

# Global Sensitivity Analysis

**Reading:** Chapter 15

**Example:** Portfolio model

$$Y = c_1 Q_1 + c_2 Q_2$$

**Note:**

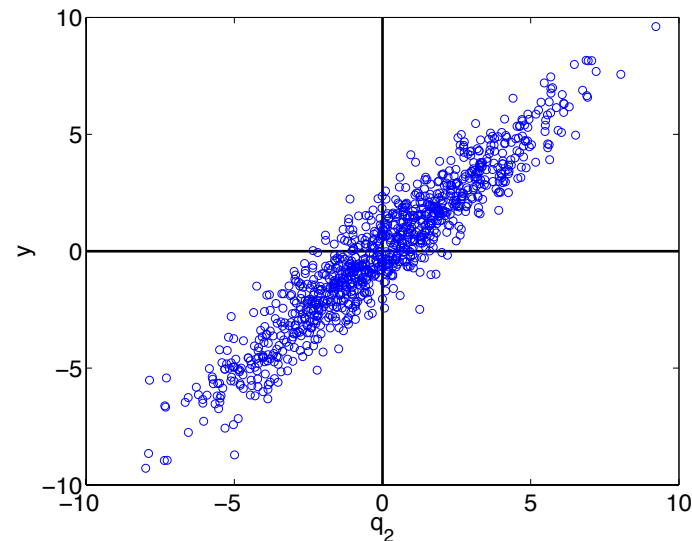
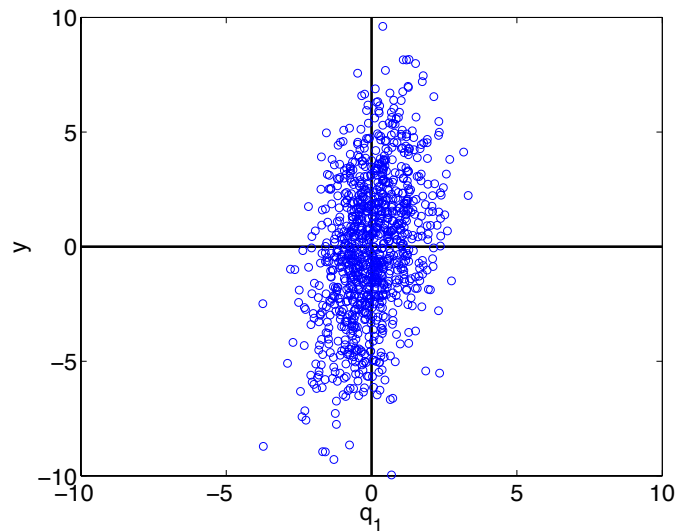
- $Q_1$  and  $Q_2$  represent hedged portfolios
- $c_1$  and  $c_2$  amounts invested in each portfolio
- $\sigma_Y^2 = c_1^2 \sigma_1^2 + c_2^2 \sigma_2^2 = 13$

Take

$$c_1 = 2, c_2 = 1$$

$$Q_1 \sim N(0, \sigma_1^2) \text{ with } \sigma_1 = 1$$

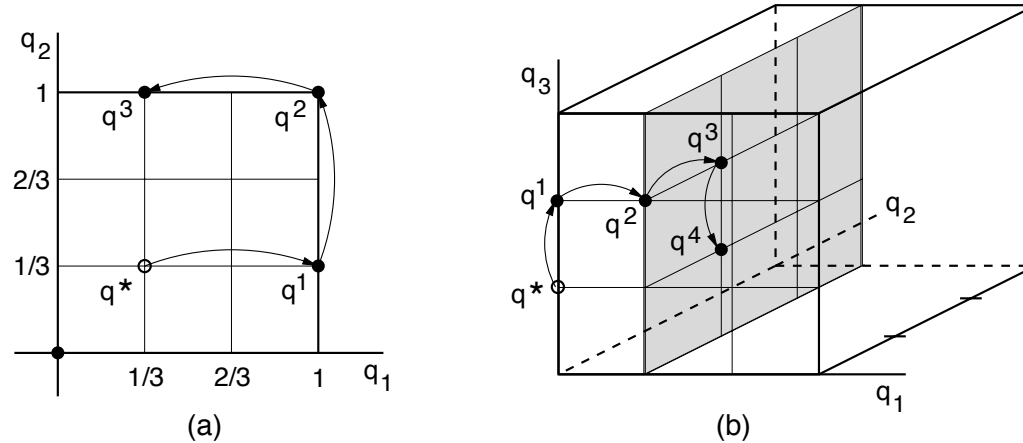
$$Q_2 \sim N(0, \sigma_2^2) \text{ with } \sigma_2 = 3$$



