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MONTE CARLO PARAMETER STUDIES
AND UNCERTAINTY ANALYSIS WITH MCNP5

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Monte Carlo Parameter Studies & Uncertainty Analyses With MCNP5

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Monte Carlo Parameter Studies & Uncertainty Analyses with MCNP5

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A software tool called *mcnp_pstudy* has been developed to automate the setup, execution, and collection of results from a series of MCNP5 Monte Carlo calculations. This tool provides a convenient means of performing parameter studies, total uncertainty analyses, parallel job execution on clusters, stochastic geometry modeling, and other types of calculations where a series of MCNP5 jobs must be performed with varying problem input specifications.



- Introduction
- mcnp_pstudy
- Examples
- Usage
 - Parameter definition
 - Parameter expansion
 - Constraints
 - Case setup & execution
 - Collecting & combining results
- Statistics
- Examples

Frequent Questions



How are calculated results affected by:

- Nominal dimensions
 - With minimum & maximum values ?
 - With as-built tolerances ?
 - With uncertainties ?
- Material densities
 - With uncertainties ?
- Data issues
 - Different cross-section sets ?
- Stochastic materials
 - Distribution of materials ?

Monte Carlo perturbation theory can handle the case of independent variations in material density, but does not apply to other cases.

Brute force approach:

Run many independent Monte Carlo calculations, varying the input parameters.



- To simplify & streamline the setup, running, & analysis of Monte Carlo parameter studies & total uncertainty analyses, a new tool has been developed: mcnp_pstudy
- Control directives are inserted into a standard MCNP input file
 - Define lists of parameters to be substituted into the input file
 - Define parameters to be sampled from distributions & then substituted
 - Define arbitrary relations between parameters
 - Specify **constraints** on parameters, even in terms of other parameters
 - Specify **repetitions** of calculations
 - Combine parameters as outer-product for parameter studies
 - Combine parameters as inner-product for total uncertainty analysis
- Sets up separate calculations
- Submits or runs all jobs
- Collects results



- Completely automates the setup/running/collection for parameter studies & total uncertainty analyses
 - Painless for users
 - 1 input file & run command can spawn 100s or 1000s of jobs
 - Fast & easy way to become the #1 user on a system
 (Added bonus: make lots of new friends in computer ops & program management.)
- Ideal for Linux clusters & parallel ASC computers:
 - Can run many independent concurrent jobs, serial or parallel
 - Faster turnaround: Easier to get many single-cpu jobs through the queues, rather than wait for scheduling a big parallel job
 - Clusters always have some idle nodes



mcnp_pstudy is written in perl

- 640 lines of perl (plus 210 lines of comments)
- Would have taken many thousands of lines of Fortran or C

Portable to any computer system

- Tested on Unix, Linux, Mac OS X, Windows
- For Windows PCs, need to execute under the Cygwin shell

Can be modified easily if needed

- To add extra features
- To accommodate local computer configuration
 - Node naming conventions for parallel cluster
 - Batch queueing system for cluster
 - Names & configuration of disk file systems (ie, local or shared)
 - Location of MCNP5 and MCNP5.mpi



MCNP input for simple Godiva calculation

MCNP input using mcnp_pstudy, Run 50 different cases Each with a distinct (odd) random seed

```
gdv
                                   gdv-A
  -18.74
                imp:n=1
                                   C @@@
                                         RNSEED = (2*int(rand(1000000))+1)
           -1
      0
           1
                imp:n=0
                                   C @@@
                                                = REPEAT 50
                                         XXX
                                      -18.74 -1
                                                  imp:n=1
      so 8.741
                                         0
                                              1
                                                  imp:n=0
1
kcode 10000 1.0 15 115
                                         so 8.741
ksrc 0 0 0
      92235 -94.73
                    92238 -5.27
                                   kcode 10000 1.0 15 115
m1
prdmp 0 0 1 1 0
                                        0 0 0
                                   ksrc
                                         92235 -94.73
                                                       92238 -5.27
                                   m1
                                  prdmp 0 0 1 1 0
                                   rand seed=RNSEED
```



 Within an MCNP input file, all directives to mcnp_pstudy must begin with

c @@@

To continue a line, use "\" as the last character

```
c @@@ xxx = 1 2 3 4 5 6 \
c @@@ 7 8 9 10
```

Parameter definitions have the form

```
c @@@ P = value or list
c @@@ P = ( arithmetic-expression )
```

Constraints have the form

```
c 000 CONSTRAINT = ( expression )
```

Control directives have the form

c @@@ OPTIONS = list-of-options

Parameter Definition



Parameters

- Like C or Fortran variables
- Start with a letter, contain only letters, integers, underscore
- Case sensitive
- Parameters are assigned values, either number(s) or string(s)
- Examples: R1, r1, U_density, U_den
- Single value

