

# 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
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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 => 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\*

[?]

- 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

[?] 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,  
spatialMoranZombi-  
fied,spatialDistanceMeanZombified,spatialEntropyZombified,spatialSlopeZom  
) ““
```

```
(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*