Local Sensitivity:  $s_i \equiv \frac{\partial Y}{\partial Q_i} \Rightarrow s_1 = 2 > s_2 = 1$

# Morris Screening

**Example:** Consider uniformly distributed parameters on  $\Gamma = [0, 1]^p$



## Elementary Effect:

$$d_i^j = \frac{f(q^j + \Delta e_i) - f(q^j)}{\Delta} \quad \begin{array}{l} i^{th} \text{ parameter} \\ j^{th} \text{ sample} \end{array}$$

$$\Delta \in \left\{ \frac{1}{\ell - 1}, \dots, 1 - \frac{1}{\ell - 1} \right\} \quad \ell \text{ is level}$$

## Global Sensitivity Measures: $r$ samples

$$\mu_i^* = \frac{1}{r} \sum_{j=1}^r |d_i^j(q)|$$

$$\sigma_i^2 = \frac{1}{r-1} \sum_{j=1}^r \left( d_i^j(q) - \mu_i \right)^2, \quad \mu_i = \frac{1}{r} \sum_{j=1}^r d_i^j(q)$$

# Morris Screening

**Strategy:** For each index  $j = 1, \dots, r$ , sample a seed values  $q^* \in \rho_Q(q)$  and specify  $p + 1$  parameter values required to approximate  $p$  elementary effects using random orientation matrix

$$B^* = [J_{p+1,1} q^* + \frac{\Delta}{2} [(2B - J_{p+1,p}) D^* + J_{p+1,p}] P^*$$

- $D^*$ :  $p \times p$  diagonal matrix with elements in  $\{-1, 1\}$
- $P^*$ : Randomly permute columns of  $p \times p$  identity
- $B$ :  $(p + 1) \times p$  strictly lower triangular matrix of ones
- $J$ :  $(p + 1) \times p$  matrix of ones

**Example:**  $k = 2$ ,  $\ell = 4$ ,  $\Delta = 2/3$ . Seed Value:  $q^* = [1/3, 1/3]$

$$D^* = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}, P^* = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, B = \begin{bmatrix} 0 & 0 \\ 1 & 0 \\ 1 & 1 \end{bmatrix}, J = \begin{bmatrix} 1 & 1 \\ 1 & 1 \\ 1 & 1 \end{bmatrix}$$

Random Orientation Matrix:

$$B^* = \begin{bmatrix} 1 & 1/3 \\ 1 & 1 \\ 1/3 & 1 \end{bmatrix}$$

**Issue:** Density choice

# Variance-Based Methods

**Sobol Representation:** For now, take  $q_i \sim \mathcal{U}(0, 1)$  and  $\Gamma = [0, 1]^p$

Take

$$f(q) = f_0 + \sum_{i=1}^p f_i(q_i) + \sum_{1 \leq i < j \leq p} f_{ij}(q_i, q_j)$$

subject to

$$\int_0^1 f_i(q_i) dq_i = \int_0^1 f_{ij}(q_i, q_j) dq_i = \int_0^1 f_{ij}(q_i, q_j) dq_j = 0$$

to ensure

$$\int_{\Gamma} f_i(q_i) f_j(q_j) dq_i dq_j = \int_{\Gamma} f_i(q_i) f_{ij}(q_i, q_j) dq_i dq_j = 0$$

Then

**Notation:**  $q_{\sim i} = [q_1, \dots, q_{i-1}, q_{i+1}, \dots, q_p]$

$$f_0 = \int_{\Gamma} f(q) dq$$

$$f_i(q_i) = \int_{\Gamma^{p-1}} f(q) dq_{\sim i} - f_0$$

$$f_{ij}(q_i, q_j) = \int_{\Gamma^{p-2}} f(q) dq_{\sim \{i,j\}} - f_i(q_i) - f_j(q_j) - f_0$$

