Sparse-Grid Stochastic Collocation Uncertainty Quantification Convergence for Multigroup Diffusion

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- 1 Deterministic Problem
- 2 UQ Methodology
- 3 Results
- 4 Ongoing Work





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Low-Enrichment Nuclear Reactor

- Homogenized Multiplying Medium
- Steady State Operation

5	5	5	5	5	5	5	5	5	5	5
5	5	5	5	5	5	5	5	5	5	5
3	3	3	3	3	3	3	5	5	5	5
3	3	3	3	3	3	3	4	5	5	5
2	1	1	1	1	2	2	3	3	5	5
2	1	1	1	1	2	2	3	3	5	5
1	1	1	1	1	1	1	3	3	5	5
1	1	1	1	1	1	1	3	3	5	5
1	1	1	1	1	1	1	3	3	5	5
1	1	1	1	1	1	1	3	3	5	5
2	1	1	1	1	2	2	3	3	5	5





Neutronics

- Diffusion Approximation
- Multigroup, G = 2
- Neglect Upscatter
- $\chi(g=1)=1$
- Solved using JFNK (GMRES) in trilinos

Uncertainties:

- Aleatoric in interaction probability
- lacktriangle Epistemic in measurements of σ





Neutron Transport

Two-Group, Two-Dimension, Diffusion Approximation

$$\begin{split} -\nabla \cdot (D_{1}(\bar{x})\nabla \phi_{1}(\bar{x})) + \left(\Sigma_{a}^{(1)}(\bar{x}) + \Sigma_{s}^{(1\to 2)}(\bar{x})\right)\phi_{1}(\bar{x}) &= \frac{1}{k(\phi)} \sum_{g'=1}^{2} \nu_{g'} \Sigma_{f}^{(g')}(\bar{x})\phi_{g'}(\bar{x}) \\ -\nabla \cdot \left(D_{2}(\bar{x})\nabla \phi_{2}(\bar{x})\right) + \Sigma_{a}^{(2)}(\bar{x})\phi_{2}(\bar{x}) &= \Sigma_{s}^{(1\to 2)}(\bar{x})\phi_{1}(\bar{x}) \end{split}$$

$$-D \left. \frac{\partial \phi^{\mathsf{in}}}{\partial x_i} \right|_{\partial D} = 0, \ \ i = 1, 2, \ \ x \in \partial_{\mathsf{top}} D \cup \partial_{\mathsf{right}} D$$
$$\left. \frac{\partial \phi}{\partial x_i} \right|_{\partial D} = 0, \ \ i = 1, 2, \ \ x \in \partial_{\mathsf{left}} D \cup \partial_{\mathsf{bottom}} D$$





Neutron Transport

Benchmark

k = 1.00007605445

Region	Group	D_g	$\Sigma_{a,g}$	$\nu \Sigma_{f,g}$	$\Sigma_s^{1,2}$
1	1	1.255	8.252e-3	4.602e-3	2.533e-2
	2	2.11e-1	1.003e-1	1.091e-1	
2	1	1.268	7.181e-3	4.609e-3	2.767e-2
	2	1.902e-1	7.047e-2	8.675e-2	
3	1	1.259	8.002e-3	4.663e-3	2.617e-2
	2	2.091e-1	8.344e-2	1.021e-1	
4	1	1.259	8.002e-3	4.663e-3	2.617e-2
	2	2.091e-1	7.3324e-2	1.021e-1	
5	1	1.257	6.034e-4	0	4.754e-2
	2	1.592e-1	1.911e-2	0	

Introduce 10% Uncertainty





UQ Methods

- Analog Monte Carlo
- Stochastic Collocation on Sparse Grids
- Anisotropic Stochastic Collocation

Uncertainty space

$$k(\textit{D}_{\textit{g}}, \Sigma^{\textit{g}}_{\textit{c}}, \nu \Sigma^{\textit{g}}_{\textit{f}}, \Sigma^{1 \rightarrow 2}_{\textit{s}}, \ldots) \rightarrow \textit{u}(\textit{Y}) \equiv \textit{u}(\textit{Y}_{1}, \textit{Y}_{2}, \ldots, \textit{Y}_{\textit{N}}), \textit{Y} \in \Gamma$$

Compare moments, $P(Y) = \prod_{n=1}^{N} \rho(Y_n)$

$$\mathbb{E}[u^r] \equiv \int_{\Omega} u(Y)^r P(Y) d\Omega$$





Monte Carlo

$$\mathbb{E}[u'] \approx \frac{1}{M} \sum_{m=1}^{M} u\left(Y^{(m)}\right)'$$





Stochastic Collocation

$$u(Y) \approx u_{h,\eta,\Lambda(L)}(Y) = \sum_{k=0}^{\eta} u(Y^{(k)}) \mathcal{L}_k(Y)$$
$$\mathcal{L}_k(Y) = \prod_{n=1}^{N} \mathcal{L}_{k_n}(Y_n)$$
$$\mathcal{L}_{k_n}(Y_n) = \prod_{j=1}^{i} \frac{Y_n - Y_n^{(i)}}{Y_n^{(k_n)} - Y_n^{(i)}}$$
$$\mathbb{E}[u(Y)] \approx \mathbb{E}[u_h(Y)] = \sum_{j=1}^{\eta} w_k \ u_h(Y^{(k)})$$





