ICATPP 2011 Como, 3-7 October 2011

Quantifying the unknown

Issues in simulation validation and their experimental impact

Maria Grazia Pia

INFN Genova, Italy

Matej Batic, Marcia Begalli, Mincheol Han, Steffen Hauf, Gabriela Hoff, Chan Hyeung Kim, Markus Kuster, Lina Quintieri, Paolo Saracco, Hee Seo, Georg Weidenspointner, Andreas Zoglauer

> State University Rio de Janeiro, Brazil Tech. Univ. Darmstadt, Germany MPI Halbleiterlabor, Munich, Germany XFEL, Hamburg, Germany INFN Sezione di Genova, Italy INFN Laboratori Nazionali di Frascati, Italy Hanyang University, Seoul, Korea UC Berkeley, USA





Intrinsic limits to simulation validation

Epistemic uncertainties

Epistemic uncertainties originate from lack of knowledge

Relatively scarce attention so far in Monte Carlo simulation Studies in deterministic simulation *(especially for critical applications)*

Possible sources in Monte Carlo simulation

- incomplete understanding of fundamental physics processes, or practical inability to treat them thoroughly
- non-existent or conflicting experimental data for a physical parameter or model (for validation)
- applying a physics model beyond the experimental conditions in which its validity has been demonstrated

Epistemic uncertainties affect the **reliability** of simulation results

Can we quantify them?

Uncertainty quantification

- Epistemic uncertainties are **difficult to quantify**
 - due to their intrinsic nature
- No generally accepted method of measuring epistemic uncertainties
 - and their contributions to reliability estimation
- Various formalisms developed in the field of deterministic simulation
 - Interval analysis
 - Dempster-Shafer theory of evidence
- Not always directly applicable in Monte Carlo simulation
 - Adapt, reinterpret, reformulate existing formalisms
 - Develop new ones specific to Monte Carlo simulation

Warm-up exercise



- EGS5, EGS
- Penelope
- MCNP(X)
- PHITS
- SHIELD-HIT
- FLUKA
- SPAR, CALOR, CEM, LAHET, INUCL, GHEISHA, Liège INCL, Bertini

- p stopping powers
- Water ionisation potential
- δ -ray production
- Multiple scattering
- Nuclear elastic
- Nuclear inelastic
 - Cross sections
 - Preequilibrium
 - Nuclear deexcitation
 - Intranuclear cascade

δ-ray or no δ-ray Preequilibrium or no preequilibrium Weisskopf-Ewing or Weisskopf-Ewing Griffin-exciton or hybrid etc.

"Validation" in the literature

- Beam energy (and energy spread) is not usually known with adequate precision in therapeutical beam lines
 - What matters in clinical applications is the range
- Typical procedure: optimize the beam parameters to be used in the simulation by fitting them to experimental data
 - Determine beam energy, energy spread etc.
 - Use optimized beam parameter values in the simulation
- This is a calibration
- This is NOT validation

T. G. Trucano, L. P. Swiler, T. Igusa, W. L. Oberkampf, and M. Pilch, "Calibration, validation, and sensitivity analysis: What's what", *Reliab. Eng. Syst. Safety, vol. 91, no. 10-11, pp. 1331-1357, 2006.*

Simulation features

Realistic proton beam line

- Geometry from Geant4 *hadrontherapy* advanced example
- G.A.P. Cirrone et al., "Implementation of a New Monte Carlo GEANT4 Simulation Tool for the Development of a Proton Therapy Beam Line and Verification of the Related Dose Distributions", *IEEE Trans. Nucl. Sci., vol. 52, no. 1, pp. 262-265, 2005*

Water sensitive volume

- proton beam: E = 63.95 MeV, σ_E = 300 keV
- Physics modeling options in the simulation application configured through a class derived from G4VModularPhysicsList
- Geant4 8.1p02, 9.1(ref-04), 9.2p03, 9.3, 9.4





"interval analysis" across different options available in Geant4

Water mean ionisation potential



Proton stopping powers





Hadronic inelastic cross sections



99% confidence interval for inelastic scattering occurrences in water (Wellisch & Axen cross sections): **1688-1849**

Occurrences with **GHEISHA-like** cross sections: **1654**

Hadronic inelastic scattering models

No visible difference in Bragg peak profiles

Wald-Wolfowitz test

p-value< 0.001 for all model options except

p-value=0.360 for Liège cascade

preequilibrium = no preequilibrium

Maria Grazia Pia, INFN Genova

p-value (reference: Precompound)

