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Adaptive-in-temperature method for fast on-the-fly sampling of thermal neutron scattering data in MCNP6

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**Nuclear Engineering Program
Rensselaer Polytechnic Institute (RPI)**

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Team Members and Collaborators

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- Yaron Danon (Co-PI, Professor)
- Camden Blake (PhD student)

➤ Collaborators

- Forrest Brown (retired) – Los Alamos National Laboratory
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Acknowledgement

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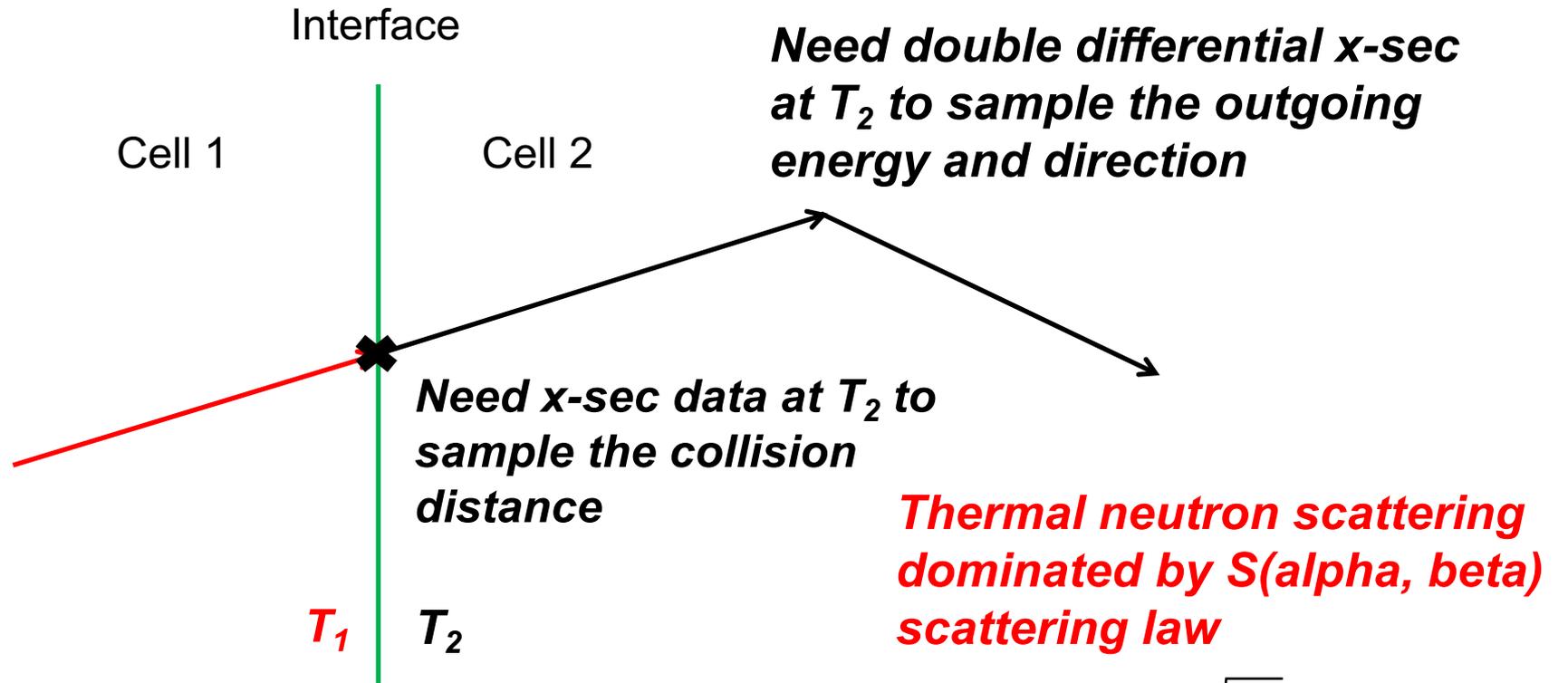
Project Objective and Motivation

- Develop thermal data libraries for selected isotopes in MCNP6 to support on-the-fly S(alpha, beta) sampling for temperature ranges applicable to nuclear criticality safety



Enhance the physics treatment in MCNP6 so that it can perform fast on-the-fly sampling of S(alpha, beta) data at arbitrary temperature

A General Random Walk Sampling in MCNP



$$\sigma(E \rightarrow E', \mu, T) = \frac{\sigma_b}{2kT} \sqrt{\frac{E'}{E}} e^{-\beta/2} S(\alpha, \beta, T)$$

α : momentum transfer β : energy transfer

Current Data Storage Format for MCNP6

- NJOY is used to create correlated energy-angle distributions for inelastic scattering.
- These consist of two parts.
 - A set of equally probable final energy bins for each incident energy.
 - A set of equally probable cosine bins for each incoming and outgoing energy pair.
- This is done for each temperature.

