
TITAN Deterministic SPECT Simulation

- The collimator in SPECT poses a challenge for deterministic modeling:
 - Spatial discretization
 - Angular discretization



- Typical dimensions include:
 - Hole diameter ~0.18 cm
 - Septa thickness ~0.02 cm
 - Length ~3.3 cm
 - Acceptance Angle ~1.6°





Determination of the *importance function* (ψ^*)

- The TITAN multigroup, parallel hybrid transport code system
- TITAN was developed by Yi and Haghighat in 2006. It is a hybrid deterministic code by partitioning the problem domain into coarse meshes and allowing the use of different transport solvers within each coarse mesh.
- TITAN is written in F90 with some features from F2003 (object oriented23), and uses MPI for parallel processing
- The current version of TITAN allows for the following solvers:
 - 1) Discrete Ordinates (S_N) Solver
 - 2) Characteristics Method (CM) Solver
 - 3) Simplified ray-tracing with fictitious quadrature set

*C. Yi and A. Haghighat, "A 3-D Block-Oriented Hybrid Discrete Ordinates and Characteristics Method," *Nuclear Science and Engineering*, **164**, pp. 221-247 (2010).







4-Stage TITAN Hybrid formulation for SPECT simulation Stage 1 Stage 4 Stage 2 Stage 1- Sn Stage 3 calculation in phantom Scattering not S_N Solver Phantom simulated Stage 2 – Selection of fictitious angular quadrature & Detector: (Not simulated) Circular OS (COS) directions Stage 3 – Sn with fictitious quadrature Stage 4 – ray tracing

Fictitious quadrature directions



Example of Benchmarking TITAN Projection Images

SIMIND Comparison

NURBS-based cardiac-torso (NCAT) phantom with Tc-99m (140 keV)



SIMIND generated projection images







Right lateral

TITAN generated projection images



Left lateral





Right lateral

Number of Projection Images	1	4	8	45	90
SIMIND Time (sec)	17	67	140	754	1508
TITAN Time (sec)	200	202	212	274	352

Times are for a single processor



Image Reconstruction

- Filtered backprojection (FBP) (Cormack 1963)
 - Analytic image reconstruction
 - Traditional standard for reconstruction due to speed and simplicity
 - Issues: filter choice, amplification of high-freq. noise, streak artifacts, cannot incorporate system details
- Algebraic reconstruction technique (ART) (Gordon *et al.* 1970)
 - Iterative constraint-based reconstruction
 - Allows the incorporation of prior knowledge
 - Issues: noisy, computationally expensive
- Maximum likelihood expectation maximization (ML-EM) (Shepp & Vardi 1982)
 - Iterative statistical reconstruction
 - For emission tomography, has recently surpassed FBP in popularity
 - Advantages include: Poisson statistics, nonnegativity constraint, incorporation of system details
 - Issues: increasing noise, computationally expensive



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ML-EM Brief Derivation

Mean number of photons detected in detector bin *d*:

$$\overline{n}_d = \mathop{\bigotimes}_{b=1}^{B} p_{b,d} \hat{I}_b$$

 $p_{b,d}$: probability that photon emitted in voxel *b* is detected in bin *d* (system matrix) \hat{f}_b : mean number of emissions in voxel *b*

Number of detected particles is a Poisson random variable, so the probability of detecting n_d^* photons in detector bin *d*:

$$P(n_d^*) = e^{-\overline{n}_d} \frac{\overline{n}_d^{n_d^*}}{n_d^*!}$$

Likelihood function:

$$L(\hat{l}) = P(n_d^* | \hat{l}) = \bigcap_{d=1}^{D} P(n_d^*) = \bigcap_{d=1}^{D} \frac{e^{-\bar{n}_d} \bar{n}_d^{n_d^*}}{n_d^*!}$$

