



Dataset of neutron energy region values for the in-reactor production of medical isotopes

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Abstract

Medical isotopes are the foundation material for nuclear medicine and are primarily produced through in-reactor irradiation. Neutron spectrum regulation is the main technical approach for enhancing the production of medical isotopes, and it requires determining the optimal neutron spectrum and quantifying the values of neutrons in different energy regions. We calculated the neutron energy region values for 20 medical isotopes (^{14}C , ^{32}P , ^{47}Sc , ^{60}Co , ^{64}Cu , ^{67}Cu , ^{89}Sr , ^{90}Y , ^{99}Mo , ^{125}I , ^{131}I , ^{153}Sm , ^{161}Tb , ^{166}Ho , ^{177}Lu , ^{186}Re , ^{188}Re , ^{92}Ir , ^{225}Ac , and ^{252}Cf). The entire energy range was divided into 238 energy regions to improve the energy spectrum resolution, and both fast and thermal reactors were simulated to enhance universal applicability. A dataset of neutron energy region values across the entire energy range was built, which identifies the positive and negative-energy regions and guides the neutron spectrum regulation process during in-reactor medical isotope production. We conducted neutron spectrum regulation based on this dataset, which effectively improved the production efficiency of medical isotopes and demonstrated the correctness and feasibility of the dataset.

Keywords Medical isotopes · High-flux isotope reactor · Neutron energy region value · Spectrum regulation

1 Introduction

Medical isotopes refer to radioactive isotopes used in medical diagnosis, treatment, and research [1]. These isotopes typically possess half-lives and radioactive properties suitable for medical applications, enabling their use in imaging, diagnosis, and radiotherapy [2]. Globally, over 100 radioactive isotopes can be applied in the medical field, and more than 30 medical isotopes are currently utilized [3], including molybdenum-99 (^{99}Mo), iodine-125 (^{125}I), iodine-131 (^{131}I), carbon-14 (^{14}C), lutetium-177 (^{177}Lu), ytterbium-90 (^{90}Y), and strontium-89 (^{89}Sr). Among them, ^{99}Mo is commonly

used for single-photon emission computed tomography (SPECT) imaging [4], ^{125}I can be used for brachytherapy and radioimmunoanalysis [5], ^{131}I is primarily used for the treatment of hyperthyroidism and thyroid cancer [6], ^{14}C is used for respiratory test drugs and pharmacokinetic studies [7], and ^{177}Lu is helpful for the targeted therapy of neuroendocrine tumors and prostate cancer [8]. Many other related applications [9, 10] have not been listed.

In-reactor irradiation is the primary approach for producing medical isotopes, accounting for 70–80% of the global medical isotope supply [11]. In-reactor medical isotope production involves complex nuclide-conversion processes. Taking the production of ^{125}I from xenon-124 (^{124}Xe) as an example [12], as shown in Fig. 1, ^{124}Xe undergoes a single (n, γ) reaction to produce xenon-125 (^{125}Xe) and then decays to ^{125}I . If the nuclear reaction continues, xenon-126 (^{126}Xe) is obtained, and this process generates unwanted impurities. Neutrons with different energy ranges exhibit different abilities to induce various nuclear reactions [13]. Nuclear reactions can be controlled by adjusting the neutron energy spectrum to enhance medical isotope production. Moreover, iodine-126 (^{126}I) with high-energy radiation is also generated, which