Current Data Storage Format for MCNP6

T_1	CDF_1	...	CDF_n	...	CDF_N	T_1	E'_1	...	E'_n	...	E'_N
E_1	$E'_{1,1}$		$E'_{1,n}$		$E'_{1,N}$	E_1	$\mu'_{CDF,1,1}$		$\mu'_{CDF,1,n}$		$\mu'_{CDF,1,N}$
...						...					
E_m	$E'_{m,1}$		$E'_{m,n}$		$E'_{m,N}$	E_i	$\mu'_{CDF,m,1}$		$\mu'_{CDF,m,n}$		$\mu'_{CDF,m,N}$
...						...					
E_M	$E'_{M,1}$		$E'_{M,n}$		$E'_{M,N}$	E_M	$\mu'_{CDF,M,1}$		$\mu'_{CDF,M,n}$		$\mu'_{CDF,M,N}$

T_1	CDF_1	...	CDF_k	...	CDF_K
(E_1, E'_1)	$\mu'_{CDF,1,1}$	$\mu'_{1,1,1}$	$\mu'_{1,1,k}$		$\mu'_{1,1,K}$
...	...				
(E_m, E'_n)	$\mu'_{CDF,m,n}$	$\mu'_{m,n,1}$	$\mu'_{m,n,k}$		$\mu'_{m,n,K}$
...	...				
(E_M, E'_N)	$\mu'_{CDF,M,N}$	$\mu'_{M,N,1}$	$\mu'_{M,N,k}$		$\mu'_{M,N,K}$

Current Sampling Procedure in MCNP6

1. Determination of target's temperature (T).
2. Find data sets at nearest temperatures ($T_i < T < T_{i+1}$).
3. Sample outgoing energy based on T_i and T_{i+1} data sets, and then do interpolation to obtain outgoing energy at T.
4. Similar procedure to obtain outgoing angle at T.

T_i	CDF ₁	...	CDF _n	CDF _{n+1}	...	CDF _N
E_1	$E'_{1,1}$		$E'_{1,n}$	$E'_{1,n+1}$		$E'_{1,N}$
:						
E_m	$E'_{m,1}$		$E'_{m,n}$	$E'_{m,n+1}$		$E'_{m,N}$
E_{m+1}	$E'_{m+1,1}$		$E'_{m+1,n}$	$E'_{m+1,n+1}$		$E'_{m+1,N}$
:						
E_M	$E'_{M,1}$		$E'_{M,n}$	$E'_{M,n+1}$		$E'_{M,N}$

T_{i+1}	CDF ₁	...	CDF _n	CDF _{n+1}	...	CDF _N
E_1	$E'_{1,1}$		$E'_{1,n}$	$E'_{1,n+1}$		$E'_{1,N}$
:						
E_m	$E'_{m,1}$		$E'_{m,n}$	$E'_{m,n+1}$		$E'_{m,N}$
E_{m+1}	$E'_{m+1,1}$		$E'_{m+1,n}$	$E'_{m+1,n+1}$		$E'_{m+1,N}$
:						
E_M	$E'_{M,1}$		$E'_{M,n}$	$E'_{M,n+1}$		$E'_{M,N}$

Data Storage Requirements by MCNP6

- To be able to sample thermal scattering, MCNP stores all the thermal data directly.
- This leads to a large amount of data that is needed.
 - For example, in the thermal energy range: 24 MB per temperature for graphite or H in water
 - For a realistic reactor, 50-100 temperatures are needed

Material	File Size (MB)
Al	25
Be (Metal)	28
H in Benzene	69.5
Be in BeO	69.5
Fe	28.9
Graphite	24.2
D in D ₂ O	30
H in ZrH	116.3
H in H ₂ O	24.9
O ₂ in UO ₂	75.1
O in BeO	57.5
H in Polyethylene	19.6
Si in SiO ₂	41.6
U in UO ₂	50.5
Zr in ZrH	56.1

How to Decrease Data Storage for MCNP6?

- Any better way to improve the storage requirement (and associated sampling procedures)?

$$\beta = \frac{E' - E}{kT}, \quad \alpha = \frac{E + E' - 2\mu\sqrt{EE'}}{AkT}$$

β : energy transfer

α : momentum transfer

$$E \rightarrow E', \text{ then } (E, E') \rightarrow \mu$$

$$E \rightarrow \beta, \text{ then } \beta \rightarrow \alpha$$

Data Storage & Sampling Methods in MC21

- Sample momentum (α) and energy (β) transfer directly.
 - α and β can then be converted into E' and μ

$$E' = E + \beta kT \qquad \mu = \frac{2E + kT(\beta - A\alpha)}{2\sqrt{E(E + \beta kT)}}$$

- MC21 implements this strategy.

