

## Critical Experiment Correlations

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### Outline

- Correlation sources and computational methodology
- Fuel rod position uncertainty, effect, and model creation
- Impact of correlations on validation



# Correlation sources and complications

- Shared fissile material common source of correlations
  - 239 LEU array cases in Handbook from PNL: only 2 unique enrichments
  - 109 of the 239 cases at 2.35 wt% <sup>235</sup>U: only 2 unique square pitch values
- Current validation approaches use trending and non-trending techniques, but both approaches assume uncorrelated data points
- The data points are clearly not completely uncorrelated, but current methods neglect the correlations
- This work develops a procedure for calculating these critical experiment correlation coefficients via Monte Carlo sampling
  - Further development is needed to incorporate correlations in validation techniques



## Correlation coefficient calculation methodology

- Random sampling of virtually all input parameters: compositions, geometry, and temperatures
- Hundreds of complete inputs are created for each experiment
- Components that are shared between experiments get the same sampled value in each realization
- Correlation coefficient is ratio of covariance to product of standard deviations of each individual experiment

$$\rho_{xy} = \frac{\operatorname{cov}(x, y)}{\sigma_x \sigma_y}$$

 Essentially this is the fraction of total uncertainty shared between the two experiments



## Fuel rod position uncertainty

- Fuel rod positions can have a huge impact on correlations for LCT systems
- Systems that are not near optimum moderation are sensitive to fuel rod placement as it controls the moderation in the system
- Two main assumptions about pin positions have been investigated:
  - 1. Pin pitch is fundamental uncertain parameter and all pin pitches in the experiment are identical
  - 2. Pin position is fundamental uncertain parameter and all pin pitches are unique based on random placement of pins
- Shared pitch assumption leads to large, shared uncertainty and thus high correlation coefficients



## LCT-042 results for different pin placements

	42-1	42-2	42-3	42-4	42-5	42-6	42-7
42-1	1	0.97	0.99	0.97	0.99	0.99	0.98
42-2	0.97	1	0.97	0.96	0.97	0.98	0.96
42-3	0.99	0.97	1	0.97	0.99	0.99	0.97
42-4	0.97	0.96	0.97	1	0.98	0.98	0.96
42-5	0.99	0.97	0.99	0.98	1	0.99	0.98
42-6	0.99	0.98	0.99	0.98	0.99	1	0.98
42-7	0.98	0.96	0.97	0.96	0.98	0.98	1

	42-1	42-2	42-3	42-4	42-5	42-6	42-7
42-1	1	0.33	0.52	0.42	0.52	0.63	0.43
42-2	0.33	1	0.32	0.48	0.42	0.37	0.19
42-3	0.52	0.32	1	0.29	0.59	0.61	0.48
42-4	0.42	0.48	0.29	1	0.46	0.31	0.26
42-5	0.52	0.42	0.59	0.46	1	0.67	0.49
42-6	0.63	0.37	0.61	0.31	0.67	1	0.46
42-7	0.43	0.19	0.48	0.26	0.49	0.46	1

Pitch sampled: all pitches are the same and are the same for all cases All fuel rod positions are sampled independently and differently in each case

Coefficients range from 0.96-0.99

Coefficients range from 0.19–0.67

So it makes a big difference, but how do I build that kind of model in SCALE?



## Mechanics of Model Building

- Perturbing each pin in the model requires unique units in KENO
- Generating hundreds of units effectively requires scripting
- SCALE 6.2 contains a TemplateEngine to automate input generation
  - Template input generated by hand (shown on future slides)
  - Template indexed so each unit gets a unique number
  - Thousands of lines of input generated in seconds
- Both KENO (20,000+ lines) and Sampler inputs generated with TemplateEngine (~70,000 – 100,000 lines)
- All inputs built using SCALE <u>NO OUTSIDE SCRIPTING</u>



### Mechanics of Model Building – KENO Template

#for(i=1; i<=459; i=i+1){</pre> unit **#{i+1000**} ' bottom end plug - below lower grid plate cylinder 2 1 0.47 0.97 0.00 origin 0.63 0.63 cuboid 4 1 1.26 0.0 1.26 0.0 0.97 0.00 unit **#{i+2000**} ' bottom end plug - in lower grid plate cylinder 2 1 0.47 0.25 0.0 origin 0.63 0.63 cylinder 4 1 0.5 0.25 0.0 origin 0.63 0.63 cuboid 1000 1 1.26 0.0 1.26 0.0 0.25 0.0

#### unit 1001

' bottom end plug - below lower grid plate cylinder 2 1 0.47 0.97 0.00 origin 0.63 0.63 cuboid 4 1 1.26 0.0 1.26 0.0 0.97 0.00 unit 2001 Output is KENO Input ' bottom end plug - in lower grid plate cylinder 2 1 0.47 0.25 0.0 origin 0.63 0.63 cylinder 4 1 0.5 0.25 0.0 origin 0.63 0.63 cuboid 1000 1 1.26 0.0 1.26 0.0 0.25 0.0

**CAK RIDGE** National Laboratory

TemplateEngine

TemplateEngine

Input

## Mechanics of Model Building – Sampler Template

TemplateEngine Input

```
read variable[origin_x_1000_#{i}_039_001]
    distribution=normal
    value = 0.63
    stddev = 0.0351
    minimum = 0.5
    maximum = 0.76
    siren="/csas5/geometry/unit[id='#{i+1000}',id='#{i+3000}',id='#{i+9000}']/cylinder/region_origin[decl='origin']/value[1:5:2]"
    cases = Case39_1 end
end variable
```

```
TemplateEngine Output
is Sampler Input
```

```
read variable[origin_x_1000_1_039_001]
  distribution=normal
  value = 0.63
  stddev = 0.0351
  minimum = 0.5
  maximum = 0.76
  siren="/csas5/geometry/unit[id='1001',id='3001',id='9001']/cylinder/region_origin[decl='origin']/value[1:5:2]"
  cases = Case39_1 end
end variable
```



## Random pin models



#### 75 realizations, looped



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## Trending results: qualitative results



 $\begin{array}{l} LCT\text{-}079\text{-}001\\ \text{High correlation} ~1.5\% \ \Delta k\\ \text{Low correlation} ~0.1 \ \% \ \Delta k \end{array}$ 



LCT-079-006 High correlation ~1.0%  $\Delta k$ Low correlation ~0.1 %  $\Delta k$ 



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- Extremely high correlation coefficients can change trending USL by  $1.0\% \Delta k$  or more (including uncertainty)
- Impact of correlations is case dependent
  - 1.5%  $\Delta k$  for LCT-079-001 vs. 1.0%  $\Delta k$  for LCT-079-006
- Results for low correlation coefficients are close to the uncorrelated results
  - ~0.1%  $\Delta k$  impact for the two LCT-079 cases (trending)



## Conclusions: All systems

- Methodology for determination of critical experiment correlations developed based on Monte Carlo sampling technique
- Important parameters for correlations can be determined, and may be different from important parameters of the systems
- Correlation coefficients for the experiments are different than the correlation coefficients for the uncertainties of the experiments
- In-depth knowledge of the critical experiments is essential for accurate correlation coefficient determination
  - Detail of current ICSBEP evaluations does not provide the required level of information