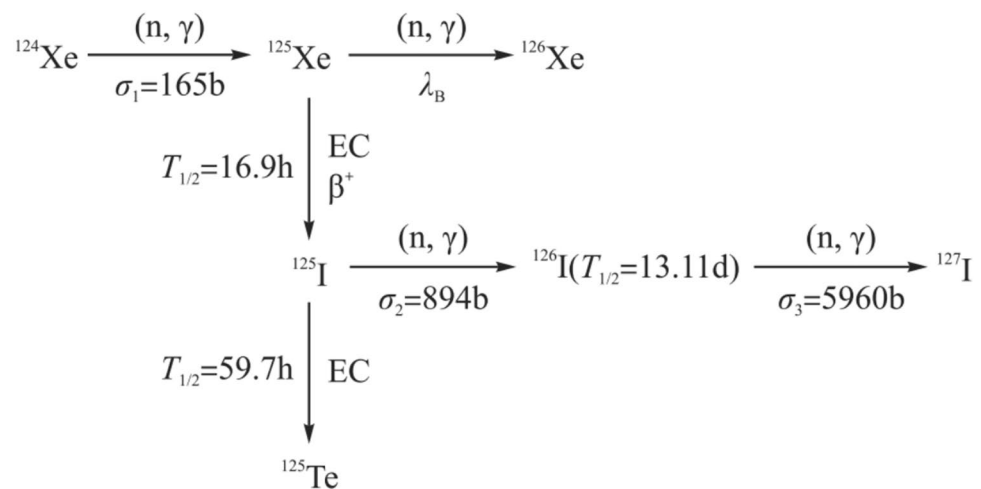
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Fig. 1 Nuclide conversion of ^{125}I production by irradiated ^{124}Xe



decreases the yield and quality of ^{125}I . Therefore, highly enriched ^{124}Xe is typically used for short-term irradiation to increase the ^{125}I yield and reduce the ^{126}I yield. Consequently, irradiation duration affects the nuclear reaction process, and a reasonable irradiation duration can effectively enhance the production of medical isotopes.

The global shortage of medical isotopes makes it crucial to enhance production efficiency. Nature has repeatedly called for attention to address the shortage of medical isotopes [14–16]. Increasing the yield of medical isotopes would benefit humans. Neutron spectrum regulation and irradiation duration optimization are primary technical approaches for improving the production efficiency of medical isotopes. However, the first step is to determine the optimal neutron spectrum for producing various medical isotopes under different irradiation durations, which necessitates the quantification of neutron values in different energy regions. Therefore, a dataset that provides neutron energy region values is essential for enhancing in-reactor medical isotope production.

We utilized the subgroup burnup analysis method [17] to quantify the value of neutrons in different energy regions for the in-reactor medical isotope production. This method considers all nuclides and neutron spectra throughout the irradiation period and constructs a dataset of neutron energy region values with a high-spectral resolution across the entire energy region. Moreover, the extreme burnup analysis method was employed to mitigate the influence of the irradiation duration on the neutron energy region values and thereby enhance the universality of the database. We performed neutron spectrum regulation based on this database, which effectively improved the production efficiency of medical isotopes, thereby demonstrating the correctness and feasibility of the database.

