

Big Data Technology and Application Institute

SENSITIVITY ANALYSIS **OF COMPUTER MODELS**

Speaker : XinLei Zhou

Outline

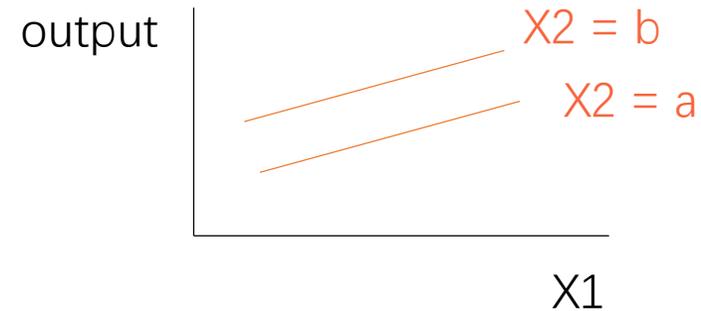
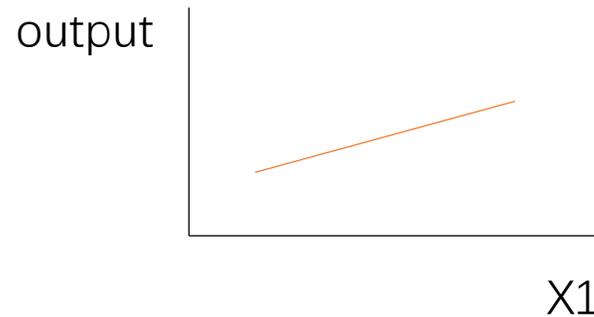
- ▶ **Introduction**
- ▶ **Function**
- ▶ **Method**

Introduction

Sensitivity analysis is a research method to measure the influence of changes in input parameters, initial conditions and boundary conditions on model output

■ simple model

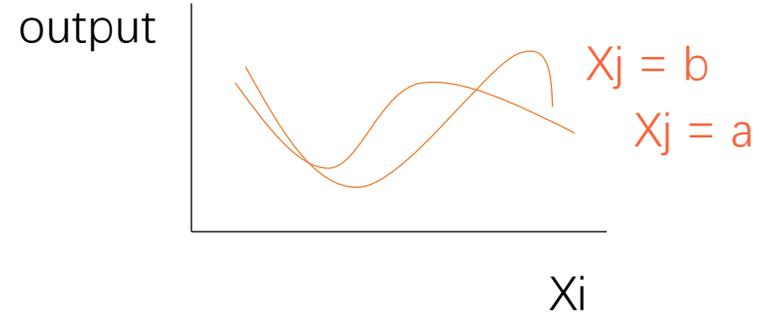
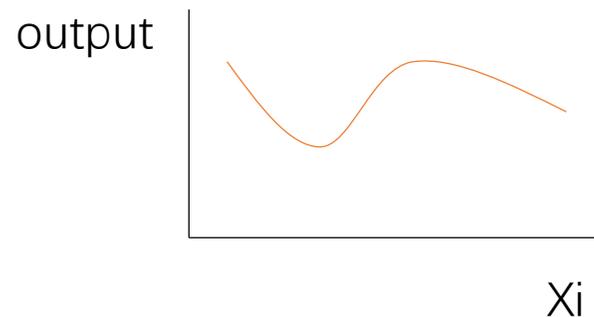
X_1 / X_2



It's easy to conclude Whether the model is sensitive to this parameter

■ complex model

$X_1 / X_2 / \dots / X_n$



It's hard to draw a conclusion directly about the influence and importance of each parameters

Function

- simplify model

Get the relative importance of each parameter and find the redundant parameters

- Increasing the reliability of the model

Analysis different parameter combinations to obtain the influence of parameters interaction

Method

Sensitivity analysis is divided into **local sensitivity analysis** and **global sensitivity analysis**

- **local sensitivity analysis**

Only one parameter at a time is analyzed and ignore the interaction between parameters

Such as : One-variable-at-a-time approach (OAT)

- **global sensitivity analysis**

Consider the effect on the results when the combination of parameters changes, and analyze the interaction between the parameters.

Such as : Regression analysis (RA) / Morris screening

OAT

- One-variable-at-a-time approach

The basic principle is to calculate the **change rate** of the output caused by small changes (eg increase or decrease by 10%) of each parameter near its best estimate.

The absolute value of the **change rate** represents the sensitivity of the parameter.

The sensitivity ranking of each parameters can be obtained easily

MS

■ Morris screening

Select one of the variables in the model and add **a tiny disturbance** within the range, while other variables remain unchanged, evaluate the output response caused by small changes in the variable

$$d_i(j) = (y^* - y) / \Delta_i$$

- i: The ith parameter (i=1...n)
- j: The jth factor combinations (j=1...t)
- y: Output value before parameter i changed
- y*: Output value after parameter i changed
- Δ_i : the amplitude of parameter i

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

The mean value is used to indicate the sensitivity and the sequence of the parameter.

The larger the value , the stronger the sensitivity of the parameters.

$$\sigma_i = \sqrt{\frac{1}{t-1} \sum_{j=1}^t [d_i(j) - \mu_i]^2}$$

The standard deviation is used to indicate the strength of the interaction between the parameters. The higher the value , the stronger the interaction of the parameters.

RA

■ Regression analysis

$$Y_i = \beta_0 + \sum_{h=1}^k \beta_h x_{i,h} + E_i \quad (i = 1, \dots, n)$$

n : number of simulated factor combinations.

E_i : fitting error of the regression model for factor combination i

Y_i : simulation response of factor combination i

β_0 : regression intercept

β_h : effect of factor h

$x_{i,h}$: value of the standardized factor h in combination i $x_{ih} = (z_{ih} - b_h)/a_h$

z_{ih} = value of non-standardized factor h $a_h = (u_h - l_h)/2$

Set the range of lowest value l_h , and upper value u_h of z_{ih} $b_h = (u_h + l_h)/2$

The relative importance of a factor is obtained by sorting the absolute values of the effects β_h

THANKS FOR

YOUR ATTENTION