Data Storage & Sampling Methods in MC21

- Coupled α and β distributions are created.
 - A table of incident energies versus CDF values that contain β values.
 - A table of α versus β mesh values that contain function values, which serve as a memory reduction technique where the function values are used to create α CDFs on-the-fly.
- This is done for each temperature.

Data Storage in MC21

T_1	CDF_1	...	CDF_j	...	CDF_N
E_1	$\beta_{1,1}$		$\beta_{1,n}$		$\beta_{1,N}$
...					
E_m	$\beta_{m,1}$		$\beta_{m,n}$		$\beta_{m,N}$
...					
E_M	$\beta_{M,1}$		$\beta_{M,n}$		$\beta_{M,N}$

T_1	β_1	...	β_N
α_1	$F(\alpha_1, \beta_1)$		$F(\alpha_1, \beta_N)$
...			
α_m	$F(\alpha_m, \beta_1)$		$F(\alpha_m, \beta_N)$
...			
α_M	$F(\alpha_M, \beta_1)$		$F(\alpha_M, \beta_N)$

$$F(\alpha_m, \beta_n) = \int_0^{\alpha_m} S(\alpha', \beta_n) e^{\beta_n/2} d\alpha'$$

$$\alpha_{max} = \frac{\sqrt{E} + \sqrt{E - \beta kT}}{AkT}$$

$$\alpha_{CDF}(\alpha | \beta_n, E) = \frac{F(\alpha, \beta_n) - F(\alpha_{min}, \beta_n)}{F(\alpha_{max}, \beta_n) - F(\alpha_{min}, \beta_n)}$$

$$\alpha_{min} = \frac{\sqrt{E} - \sqrt{E - \beta kT}}{AkT}$$

C. T. Ballinger, "The Direct S(alpha, beta) Method for Thermal Neutron Scattering," Knolls Atomic Power Laboratory, Schenectady, NY, 1994.
 T. M. Sutton, et al., "The MC21 Monte Carlo Transport Code (LM-06K144)," (M&C + SNA 2007), LaGrange Park, IL, 2007.

Sampling Procedure in MC21

1. Determination of target's temperature (T).
2. Find data sets at nearest temperatures ($T_i < T < T_{i+1}$).
3. Sample β based on T_i and T_{i+1} data sets, and then do interpolation to obtain β at T.
4. Similar procedure to obtain α at T.
5. Convert α and β to outgoing energy and scattering angle at T.

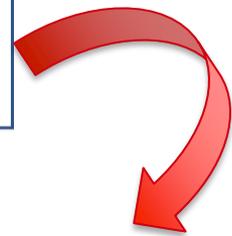
T_i	CDF_1	...	CDF_n	CDF_{n+1}	...	CDF_N
E_1	$\beta_{1,1}$		$\beta_{1,n}$	$\beta_{1,n+1}$		$\beta_{1,N}$
:						
E_m	$\beta_{m,1}$		$\beta_{m,n}$	$\beta_{m,n+1}$		$\beta_{m,N}$
E_{m+1}	$\beta_{m+1,1}$		$\beta_{m+1,n}$	$\beta_{m+1,n+1}$		$\beta_{m+1,N}$
:						
E_M	$\beta_{M,1}$		$\beta_{M,n}$	$\beta_{M,n+1}$		$\beta_{M,N}$

T_{i+1}	CDF_1	...	CDF_n	CDF_{n+1}	...	CDF_N
E_1	$\beta_{1,1}$		$\beta_{1,n}$	$\beta_{1,n+1}$		$\beta_{1,N}$
:						
E_m	$\beta_{m,1}$		$\beta_{m,n}$	$\beta_{m,n+1}$		$\beta_{m,N}$
E_{m+1}	$\beta_{m+1,1}$		$\beta_{m+1,n}$	$\beta_{m+1,n+1}$		$\beta_{m+1,N}$
:						
E_M	$\beta_{M,1}$		$\beta_{M,n}$	$\beta_{M,n+1}$		$\beta_{M,N}$

A Little Recap

T_1	CDF_1	...	CDF_n	...	CDF_N	T_1	E'_1	...	E'_n	...	E'_N
E_1	$E'_{1,1}$		$E'_{1,n}$		$E'_{1,N}$	E_1	$\mu'_{CDF,1,1}$		$\mu'_{CDF,1,n}$		$\mu'_{CDF,1,N}$
...						...					
E_m	$E'_{m,1}$		$E'_{m,n}$		$E'_{m,N}$	E_i	$\mu'_{CDF,m,1}$		$\mu'_{CDF,m,n}$		$\mu'_{CDF,m,N}$
...						...					
E_M	$E'_{M,1}$		$E'_{M,n}$		$E'_{M,N}$	E_M	$\mu'_{CDF,M,1}$		$\mu'_{CDF,M,n}$		$\mu'_{CDF,M,N}$

