

Design of Experiments and Sensitivity analysis

Course and practical application

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OpenMOLE

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- 1 Introduction
- 2 Basic experiments
- 3 DOE Samplings
- 4 Sensitivity analysis
- 5 Application in OpenMOLE
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- Interactive model exploration by hand and the need for preliminary experiments - The Design of Experiments as the definition of tasks to extract information from the simulation model - Example: NetLogo behavior space: basic grid DOE - Sensitivity analysis as an advanced DOE

Remark 1: terminology strongly depends on disciplines and practices

*Remark 2: these are generally **preliminary experiments** to prepare more elaborated, question-related, experiments*

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- *Provide explicitly sampling points on which the model (or its replication task) will be run: notion of ****direct sampling**** in OpenMOLE (corresponds to DOE in the literature)*
- full samplings - elaborated sampling for high dimensions given a low computational budget (****the curse of dimensionality****)

All factors have nominal values and a discrete variation set, in which each is varied while others remaining fixed

- *when model is slow - or computational budget highly limited* - *does not capture interaction between parameters, and highly dependent on nominal values* - *seen as a bad practice* BUT *useful for models taking significant time, and prone to thematic interpretation*

Example where One-At-a-Time fails

Ensemble product of discrete variation ranges for factors (usually a regular grid but not necessarily)

quickly limited by the curse of dimensionality - in practice still powerful with a quick model and a low number of parameters

naive approach, i.e. done by many "simulation-newcomers" such as economics or some parts of physics

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Computational limitations \implies need specific methods to efficiently sample the parameter space

The field of Design of Experiments has proposed different methods for numerical experiments given limited computational resources

Examples: Sobol sequence (quicker convergence of integral estimation), Latin Hypercube Sampling, Orthogonal sampling

Minimizing discrepancy: intuitively being spread evenly across the parameter space (def of discrepancy)

|x| | | | | | :-: | :-: | :-: | :-: | :-: | :-: | |x| | | | |x| | | |x| | | |x| |

Latin cube: one point in each row and column; hypercube generalization in any dimension

Quasi-random sequences with low discrepancy (also Halton sequences e.g.)

Estimate integral in $1/N$ instead of $1/\sqrt{N}$ with random sampling
Constructed recursively (using bit representations).

TODO illustration in 2D

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How to summarize model sensitivity and isolate principal factors ?

Examples: Morris and Saltelli methods

Idea : Sample trajectories in the parameter space in a One-At-a-Time manner. Screening method isolating *elementary effects*

[Saltelli et al., 2004]

- isolate local effects of factors - more efficient than point sampling to get individual effects - useful as a first experiment to understand the relative influence of factors

[Campolongo et al., 2011] propose to extend the method with Sobol sequences

Estimation of relative and conditional variances

$$ST_i = \frac{E_{\mathbf{X} \sim i} [\text{Var}(Y | \mathbf{X} \sim i)]}{\text{Var}(Y)}$$

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```
val explo = DirectSampling( evaluation = model, sampling =  
... )
```

- One-factor sampling

```
sampling = OneFactorSampling( (x1 in (0.0 to 1.0 by 0.2))  
nominal 0.5, (x2 in (0.0 to 1.0 by 0.2)) nominal 0.5 )
```

- Grid sampling

```
sampling = (x1 in (0.0 to 1.0 by 0.5)) x (x2 in (0.0 to 1.0  
by 0.5))
```

- LHS Sampling

```
sampling = LHS(100,x1 in (0.0,1.0),x2 in (0.0,1.0))
```

- Sobol sampling

```
sampling = SobolSampling(100,x1 in (0.0,1.0),x2 in  
(0.0,1.0))
```

(method in itself)

```
val sen = SensitivitySaltelli( //evaluation = (model on
env), evaluation = (model on env by 1000), samples =
100000, inputs = Seq(humanFollowProbability in (0.0,1.0),
humanInformedRatio in (0.0,1.0),humanInformProbability in
(0.0,1.0)), outputs = Seq(peakTime, peakSize,
totalZombified,halfZombified,
spatialMoranZombified,spatialDistanceMeanZombified,spatialEntr
)
```

(example from market)

```
SensitivityMorris( evaluation = modelExec on envLocal hook  
storeSimuCSV, inputs = Seq(inputNumberOfCars in (1.0,  
41.0), inputAcceleration in (0.0, 0.0099),  
inputDeceleration in (0.0, 0.099) ), outputs =  
Seq(outputSpeedMin, outputSpeedMax), repetitions = 100,  
levels = 5)
```

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Your turn to run some direct samplings and/or sensitivity analysis

- ▶ given the described zombie model, what first experiment beyond stochasticity would be relevant ?
- ▶ write a script
- ▶ explore results (using e.g. the OpenMOLE GUI plots)

Resources: - one script running directsampling - example of grid explo results - example of Saltelli

Cooperation model

 Campolongo, F., Saltelli, A., and Cariboni, J. (2011).

From screening to quantitative sensitivity analysis. a unified approach.

Computer Physics Communications, 182(4):978–988.

 Saltelli, A., Tarantola, S., Campolongo, F., and Ratto, M. (2004).

Sensitivity analysis in practice: a guide to assessing scientific models.

Chichester, England.