# Variance-Based Methods

## Notation:

$$\mathbb{E}(Y|q_i) = \int_{\Gamma^{p-1}} f(q) dq_{\sim i}$$

$$\mathbb{E}(Y|q_i, q_j) = \int_{\Gamma^{p-2}} f(q) dq_{\sim \{ij\}}$$

## Note:

$$f_0 = \mathbb{E}(Y)$$

$$f_i(q_i) = \mathbb{E}(Y|q_i) - f_0$$

$$f_{ij}(q_i, q_j) = \mathbb{E}(Y|q_i, q_j) - f_i(q_i) - f_j(q_j) - f_0.$$

## Total Variance:

$$D = \text{var}(Y) = \int_{\Gamma} f^2(q) dq - f_0^2$$

## Partial Variances:

$$D_i = \int_0^1 f_i^2(q_i) dq_i$$

$$D_{ij} = \int_0^1 \int_0^1 f_{ij}^2(q_i, q_j) dq_i dq_j.$$

## Sobol Indices:

$$S_i = \frac{D_i}{D}, \quad S_{ij} = \frac{D_{ij}}{D}, \quad i, j = 1, \dots, p$$

$$S_{T_i} = S_i + \sum_{j=1}^p S_{ij}$$

# Variance-Based Methods

**Variance Interpretations:** Since

$$\mathbb{E}[\mathbb{E}(Y|q_i)] = \int_0^1 \left[ \int_{\Gamma^{p-1}} f(q) dq_{\sim i} \right] dq_i = f_0,$$

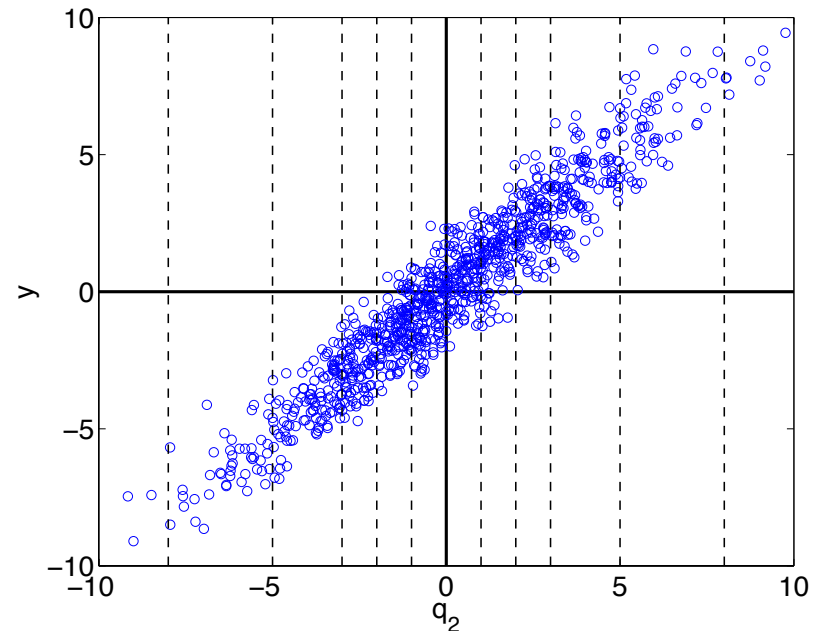
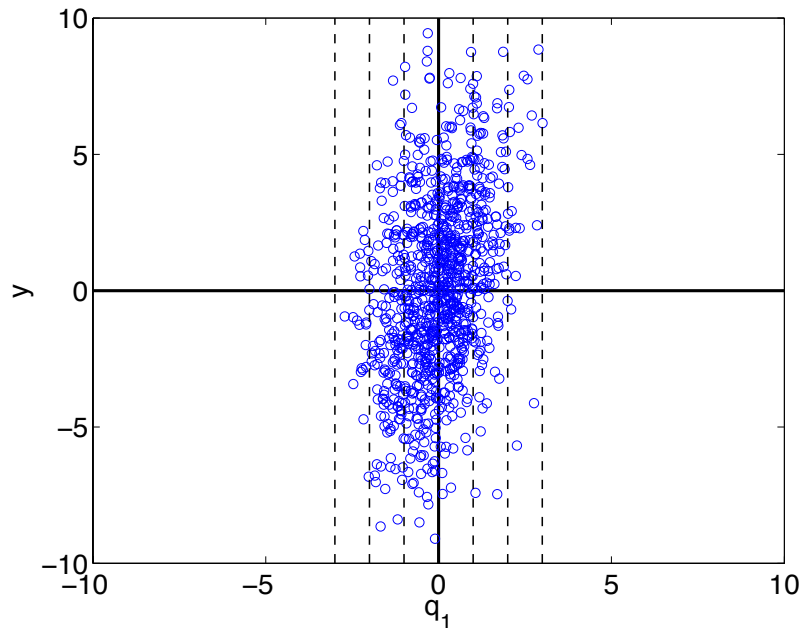
it follows that

