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Why should all modelers do sensitivity analysis? SA for complex models



Aim of presentation

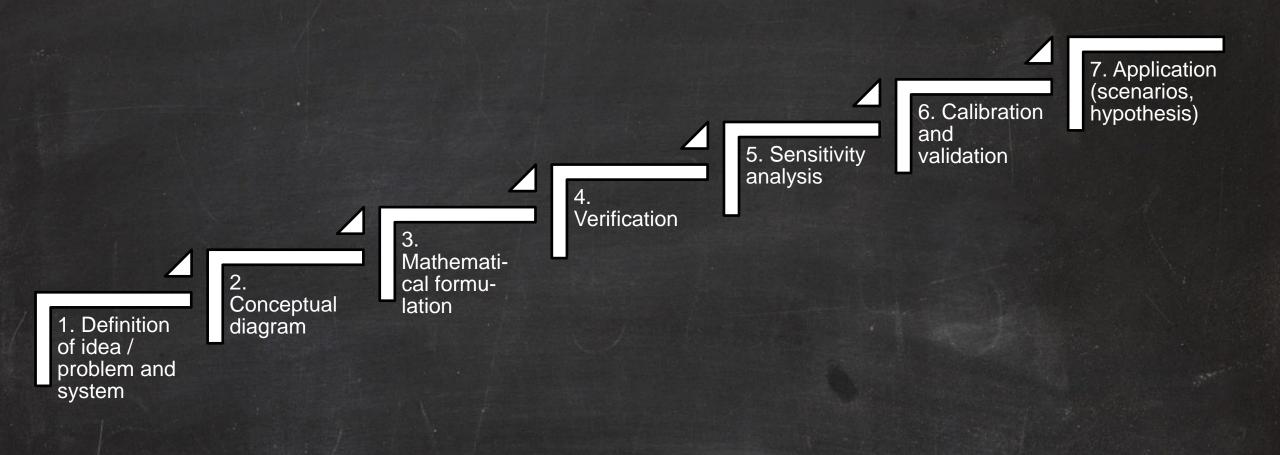
Learning outcomes:

Participants will get an introduction to

- The importance of sensitivity analysis
- SA connection to uncertainty analysis
- Local and global SA methods
- Potential SA errors
- SA methods with parsac
- SA examples

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The modelling process

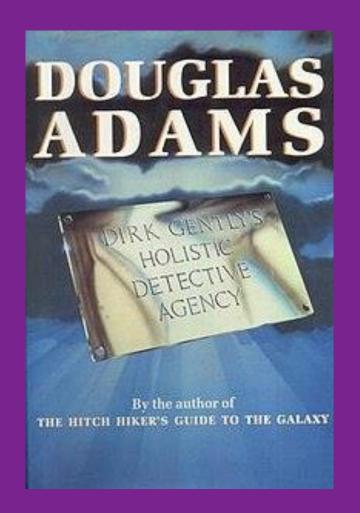




What modelling is not

Well, Gordon's great insight was to design a program which allowed you to specify in advance what decision you wished it to reach, and only then to give it all the facts. The program's task, [...], was to construct a plausible series of logical-sounding steps to connect the premises with the conclusion.

Douglas Adams, 1987, p. 69, *Dirk Gently's Holistic Detective Agency*





There are many opinions (and feelings) about models

Because models influence decisions, which in turn influence people's lives.



"The entire environmental regulation is dependent on a model of emissions that does not match reality ..."



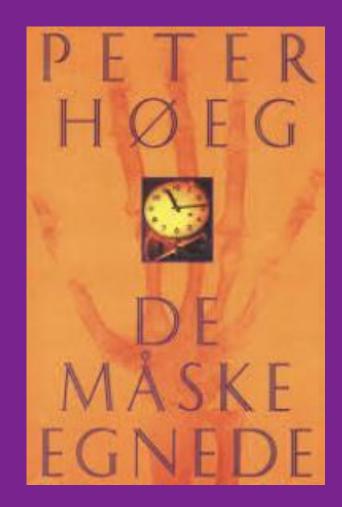


Why understand uncertainty?

That is what we meant by science.

That both question and answer are tied up with uncertainty, and that they are painful. But that there is no way around them. And that you hide nothing; instead, everything is brought out into the open.

