# State of the art Methods and Tools in VVUQ

Diana Suleimenova

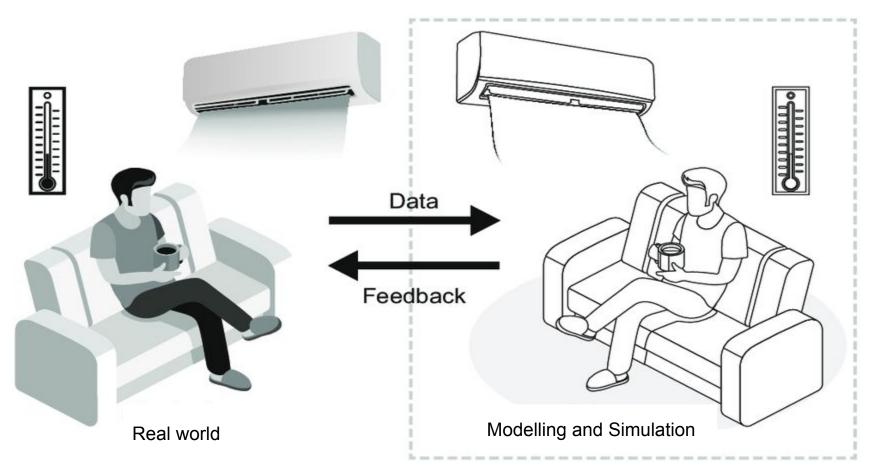
Department of Computer Science, Brunel University London

The Role of Uncertainty in Mathematical Modelling of Pandemics 8 February 2022









Source: Deguchi et al. (2020)

## Concepts and terminology

#### Society of Computer Simulation (SCS):

#### Qualification

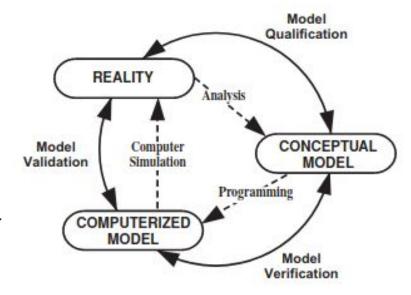
 "Determination of adequacy of the conceptual model to provide an acceptable level of agreement for the domain of intended application."

#### Verification

 deals with the relationship between the conceptual model and the computerized model.

#### Validation

 deals with the relationship between the computerized model and reality.

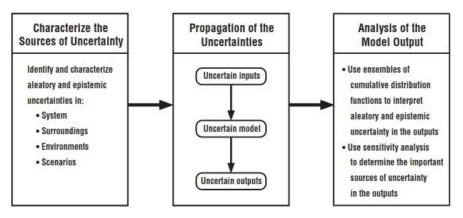


The role of Verification and Validation within the phases of modelling and simulation (Schlesinger, 1979).

## Concepts and terminology (cont.)

#### Uncertainty:

- Aleatory
  - "Uncertainty due to inherent randomness."
  - describes the natural/intrinsic variability of a quantity of interest (Qol).
- Epistemic
  - "Uncertainty due to lack of knowledge."
  - describes the lack of knowledge and is potentially reducible by acquiring more knowledge.



Three basic steps in an uncertainty analysis (Oberkampf and Roy, 2010).

### **VVUQ** Tools









Name	Language	Main Features	License	Cross- platform	
Dakota	Dakota C/C++[2] General purpose: uncertainty propagation, surrogate modeling, sensitivity analysis, model calibration, reliability analysis, risk analysis, external code wrapping		LGPL	Yes	
DiceDesign	R	Construction of experimental designs	GPLv3	Yes	
DiceKriging	R	GP (Kriging) metamodeling G		Yes	
DiceOptim	R	GP (Kriging)-based optimization	GPLv3	Yes	
FERUM	MATLAB	Reliability analysis (FORM, SORM, Subset simulation, etc.), reliability-based design optimization (RBDO), and global sensitivity analysis	GPLv3	Yes	
mistral	R	Reliability analysis library (FORM, Importance Sampling, Subset Simulation, etc.)	CeCILLv2	Yes	
MUQ	C++/Python	General purpose: surrogate modeling (PCE, GP), constrained optimization, Bayesian inversion)	GPLv2	Yes	
OpenCossan	MATLAB	General purpose: uncertainty propagation, surrogate modeling, sensitivity analysis, reliability, robust optimization	LGPL	Yes	
OpenTURNS	C++/Python	General purpose: uncertainty propagation, surrogate modeling, sensitivity analysis, reliability, optimization	LGPL	Yes	
sensitivity	R	Sensitivity analysis (Sobol' indices, FAST, PCC, etc.) with support for multidimensional outputs	GPLv2	Yes	
SIMLab	100	GUI-based sensitivity analysis (Sobol' indices, FAST, Morris, etc.)	Freeware	No (Windows)	
SUMO Toolbox	MATLAB	Surrogate modeling (GP, SVM, neural networks, etc.) and surrogate-based optimization	AGPLv3	Yes	
UQLab	MATLAB	General purpose: uncertainty propagation, surrogate modeling, sensitivity analysis, reliability analysis, Bayesian inversion, robust optimization, external code wrapping	3-Clause BSD	Yes	
UQpy	Python	Uncertainty propagation, stochastic processes	MIT	Yes	
UQ Toolkit (UQTk)	C++/Python	Uncertainty propagation, surrogate modelling, sensitivity analysis, Bayesian inversion, external code wrapping	LGPL	Yes	