T_1	CDF_1	...	CDF_k	...	CDF_K
(E_1, E'_1)	$\mu'_{CDF,1,1}$	$\mu'_{1,1,1}$		$\mu'_{1,1,k}$	$\mu'_{1,1,K}$
...	...				
(E_m, E'_m)	$\mu'_{CDF,m,n}$	$\mu'_{m,n,1}$		$\mu'_{m,n,k}$	$\mu'_{m,n,K}$
...	...				
(E_M, E'_M)	$\mu'_{CDF,M,N}$	$\mu'_{M,N,1}$		$\mu'_{M,N,k}$	$\mu'_{M,N,K}$



T_1	CDF_1	...	CDF_j	...	CDF_N	T_1	β_1	...	β_N
E_1	$\beta_{1,1}$		$\beta_{1,n}$		$\beta_{1,N}$	α_1	$F(\alpha_1, \beta_1)$		$F(\alpha_1, \beta_N)$
...						...			
E_m	$\beta_{m,1}$		$\beta_{m,n}$		$\beta_{m,N}$	α_m	$F(\alpha_m, \beta_1)$		$F(\alpha_m, \beta_N)$
...						...			
E_M	$\beta_{M,1}$		$\beta_{M,n}$		$\beta_{M,N}$	α_M	$F(\alpha_M, \beta_1)$		$F(\alpha_M, \beta_N)$

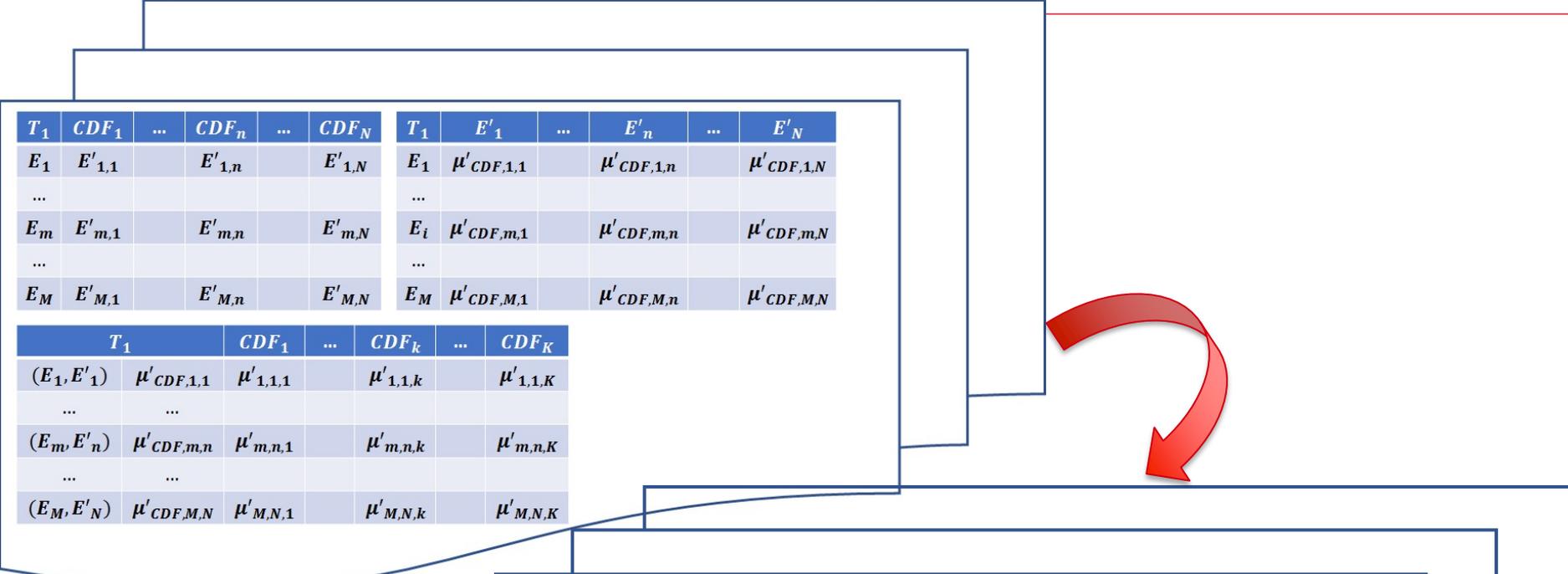
$$F(\alpha_m, \beta_n) = \int_0^{\alpha_m} S(\alpha', \beta_n) e^{\beta_n/2} d\alpha'$$

Current Data Storage Techniques

- Key Points:
 - Both MCNP6 and MC21 interpolate scattering data in order to account for temperature effects.
 - Therein, a large amount of data is needed to ensure accuracy.
- Any further improvement to decrease data storage?
 - We (RPI) have (or had*) a solution!