Stochastic Collocation: Index Set $\Lambda(L)$

Tensor Product:

$$\Lambda_{\mathsf{TP}}(L) = \Big\{ \bar{p} = [p_1, ..., p_N] : \max_{1 \le n \le N} p_n \le L \Big\}, \eta = (L+1)^N$$

Total Degree:

$$\Lambda_{TD}(L) = \left\{ \bar{p} = [p_1, ..., p_N] : \sum_{n=1}^{N} p_n \le L \right\}, \eta = {L + N \choose N}$$

Hyperbolic Cross:

$$\Lambda_{HC}(L) = \left\{ \bar{p} = [p_1, ..., p_N] : \prod_{n=1}^N p_n + 1 \le L + 1 \right\}, \eta \le (L+1)(1 + \log(L+1))^{N-1}$$





Stochastic Collocation: Index Set $\Lambda(L)$

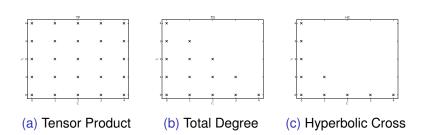


Figure: Index Set Examples: N = 2, L = 4





Stochastic Collocation on Sparse Grids

$$u(Y) \approx S_{N,\Lambda(L)}[u](Y) = \sum_{\substack{i \in \Lambda(L) \\ j=\{0,1\}^N, \\ i+j\in\Lambda(L)}} c(i) \bigotimes_{n=1}^N \mathcal{U}_{n,p(i_n)}[u](Y),$$

$$\bigotimes_{n=1}^{N} \mathcal{U}_{n,p(i_n)}[u](Y) \equiv \sum_{k_1=0}^{p(i_1)} \cdots \sum_{k_N=0}^{p(i_N)} u_h \Big(Y^{(k_1)}, \cdots, Y^{(k_N)}\Big) \prod_{n=1}^{N} \mathcal{L}_{k_n}(Y_n),$$

$$= \sum_{k=0}^{p(\vec{i})} u_h \Big(Y^{(k)}\Big) \mathcal{L}_k(Y),$$

Stochastic Collocation on Sparse Grids

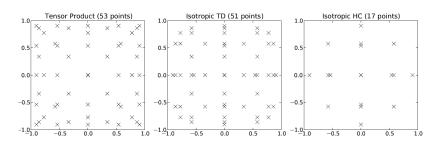


Figure: Sparse Grids, N = 2, L = 4, p(i) = i, Legendre points





Stochastic Collocation on Sparse Grids

		TP	TD		HC	
Ν	L	$ \Lambda(L) $	$ \Lambda(L) $	η	$ \Lambda(L) $	η
3	4	125	35	165	16	31
	8	729	165	2,097	44	153
	16	4,913	969	41,857	113	513
	32	35,737	6,545	1,089,713	309	2,181
5	2	293	21	61	11	11
	4	3,125	126	781	31	71
	8	59,049	1,287	28,553	111	481

Table: Index Set and Collocation Size Comparison





Anisotropic Sparse Grids

$$\tilde{\Lambda}_{\mathsf{TD}}(L) = \left\{ \bar{p} = [p_1, ..., p_N] : \sum_{n=1}^{N} \alpha_n p_n \leq |\vec{\alpha}|_1 L \right\}$$

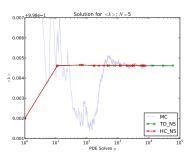
$$\tilde{\Lambda}_{HC}(L) = \left\{ \bar{p} = [p_1, ..., p_N] : \prod_{n=1}^{N} (p_n + 1)^{\alpha_n} \le (L + 1)^{|\vec{\alpha}|_1} \right\}$$



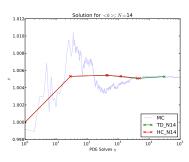


Results

< k > Solutions



(a)
$$< k >$$
, N=5

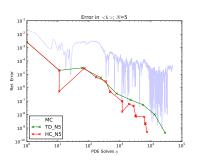


(b)
$$< k >$$
, N=14

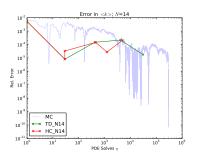


Results

Convergence



(a) Error in $\langle k \rangle$, N=5



(b) Error in $\langle k \rangle$, N=14





Continuing Efforts

- Increased Material Complexity
- Algorithmic Anisotropic Grids
- HDMR
- Multiphysics (Neutronics, Thermal Hydraulics, Materials)