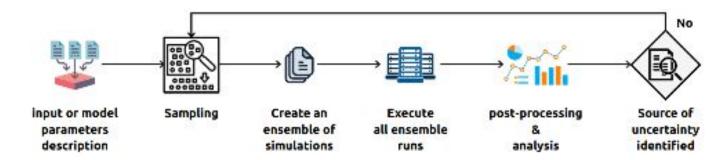


## **EasyVVUQ**



- EasyVVUQ aims to make it as easy as possible to implement advanced techniques for uncertainty quantification for existing application codes (or workflows).
- Primary focus on non-intrusive UQ techniques, where many model instances are run to quantify uncertainties and analyse parameter sensitivities.

https://easyvvug.readthedocs.io

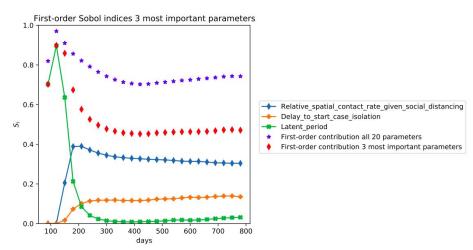


Basic UQ workflow

## EasyVVUQ samplers



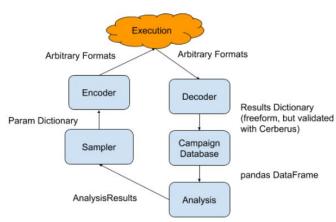
- Monte-Carlo
- Quasi Monte-Carlo
- Polynomial Chaos Expansion (PCE)
- Stochastic Collocation (SC)
- Adaptive SC sparse grids
- All of these can be used to calculate
   Sobol Sensitivity Indices



## EasyVVUQ Enhances Workflow



- Integrated tightly with the rest of Python scientific computing infrastructure.
  - E.g. access through a wide range of sampling approaches using ChaosPy.
- Adapts to users existing code via extension of encoder and decoder classes.
- Minimal boiler-plate code required (minimal set-up overhead).
- Takes care of bookkeeping (creating folders for storing results for example),
   collects other meta-info in a database.
- Execution can be done:
  - Locally (no tools required)
  - Supercomputers
     (using FabSim3, Dask and/or QCG-PJ)
  - Cloud HPC (using Kubernetes).





#### FabSim3



- User-developer oriented automation/curation toolkit
  - Aim: reduce human effort required to create/modify/repurpose/salvage complex computational workflows.
- Uses generic pattern code (reusable in other tools) for UQ and V&V.
- Supports range of backends:
  - Localhost.
  - o Range of supercomputers.
  - QCG broker & Pilot Jobs.
- FabSim3 is generic, but plugins provide application-specific advantages.

#### FabSim3 relies on bash one-liners, e.g.:

fabsim localhost run\_amazingly\_complex\_app fabsim eagle validate\_flee:mali,cores=24,replicas=5 fabsim localhost install\_plugin:FabFlee