Peter Høeg, 1993, Borderlines





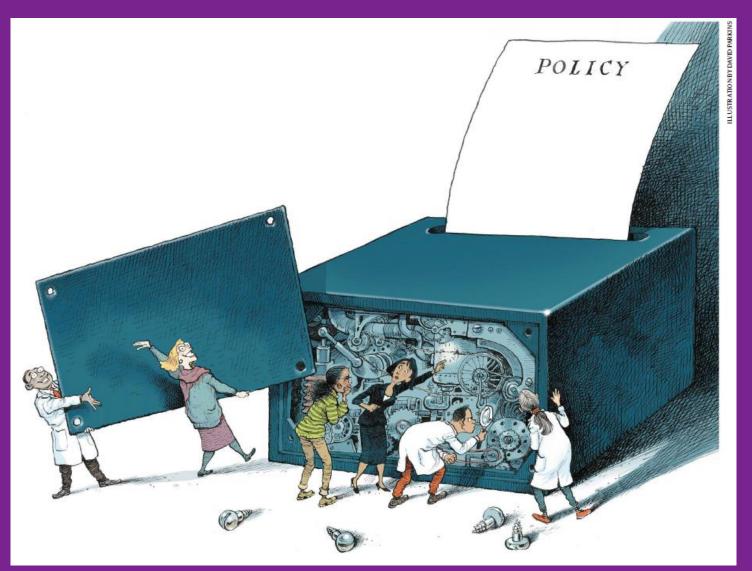


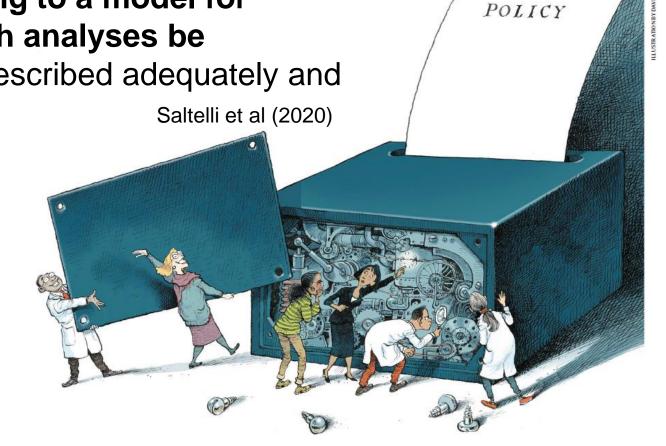
Illustration from Nature article by Saltelli et al (2020)



SA should be common practice

"However, these global uncertainty and sensitivity analyses are often not done. Anyone turning to a model for insight should demand that such analyses be conducted, and their results be described adequately and made accessible."

SA is also recommended by both the EU Commission and the US Environmental Proctection Agency.





Best practice for SA

Preanalysis

- 1) Establish goals for analysis
- 2) Decide on input factors
- 3) Decide on distribution function for each input factor
- 4) Choose SA strategy and method(s)

Step 1

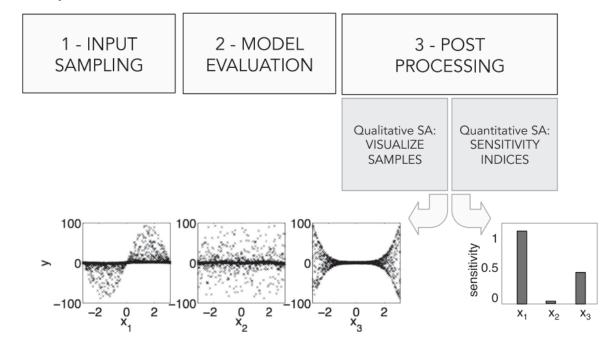
• 5) Generate input sample

Step 2

• 6) Execute model on generated input sample and produce output

Step 3

• 7) Analyze model outputs with SA methods, evaluate and draw conclusion of SA



Pianosi et al (2016)



Factor prioritization

objective: to identify most the most important factor

Factor fixing

objective: fixing of non-influential factors (screening)

Variance cutting

objective: reduce variance of output from its unconditional value to a lower preestablished threshold value.

Factors mapping

objective: which factor is most responsible for producing realisations of output in the region of interest?



Caution: Here be errors!

Type I error

Classify an un-important parameter/factor as influential.

Type II error

Classify an important parameter/factor as noninfluential.

Type III error

Wrong framing of the sensitivity analysis.



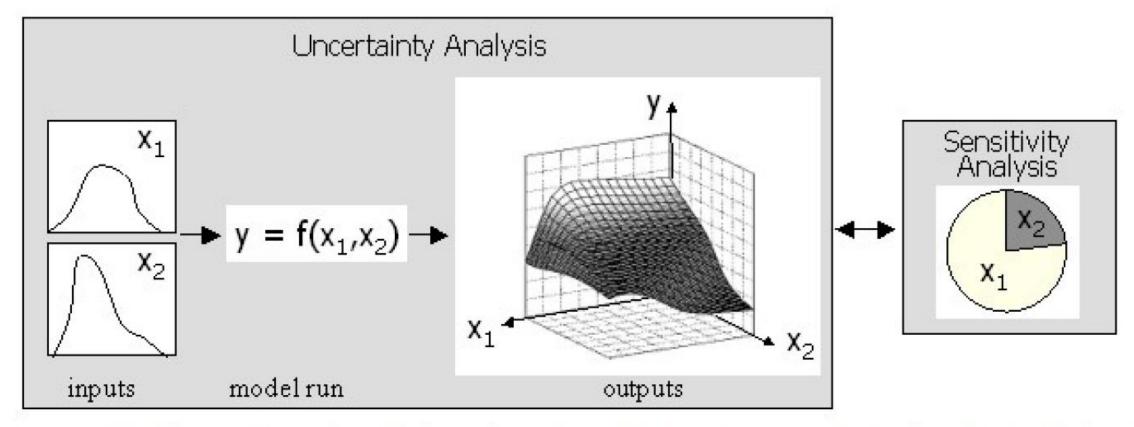


Figure D.5.1. Uncertainty and sensitivity analyses. Uncertainty analysis investigates the effects of lack of knowledge or potential errors of model inputs. Sensitivity analysis evaluates the respective contributions of inputs x_1 and x_2 to output y.



Sensitivity analysis: local vs global

Local sensitivity analysis

- Evaluates the effect of changing one parameter (from baseline) on the model output. This is repeated for all parameters one-at-a-time (OAT). All parameters considered in these OAT analyses are taken to be independent from one another.
- Therefore, local SA should only be used when the model has proved to be linear or at least additive. As local SA does not allow to take parameter interactions into account

Global sensitivity analysis

 Simultaneously evaluates both the effect of each individual parameter and its interaction with other parameters on model output. This is because all parameters are varied simultaneously over the entire parameter space.