2 Models and methods

2.1 Reactor models

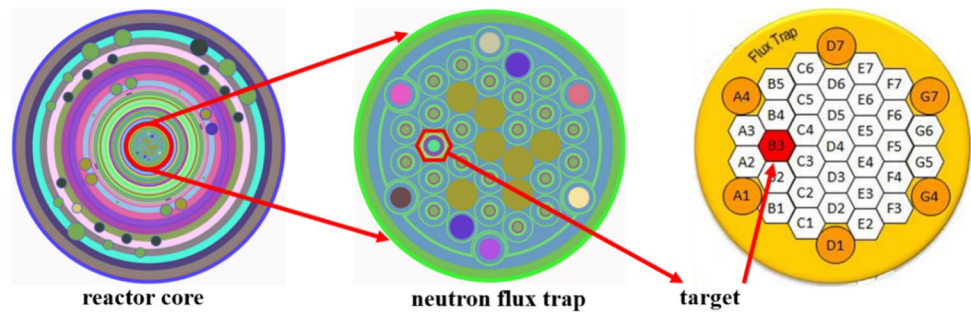
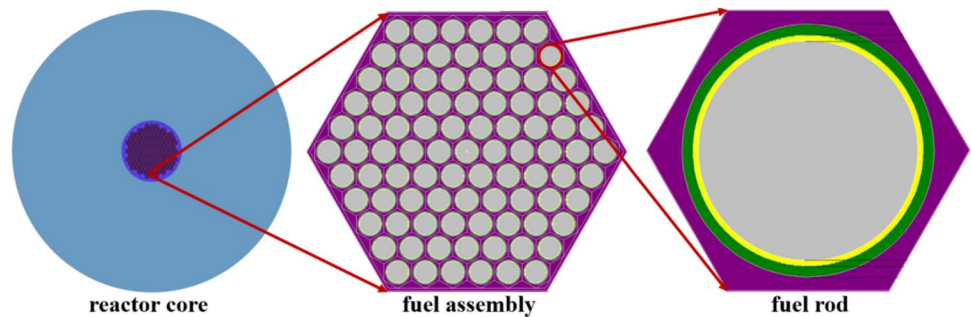
The dataset was simulated in the high-flux isotope reactor (HFIR) [18], which is a multipurpose isotope production and testing reactor with a current operating power of 85 MW. The HFIR operates on a cycle of 22–24 d and is capable of the world's highest steady-state neutron thermal flux ($2.6 \times 10^{15} \text{ cm}^{-2} \text{ s}^{-1}$), making it a significant global production base for medical isotopes. We inserted a target assembly with a length of 50 cm and diameter of 0.85 cm inside the neutron trap to produce twenty medical isotopes (^{14}C , ^{32}P , ^{47}Sc , ^{60}Co , ^{64}Cu , ^{67}Cu , ^{89}Sr , ^{90}Y , ^{99}Mo , ^{125}I , ^{131}I , ^{153}Sm , ^{161}Tb , ^{166}Ho , ^{177}Lu , ^{186}Re , ^{188}Re , ^{92}Ir , ^{225}Ac , ^{252}Cf). The reactor Monte Carlo (RMC) code was utilized for modeling [19], as illustrated in Fig. 2. Moreover, because HFIR is a thermal reactor, we also performed the same simulations in a high-flux fast reactor (HFFR) [17] to enhance the universality of the database, as shown in Fig. 3.

2.2 Calculation methods

The yields of medical isotopes can be obtained by performing a Monte Carlo burnup calculation that couples the Monte Carlo criticality and point burnup calculations. The point burnup equation describes the transmutation of nuclides with time and can be written as [20]:

$$\frac{dn_i}{dt} = \sum_{j \neq i} b_{j,i}^{\text{eff}} \lambda_j^{\text{eff}} n_j - \lambda_i^{\text{eff}} n_i \quad (1)$$

where n_i is the density of the i th nuclide, λ_i^{eff} is the effective decay constant of the i th nuclide, and $b_{i,j}^{\text{eff}}$ is the branching

Fig. 2 (Color online) modeling diagram of HFIR**Fig. 3** (Color online) modeling diagram of the fast reactor

ratio for transmuting the i th nuclide to the j th nuclide. Here, λ_i^{eff} and $b_{i,j}^{\text{eff}}$ can be calculated from the following formula:

$$\begin{cases} \lambda_i^{\text{eff}} = \lambda_i + \phi \sum_j \sigma_{i,j} \\ b_{i,j}^{\text{eff}} = (b_{i,j} \lambda_i + \sigma_{i,j} \phi) / \lambda_i^{\text{eff}} \end{cases} \quad (2)$$

where λ_i is the decay constant of the i th nuclide, ϕ is the neutron flux, and $\sigma_{i,j}$ is the one-group cross section where the reaction of the i th nuclide generates the j th nuclide.

One-group cross sections are required for the Monte Carlo burnup calculation, which is computed by the Monte Carlo criticality calculation:

$$(L + C - S)\phi = \frac{1}{k_{\text{eff}}} F\phi \quad (3)$$

where L is the leakage operator, C is the collision operator, S is the scattering operator, F is the fission operator, and k_{eff} is the effective multiplication factor.

The neutron flux and nuclear reaction rates are determined by solving Eq. (3) and are used to determine the one-group cross sections:

$$\sigma_{i,j} = R_{i,j} / \phi \quad (4)$$

where $R_{i,j}$ is the nuclear reaction rate where the i th nuclide generates the j th nuclide.

