# Sensitivity Analysis - Exercises EMSE, Major "Data Science"

O. Roustant, January 2015

In the following, we assume that  $X_1, \ldots, X_d$  are independent random variables with probability measures  $\nu_1, \ldots, \nu_d$ . We denote:

- $X = (X_1, \dots, X_d);$
- $\nu = \nu_1 \otimes \ldots \otimes \nu_d$  the probability measure of X;
- $\Delta = \Delta_1 \times \ldots \times \Delta_d$ , the integration domain;
- $m_i^{(1)} = E(X_i), m_i^{(2)} = E(X_i^2)$ : the first and second order moments.

### 1 Additive functions - 1st order polynomials & SRCs.

Consider an additive function:

$$f(x) = \beta_0 + f_1(x_1) + \ldots + f_d(x_d),$$

where the  $f_i(X_i)$ 's are centered (with respect to the measure  $\mu_i$ ) and square-integrable.

- 1. What should be the ANOVA decomposition of f? Prove it and compute all Sobol indices.
- 2. Deduce from 1 the ANOVA decomposition of a first order polynomial:

$$f(x) = \beta_0 + \beta_1 x_1 + \ldots + \beta_d x_d$$

and all Sobol indices. The results can be expressed in function of the  $m_i^{(1)}$ 's.

3. Deduce that the Sobol indices of a 1st order polynomial are equal to the squared SRCs.

# 2 Models with product interactions

Consider a 2-dimensional function of the form:

$$f(x) = \beta_0 + f_1(x_1) + f_2(x_2) + \lambda g_1(x_1)g_2(x_2)$$

where the  $f_i(X_i)$ 's and  $g_i(X_i)$ 's are centered and square-integrable.

- 1. What should be the ANOVA decomposition of f? Prove it and compute all Sobol indices.
- 2. Deduce from 1 the ANOVA decomposition of a first order polynomial with interactions:

$$f(x) = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_{12} x_1 x_2$$

as well as the Sobol indices in the following cases:

- (a)  $X_1$  and  $X_2$  are centered;
- (b) In general: Express the results in function of the  $m_i^{(1)}$ 's.
- 3. Same question with a second order polynomial with interactions. Then express the results also with the  $m_i^{(2)}$ 's.

#### 3 A 3-dimensional test-function

The *Ishigami* function is defined over  $\Delta := [-\pi; \pi]^3$  by:

$$f(x) = \sin(x_1) + A\sin^2(x_2) + Bx_3^4\sin(x_1)$$

with A = 7 and B = 0.1.

- 1. Visualize the function, by cutting it at different levels of  $x_2$ .
- 2. Compute the ANOVA decomposition and Sobol indices when  $\mu$  is uniform over  $\Delta$ .
- 3. Estimate the main effects by using simulations of the inputs. Plot them on the same figure and compare with the theoretical ones. Same question with the 2-dimensional projection over  $(x_1, x_3)$  and the second order interaction  $f_{1,3}(x_1, x_3)$ .

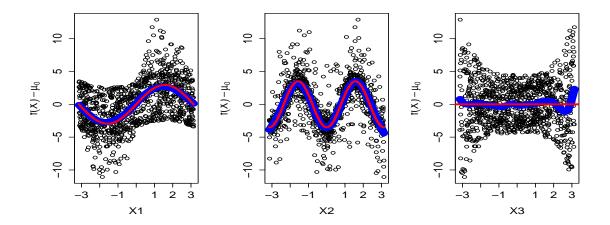


Figure 1: Main effects of Ishigami function: Theoretical (straight line) and estimated (bold line)

# 4 1-mean separable functions

Consider a separable (or "tensor-product") function of the form:

$$f(x) = \prod_{i=1}^{d} (1 + f_i(x_i))$$

where the  $f_i(X_i)$ 's are centered and square-integrable.