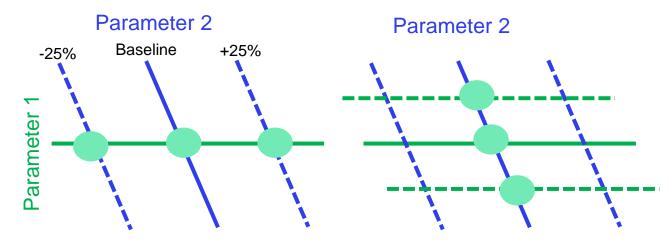
For an example of why local SA is usually perfunctory for complex environmental models see:

Saltelli and Annoni (2010)

How to avoid a perfunctory sensitivity analysis



Illustration of local vs global SA



Step 1

Value of parameter 1 is changed with +/- 25% from baseline. Model is evaluated 3 times.

Step 2

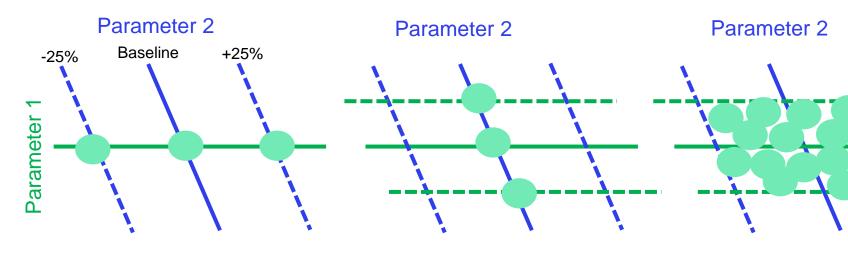
Value of parameter 2 is changed with +/- 25% from baseline. Model is evaluated 3 times.

Local SA

(steps repeated as many times as number of parameters)



Illustration of local vs global SA



Step 1

Value of parameter 1 is changed with +/- 25% from baseline. Model is evaluated 3 times.

Step 2

Value of parameter 2 is changed with +/- 25% from baseline. Model is evaluated 3 times.

Entire parameter range of par1 and par2 is sampled by a sampling scheme and model is evaluated as many times as required by global SA method.

Local SA

(steps repeated as many times as number of parameters)

Global SA

For another example of why local SA is usually perfunctory for complex environmental models see:
Saltelli and Annoni (2010) How to avoid a perfunctory sensitivity analysis



Overview of global SA approaches

- 1. Multiple-start perturbation approaches, where global sensitivity is obtained by aggregation of 'OAT' sensitivities obtained at different baseline points (e.g. the Elementary Effects Test or method of Morris);
- 2. Correlation and regression approaches, where sensitivity is measured by the correlation between input and output samples;
- 3. Regional sensitivity analysis (or Monte Carlo filtering) methods, where sensitivity is related to variations in the distributions of input factors induced by conditioning the outputs; and
- 4. Variance-based and density-based approaches, where sensitivity is linked to variations in the output distribution induced by conditioning the inputs.

 Wassers & Diagoni (2010): What has Clabal Consitiuits.

Wagener & Pianosi (2019): What has Global Sensitivity Analysis ever done for us? A systematic review to support scientific advancement and to inform policymaking in earth system modelling



Global SA approaches

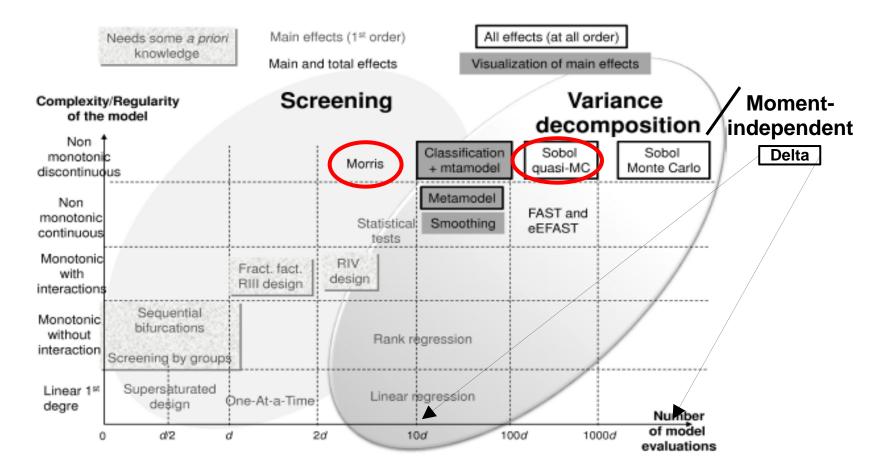


Fig. 6. SA methods graphical synthesis.