- * [1] A. T. Pavlou and W. Ji, "On-the-Fly Sampling of Temperature-Dependent Thermal Neutron Scattering Data for Monte Carlo Simulations," *Annals of Nuclear Energy*, 71, 411-426, 2014.
- * [2] A. T. Pavlou, "An Adaptive-In-Temperature Method for On-The-Fly Sampling of Thermal Neutron Scattering Data in Continuous-Energy Monte Carlo Codes," Rensselaer Polytechnic Institute, Troy, NY, 2015.

From (E, E', μ') Space to (α, β) Space



T_1	CDF_1	...	CDF_n	...	CDF_N	T_1	E'_1	...	E'_n	...	E'_N
E_1	$E'_{1,1}$		$E'_{1,n}$		$E'_{1,N}$	E_1	$\mu'_{CDF,1,1}$		$\mu'_{CDF,1,n}$		$\mu'_{CDF,1,N}$
...						...					
E_m	$E'_{m,1}$		$E'_{m,n}$		$E'_{m,N}$	E_i	$\mu'_{CDF,m,1}$		$\mu'_{CDF,m,n}$		$\mu'_{CDF,m,N}$
...						...					
E_M	$E'_{M,1}$		$E'_{M,n}$		$E'_{M,N}$	E_M	$\mu'_{CDF,M,1}$		$\mu'_{CDF,M,n}$		$\mu'_{CDF,M,N}$

T_1	CDF_1	...	CDF_k	...	CDF_K
(E_1, E'_1)	$\mu'_{CDF,1,1}$	$\mu'_{1,1,1}$		$\mu'_{1,1,k}$	$\mu'_{1,1,K}$
...	...				
(E_m, E'_m)	$\mu'_{CDF,m,n}$	$\mu'_{m,n,1}$		$\mu'_{m,n,k}$	$\mu'_{m,n,K}$
...	...				
(E_M, E'_M)	$\mu'_{CDF,M,N}$	$\mu'_{M,N,1}$		$\mu'_{M,N,k}$	$\mu'_{M,N,K}$

T_1	CDF_1	...	CDF_j	...	CDF_N	T_1	β_1	...	β_N
E_1	$\beta_{1,1}$		$\beta_{1,n}$		$\beta_{1,N}$	α_1	$F(\alpha_1, \beta_1)$		$F(\alpha_1, \beta_N)$
...						...			
E_m	$\beta_{m,1}$		$\beta_{m,n}$		$\beta_{m,N}$	α_m	$F(\alpha_m, \beta_1)$		$F(\alpha_m, \beta_N)$
...						...			
E_M	$\beta_{M,1}$		$\beta_{M,n}$		$\beta_{M,N}$	α_M	$F(\alpha_M, \beta_1)$		$F(\alpha_M, \beta_N)$

$$F(\alpha_m, \beta_n) = \int_0^{\alpha_m} S(\alpha', \beta_n) e^{\beta_n/2} d\alpha'$$

A Better Strategy Than MCNP6 & MC21

T_1	CDF_1	...	CDF_n	...	CDF_N	T_1	E'_1	...	E'_n	...	E'_N
E_1	$E'_{1,1}$		$E'_{1,n}$		$E'_{1,N}$	E_1	$\mu'_{CDF,1,1}$		$\mu'_{CDF,1,n}$		$\mu'_{CDF,1,N}$
...
E_m	$E'_{m,1}$		$E'_{m,n}$		$E'_{m,N}$	E_m	$\mu'_{CDF,m,1}$		$\mu'_{CDF,m,n}$		$\mu'_{CDF,m,N}$
...
E_M	$E'_{M,1}$		$E'_{M,n}$		$E'_{M,N}$	E_M	$\mu'_{CDF,M,1}$		$\mu'_{CDF,M,n}$		$\mu'_{CDF,M,N}$

T_1	CDF_1	...	CDF_k	...	CDF_K
(E_1, E'_1)	$\mu'_{CDF,1,1}$	$\mu'_{1,1,1}$		$\mu'_{1,1,k}$	$\mu'_{1,1,K}$
...	...				
(E_m, E'_m)	$\mu'_{CDF,m,1}$	$\mu'_{m,n,1}$		$\mu'_{m,n,k}$	$\mu'_{m,n,K}$
...	...				
(E_M, E'_M)	$\mu'_{CDF,M,1}$	$\mu'_{M,n,1}$		$\mu'_{M,n,k}$	$\mu'_{M,n,K}$