$$D_i = \text{var}[\mathbb{E}(Y|q_i)] \Rightarrow S_i = \frac{\text{var}[\mathbb{E}(Y|q_i)]}{\text{var}(Y)}$$

Similarly

$$S_{T_i} = 1 - \frac{\text{var}[\mathbb{E}(Y|q_{\sim i})]}{\text{var}(Y)} = \frac{\mathbb{E}[\text{var}(Y|q_{\sim i})]}{\text{var}(Y)}$$

**Note:** Implementation algorithm discussed in Section 15.1.3.



# Global Sensitivity Analysis

**Example:** Sobol function

$$Y = \prod_{i=1}^p g_i(Q_i) \quad , \quad g_i(Q_i) = \frac{|4Q_i - 2| + a_i}{1 + a_i}$$

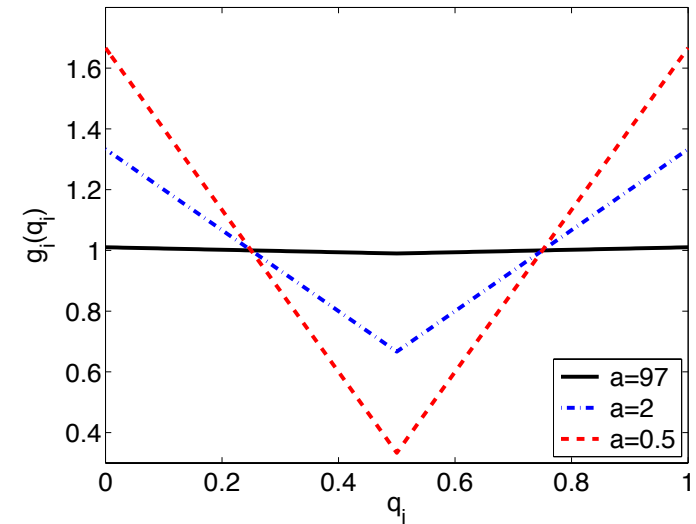
where  $a_i \geq 0$  are fixed, deterministic coefficients that determine relative importance of parameters.