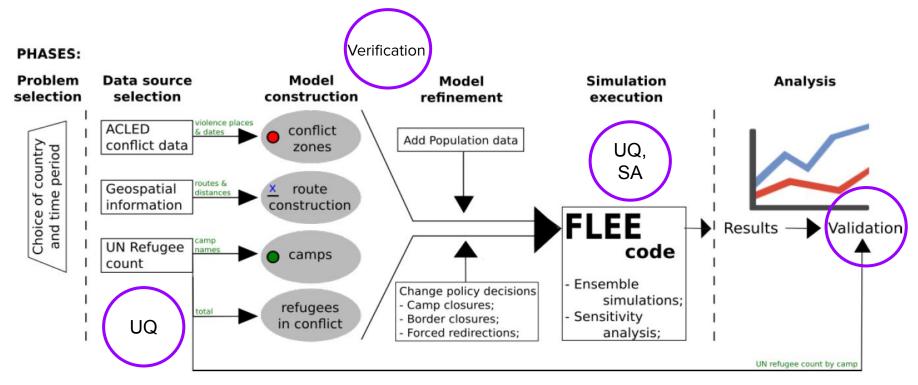
## V&V patterns in FabSim3



#### Four prominent V&V patterns:

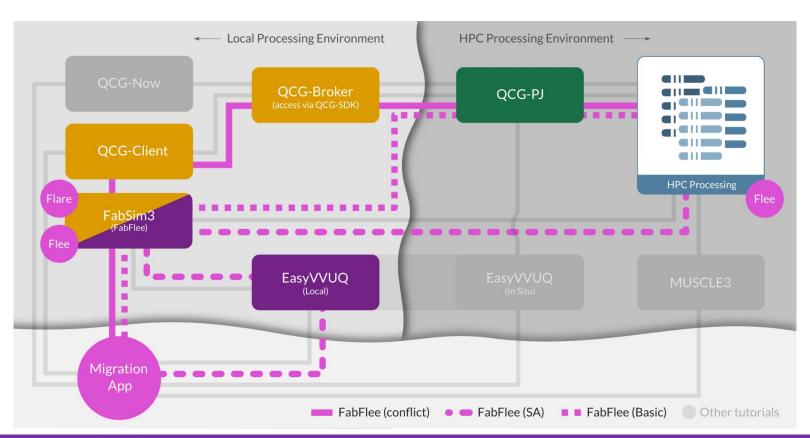
- 1. Stable Intermediate Forms (SIF):
  - o monitors iterative simulation development, evaluating each intermediate step against the baseline.
- 2. Level of Refinement (LoR):
  - o focuses asymptotic behaviour of Qols upon changing certain model parameters, e.g. increasing grid resolution and the convergence of Sobol indices.
- 3. Ensemble Output Validation (EoV):
  - employs a sample testing function to compare output from different model simulations, where each simulation has its own output directory.
- 4. Quantities of Interest (QoI):
  - o focuses on extracting a distribution of Qols from the simulations by applying a similarity measure (e.g. Wasserstein distance).

## Example: VVUQ in Multiscale Migration Prediction



Suleimenova et al., (2021). Phil. Trans A. vol. 379 (2197).

## **VECMAtk Migration Tube Map**



## In-depth: Performing UQ analysis

FabFlee	
FabSim3	plugin

• **Step 1** Definition of the space parameters.

#### EasyVVUQ

- **Step 2** Identification and assessment of uncertain input distributions.
- **Step 3** Generation of required samples and runs.

#### Localhost/HPC

• **Step 4** Executions: evaluation of the application.

#### **EasyVVUQ**

Step 5 Aggregation of the outputs.

• **Step 6** Analysis: calculation of UQ and SA measures.

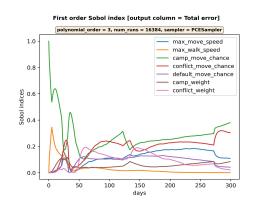
#### FabFlee FabSim3 plugin

• Step 7 Exploitation of the results: visualization, updating database, etc ...

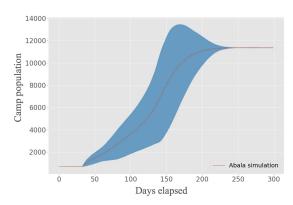
## Migration VVUQ analysis

#### Semi-intrusive UQ analysis:

- SA-based dimension reduction: investigated 6-7 parameters with polynomial order of 3 using stochastic collocation and polynomial chaos expansion samplers with Sobol's method.
- **Efficient sampling:** measure the uncertainty caused by a stochastic model for evolving violent events on the output (populations in camps).



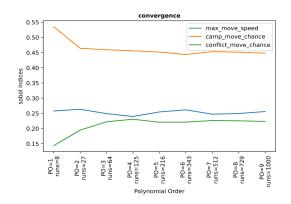
Simulations	Parameters	Mali	Burundi	South Sudan	CAR
SA iteration 1	6	0.4914	0.3832	0.4926	0.3809
SA iteration 2	7	0.4101	0.3361	0.4753	0.3257
Default values	6	0.3122	0.2571	0.5234	0.3378



## Migration VVUQ analysis

#### Validation and Verification analysis:

- Level of Refinement (LoR): the Sobol indices for each parameters with different polynomial orders;
- Ensemble Output Validation (EoV): implemented within FabSim3 to calculate the validation metric and mean score of ensemble runs.



```
~/FabSim3/results/mali_training_vecma_1/RUNS/run_1
Validation run_1: 0.3392997798547688
    ~/FabSim3/results/mali_training_vecma_1/RUNS/run_7
Validation run_7: 0.2797604227928289
...
    ~/FabSim3/results/mali_training_vecma_1/RUNS/run_6
Validation run_6: 0.2841636513726309
Mean score: 0.29770008098431655
```

## Questions?

Thank you for your attention.