T_1	CDF_1	...	CDF_j	...	CDF_N	T_1	β_1	...	β_N
E_1	$\beta_{1,1}$		$\beta_{1,n}$		$\beta_{1,N}$	α_1	$F(\alpha_1, \beta_1)$		$F(\alpha_1, \beta_N)$
...			
E_m	$\beta_{m,1}$		$\beta_{m,n}$		$\beta_{m,N}$	α_m	$F(\alpha_m, \beta_1)$		$F(\alpha_m, \beta_N)$
...			
E_M	$\beta_{M,1}$		$\beta_{M,n}$		$\beta_{M,N}$	α_M	$F(\alpha_M, \beta_1)$		$F(\alpha_M, \beta_N)$

$$F(\alpha_m, \beta_n) = \int_0^{\alpha_m} S(\alpha', \beta_n) e^{\beta_n/2} d\alpha'$$

	CDF_1	...	CDF_n	...	CDF_N
E_1	$\beta(T)_{1,1}$		$\beta(T)_{1,n}$		$\beta(T)_{1,N}$
:					
E_m	$\beta(T)_{m,1}$		$\beta(T)_{m,n}$		$\beta(T)_{m,N}$
:					
E_M	$\beta(T)_{M,1}$		$\beta(T)_{M,n}$		$\beta(T)_{M,N}$

	CDF_1	...	CDF_N
β_1	$\alpha(T)_{1,1}$		$\alpha(T)_{1,N}$
:			
β_m	$\alpha(T)_{m,1}$		$\alpha(T)_{m,N}$
:			
β_M	$\alpha(T)_{M,1}$		$\alpha(T)_{M,N}$

A Better Strategy Than MCNP6 & MC21

- These $\beta(T)_{m,n}$ and $\alpha(T)_{m,n}$ enable “**on-the-fly**” treatment of temperature-dependence (vs. interpolation between temperatures).

	CDF_1	...	CDF_n	CDF_{n+1}	...	CDF_N
E_1	$\beta(T)_{1,1}$		$\beta(T)_{1,n}$	$\beta(T)_{1,n+1}$		$\beta(T)_{1,N}$
:						
E_m	$\beta(T)_{m,1}$		$\beta(T)_{m,n}$	$\beta(T)_{m,n+1}$		$\beta(T)_{m,N}$
E_{m+1}	$\beta(T)_{m+1,1}$		$\beta(T)_{m+1,n}$	$\beta(T)_{m+1,n+1}$		$\beta(T)_{m+1,N}$
:						
E_M	$\beta(T)_{M,1}$		$\beta(T)_{M,n}$	$\beta(T)_{M,n+1}$		$\beta(T)_{M,N}$

A Better Strategy Than MCNP6 & MC21

	CDF_1	...	CDF_n	...	CDF_N		CDF_1	...	CDF_N
E_1	$\beta(T)_{1,1}$		$\beta(T)_{1,n}$		$\beta(T)_{1,N}$	β_1	$\alpha(T)_{1,1}$		$\alpha(T)_{1,N}$
:						:			
E_m	$\beta(T)_{m,1}$		$\beta(T)_{m,n}$		$\beta(T)_{m,N}$	β_m	$\alpha(T)_{m,1}$		$\alpha(T)_{m,N}$
:						:			
E_M	$\beta(T)_{M,1}$		$\beta(T)_{M,n}$		$\beta(T)_{M,N}$	β_M	$\alpha(T)_{M,1}$		$\alpha(T)_{M,N}$

- How do we define these $\beta(T)_{m,n}$ and $\alpha(T)_{m,n}$?
- This is where new research direction can jump in.
 - Data Analytics Methods?
 - Machine Learning Methods?

On-The-Fly Strategy Data Storage Format

	CDF_1	...	CDF_N
E_1	$[b_{1,1,1} \dots b_{1,1,K}]$		$[b_{1,N,1} \dots b_{1,N,K}]$
...			
E_m	$[b_{m,1,1} \dots b_{m,1,K}]$		$[b_{m,N,1} \dots b_{m,N,K}]$
...			
E_M	$[b_{M,1,1} \dots b_{M,1,K}]$		$[b_{M,N,1} \dots b_{M,N,K}]$

	CDF_1	...	CDF_N
β_1	$[a_{1,1,1} \dots a_{1,1,L}]$		$[a_{1,N,1} \dots a_{1,N,L}]$
...			
β_m	$[a_{m,1,1} \dots a_{m,1,L}]$		$[a_{m,N,1} \dots a_{m,N,L}]$
...			
β_M	$[a_{M,1,1} \dots a_{M,1,L}]$		$[a_{M,N,1} \dots a_{M,N,L}]$