Log-likelihood will have the same maximum location: $\ln(L(\hat{l})) = \bigotimes_{d=1}^{D} \left(-\overline{n}_{d} + n_{d}^{*} \ln(\overline{n}_{d}) - \ln(n_{d}^{*}!) \right)$ $= \bigotimes_{d=1}^{D} \stackrel{e}{\ominus} - \bigotimes_{b=1}^{B} p_{b,d} \hat{l}_{b} + n_{d}^{*} \ln(\bigotimes_{b=1}^{B} p_{b,d} \hat{l}_{b}) - \ln(n_{d}^{*}!) \stackrel{i}{\bigcup}$ $\stackrel{i}{\bigcup}$

Take derivative and set to zero to find maximum:

$$\frac{\sqrt{\ln(L(\vec{l}))}}{\sqrt{n}_{d}} = -\overset{D}{\underset{d=1}{a}} p_{b,d} + \overset{D}{\underset{d=1}{a}} \frac{n_{d}^{*}}{\overset{D}{\underset{b'=1}{a}} p_{b',d} \hat{l}_{b'}} p_{b,d} = 0$$

Multiply by \hat{I}_{b} and solve: $\hat{I}_{b}^{(i+1)} = \frac{\hat{I}_{b}^{(i)}}{\hat{a}_{d=1}^{D} p_{b,d}} \hat{a}_{d=1}^{D} \frac{n_{d}^{*}}{\hat{a}_{b'=1}^{B} p_{b',d}} \hat{I}_{b'}^{(i)}} p_{b,d}, \ b = 1, \square, B$ 46



ML-EM can be viewed as a series of projections and backprojections





TITAN with Image Reconstruction (TITAN-IR)



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Analyzing TITAN-IR

- 1) 2-D elliptical water phantom with two circles of high intensity source (i.e., lesions)
- 2) Jaszczak: 3-D quality

assurance phantom, cold sphere region

3) NCAT: NURBS-based cardiactorso, 3-D heterogeneous phantom









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Reconstruction Analysis

- Visually display reconstructed images
- Plot profiles through important areas of reconstructed images •
- Quality metrics: ۲
 - Mean relative error (MRE)

MRE =
$$\frac{1}{N_d} \mathop{a}\limits_{d=1}^{N_d} \frac{\left| \hat{n}_d^{(i)} - n_d^* \right|}{n_d^*}$$

Mean squared error (MSE) N_{J}

MSE =
$$\frac{1}{N_d} \sum_{d=1}^{N_d} (\hat{n}_d^{(i)} - n_d^*)^2$$

 $\hat{n}_{d}^{(i)}$ = counts in detector bin *d* at iteration *i* n_d^* = measured counts in detector bin d

$$\overline{I}_l$$
 = average source intensity in lesion
 \overline{I}_0 = average reference background intensity

• Contrast
$$C_l = -$$

$$C_l = \frac{\overline{I}_l - \overline{I}_0}{\overline{I}_0}$$

• Noise =
$$\frac{1}{\overline{I_0}} \overset{\mathfrak{A}}{\underset{e}{\varsigma}} \frac{\overset{\mathfrak{A}}{\overset{}}}{\underset{i=1}{\overset{N_V}{\underset{i=1}{\circ}}} (I_i - \overline{I_0})^2 \overset{"o"}{\overset{}}{\underset{i=1}{\overset{\circ}{\underset{i=1}{\circ}}} (I_i - \overline{I_0})^2 \overset{"o"}{\overset{}}{\underset{i=1}{\overset{\circ}{\underset{i=1}{\circ}}} (I_i - \overline{I_0})^2 \overset{"o"}{\overset{}}{\underset{i=1}{\overset{\circ}{\underset{i=1}{\circ}}} (I_i - \overline{I_0})^2 \overset{"o"}{\overset{"o"}{\underset{i=1}{\circ}}} (I_i - \overline{I_0})^2 \overset{"o"}{\underset{i=1}{\circ}} (I_i - \overline{I_0})^2 \overset{"o"}{\underset{i$$