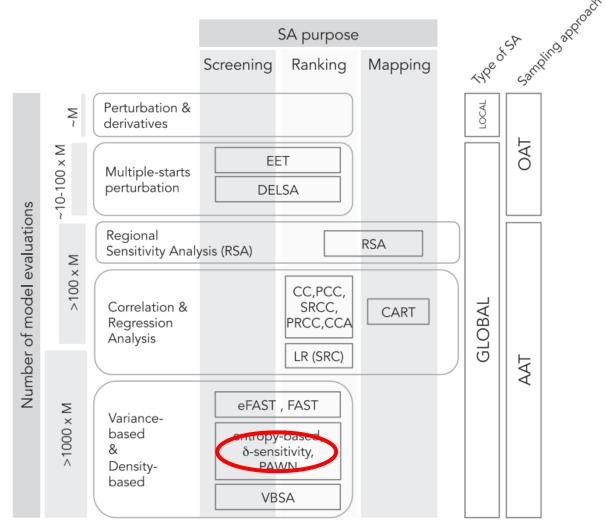
Fig from loos & Lemaitre (2014): A review on global sensitivity analysis methods with own addition of moment-independent method

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Global SA approaches



Pianosi et al. (2016)

Fig. 3. Classification system of Sensitivity Analysis methods based on computational complexity (vertical axis; M is the number of input factors subject to SA) and purposes of the analysis. Some of the most widely used methods are reported (acronyms are defined in corresponding paragraphs of Sec. 3). Types of SA and sampling approaches are defined in Sec. 2.2. Figures about computational complexity are indicative, for a further discussion see Sec. 4.5.

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Variance-based method: Sobol

First-order effect

XX

Second-order effect

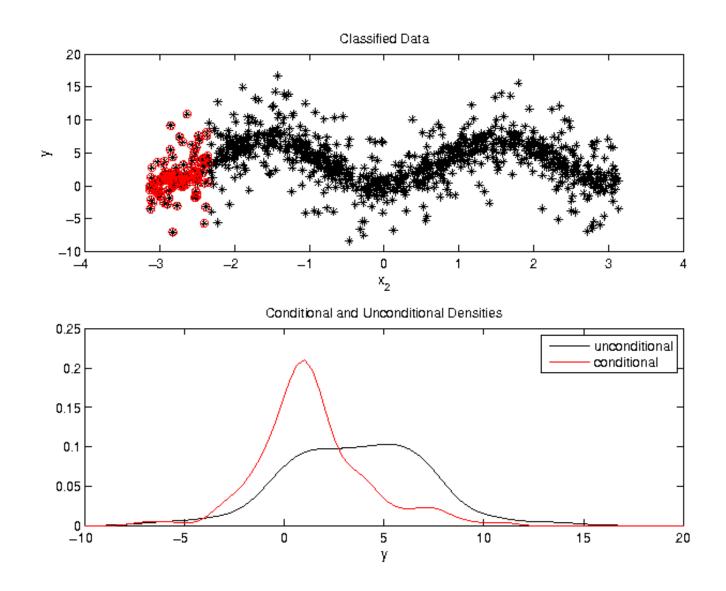
XX.

Total order effect

XX.

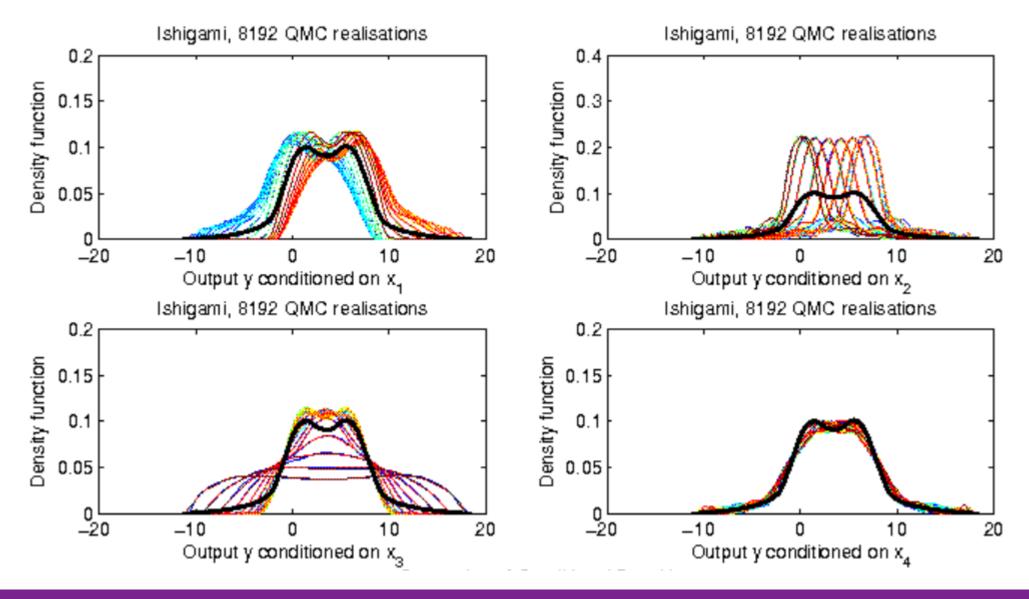


Moment-Independent method: Delta



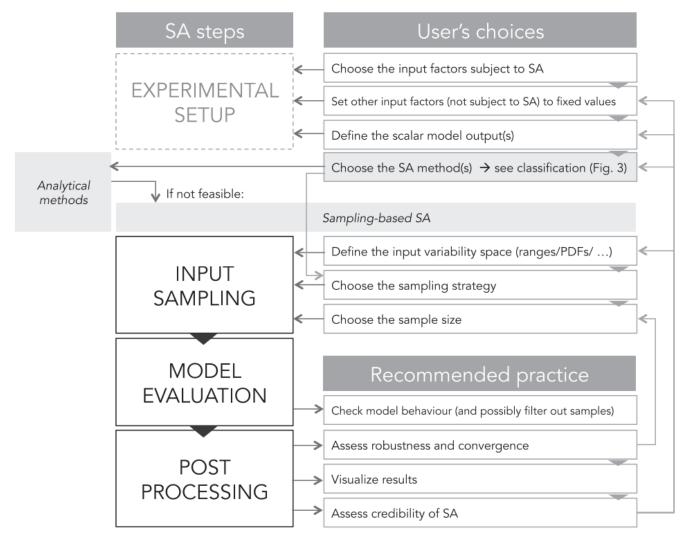


Moment-Independent method: Delta





SA is an iterative process



Pianosi et al (2016) Fig. 4. Workflow for the application of Sensitivity Analysis, choices to be made and recommended practice for their revision.