- 1. What is the ANOVA decomposition of f? Compute all Sobol indices.
- 2. Considering the uniform measure over  $[0,1]^d$ , deduce from 1 the ANOVA decomposition and the Sobol indices of the famous g-Sobol function:

$$f(x) = \prod_{i=1}^{d} \frac{|4x_i - 2| + a_i}{1 + a_i}$$

where the  $a_i$ 's are non-negative real numbers.

3. Let d=2, and  $a_1=1, a_2=5$ . Perform simulations of the inputs (denote them  $X^{*k}$ ,  $k=1,\ldots,N$ ), and estimate the main effects. Draw them on the same figure, and compare with the theoretical ones.

### 5 Closed and total indices. Pick-and-freeze formulas (Sobol)

In this exercise, we consider a function f such that f(X) is in  $L^2(\nu)$ , and denote by  $\sigma^2 = \text{var}(f(X))$  the overall variance,  $D_I = \text{var}(E[f(X)|X_I])$ , and  $S_I = D_I/\sigma^2$  the Sobol indices.

1. Prove that the un-normalized Sobol index of  $X_1$  is given by:

$$D_1 = \mathbf{cov}(f(X_1, X_{-1}), f(X_1, Z_{-1})),$$

where  $Z_{-1}$  is an independent copy of  $X_{-1}$  (same distribution and independent of the  $X_i$ 's). Hint: Conditionaly to  $X_1$ , what can you say of  $f(X_1, X_{-1})$  and  $f(X_1, Z_{-1})$ ?

2. Deduce that:

$$D_1 = \int_{\Delta \times \Delta_{-1}} f(x_1, x_2, \dots, x_d) f(x_1, z_2, \dots, z_d) d\nu(x) d\nu_{-1}(z_{-1}) - (\mu_0)^2$$

Explain how to compute numerically  $D_1$  and  $S_1$ . Justify the word "pick-and-freeze".

- 3. Generalize to "closed indices"  $D_I^C := \sum_{J \subset I} D_J$  for a subset I.
- 4. Explain how Sobol indices are obtained from closed indices: For instance, express  $D_{1,2}$  as a linear combination of closed indices.
- 5. Prove the pick-and-freeze formula for "total indices",  $D_i^T = \sum_{J \supset \{i\}} D_J$ :

$$D_1^T = \frac{1}{2} \mathbf{var}(f(X_1, X_{-1}) - f(Z_1, X_{-1})),$$

or equivalently:

$$D_1^T = \frac{1}{2} \int_{\Delta \times \Delta_1} [f(x_1, x_2, \dots, x_d) - f(z_1, x_2, \dots, x_d)]^2 d\nu(x) d\nu_1(z_1)$$

Hint: Prove first that  $D_1^T = \sigma^2 - D_{-1}^C$ .

6. Generalize to a subset I, with  $D_I^T := \sum_{J \cap I \neq \emptyset} D_J$ .

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### 6 Computer lab

The aim of the lab is to perform a global SA of the 3-dimensional Ishigami (toy) function, directly on it, or on a Kriging metamodel.

- 1. Consider the Ishigami function of Exercise 3. Perform question 3.
- 2. Estimate the Sobol indices of Ishigami, with R package sensitivity (function fast99).
- 3. In this question we build a Kriging metamodel for Ishigami. We first consider a design size of n = 30.
  - (a) Build a maximin Latin hypercube design X with **DiceDesign** (function maximinESE\_LHS). Visualize the space-filling properties (compare visually with the uniform design serving as a initial design for maximinESE\_LHS).
  - (b) Compute the response y at X, and construct a Kriging model with **DiceKriging**. Look at the validation results.
- 4. Perform a sensitivity analysis on the Kriging metamodel. For that goal, use sensitivity::fast99 function on the following wrapper:

```
kriging.mean <- function(Xnew, m) {
predict(m, Xnew, "UK", se.compute = FALSE, checkNames = FALSE)$mean
}
where m is the name of the metamodel object of question 3.b.</pre>
```

- 5. Compare the results of 2 and 4.
- 6. Redo questions 3-5 with n = 60, 90. Conclusion?