| Geant4 | Test | Hadronic | Kolmogorov | Anderson | Cramer |
|---------|--------|----------------|------------|----------|-----------|
| version | range | model | Smirnov | Darling | von Mises |
| | | Bertini | 1 | 1 | 1 |
| | | LEP | 0.954 | 0.988 | 0.984 |
| | Whole | Liège | 1 | 1 | 1 |
| | | CHIPS | 1 | 1 | 1 |
| 9.3 | | GEM | 1 | 1 | 1 |
| | | Fermi break-up | 1 | 1 | 1 |
| | | Binary | 0.954 | 0.938 | 0.973 |
| | | Bertini | 1 | 1 | 1 |
| | Left | LEP | 0.945 | 0.961 | 0.979 |
| | branch | Liège | 1 | 1 | 1 |
| | | CHIPS | 1 | 1 | 1 |
| | | GEM | 1 | 1 | 1 |
| | | Fermi break-up | 1 | 1 | 1 |
| | | Binary | 0.945 | 0.858 | 0.962 |
| | | Bertini | 1 | 0.986 | 0.972 |
| | Right | LEP | 1 | 0.986 | 0.972 |
| | branch | Liège | 1 | 0.986 | 0.972 |
| | | CHIPS | 1 | 0.986 | 0.972 |
| | | GEM | 1 | 0.986 | 0.972 |
| | | Fermi break-up | 1 | 0.986 | 0.972 |
| | | Binary | 1 | 0.986 | 0.972 |
| 9.1 | Left | Bertini | 0.981 | 0.901 | 0.980 |
| | branch | LEP | 0.945 | 0.949 | 0.937 |
| 8.1 | Left | Bertini | 1 | 1 | 1 |
| | branch | LEP | 0.996 | 0.814 | 0.847 |



reference: Precompound



400

350

300

250

200

Entries

secondary p

Precompound

Bertini

LEP

Maria Grazia Pia, INFN Genova

Number of entries

reference: default Evaporation



Maria Grazia Pia, INFN Genova

Relative difference (%)

Cascade-preequilibrium

Precompound model activated through Binary Cascade w.r.t. standalone Precompound model



Transition between intranuclear cascade and preequilibrium determined by empirical considerations

In Geant4 Binary Cascade model cascading continues as long as there are particles above a 70 MeV kinetic energy threshold (along with other conditions required by the algorithm)

Some get lost on the way...



95% confidence intervals

Calibration: 50 and 200 GeV

Multiple scattering

G4MultipleScattering G4hMultipleScattering

| | RangeFactor | StepLimit | LatDisplacement | skin | geomFactor | Model |
|---------|-------------|-----------|-----------------|------|------------|------------|
| 8.1 | 0.02 | 1 | | | | UrbanMSC |
| 9.1 | 0.02 | 1 | 1 | 0 | 2.5 | UrbanMSC |
| 9.2p0.3 | 0.02 | 1 | 1 | 3 | 2.5 | UrbanMSC |
| 9.3 | 0.04 | 1 | 1 | 3 | 2.5 | UrbanMsc92 |
| 9.3 hMS | 0.2 | 0 | 1 | 3 | 2.5 | UrbanMsc90 |

Reference:



Total deposited energy



Equivalent across different hadronic models in the same release Differences across releases



Maria Grazia Pia, INFN Genova

Hadronic physics models

How well do we know basic physics parameters in the simulation?

Atomic electron binding energies

| EGS4 EGSnrc | Tol 1978 (Shirley-Uppsala) | | | | |
|----------------|--|--|--|--|--|
| EGS5 | Tol 1996 (Larkins(Sevier 1972)) | | | | |
| FLUKA | Unknown | | | | |
| GEANT 3 | Modified Bearden&Burr | | | | |
| Geant4 | EADL, Carlson+Williams, Tol 1978, (Bearden&Burr) | | | | |
| MCNP | Carlson | | | | |
| Penelope | Carlson, Tol 1978 | | | | |
| | | | | | |
| GUPIX | Sevier 1979 X-ray Data Williams | | | | |
| ISICS | Bearden & Burr, Williams Booklet | | | | |



Direct validation

Whole periodic table: ~1500 shells



Systematics

- Experimental configuration of measurement
- Chemical and physical effects (not accounted for in experimental error)

Usual GoF tests fail

- Underestimated uncertainties, small differences

Effects on simulation



X-ray energies

Comparison with experimental data (review by Deslattes et al.) Compton scattering

Doppler broadened photon spectrum



Benefits of quantifying uncertainties

- Epistemic uncertainties are reducible
 - Can be reduced or suppressed by extending knowledge
 - New experimental measurements
- Uncertainty quantification gives us guidance about
 - What to measure
 - What experimental **precision** is needed/adequate
 - Priorities: which uncertainties generate the worst systematic effects
- Measurements are not always practically possible
 - Uncertainty quantification to control systematics

Especially important in critical applications

Conclusions

- Evaluation of systematic effects associated with epistemic uncertainties
 The impact of
 - Sensitivity analysis (~interval analysis)
 - More refined methods: Dempster-Shafer
 - Methods specific to Monte Carlo simulation?

The **impact** of epistemic uncertainties depends on the **experimental** application environment

- Complementary statistical methods contribute to identify and quantify effects
 - Qualitative appraisal is not adequate
- Epistemic uncertainties are reducible
 - Can be reduced or suppressed by extending knowledge
 - New experimental measurements
- Uncertainty quantification gives us **guidance** about
 - What to measure
 - What experimental precision is needed/adequate
 - Priorities: which uncertainties generate the worst systematic effects

INFN UQ

- Project on Geant4 uncertainty quantification
- Discussed this week in INFN Computing Committee
- Complementary expertise
 - Geant4 developers
 - Experimental groups
 - Theory
 - Mathematics/statistics
- Collaboration is welcome!