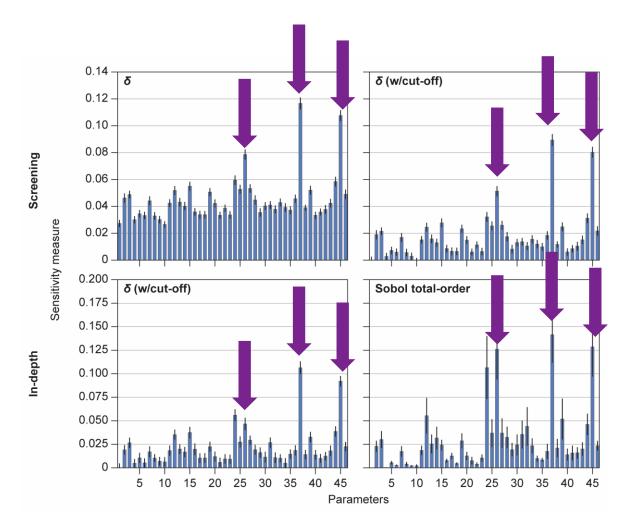
SA with parsac



Andersen et al (2020) *Environ. Model. Softw.*

calibrated value -Generate samples SCREENING: BORGONOVO'S & ANALYSIS with LHS LHS Summer mean Spin-up period **Execute model simulations** Compute sensitivity indices Cut-off threshold Apply cut-off threshold δ (LHS) Sobol (Saltelli) IN-DEPTH: BORGONOVO'S 5 AND SOBOL ANALYSIS calibrated value calibrated value Generate samples with LHS and Saltelli LHS Saltelli sampling Summer mean Spin-up period **Execute model simulations** Summer mean Summer mean Sensitivity indices Sensitivity indices Compute sensitivity indices

SA with parsac

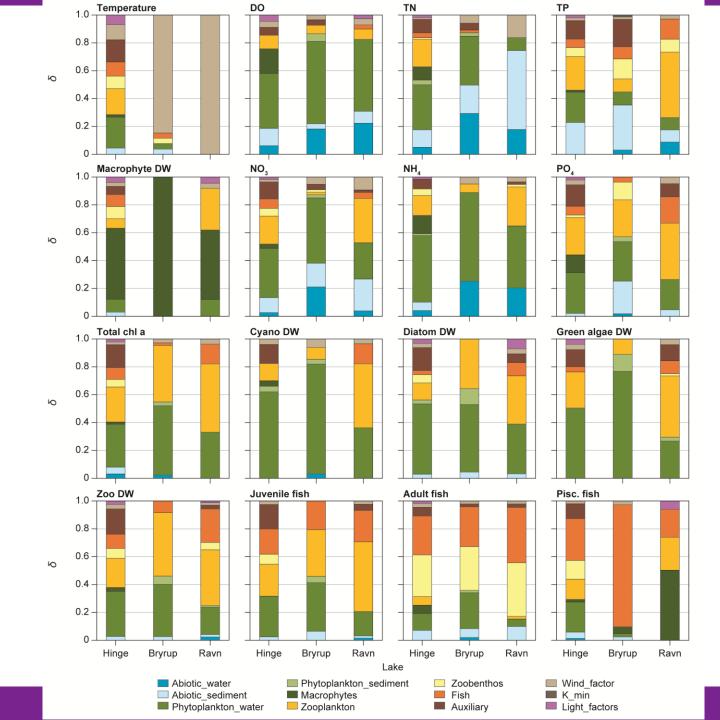


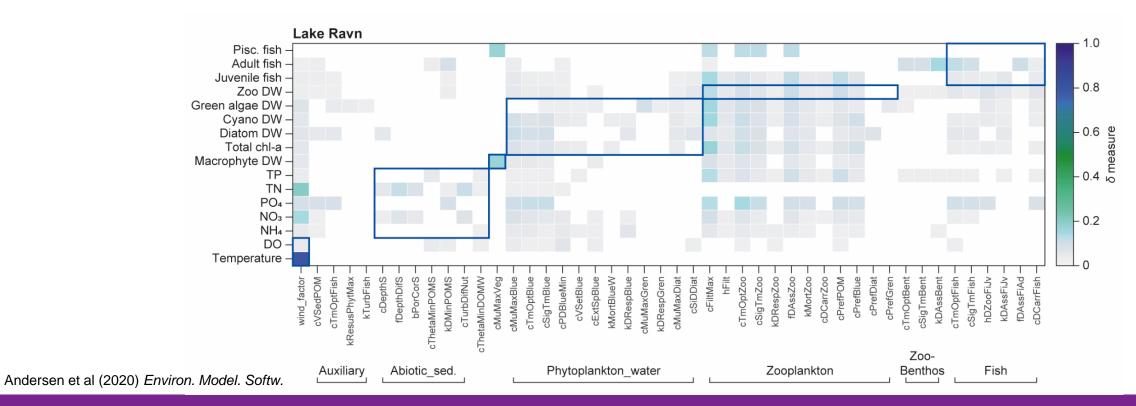
Andersen et al (2020) Environ. Model. Softw.

