

# Introduction to Sensitivity and Uncertainty Analysis in Reactor Physics

**Maria Pusa** 

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#### **Outline**

- Sensitivity analysis
- Uncertainty analysis
- Methods
- Application to reactor physics
- Example calculation



### Sensitivity

- Starting point: mathematical model containing uncertain parameters and response dependent on this model
- Question: If one of the parameters is perturbed, how will this affect the response?
- Mathematical definition:
  - Simplest case: local sensitivity of response R with respect to parameter  $\alpha$  at point  $\alpha=\alpha^0$  is the derivative

$$s_{\alpha} = \left(\frac{dR}{d\alpha}\right)_{\alpha = \alpha^0} \tag{1}$$

 This generalizes easily to more general mathematical systems (e.g. parameters that are functions and responses that are functionals)



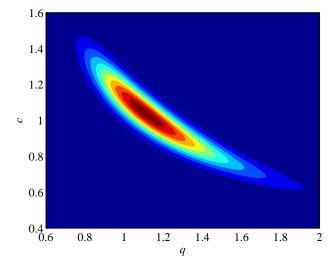
### **Sensitivity Analysis**

- Objective: Compute derivatives with respect to all parameters of interest
- Brute-force approach:
  - Variate the parameters one-by-one and compute the response
  - Inefficient when there are several parameters
- Deterministic approach:
  - Formulate the problem mathematically and compute the derivatives
  - Very efficient if a mathematical concept called adjoint is utilized



### **Uncertainty**

- Starting point: a mathematical model containing uncertain parameters and response dependent on this model
- Question: How to quantify the uncertainty related to the parameters?
  - Bayesian probability definition: knowledge about a parameter presented as probability distribution
  - Variance (one parameter) or covariance (several parameters) of the distribution may be chosen as the descriptive statistic for the uncertainty





### **Uncertainty Analysis**

- Objective: Compute the probability distribution of the response based on the probability distributions of the uncertain parameters
- Determination of the exact distribution usually extremely difficult
  - ⇒ Compute only variance/covariance due to uncertain parameters OR estimate distribution based on simulations
- Inaccuracy related to numerical methods or approximation errors not included in classical uncertainty analysis



## **Uncertainty Analysis Methods**

- Deterministic approach:
  - 1. Calculate response sensitivity vector *s*
  - 2. Linearize response

$$R \approx s\alpha$$
 (2)

3. Compute respective variance/covariance

$$\operatorname{Cov}[R] \approx \operatorname{Cov}[s\alpha] = s\operatorname{Cov}[\alpha]s^T$$
 (3)

- Statistical approach
  - 1. Sample points from distribution  $p(\alpha)$
  - 2. Compute R corresponding to each sample
  - 3. Compute uncertainty estimates based on simulated p(R)



## **Application to Reactor Physics**

- *Mathematical model*: transport (or diffusion) equation, potentially combined with a depletion model
- Responses: multiplication factor, reaction rates, homogenized cross-sections etc.
- *Uncertain parameters:* neutron cross-sections, initial nuclide concentrations, system dimensions etc.



### **Application to Reactor Physics: Adjoint-based Approach**

- + Computationally very efficient
- + Yields detailed sensitivity profiles
- Best-suited for deterministic codes
- Requires extensive modifications in the code
- Has not been applied to depletion problems



## **Application to Reactor Physics: Statistical Approach**

- + Well-suited for both deterministic and Monte Carlo codes
- + Code can be treated as a black box (depletion does not cause any difficulties!)
- + Yields additional information about the distribution p(R) (besides variance/covariance)
- Computationally expensive
- Does not yield sensitivity information



## **S&U** analysis with Monte Carlo method

- Statistical approach
  - Sample from Gaussian distribution based on covariance data
  - Total Monte Carlo:
    - \* A.J. Koning and D. Rochman, *Towards sustainable nuclear energy:*Putting nuclear physics to work, Ann. Nucl. Energy, **35**, 11, 2024–2030 (2008)
  - Suitable for burnup calculations
- Adjoint-based approach
  - exploit the physical interpretation of adjoint:
    - \* Brian C. Kiedrowski, Forrest B. Brown and Paul P. H. Wilson, Adjoint-Weighted Tallies for k-Eigenvalue Calculations with Continuous-Energy Monte Carlo, Nucl. Sci. Eng., 168, 3, 226–141 (2011)
  - Suitable for problems covered by generalized perturbation theory

## **Example of S&U Calculation**

Calculation code: CASMO-4

Source of uncertainty: neutron cross-sections

S&U analysis method: Adjoint-based

Test case: a 7×7 BWR assembly [1]