As shown in Eqs. (1)–(4), the Monte Carlo burnup calculation can only quantify the production efficiency of the entire neutron spectrum, not each energy region, which leads

to the unknown energy regions that are favorable and unfavorable for in-reactor medical isotope production. Therefore, two methods have been proposed for the high-spectral resolution quantification of neutron energy region values: (1) the subgroup burnup analysis method and (2) the extremum burnup analysis method.

(1) **Subgroup burnup analysis method:** The neutron flux in each energy region is reduced during the Monte Carlo burnup calculation, and the influence of the flux reduction on the yield is observed, quantifying the neutron energy region values (labeled I_i), which are calculated using the following formula:

$$I_i = \frac{\Delta Y_i / Y}{\Delta \phi_i} = \frac{(Y - Y_i) / Y}{\phi_i - \phi'_i} = \frac{(Y_i - Y_i) / Y_i}{M \cdot \phi_i} \quad (5)$$

where the subscript “ i ” is the index of an energy region; Y and Y_i are the yields of medical isotopes before and after flux reduction, respectively; ϕ and ϕ' are the neutron flux in the i th energy region before and after flux reduction, respectively; and M is the ratio of flux reduction, which takes values of one-eighth (1/8), one-quarter (1/4), one-half (1/2), and one (1/1). Previous research found [17] that M has a minor effect on these results. Therefore, we adopted $M = 8$. We achieved neutron flux reduction in a specific neutron energy region by modifying the source code of the RMC program.

The subgroup burnup analysis method calculates the sensitivity of the relative change in the yield to the absolute change in the neutron flux in each energy region. To ensure the high-spectral resolution of the dataset, the entire energy range was divided into 238 energy regions [21]. Moreover,

the subgroup burnup analysis method has no restrictions on the energy division and can achieve a higher spectral resolution. Taking the 90 d irradiation production of californium-252 (^{252}Cf) in the HFIR as an example [17], we obtained a neutron value curve across the entire energy range by calculating the neutron values for all 238 energy regions, as shown in Fig. 4. We also provided the top-10 maximum and minimum values along the neutron value curve, as shown in Table 1.

(2) **Extremum burnup analysis method:** The irradiation duration significantly affected the neutron energy region values calculated by the subgroup burnup analysis method. For example, in some energy regions, if the irradiation time exceeds what is required for the yield to reach its peak, then a phenomenon in which the yield increases first and then decreases throughout the irradiation process occurs. This phenomenon is known as “over-irradiation.” In other energy regions, irradiation for this duration may not be sufficient to cause the yield to reach its peak. To address this issue,