SCREENING: BORGONOVO'S & ANALYSIS

IN-DEPTH: BORGONOVO'S 5 AND SOBOL ANALYSIS









SA challenges to be aware of



Here be dragons!



Type III errors can be FOOLISH





Type III errors can be DANGEROUS

US government decided to build a repository for radioactive waste in the Yucca Mountains.

A model named Total System Performance Assessment (TSPA) was used for the safety analysis computations. TSPA is composed of 286 sub-models and should predict thousand of years into the future.

Key assumption: Low permeability of the geological formation.

But when construction began, new evidence led to an upward revision of four orders of magnitude of the permeability parameter!

"According to the authors the error was due to the modelling of the granite formation as a homogeneous medium, while a fissures and faults model of the same formation would have been more realistic."

"It is important [...] to recognize that the sensitivity of the parameter in the equation is what is being determined, not the sensitivity of the parameter in nature. [...] If the model is wrong or if it is a poor representation of reality, determining the sensitivity of an individual parameter in the model is a meaningless pursuit."

Pilkey and Pilkey-Jarvis, (2007), p.25





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Unquantifiable uncertainties...

Main dimensions of uncertainty in complex environmental models are

- -Technical (inexactness)
- –Methodological (unrealiability)
- –Epistemological (ignorance)
- -Societal (social robustness)

"Quantitative methods only address the technical dimension only. They can, however, be complemented with new qualitative approaches addressing aspects of uncertainty that are hard to quantify and were largely unadressed in the past." van der Sluijs et al, 2005



Sensitivity auditing

An entire research field dedicated to understanding uncertainty in models. This is (maybe even more) important when models are used beyond research interest e.g. in consultancy and to inform policy. Sensitivity auditing belongs to the tradition of **Post-Normal Science**.

- 1. Is the model redundant?
- 2. Are there important implicit assumptions?
- 3. Is uncertainty instrumentally amplified or compressed?
- 4. Was a <u>sensitivity analysis</u> performed prior to publication of the inference?
- 5. Is the model transparent?
- 6. Does the model address the right question?
- 7. Was <u>sensitivity analysis</u> performed holistically?

A brief introduction to sensitivity audting can be find in Saltelli & Funtowicz (2015): Evidence-based policy at the end of the Cartesian dream: The case of mathematical modelling



There are many opinions (and feelings) about models

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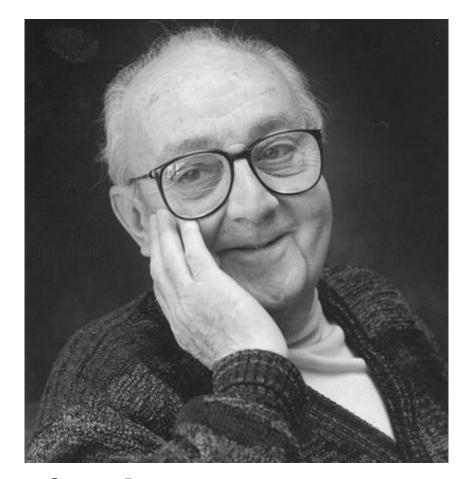
"The entire environmental regulation is dependent on a model of emissions that does not match reality ..."





"All models are wrong"

"So the questions you need to ask is not "Is the model true?" (it never is) but "Is the model good enough for this particular application?"

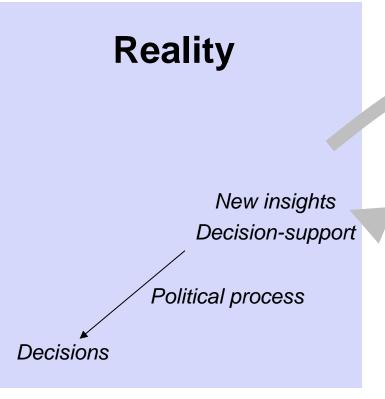


George Box



₩ Welcome to Model Land

Which assumptions did you take when entering?





Model Land

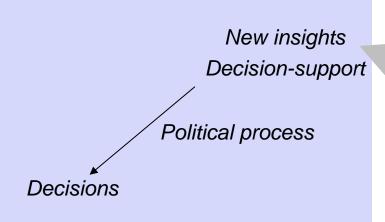
All assumptions are true. In here, mathematics rule.



Welcome to Model Land

Which assumptions did you take when entering?

Reality





How do we provide **confidence** in models? How to handle (deep) **uncertainties**?

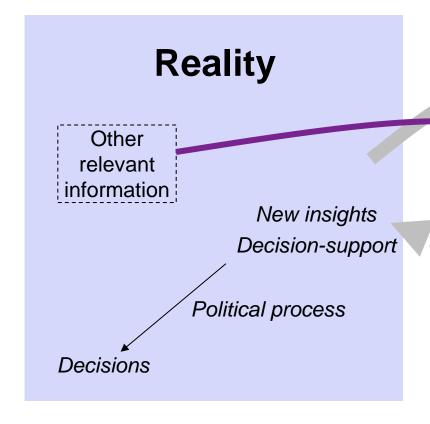
Model Land

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Welcome to Model Land

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Model validation

How do we provide **confidence** in models?
How to handle (deep) **uncertainties**?