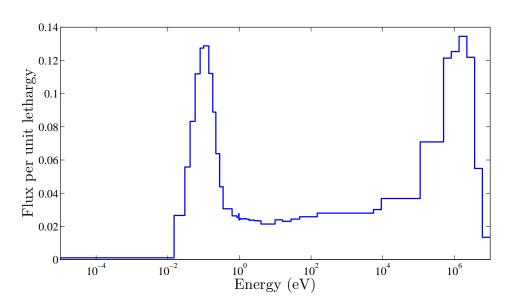
Rod type	<sup>235</sup> U (wt.%)	Gd <sub>2</sub> O <sub>3</sub> (wt.%)	No. of rods
1	2.93	0	26
2	1.94	0	12
3	1.69	0	6
4	1.33	0	1
5A	2.93	3.0	3
6B	2.93	3.0	1

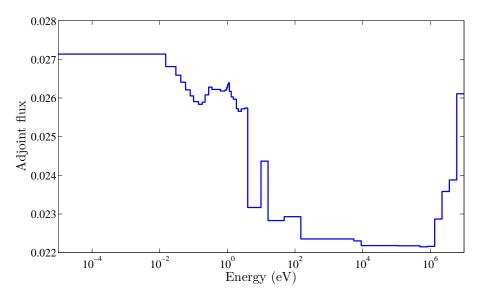
	4	3	3	2	2	2	3	
	3	2	1	1	1	1	2	
	3	1	5A	1	1	5A	1	
	2	1	1	1	1	1	1	
	2	1	1	1	6B	1	1	

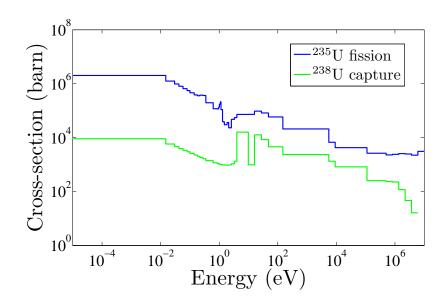
K. Ivanov et al., Benchmark for uncertainty analysis in modeling (UAM) for design, operation, and safety analysis of LWRs, Volume I: Specification and



# **Example: flux and adjoint flux**









# Example: $k_{inf}$ S&U profiles (1)

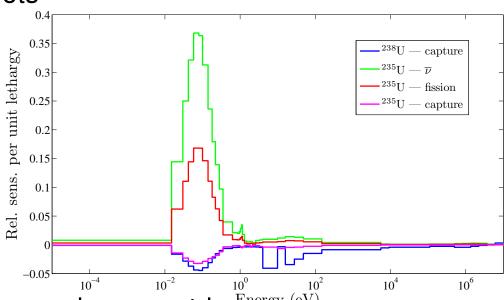
- $k_{inf} = 1.1055$
- $\Delta k_{\rm inf}/k_{\rm inf}=0.5076\%$

Nuclide	Param.pair	Rel. sensitivity	Rel. uncertainty
<sup>238</sup> U	$\sigma_{ m c}$ , $\sigma_{ m c}$	$-2.448 \times 10^{-1}$	$3.198 \times 10^{-1}$
$^{235}U$	u , $ u$	$9.161 \times 10^{-1}$	$2.720 \times 10^{-1}$
$^{235}U$	$\sigma_{ m c}$ , $\sigma_{ m c}$	$-1.010 \times 10^{-1}$	$1.423 \times 10^{-1}$
$^{235}U$	$\sigma_{ m f}$ , $\sigma_{ m f}$	$4.157 \times 10^{-1}$	$1.416 \times 10^{-1}$
$^{238}U$	$\sigma_{ ext{s}}$ , $\sigma_{ ext{s}}$	$-1.499 \times 10^{-2}$	$1.320 \times 10^{-1}$
$^{235}U$	$\sigma_{ m c}$ , $\sigma_{ m f}$		$1.242 \times 10^{-1}$
$^{235}U$	$\chi$ , $\chi$	$9.161 \times 10^{-1}$	$1.030 \times 10^{-1}$
$^{238}U$	u , $ u$	$6.107 \times 10^{-2}$	$7.102 \times 10^{-2}$
$^{1}H$	$\sigma_{ m c}$ , $\sigma_{ m c}$	$-1.072 \times 10^{-1}$	$5.362 \times 10^{-2}$
<sup>1</sup> H	$\sigma_{ ext{ iny S}}$ , $\sigma_{ ext{ iny S}}$	$1.263 \times 10^{-1}$	$5.061 \times 10^{-2}$

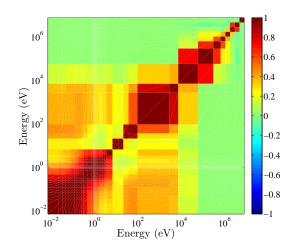


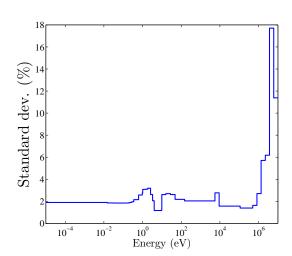
## Example: $k_{inf}$ S&U profiles (2)

Sensitivity plots



• <sup>238</sup>U capture covariance matrix Energy (eV)