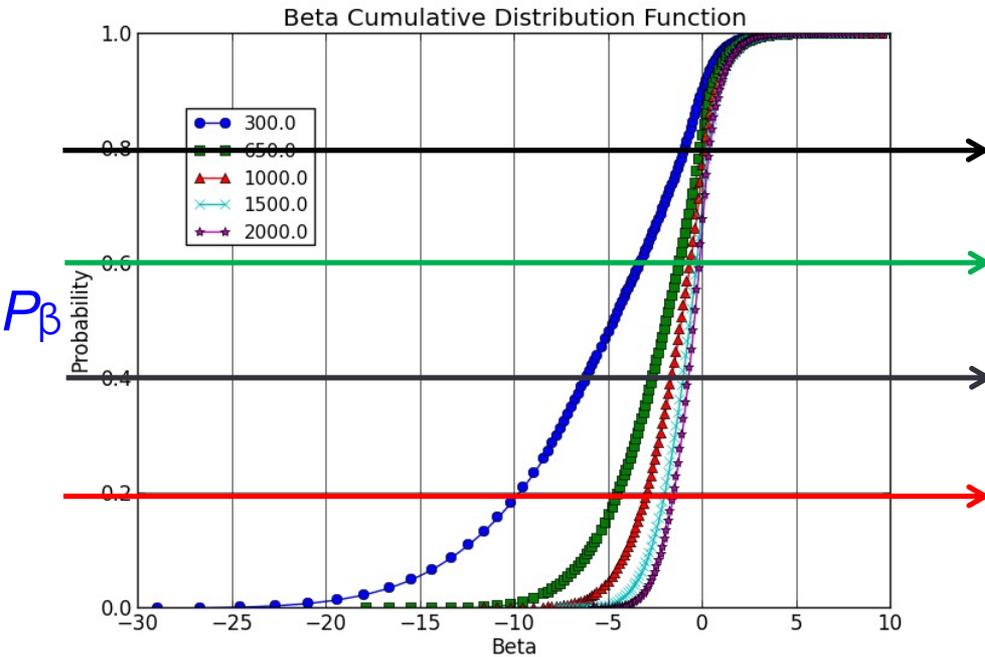
$$\beta(T)_{m,n} = \sum_{k=1}^K b_{m,n,k} X^k(T) + const.$$

$$\alpha(T)_{m,n} = \sum_{l=1}^L a_{m,n,l} X^l(T) + const.$$

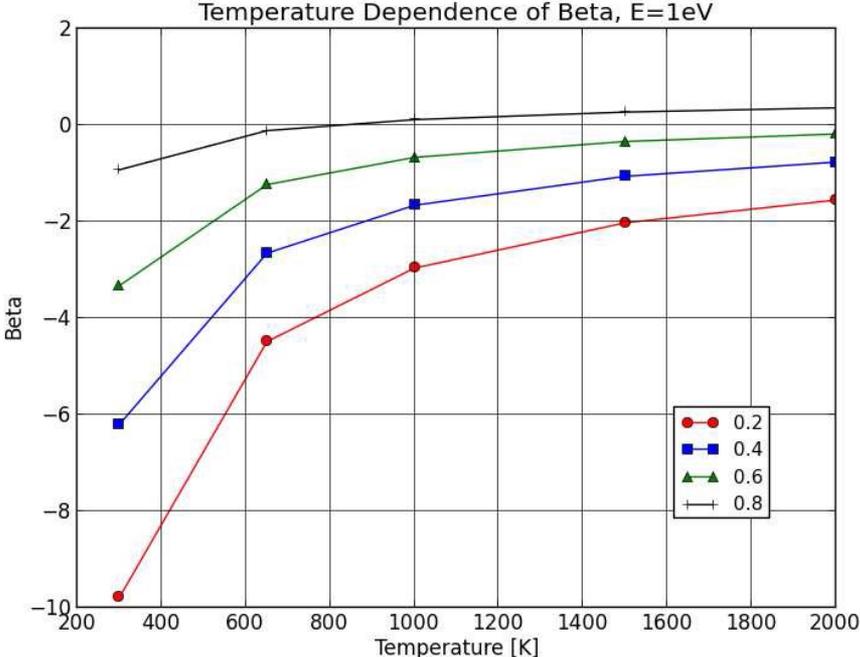
Where K and L are the orders of the fits.