2) Jaszczak Cold Sphere Phantom

- 6 cold spheres with radii of 0.635, 0.795, 0.955, 1.27, 1.59, and 1.9 cm
- 185 MBq Tc-99m source (140 keV)
- Reference projection data obtained at 64 angles over 360° using SIMIND
- System matrix *p(b,d)*
 - Generated by Image Reconstruction Toolbox in MATLAB (models attenuation but not scatter)
 - Dimensions of (64x64x32) by (64x32x64)
- Initial guess is a uniform source distribution
- Three cases of projection data:
 - 1) No noise & no collimator blur
 - 2) Noisy & no collimator blur
 - 3) Noisy collimated data







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Jaszczak Cold Sphere Phantom: Noisy Collimated Projection Data

Collimator	Hole Diameter	Septa Thickness	Length	Acceptance Angle
GE-LEGP*	0.25 cm	0.03 cm	4.10 cm	1.83°
SE-LEHR ⁺	0.111 cm	0.016 cm	2.405 cm	1.39°

*General Electric – Low energy, general purpose collimator

⁺Siemens – Low energy, high resolution collimator



Jaszczak Cold Sphere Phantom: Noisy Collimated Projection Data



Contrast in each cold sphere (radius) for noisy GE-LEGP (1.83°) projection data

Contrast in each cold sphere (radius) for noisy SE-LEHR (1.39°) projection data



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Jaszczak Cold Sphere Phantom: Noisy Collimated Projection Data



Reconstruction of noisy GE-LEGP data



Reconstruction of noisy SE-LEHR data



Comparison of TITAN-IR with Other Methods Based on Jaszczak Phantom:

- Filtered backprojection (FBP)
 - Traditional standard for image reconstruction
 - Implemented in MATLAB and includes the Chang attenuation correction*
- ML-EM with System Matrix (SM) only
 - Standard ML-EM reconstruction method
 - Algorithm written in Fortran 90
 - Uses the same system matrix that TITAN-IR uses for backprojection

*L.-T.Chang, "A method for attenuation correction in radionuclide computed tomography," IEEE Trans. Nucl. Sci., 1978



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Comparison of Methods with Jaszczak Phantom

Algorithm	Noiseless, no collimator blur	Noisy, no collimator blur	Noisy GE-LEGP	Noisy SE-LEHR
FBP	••••			
ML-EM with SM only	••••			
TITAN-IR				56

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Comparison of Methods with Jaszczak Phantom

Contrast in reconstruction of noisy SE-LEHR projection data





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Computation Time

Jaszczak phantom: Noisy GE-LEGP projection data

Computing environment [Up to 16 cores, Intel Xeon E5 2.6 GHz processors, 16 GB per core]

Processor Cores	Wall Clock Time (s)	Speedup
1	1665.7	-
2	905.1	1.8
4	524.3	3.2
8	341.4	4.9
16	172.0	9.7



Spent fuel Pool & Cask Modeling

- Standard approach Full Monte Carlo calculations face difficulties in this area
 - Convergence is difficult due to undesampling (due to absorbers)
 - Convergence can also be difficult to detect
 - Computation times are very long, especially to get detailed information
 - Changing pool configuration requires complete recalculation



The RAPID (Real-time Analysis for Particle transport and In-situ Detection) code system





RAPIDTM Code System



- RAPID is capable of calculating the system eigenvalue k_{eff}, pin-wise axially-dependent 3D fission density distribution, and detector response.
- RAPID is comprised of **six stages**:

Pre-calculation

Stage 1 – Calculation of material concentration Stage 2 - Calculation of fission matrix (FM) coefficients Stage 3 - Calculation of field-of-view (FOV) Stage 4 – Calculation of importance function

Calculation

Stage 5 - Processing of FM coefficients & Solution of a linear system of equations (i.e., FM formulation)