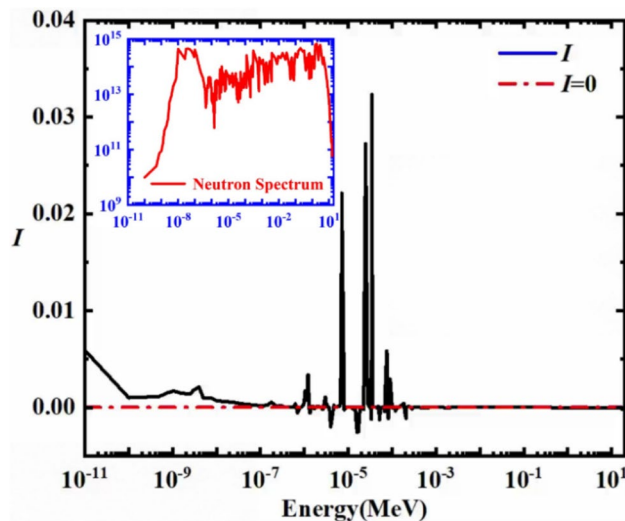


Fig. 4 Neutron values curve for ^{252}Cf production in HFIR

Table 1 Maximum and minimum values along the neutron value curve

Largest		Smallest	
Energy regions (MeV)	Values	Energy regions (MeV)	Values
$[9.00 \times 10^{-5}, 1.00 \times 10^{-4}]$	2.95×10^{-3}	$[1.60 \times 10^{-5}, 1.70 \times 10^{-5}]$	-2.57×10^{-3}
$[1.20 \times 10^{-6}, 1.23 \times 10^{-6}]$	3.29×10^{-3}	$[1.70 \times 10^{-5}, 1.85 \times 10^{-5}]$	-2.51×10^{-3}
$[1.23 \times 10^{-6}, 1.25 \times 10^{-6}]$	3.39×10^{-3}	$[4.00 \times 10^{-6}, 4.75 \times 10^{-6}]$	-1.99×10^{-3}
$[7.20 \times 10^{-5}, 7.60 \times 10^{-5}]$	3.50×10^{-3}	$[1.51 \times 10^{-5}, 1.60 \times 10^{-5}]$	-1.30×10^{-3}
$[7.60 \times 10^{-5}, 8.00 \times 10^{-5}]$	5.85×10^{-3}	$[5.20 \times 10^{-5}, 5.34 \times 10^{-5}]$	-1.28×10^{-3}
$[1.00 \times 10^{-11}, 1.00 \times 10^{-10}]$	5.92×10^{-3}	$[3.38 \times 10^{-5}, 3.46 \times 10^{-5}]$	-1.21×10^{-3}
$[7.00 \times 10^{-6}, 7.15 \times 10^{-6}]$	1.11×10^{-2}	$[8.20 \times 10^{-5}, 9.00 \times 10^{-5}]$	-1.13×10^{-3}
$[7.15 \times 10^{-6}, 8.10 \times 10^{-6}]$	2.21×10^{-2}	$[2.08 \times 10^{-4}, 2.10 \times 10^{-4}]$	-1.12×10^{-3}
$[2.50 \times 10^{-5}, 2.75 \times 10^{-5}]$	2.72×10^{-2}	$[7.00 \times 10^{-7}, 7.50 \times 10^{-7}]$	-5.70×10^{-4}
$[3.46 \times 10^{-5}, 3.55 \times 10^{-5}]$	3.23×10^{-2}	$[1.35 \times 10^{-6}, 1.40 \times 10^{-6}]$	-5.70×10^{-4}

we employed the extreme burnup analysis method to eliminate the influence of the irradiation duration on the neutron energy region values. The extreme burnup analysis method determines the extreme value in terms of time, which is achieved by calculating the maximum derivative between the neutron energy region value and irradiation time, as shown in Eq. (6):

$$E_j = \frac{\Delta I_j}{\Delta t} = \frac{I_{j+1} - I_j}{\Delta t} \quad (6)$$

where the subscript “j” is the index of a time step, I_j is the neutron energy region value in the jth time step, and Δt represents the time step.

We divided the irradiation duration (90 days) into 20 equal time steps and performed calculations using the subgroup burnup analysis method for each time step, resulting in 20 neutron energy region values for all 238 energy regions. We then calculated the maximum derivative $E_{j, \max}$ between the neutron energy region value and irradiation time and used it as the final result to quantify the neutron values in each energy region. The extremum burnup analysis method also established an extreme neutron value curve, as shown in Fig. 5. The top-10 maximum and minimum values along the curve are listed in Table 2.

The maximum values indicate that neutrons in these energy regions are conducive. Hence, increasing the neutron flux in these energy regions (referred to as positive-energy regions) increases the yield of medical isotopes. Conversely, the minimum values suggest that neutrons in these energy regions are detrimental and that increasing the neutron flux in these regions (referred to as negative-energy regions) decreases the yield.

