

Global Sensitivity Analysis of Fields: A Demonstration for Hydrogen Autoignition

Jonathan Wang*, Ph.D. in Theoretical & Applied Mechanics, est. May 2019

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July 27, 2016

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Abstract

Complex chemical mechanisms in combustion models typically involve numerous reactions with uncertain reaction rates; global sensitivity analysis (GSA) facilitates the propagation of uncertainty by reducing the dimensionality of the parameter space. GSA has thus far been focused on scalars; in this work, we employ the Karhunen-Loève expansion (KLE) and polynomial chaos expansion (PCE) to compute Sobol' sensitivity indices for a field. We demonstrate the approach for a simplified problem that mimics autoignition and emphasize that field GSA provides important information that scalar GSA does not. Applications to other combustions systems are also discussed.

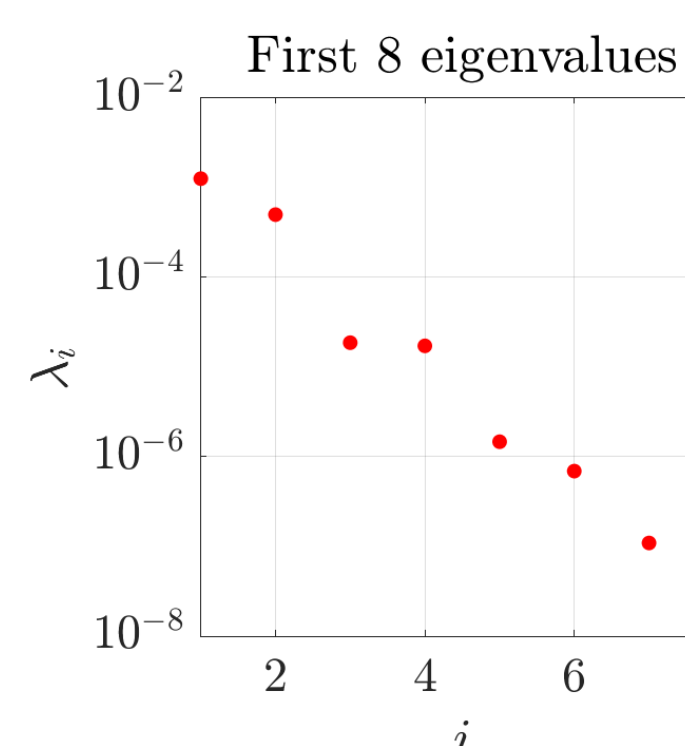
Background

- Because propagating uncertainty through large reacting flow simulations is often infeasible with detailed chemical mechanisms, reduced models are typically used.
- Field GSA exposes the nature of the model reduction or neglected uncertainty in a way that scalar GSA does not; for example, it can highlight the region of a flame in which a particular reaction introduces significant uncertainty.
- Field GSA poses a computational and technical challenge: naïvely building PCEs separately at each physical point is expensive and raises questions of how to represent the correlation structure between points.
- The KLE addresses these challenges by providing a compact representation of a stochastic process that preserves correlation structure and establishing a framework in which Sobol' sensitivity indices can be computed from PCEs.

Karhunen-Loève expansion (KLE):

- The KLE modes decay exponentially, and the KLE is able to recover the covariance kernel with only the first two terms.
- However, accurately representing the full stochastic behavior of F —in particular, the joint PDFs—requires additional terms.

$$F(x; \vec{\xi}) = \frac{a}{2} \tanh(r(x - b) + c); \quad b \sim \mathcal{N}(0.5, 0.02); \quad r \sim \mathcal{N}(10, 2)$$



- The KLE compresses field data by projection onto a deterministic and stochastic basis.

$$F(x, \vec{\xi}) = \sum_{n=1}^N \sqrt{\lambda_n} \eta_n(\vec{\xi}) u_n(x)$$

- $u_n(x)$ are orthogonal eigenvectors (deterministic).
- $\eta_n(\vec{\xi})$ are uncorrelated random variables (stochastic).