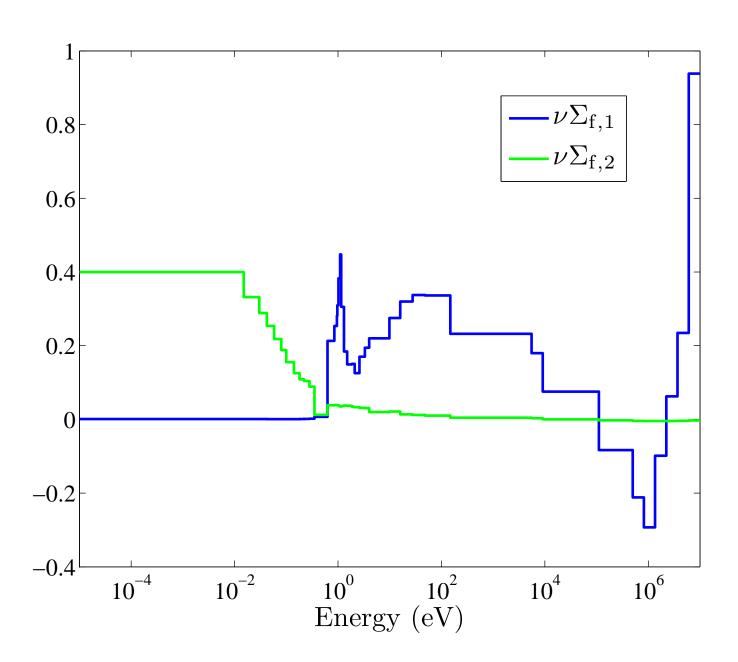


# **Example: Homogenized two-group cross-section** uncertainties

Response R	Value	Relative uncertainty $\frac{\Delta R}{R}$ (%)
$ u\Sigma_{\mathrm{f},1}$	$4.976 \times 10^{-3}$	$8.399 \times 10^{-1}$
$ u\Sigma_{\mathrm{f},2}$	$6.922 \times 10^{-2}$	$4.490 \times 10^{-1}$
$\Sigma_{\mathrm{a},1}$	$7.283 \times 10^{-3}$	$7.526 \times 10^{-1}$
$\Sigma_{ m a,2}$	$5.494 \times 10^{-2}$	$2.122 \times 10^{-1}$
$\Sigma_{ m c,1}$	$5.348 \times 10^{-3}$	$1.098 \times 10^{0}$
$\Sigma_{\mathrm{c,2}}$	$2.653 \times 10^{-2}$	$5.066 \times 10^{-1}$
$\Sigma_{\rm f,1}$	$1.935 \times 10^{-3}$	$5.563 \times 10^{-1}$
$\Sigma_{ m f,2}$	$2.841 \times 10^{-2}$	$3.244 \times 10^{-1}$



# **Example:** generalized adjoints

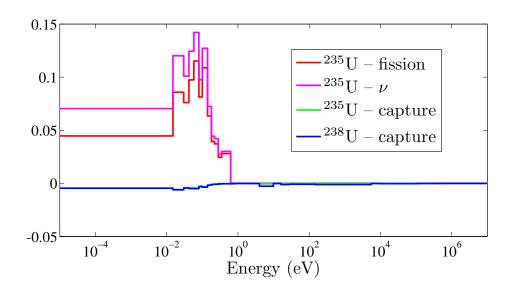




## **Example:** S&U profiles for $\nu\Sigma_{\rm f,2}$

•  $\nu \Sigma_{\rm f,2} = 6.922 \times 10^{-2}$ , relative uncertainty  $4.490 \times 10^{-1}\%$ 

Nuclide	Param. pair	Sensitivity	Contribution to $\frac{\Delta R}{R}$ (%)
<sup>235</sup> U	$\overline{ u},\overline{ u}$	$9.996 \times 10^{-1}$	$3.105 \times 10^{-1}$
<sup>235</sup> U	$\sigma_{ m f},\sigma_{ m f}$	$7.985 \times 10^{-1}$	$2.893 \times 10^{-1}$
<sup>235</sup> U	$\sigma_{ m f},\sigma_{ m c}$	$7.985 \times 10^{-1}$	$1.134 \times 10^{-1}$
<sup>238</sup> U	$\sigma_{ m c},\sigma_{ m c}$	$-4.406 \times 10^{-2}$	$7.257 \times 10^{-2}$
<sup>235</sup> U	$\sigma_{ m c},\sigma_{ m c}$	$-3.599 \times 10^{-2}$	$5.613 \times 10^{-2}$





#### **Summary**

- Sensitivity analysis
  - Adjoint-based approach
  - Brute force method
- Uncertainty analysis
  - Deterministic (requires sensitivities)
  - Statistical sampling
- S&U analysis with Monte Carlo method
  - Statistical sampling based on covariance data
  - Total Monte Carlo
  - Adjoint-based (exploit physical interpretation of adjoint)