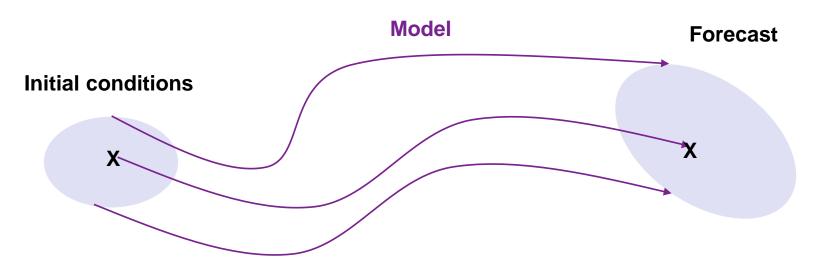
Model Land

All assumptions are true. In here, mathematics rule.

Sensitivity analysis



Butterflies and Hawkmoths



Butterfly Effect

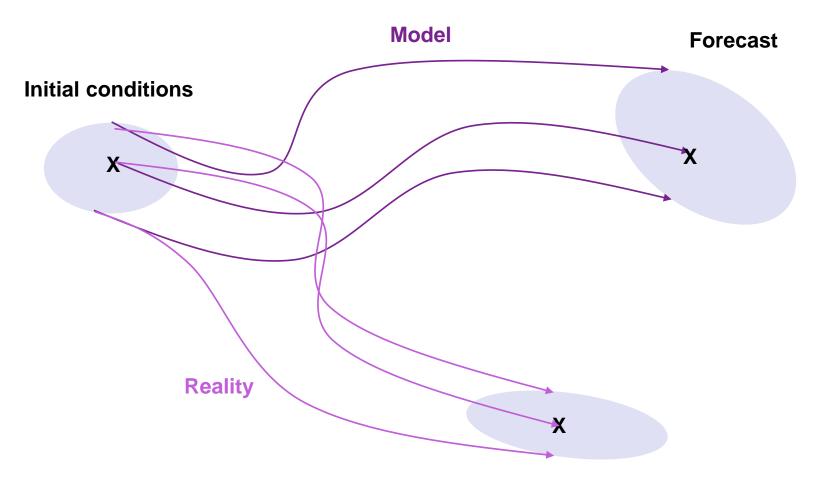


In Model Land

Uncertainties in initial conditions
Limits timescale



Butterflies and Hawkmoths



Butterfly Effect



In Model Land

Uncertainties in initial conditions
Limits timescale

Hawkmoth Effect



In the Gap

Structural uncertainties in model Limits timescale **and** accuracy

Heavily inspired by slides from Erica Thompson and her book *Escape from Model Land*



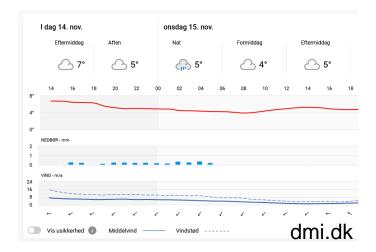
When doing scenarios: Weather- or climate-like predictions?

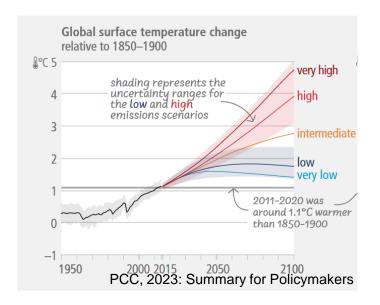
Weather-like prediction tasks:

- similar decisions, in similar contexts, are made frequently (e.g. daily, monthly);
- a size-able forecast-outcome data archive available:
- the models involved have a long lifetime relative to the lead time of the forecast.

Climate-like prediction tasks:

- the decision is effectively one-off, and the decision not to act has potential significant costs;
- the forecast-outcome archive is effectively empty (task is extrapolation)
- the model has a short lifetime compared to the lead time of the forecast.

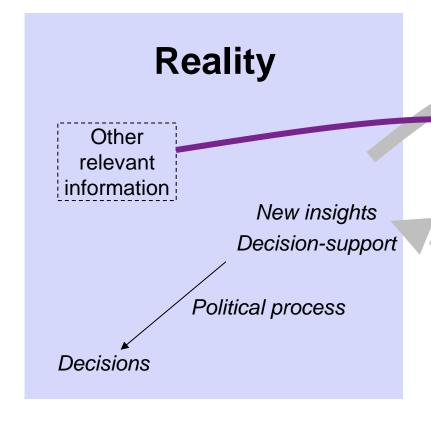






Welcome to Model Land

Which **assumptions** did you take when entering?





Model validation

How do we provide **confidence** in models?
How to handle (deep) **uncertainties**?

Model Land

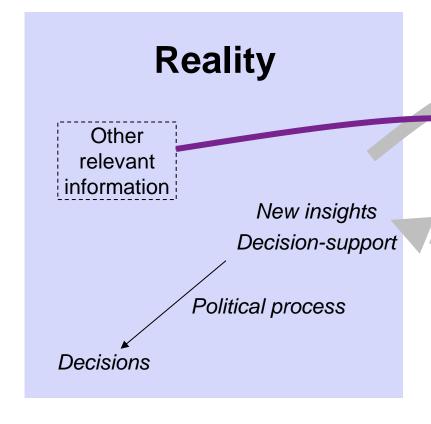
All assumptions are true. In here, mathematics rule.

Sensitivity analysis



Welcome to Model Land

Which assumptions did you take when entering?





Model validation

How do we provide **confidence** in models?
How to handle (deep) **uncertainties**?

Model Land

All assumptions are true. In here, mathematics rule.

