ii) a lab-scale swirled spray flame combustor using a multi-mixture-fraction Flamelet-Progress Variable (FPV) approach to capture preferential vaporization effects.

References

- [1] Liberatori, J. et al.. *BayeSAF: Emulation and Design of Sustainable Alternative Fuels via Bayesian Inference and Descriptors-Based Machine Learning*. Fuel (under Review).
- [2] Liberatori, J. et al.. *Deep learning-based forward UQ on Bayes-informed sustainable aviation fuel surrogates: a case study on bio-jet fuel vaporization*. International Journal of Heat and Mass Transfer (under Review).
- [3] Blandino, M. et al.. *Multicomponent HyChem Kinetic Mechanism Generation Using Trust-Region Bayesian Optimization*. MCS13 Corfu, Greece, June 1–5, 2025.