

# Cooperative game theory and importance quantification

Marouane El Idrissi<sup>123</sup>, Nicolas Bousquet<sup>234</sup>, Fabrice Gamboa<sup>1</sup>, Bertrand Iooss<sup>123</sup>,  
Jean-Michel Loubes<sup>1</sup>

<sup>1</sup>*Institut de Mathématiques de Toulouse, Université Toulouse III.*

<sup>2</sup>*EDF R&D, Lab Chatou.*

<sup>3</sup>*SINCLAIR AI Laboratory.*

<sup>4</sup>*LPSM, Sorbonne Université.*

## Abstract

Being able to quantify the importance of random inputs of an input-output black-box model is at the cornerstone of the fields of sensitivity analysis (SA) and explainable artificial intelligence (XAI). To perform this task, methods such as *Shapley effects* and *SHAP* have received a lot of attention. The former offers a solution for output variance decomposition with non-independent inputs, and the latter proposes a way to decompose predictions of predictive models. Both of these methods are based upon the *Shapley values*, an allocation mechanism from the *cooperative game theory*.

This presentation aims at shedding light on the underlying mechanism behind the paradigm of cooperative games for input importance quantification. To that extent, a link is drawn between an extension of the *Möbius inversion formula* to boolean lattices leading to coalitional decompositions [1]. Allocations can be seen as aggregations of such decomposition, leading to a more general view of the importance quantification problem.

This generalization is leveraged in order to solve a problem in the context of global SA with dependent inputs. The Shapley effects are known not to be able to detect exogenous inputs (i.e., variables not in the model). Using a different allocation, namely the *proportional values*, lead to interpretable importance indices with the ability to identify such inputs [2].

These indices are illustrated on a classical problem of surrogate modeling of a costly numerical model: the transmittance performance of an optical filter. It allows for clear and interpretable decision rules for feature selection and dimension reduction.

**Keywords:** Uncertainty quantification ; Sensitivity analysis ; Interpretability ; Importance quantification

**AMS subject classifications:** Primary: 62P30 ; Secondary: 91A12, 62H99

## References

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