- The PCE efficiently represents a random variable using a smooth distribution.

$$\eta_n(\vec{\xi}) = \sum_{k=0}^{P-1} \eta_{nk} \Psi_k(\vec{\xi})$$

- $\vec{\xi}$ are (Gaussian) iid germs.
- $\Psi_k(\vec{\xi})$ are (Hermite) polynomials.
- η_{nk} are computed by projection with Monte Carlo integration.

- Sobol' indices quantify the relative contribution of uncertainty from each parameter.

$$\mathcal{S}_i(x) \equiv \frac{V(E(F|\xi_i))}{V(F)} = \frac{\sum_{\alpha \in \mathbb{I}_i} F_\alpha(x) \langle \Psi_\alpha^2 \rangle}{\sum_{\alpha} F_\alpha(x) \langle \Psi_\alpha^2 \rangle}, \quad \text{where}$$

$$F(x, \vec{\xi}) = \sum_{k=0}^{P-1} F_k(x) \Psi_k(\vec{\xi})$$

$$F_k(x) = \sum_{n=1}^N \sqrt{\lambda_n} \eta_{nk} u_n(x)$$

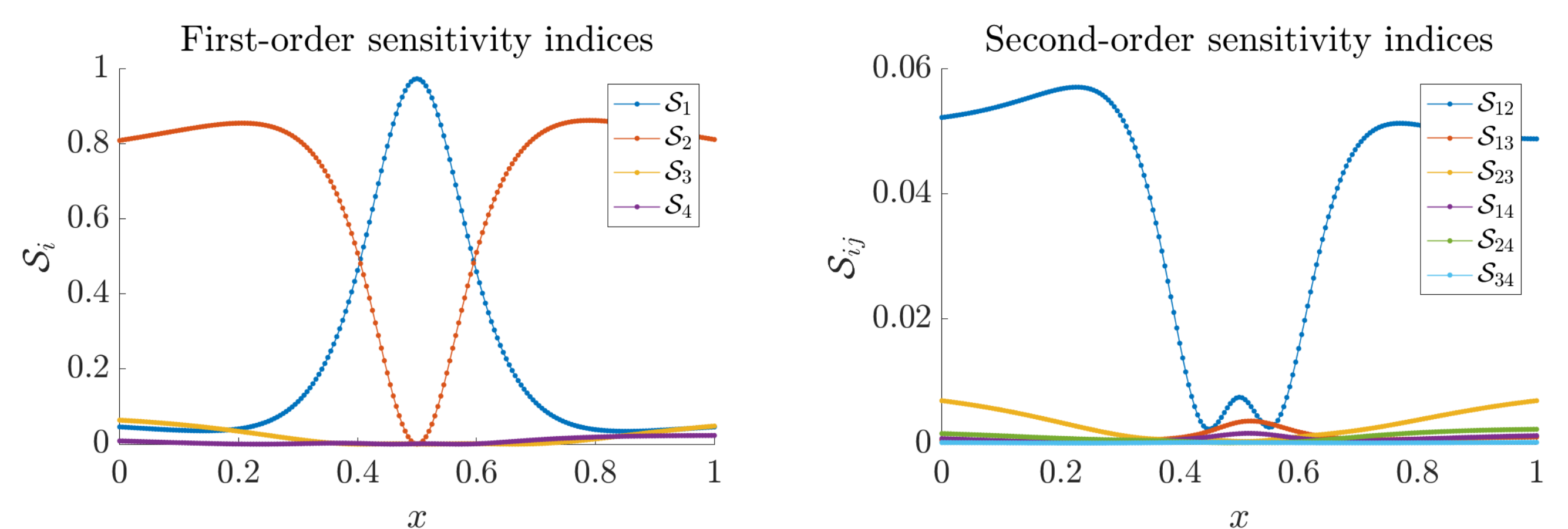
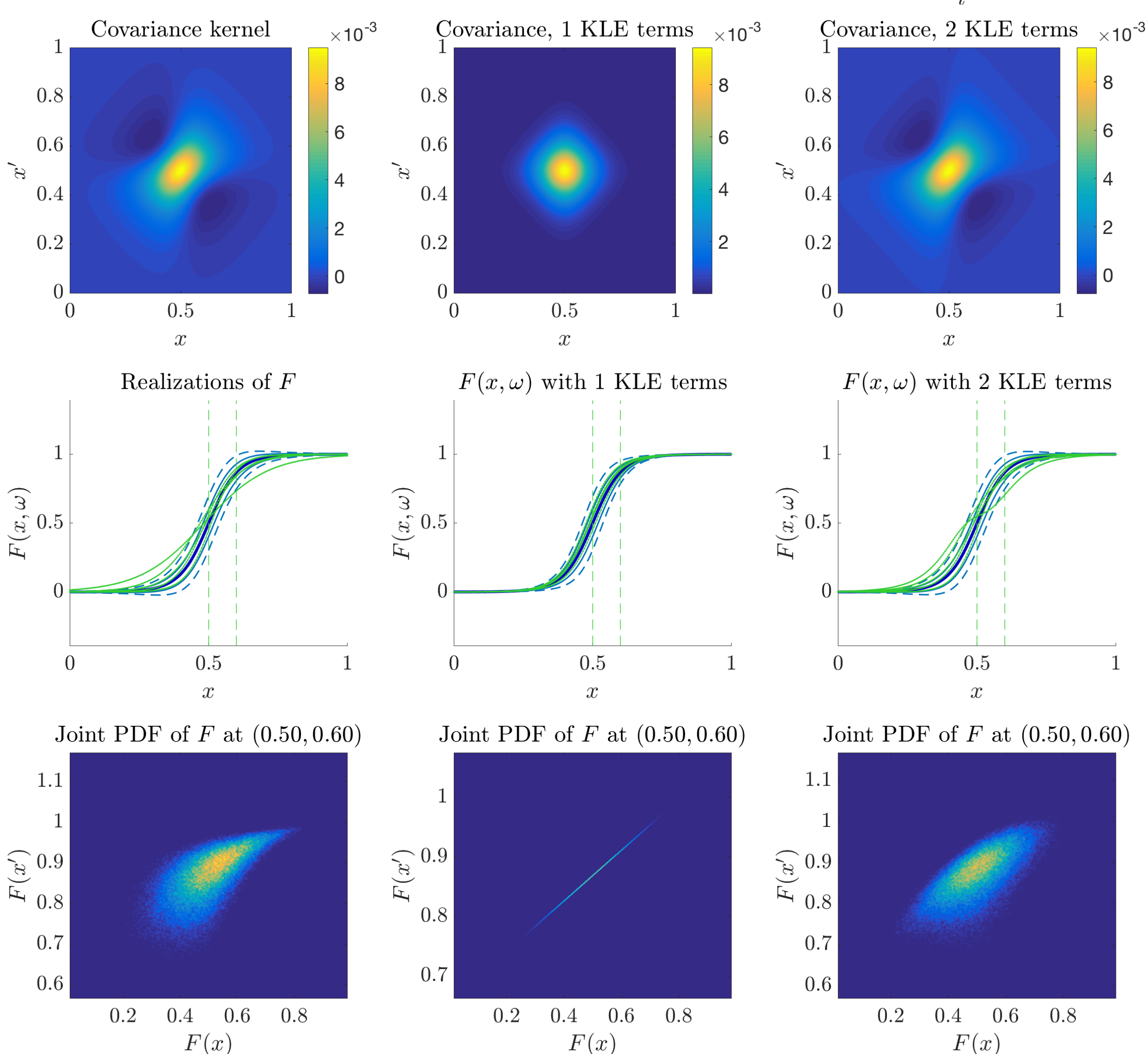
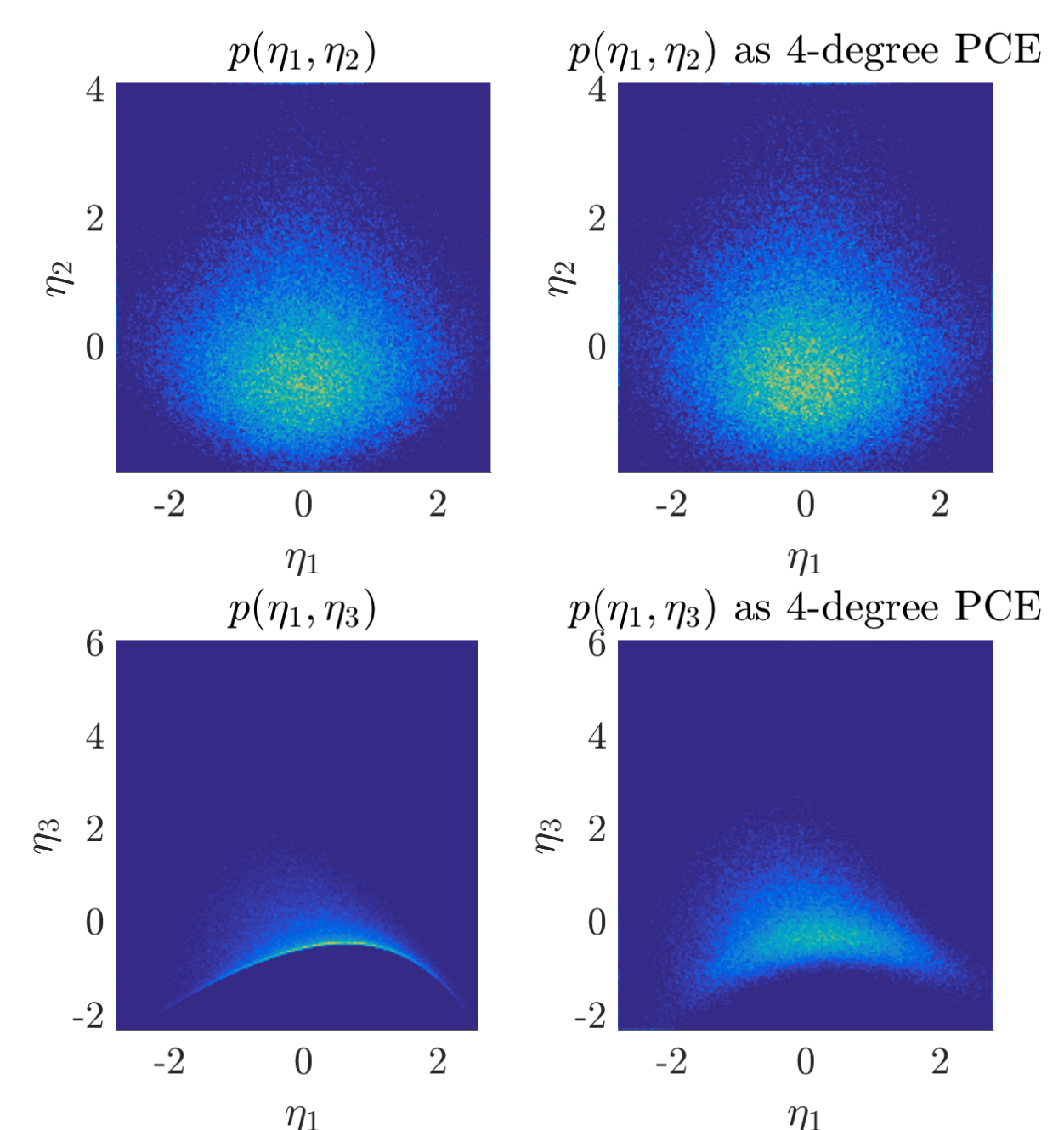
Results

Polynomial chaos expansion (PCE):

- Expanding a process with two independent stochastic inputs results in two independent KL modes.
- The sharp features of the PDFs are difficult for the PCE to resolve and suggest a strong dependence between higher KLE modes. It is yet unclear what that relationship is or how it should be interpreted.

Sobol' sensitivity indices:

- b controls the location of the inflection point.
- r controls the slope of the profile.



Conclusions and Future Work

- This approach for field GSA is demonstrated for a simplified problem, but deciding where to truncate the expansion requires further investigation into the dependence among the KLE modes.
- Field GSA is more informative than scalar GSA in that it indicates not only the parameters that contribute most to the uncertainty of a prediction, but also where that contribution takes place in the physical domain.
- This method is now being applied to the temperature-time profile of an autoignition simulation using a reduced hydrogen combustion mechanism.
- Future work includes application to 2D and 3D multiphysics simulations to study plasma-assisted combustion.