Based on the subgroup and extremum burnup analysis methods, we obtained high-resolution neutron energy region values across the entire energy range and built a dataset to determine the positive- and negative-energy regions. Increasing the neutron flux in positive-energy regions and

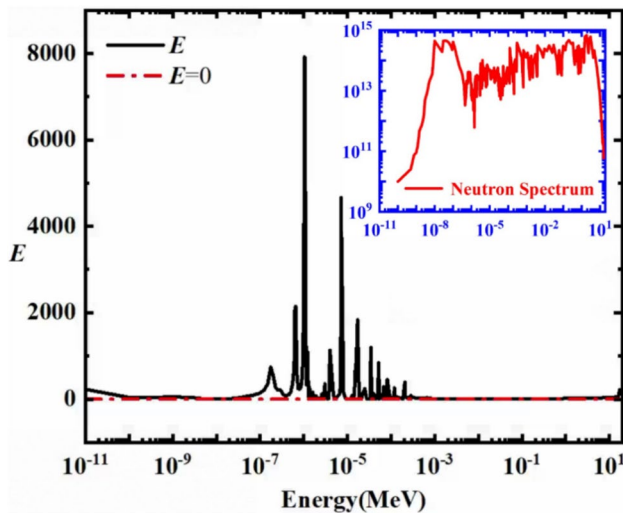


Fig. 5 Extreme neutron value curve for ^{252}Cf production in HFIR

reducing it in negative-energy regions can improve the yield of medical isotopes. Therefore, this dataset can help guide the neutron spectrum regulation process for in-reactor medical isotope production.

3 Data records

The dataset employs two methods (the subgroup burnup and extremum burnup analysis methods) to calculate the neutron energy region values across 238 energy regions for 20 medical isotopes (^{14}C , ^{32}P , ^{47}Sc , ^{60}Co , ^{64}Cu , ^{67}Cu , ^{89}Sr , ^{90}Y , ^{99}Mo , ^{125}I , ^{131}I , ^{153}Sm , ^{161}Tb , ^{166}Ho , ^{177}Lu , ^{186}Re , ^{188}Re , ^{92}Ir , ^{225}Ac , ^{252}Cf) in two reactor models (high-flux isotope and high-flux fast reactors). Consequently, there were a total of $2 \times 2 \times 20 \times 238 = 19,040$ neutron energy region values. We organized this dataset into a file named NERV.txt (meaning neutron energy region values), according to the format

specified in Table 3. This dataset is available at <https://doi.org/10.57760/sciencedb.j00186.00468>.

By downloading the NERV.txt file from the repository, readers can obtain the neutron energy region value in a specific energy region, calculated using either the subgroup burnup or extreme burnup analysis method, for the production of a certain medical isotope in HFFR or HFIR, according to the fixed format outlined in Table 3. For instance, if we wish to retrieve the neutron value for the 100th energy region, calculated by the extreme burnup analysis method, for the production of ^{131}I in the HFFR reactor. These data should be located in the 10th column of the 1150th row in the NERV.txt file, with a value of 5.17959×10^{-15} .

4 Validation and usage

4.1 Technical validation

We constructed numerous neutron spectra by dispersing nuclides into target materials and verified whether the yields of the medical isotopes under these neutron spectra positively correlate with the total value of the neutron spectra (labeled as “ T ”), which is calculated by:

$$T = \sum_{i=1}^N \bar{\phi}_i \cdot I_i \quad (7)$$

where N represents the number of energy regions, which equals 238; $\bar{\phi}_i$ denotes the normalized neutron flux in the i th energy region; and I_i indicates the neutron energy region value of the i th energy bin.

Taking Rhenium-188 (^{188}Re) as an example, we selected ten nuclides (^4He , ^7Li , ^{50}Cr , ^{64}Ni , ^{73}Ge , ^{75}As , ^{130}Ba , ^{145}Nd , ^{165}Ho , ^{176}Hf) with three added amounts (10^{18} , 5×10^{18} , and 1×10^{19} atom/cm 3), resulting in 30 different neutron spectra.