Stage 6 – Calculation of Detector response



Determination of fission Matrix (FM) Coefficients

• Eigenvalue formulation

$$F_i = \frac{1}{k} \sum_{j=1}^N a_{i,j} F_j$$



- *k* is eigenvalue
- *F_j* is fission source, *S_j* is fixed source in cell j
- $a_{i,j}$ is the number of fission neutrons produced in cell *i* due to a fission neutron born in cell *j*.
- Subcritical multiplication formulation

$$F_{i} = \sum_{j=1}^{N} (a_{i,j}F_{j} + b_{i,j}S_{j}),$$

• $b_{i,j}$ is the number of fission neutrons produced in cell *i* due to a source neutron born in cell *j*.



FM Coefficients Determination : a Multi-layer approach

• Brute force approach:

• For a typical <u>spent nuclear fuel pool</u> with a sub-region of 9x9 assemblies:

 $N = 9 \times 9 \times 264 = 21,384$ total fuel pins

- Considering 24 axial segments per rod, then N = 513,216
- Standard FM would require N = 513,216 separate fixedsource calculations to determine the coefficient matrix
 - A matrix of size N x N = 2.63391E+11 total coefficients (> 2 TB of memory is needed)
- The straightforward approach is clearly NOT feasible
- Multi-layer, regional approach ((in the process of filing for a patent))
 - Determine coefficients as a function of different parameters (Stage 1)
 - Process coefficients for problem of interest (**Stage 3**)

9x9 array of assemblies in a pool





Rapid : Code System Structure





I²S-LWR – Reference Model

I²S-LWR FUEL ASSEMBLY

- 19x19 fuel lattice
 - 335 fuel rods, 24 control/guide tubes, 1 instrumentation tube
- U_3Si_2 fuel enriched to 4.95 wt-% ²³⁵U

SPENT FUEL POOL

- Based on AP1000 SFP
- Consider a 9x9 segment of SFP (81 assemblies)
- Storage cell walls made of Metamic[®] (B4C-Al) between SS plates





Pre-Calculation – p³RAPID Stage 1: Burnup Calculation with SCALE/TRITON

- **Need :** Material composition & Intrinsic source
- Use: SCALE 6.1 TRITON
 - The TDEPL option used to invoke NEWT 2D & ORIGEN
- For:
 - enrichment of 4.95 wt-%; burnups: 37, 59 GWd/MTHM; and, Cooling Times: 14 days, 1 & 9 years
 - Quarter assembly model used.
 - 49 different fuel materials (considering octal symmetry)

	20 38 42
	29 37 45
	24 28 36 45
	22 27 34 43
8 11 15	21) 26 33 42



Pre-Calculation – P³RAPID Stage 2: Coefficient calculations

• Using information from Stage 1,



• A database of FM coefficient is prepared



Comparison of RAPID with MCNP reference calculation





Test cases

- Performed eigenvalue calculations for a 2x2 segment of the reference SFP.
 - 4 test cases are defined, each containing different combinations of burnups/cooling times
 - Fuel region of the model partitioned into 32,256 fission regions (tallies)
- Reference MCNP eigenvalue parameters are:
 - 10⁶ particles per cycle,
 - 400 skipped cycles
 - 400 active cycles



Description of Test cases – Pool segments



*'0 year' cooling time refers to ~14 days



Comparison of Eigenvalues

	MCNP	RAPID	Rel. Diff.
Case	k _{eff}	k _{eff}	RAPID vs. MCNP (pcm)
1	0. 79998 (± 4 pcm)	0.80020	28
2	0.79511(± 4 pcm)	0.79532	26
3	0.60444(± 3 pcm)	0.60425	-31
4	0.58330(± 3 pcm)	0.58322	-14



Comparison Radial fission densities (FD)





MCNP Predictions (CASE 1)



1-σ Relative Uncertainty





RAPID VERSUS MCNP(CASE 1)