C @@@ P1 = value

List of values

C @@@ P2 = value1 value2 ... valueN

List of N random samples from a Normal probability density

C @@@ P3 = normal N ave dev

List of N random samples from a Uniform probability density

C @@@ P4 = uniform N min max

Parameter Definition



Arithmetic expression

```
C @@@ P5 = (arithmetic-statement)
```

- Can use numbers & previously defined parameters
- Can use arithmetic operators +, -, *, /, % (mod), ** (exponentiation)
- Can use parentheses ()
- Can use functions: sin(), cos(), log(), log10(), exp(), int(), abs(), sqrt()
- Can generate random number in (0,N): rand(N)
- Must evaluate to a single number
- Examples:

```
c @@@ FACT = normal 1 1.0 .05
c @@@ UDEN = ( 18.74 * FACT )
c @@@ URAD = ( 8.741 * (18.74/UDEN)**.333333 )
```

Repetition (list of integers, 1..N)

```
C 000 P6 = repeat N
```

Parameter Definition



Examples

```
C rod height in inches, for search
C @@@ HROD = 5 10 15 20 25 30 35
                                       40 45 50
  nominal dimension, with uncertainty
C @@@ X1 = normal 25
                        1.234 .002
C dimension, with min & max
C @@@ X2 = uniform 25
                        1.232 1.236
  try different cross-sections
  aaa
      U235 = 92235.42c 92235.49c 92235.52c \
  aaa
             92235.60c 92235.66c
C different random number seeds (odd)
  aaa
      SEED = (2*int(rand(1000000)) + 1)
```



Random Sampling of Parameters

For parameters sampled from a **Uniform** probability density, each sample is obtained as

$$P = xmin + (xmax-xmin)*rand()$$

 For parameters sampled from a Normal probability density, each sample is obtained using the Box-Muller scheme

$$P = ave + dev * sqrt(-2*log(rand()) * sin(2*pi*rand())$$

Other probability densities could easily be added

Arithmetic Expressions & Constraints

- Evaluated within perl, using the eval function
- Must conform to perl rules for arithmetic

Parameter Expansion



 After all parameters are defined, mcnp_pstudy expands them into sets to be used for each separate MCNP calculation

Outer product expansion: All possible combinations.

Parameters specified first vary fastest.

Inner product expansion: Corresponding parameters in sequence.
 If not enough entries, last is repeated.

Example: c @ @ @ A = 1 2

c @ @ @ B = 3 4

c @ @ @ C = 5

Outer: Case 1: A=1, B=3, C=5

Case 2: A=2, B=3, C=5

Case 3: A=1, B=4, C=5

Case 4: A=2, B=4, C=5

Inner: Case 1: A=1, B=3, C=5

Case 2: A=2, B=4, C=5

Constraint Conditions



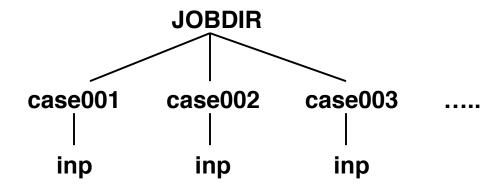
- After all parameters are defined & expanded, constraint conditions are evaluated
- Constraints involve comparison operators (>, <, >=, <=, ==, !=) or logical operators (&& (and), II (or), ! (not)), and may involve arithmetic or functions
- Constraints must evaluate to True or False
- If a any constraint is not met, the parameters for that case are discarded & re-evaluated until all of the constraints are satisfied

Example

```
C pick a random direction
C @@@ ANGLE = ( 6.2831853 * rand(1) )
C @@@ UUU = ( cos(ANGLE) )
C @@@ VVV = ( sin(ANGLE) )
C
C same, using CONSTRAINT to implement rejection scheme
c @@@ RN1 = ( 2.*rand(1) - 1. )
C @@@ RN2 = ( 2.*rand(1) - 1. )
C @@@ CONSTRAINT = ( RN1**2 + RN2**2 < 1.0 )
C @@@ UUU = ( RN1 / srqt(RN1**2 + RN2**2) )
C @@@ VVV = ( RN2 / sqrt(RN1**2 + RN2**2) )</pre>
```



Directory structure for MCNP5 jobs



- Unix filesystem conventions followed
 JOBDIR/case001/inp, JOBDIR/case002/inp, etc.
- Values of parameters are substitued into the original MCNP5 input file to create the input files for each case
 - Parameters substituted only when exact matches are found
 - Example: **UDEN** matches **UDEN**, and not **UDEN1**, **UDENS**, **uden**

Job Options



Specifying options for running jobs

Can be specified on the mcnp pstudy command-line

```
mcnp pstudy -inner -setup -i inp01
```

Within the INP file

c @@@ OPTIONS = -inner

Common options

The INP filename is str, default = inp -i str Use str as the name of the job directory -jobdir str Use str as the name for case directories -case str Append str to the MCNP5 run command, -mcnp opts str may be a string such as 'o=outx tasks 4' -bsub opts str str is appended to the LSF bsub command -inner Inner product approach to case parameter substitution Outer product approach to case parameter substitution -outer Create the cases & INP files for each -setup Run the MCNP5 jobs on this computer -run Submit the MCNP5 jobs using LSF bsub command -submit Collect results from the MCNP5 jobs -collect