**Note:** For  $Q_i \sim \mathcal{U}(0, 1), i = 1, \dots, p$ .

$$D_i = \text{var}[\mathbb{E}(Y|q_i)] = \frac{1}{3(1 + a_i)^2}$$

$$D_{ij} = \text{var}[\mathbb{E}(Y|q_i, q_j)] - D_i - D_j = D_i D_j$$

$$D = \text{var}(Y) = -1 + \prod_{i=1}^p (1 + D_i)$$



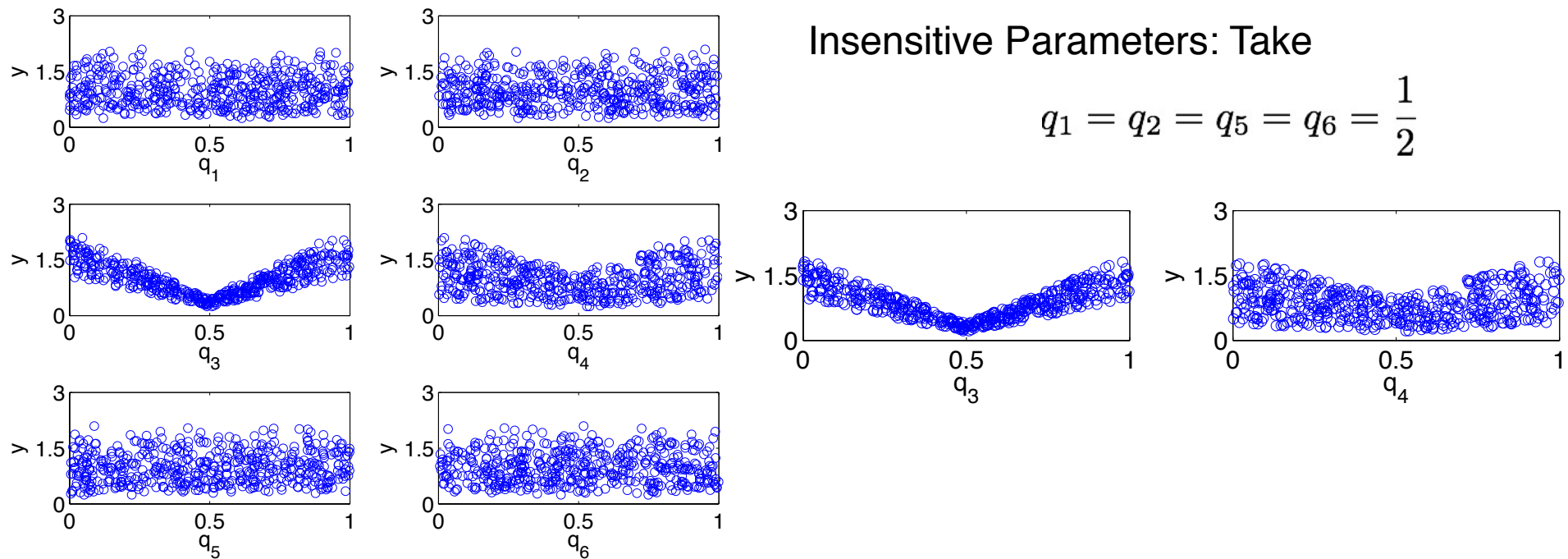
# Global Sensitivity Analysis

Sobol Indices:

	$Q_1$	$Q_2$	$Q_3$	$Q_4$	$Q_5$	$Q_6$
$a_i$	<b>78</b>	<b>12</b>	<b>0.5</b>	<b>2</b>	<b>97</b>	<b>33</b>
$D_i$	$5.3 \times 10^{-5}$	$2.0 \times 10^{-3}$	$1.5 \times 10^{-1}$	$3.7 \times 10^{-2}$	$3.5 \times 10^{-5}$	$2.9 \times 10^{-4}$
$S_i$	$2.8 \times 10^{-4}$	$1.0 \times 10^{-2}$	$7.7 \times 10^{-1}$	$1.9 \times 10^{-1}$	$1.8 \times 10^{-4}$	$1.5 \times 10^{-3}$
$S_{T_i}$	$3.3 \times 10^{-4}$	$1.2 \times 10^{-2}$	$8.0 \times 10^{-1}$	$2.2 \times 10^{-1}$	$2.1 \times 10^{-4}$	$1.8 \times 10^{-3}$

Morris Indices: With  $\ell = 4, \Delta = \frac{2}{3}, r = 4$

	$Q_1$	$Q_2$	$Q_3$	$Q_4$	$Q_5$	$Q_6$
$\mu_i$	-0.006	-0.078	-0.130	-0.004	0.012	-0.004
$\mu_i^*$	0.056	0.277	1.760	1.185	0.035	0.099
$\sigma_i$	0.064	0.321	2.049	1.370	0.041	0.122





# Global Sensitivity Analysis

**Example:** Spring model

$$m \frac{d^2 z}{dt^2} + kz = 0$$

$$z(0) = 1, \quad \frac{dz}{dt}(0) = 0$$

Responses: For  $q = [k, m]$ , consider

$$y = f(q) = \cos \left( \sqrt{\frac{k}{m}} \cdot \frac{\pi}{2} \right)$$

$$y = \int_0^{\pi/2} \cos \left( \sqrt{\frac{k}{m}} t \right) dt = \sqrt{\frac{m}{k}}$$