24.00

20.00

16.00

12.00

8.00

4.00

0.00

-4.00

-8.00

MCNP Predictions (CASE 2)



1-σ Relative Uncertainty





RAPID VERSUS MCNP (CASE 2)








MCNP Predictions (CASE 3)



$1 - \sigma$ Relative Uncertainty





RAPID VERSUS MCNP (CASE 3)





% Relative Difference





MCNP Predictions (CASE 4)



1-σ Relative Uncertainty





RAPID VERSUS MCNP (CASE 4)





% Relative Difference





Computation Time

	MCNP		RAPID		
Case	Cores	Time [min)	Cores	Time [min]	Speedup
1	16	1020 (17 hrs)	1	0.50	2044
2	16	1013 (17 hrs)	1	0.51	1980
3	16	1082 (18 hrs)	1	0.50	2163
4	16	1149 (19 hrs)	1	0.50	2284



Comparison of RAPID to MCNP reference models

- Single assembly & full cask models -



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RAPID vs. MCNP – Single assembly model



- RAPID calculated and MCNP system eigenvalue (k_{eff}) and pin-wise, axiallydependent fission density distribution, i.e, 6,336 tallies, are compared.
- Significant speedup is obtained using RAPID on just a single computer core.

Case	MCNP	RAPID
k _{eff}	1.18030 (± 2 pcm)	1.18092
k_{eff} relative difference	-	53 pcm
Fiss. density adjusted rel. uncertainty	0.48%	-
Fission density relative diff.	-	0.65%
Computer	16 cores	1 core
Time	666 min (11.1 hours)	0.1 min (6 seconds)
Speedup	-	6,666



RAPID vs. MCNP – Full cask model



- RAPID calculated and MCNP system eigenvalue (k_{eff}) and pin-wise, axiallydependent fission density distribution, i..e, **202,752** tallies, are compared.
- The speedup increases with the dimension of the model.

Case	MCNP	RAPID
k _{eff}	1.14545 (± 1 pcm)	1.14590
Relative Difference	-	39 pcm
Fission density rel. uncertainty	1.15%	-
Fission density relative diff.	-	1.56%
Computer	16 cores	1 core
Time	13,767 min (9.5 days)	0.585 min (35 seconds)
Speedup	-	23,533



GBC-32 3D fission density distribution





Determination of neutron Dose



• Given the *neutron dose-to-flux ratio* $(f_n) \left(\frac{\frac{mrem}{hr}}{\frac{\#}{cm^2-s}}\right)$, then

$$Dose = \langle \psi f_n \rangle$$

 Then, Dose is calculated using the adjoint-function methodology by

 $Dose = \langle \psi^* S \rangle$

Where,

$$H^*\psi^*=f_n$$



TITAN Calculation Model

- Multigroup cross-section (energy group structure, Pn order)
- Calculation model (segment of the cask), Field-Of-View (FOV)?
- Spatial meshing, angular quadrature order, finite-differencing formulation, convergence



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TITAN calculation

- 1. 2-assembly model
- 2. size = 91.24x71.27x40 cm³
- 3. 2.5cm x 2.5cm x 5cm voxel air detector
- 4. # meshes = 386,286,
- S10 angular quadrature set
 P3
- 5. 19-group [BUGLE-96 library; groups 3-21]
- 6. 8 cores
- 7. 70 min





Dose calculation

• The dose formulation (i.e., detector response) is expressed by:

$$D = \sum_{g=3}^{21} \sum_{i=1}^{N_{cell}} \psi_{i,g}^* (\chi_g S_i)$$

- χ_g is the Watt spectrum for energy group g, $\psi_{i,g}^*$ is the importance function of cell *i* for group *g*, and *S_i* is the RAPID calculated neutron source in cell *i*.
- The calculated dose is:

Dose per unit source =
$$7.79 \cdot 10^{-12} \left[\frac{rem}{hr} \right]$$



Detector field-of-view



• More than 90% of the dose evaluated at the canister's surface is due to the outermost row of assemblies, near the boundary.