Running or Submitting Jobs



- Jobs can be run on the current system, or can be submitted to a batch queueing system (e.g., LSF)
- Tally results & K-effective can be collected when jobs finish

Examples:

```
bash: mcnp_pstudy -inner -i inp01 -setup
bash: mcnp_pstudy -inner -i inp01 -run
bash: mcnp_pstudy -inner -i inp01 -collect

bash: mcnp_pstudy -inner -i inp01 -setup -run -collect

bash: mcnp_pstudy -inner -i inp01 -setup -submit
... wait till all jobs complete...
bash: mcnp_pstudy -inner -i inp01 -collect
```

Combining Results



 Tally results & K-effective from separate cases can be combined using batch statistics:

$$\overline{X} = \frac{1}{M} \cdot \sum_{k=1}^{M} X_k \qquad \sigma_{\overline{X}} = \sqrt{\frac{1}{M-1} \cdot \left[\frac{1}{M} \sum_{k=1}^{M} X_k^2 - \overline{X}^2 \right]}$$

where **M** is the number of cases & X_k is some tally or Keff for case **k**

 Variance due to randomness in histories decreases as 1/M, but variance due to randomness in input parameters is constant

$$\sigma \frac{2}{X} \approx \sigma \frac{2}{X}$$
, Monte + $\sigma \frac{2}{X}$, Initial Carlo Conditions

Varies as 1/M

~ Constant



Vary the fuel density randomly & adjust radius for constant mass, for 50 cases

Vary fuel density & mass independently, for 50 cases

```
qdv-D
                                              qdv-E
c vary fuel density - normal, 5%sd,
                                              c vary fuel radius - normal, 5%sd
c adjust the radius to keep constant mass
                                              c vary fuel density- normal, 5%sd
C
                                              C
c @@@ FACT= normal 50 1.0 .05
                                              c @@@ OPTIONS = -inner
c @@@ UDEN= ( 18.74*FACT )
c @@@ URAD= ( 8.741*(18.74/UDEN)**.333333 )
                                              c @@@ DFACT = normal 50
                                              c @@@ UDEN = ( DFACT * 18.74 )
C
                      imp:n=1
1
         -UDEN
                 -1
2
                      imp:n=0
                                              c @@@ UFACT = normal 50
                  1
                                              c @@@ URAD
                                                          = (UFACT * 8.741)
         URAD
1
      so
                                              1
                                                      -UDEN
                                                              -1
                                                                    imp:n=1
kcode 10000 1.0 15
                                                                    imp:n=0
                      115
                                                               1
ksrc
      0. 0. 0.
                     92238 - 5.27
      92235 - 94.73
m1
                                              1
                                                      URAD
                                                   so
prdmp 0 0 1 1 0
                                              kcode 10000 1.0
                                                                15
                                                                    115
                                              ksrc 0.0.0.
                                                    92235 -94.73
                                                                   92238 -5.27
                                              m1
                                              prdmp 0 0 1 1 0
```



Table 1. Results from varying parameters in the Godiva problem

Problem	Description	K-effective	σ _{K-eff}
base	Base case, discard 15 initial cycles, retain 100 cycles with 10K histories/cycle, 1M total histories	0.9970	0.0005
А	Repeat the base problem 50 times, 50M total histories	0.9972	0.0001
В	Vary the fuel density only: sample from a normal distribution with 5% std.dev, 50M total histories	0.9961	0.0061
С	Vary the fuel radius only: sample from a normal distribution with 5% std.dev, 50M total histories	1.0057	0.0051
D	Vary the enrichment only, sample from a normal distribution with 5% std.dev, 50M total histories	0.9890	0.0027
E	Sample the fuel density from a normal distribution with 5% std.dev, and adjust the fuel radius to keep constant fuel mass, 50M total histories	0.9966	0.0042
F	Sample the fuel density from a normal distribution with 5% std.dev, and independently sample the radius from a normal distribution with 5% std.dev, 50M total histories	1.0073	0.0076



Parameter studies

- Run a series of cases with different control rod positions
- Run a series of cases with different soluble boron concentrations
- Run a series of cases sampling certain dimensions from a Uniform or Normal probability density
- Run a series of cases substituting different versions of a cross-section

Total uncertainty analysis

Run a series of cases varying all input parameters according to their uncertainties

Parallel processing using a "parallel jobs" approach

- Running N separate jobs with 1 cpu each will be more efficient than running 1 job with N cpus
- Eliminates queue waiting times while cpus are reserved
- Take advantage of cheap Linux clusters

Simulation of stochastic geometry

 Run a series of cases with portions of geometry sampled randomly, with a different realization in each case

Conclusions



- mcnp_pstudy works
 - In use regularly at LANL for a variety of real applications
 - Developed on Mac & PC, runs anywhere
 - Easy to customize, if you have special needs
- To get it:

– MCNP5 website: www-xdiv.lanl.gov/x5/MCNP

FB Brown, JE Sweezy, RB Hayes, "Monte Carlo Parameter Studies and Uncertainty Analyses with MCNP5", PHYSOR-2004, Chicago, IL (April, 2004)