One typical example: $\beta(T) \approx \sum_{n=0}^N b_n T^{-n}$, $\alpha(T) \approx \sum_{m=0}^M a_m T^{-m}$

On-The-Fly Strategy Data Generation



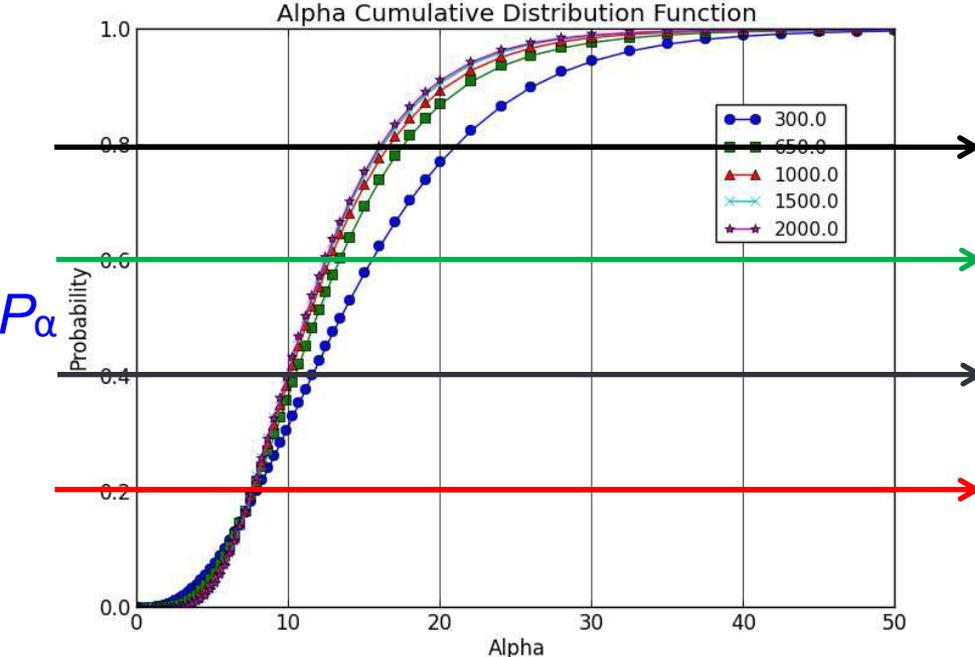
§ Beta PDFs for graphite at 1.0 eV and various temperatures



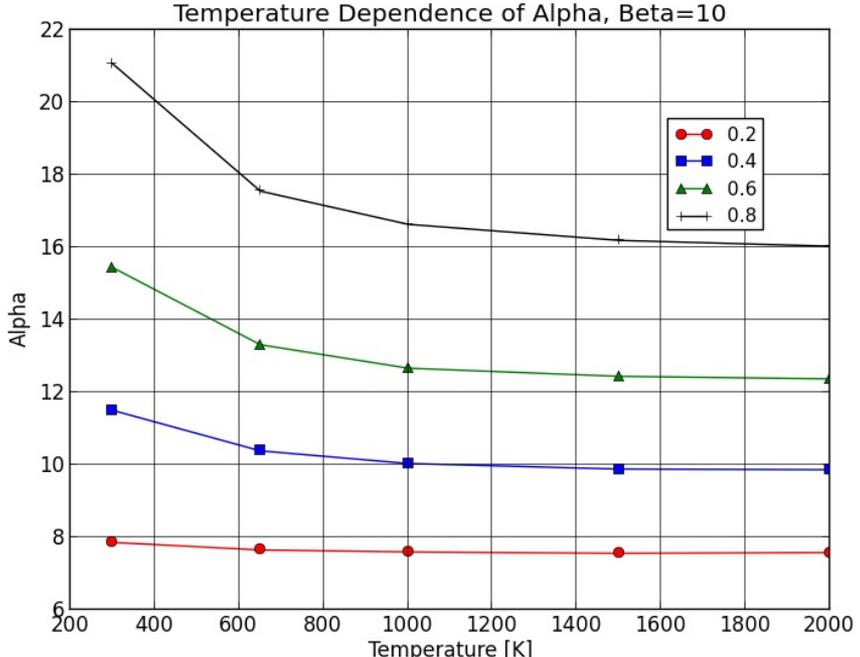
§ Beta CDFs for graphite at 1.0 eV and various temperatures

β temperature fit meshes: E , P_β

On-The-Fly Strategy Data Generation



§ Alpha PDFs for graphite at $\beta = 10$ and various temperatures



§ Alpha CDFs for graphite at $\beta = 10$ and various temperatures

α temperature fit meshes: β , P_α

Current and Future ...

- Contract with LANL was finalized (delayed by a year) ...
- On-the-fly strategy driven libraries are being developed based on ENDF/B-VIII.0.
- Will report more details on the next review meeting.
- Any questions?



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