Experimental Benchmarking of RAPID

Phase 1





Benchmark facility - US Naval academy Subcritical (USNA-SC)

- A cylindrical pool with natural uranium (fuel) and light water (moderator)
- There are a total of 268 fuel rods, arranged in a hexagonal lattice
- Fuel: hollow aluminum tubes containing 5 annular fuel slugs



• Neutron source: PuBe



Whole Core - Total Neutron Flux





Whole Core : Neutron flux Distribution



NSEL Noder Sterce and Engineering Laborato Verini Tech Research Cartes, Atlinguo,

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Whole Core : Relative Uncertainties of Neutron Flux





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Experiments

- Count rate in a ³He proportional counter was measured by placing the counter within the annulus of each fuel pin
- Neutron counts are determined in fuel pins along three radial profiles (11, 12, & 13) shown in the figure.





Comparison of reaction rates (Counts) of ³He detector [experiment vs calculation]

Estimated Detector Efficiency based on least-squares minimization

$$Eff = \frac{\sum_i c_i m_i}{\sum_i c_i^2}$$

Where, m_i = Measured response at position *i* c_i = Calculated response at position *i*



$f = \frac{c}{m}$ (Ratio of calculated to measured responses)





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Virtual Reality - Phase 1

- Two of my graduate students (Nate Roskoff and Val Maslcolino) are working with me and Drs. Polys and Rajamohan and a student from School of Arts and Design in this project
- Tasks
 - Development of a **connectivity environment** between the visualization systems in *Blacksburg and Arlington*. This will make possible seamless interaction in a virtual environment between collaborators that are geographically separated.
 - Development of a VRS for a spent nuclear fuel pool. This virtual model tool includes our RAPID code system for monitoring the pool in real time.
- We have developed software using Paraview and x3dom packages (examples are available at <u>http://nsel.ncr.vt.edu/vrs.html</u>)







Virtual Spent Fuel Pool, Virtual pool-assembly & real-time RAPID • calculation





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Conclusions

- MRT methodology allows for development of real-time tools for analysis of nuclear systems
- Thus far, we have developed
 - INSPCT-S, AIMS, TITAN-IR, & RAPID
- We have demonstrated that indeed we obtain accurate, detailed solution in real time!
- This is especially true for RAPID code system that has been applied to the simulation of spent fuel pools and casks
 - Pin-wise, axially dependent fission density is determined in 35 seonds
- Further, it is demonstrated that after about 18 hours of calculation MCNP has not fully convergence near the absorber racks, i.e., difficulties with undersampling.
 - Standard eigenvalue Monte Carlo has difficulties with HDR, undersampling, and correlation
 - The FM approach used in RAPID is a solution to above difficulties
 - The RAPID MRT algorithm is able to **overcome the main issues** related to Monte Carlo eigenvalue calculations such as source convergence and cycle-to-cycle correlation



Ongoing & Future Studies

- Continued sensitivity analysis of RAPID for different burnups
- Complete experimental benchmarking of RAPID using the U.S. Naval Academy's subcritical facility
 - Preliminary experimental benchmarking results were presented at the recent INMM meeting, July 2016.
- Initiated determination of statistical uncertainties associated RAPID calculated eigenvalue and fission density
 - i.e., Propagation of the uncertainties of the FM coefficients
- External dose/detector response calculation has been implemented in RAPID using the TITAN-calculated importance function methodology*
- Extend RAPID for material identification
- An automated methodology for the determination of the FOV of a detector is under development.
- TITAN dose calculation will be benchmarked against a reference A³MCNP (Automated Adjoint Accelerated MCNP) code prediction.
- Developing virtual reality system for a spent fuel pool



Thanks!

Questions?







A new book

Monte Carlo Methods for Particle Transport



Alireza Haghighat





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