Sensitivity analysis



Supplementary reading

- Saltelli et al (2020) Five ways to ensure that models serve society: a manifesto, Nature
- Borgonovo et al (2017) Making the most out of a hydrological model data set: Sensitivity analyses to open the model black-box, <u>Water Resources Research</u>
- Razavi, S., Jakeman, A., Saltelli, et al (2021) The Future of Sensitivity Analysis: An
 essential discipline for systems modeling and policy support, <u>Environmental Modelling</u>
 and Software



Additional notes

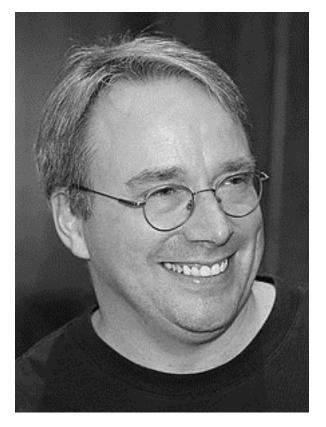


Do open source (when possible)

"Given enough eyeballs, all bugs are shallow"

Linus' Law

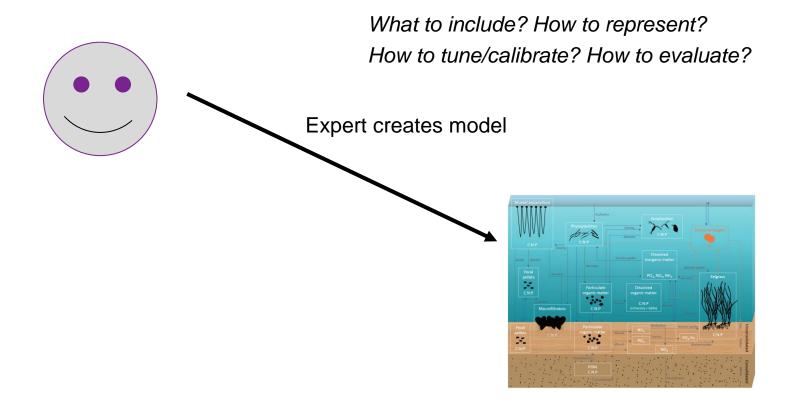
(by Eric S. Raymond)



Linus Thorvalds

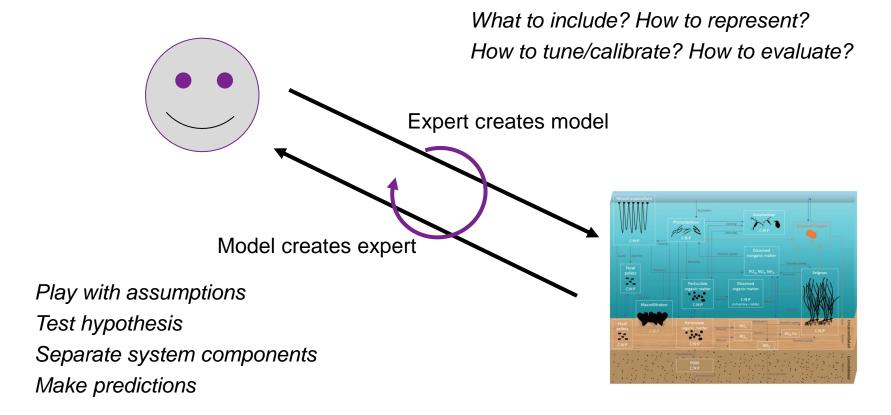


What is the relationship between models and experts?



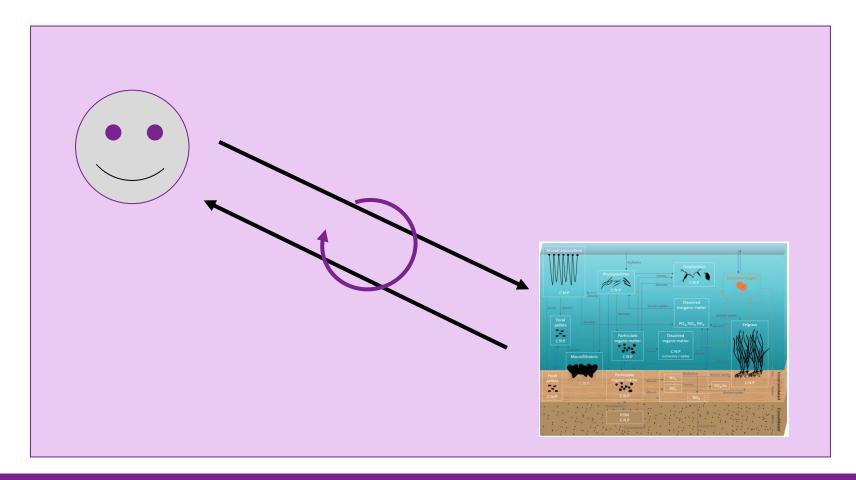


What is the relationship between models and experts?





Be mindful of expert system





Be mindful of expert system

"[...] models are complex amalgam of theoretical and phenomenological laws (and the governing equations and algorithms that represent them), empirical input parameters, and a model conceptualization. When a model generates a prediction, of what precisely is the prediction a test? The laws? The input data? The conceptualization? Any part (or several parts) of the model might be in error, and there is no simple way to determine which one it is'."

Naomi Oreskes, 2000

Oreskes, N., 2000, Why predict? Historical perspectives on prediction in Earth Science, in Prediction, Science, Decision Making and the future of Nature, Sarewitz et al., Eds., Island Press, Washington DC.