Table 2 Maximum and minimum values along the extreme neutron values curve

Largest		Smallest	
Energy regions (MeV)	Values	Energy regions (MeV)	Values
$[1.01 \times 10^{-6}, 1.02 \times 10^{-6}]$	3.77×10^4	$[5.00 \times 10^{-9}, 7.50 \times 10^{-9}]$	-1.10×10^1
$[1.09 \times 10^{-6}, 1.10 \times 10^{-6}]$	4.56×10^4	$[7.50 \times 10^{-9}, 1.00 \times 10^{-8}]$	-4.03×10^{-1}
$[7.15 \times 10^{-6}, 8.10 \times 10^{-6}]$	4.66×10^4	$[1.85 \times 10^{-5}, 1.90 \times 10^{-5}]$	7.23×10^{-2}
$[1.08 \times 10^{-6}, 1.09 \times 10^{-6}]$	5.28×10^4	$[5.00 \times 10^{-6}, 5.40 \times 10^{-6}]$	1.80×10^{-1}
$[1.02 \times 10^{-6}, 1.03 \times 10^{-6}]$	5.53×10^4	$[2.70 \times 10^{-1}, 3.30 \times 10^{-1}]$	2.28×10^{-1}
$[1.07 \times 10^{-6}, 1.08 \times 10^{-6}]$	6.52×10^4	$[3.30 \times 10^{-1}, 4.00 \times 10^{-1}]$	3.55×10^{-1}
$[1.03 \times 10^{-6}, 1.04 \times 10^{-6}]$	7.21×10^4	$[2.00 \times 10^{-1}, 2.70 \times 10^{-1}]$	3.69×10^{-1}
$[1.06 \times 10^{-6}, 1.07 \times 10^{-6}]$	7.43×10^4	$[1.50 \times 10^{-1}, 2.00 \times 10^{-1}]$	5.32×10^{-1}
$[1.04 \times 10^{-6}, 1.05 \times 10^{-6}]$	7.60×10^4	$[4.00 \times 10^{-1}, 4.20 \times 10^{-1}]$	7.23×10^{-1}
$[1.05 \times 10^{-6}, 1.06 \times 10^{-6}]$	7.92×10^4	$[1.28 \times 10^{-1}, 1.50 \times 10^{-1}]$	7.41×10^{-1}

Table 3 Format of the dataset of neutron energy region values

Fixed format	Example
Isotope name (character×10 to the left)	C-14.....
HFIR subgroup (character×13 to the left)	HFIR-Subgroup
24 lines with 10 values per line (character×13×10 to the left)×24	0.00000E+00-0.00000E+00...
HFIR extremum (character×13 to the left)	HFIR-Extremum
24 lines with 10 values per line (character×13×10 to the left)×24	0.00000E+00-0.00000E+00...
HFFR subgroup (character×13 to the left)	HFFR-Subgroup
24 lines with 10 values per line (character×13×10 to the left)×24	0.00000E+00-0.00000E+00...
HFFR extremum (character×13 to the left)	HFFR-Extremum
24 lines with 10 values per line (character×13×10 to the left)×24	0.00000E+00-0.00000E+00...
Isotope Name (character×10 to the left)	P-32.....
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The correlation coefficient between ΔY and ΔT was calculated as follows:

$$R(\Delta Y, \Delta T) = \frac{\text{Cov}(\Delta Y, \Delta T)}{\sqrt{\text{Var}[\Delta Y]\text{Var}[\Delta T]}} \quad (8)$$

where ΔY represents the change in the yield of ^{188}Re before and after the addition of nuclides, ΔT is the change in the total value of the neutron spectra before and after the addition of nuclides, $\text{Cov}(x, y)$ is the covariance between x and y , $\text{Var}[x]$ is the variance of x , and $R(x, y)$ is the correlation coefficient between x and y , shown in Table 4.

As shown in Table 4, ΔY exhibits a positive correlation with ΔT , with a correlation coefficient of 0.9409, meaning that as the total value of the neutron spectra increases, the yield of the medical isotopes also increases. Therefore, a dataset of the neutron energy region values can guide neutron spectrum regulation to enhance in-reactor medical isotope production.

4.2 Applications

The dataset of the neutron energy region values identified the positive- and negative-energy regions for producing various medical isotopes. Theoretically, the production efficiency of medical isotopes can be increased by reducing

the neutron flux in negative-energy regions. We adopted the neutron spectrum filtering technology to reduce the neutron flux in the negative-energy regions, which disperses nuclides with resonance peaks in the negative-energy regions to the target material. The effects of adding filter nuclides on the final yields, using the production of ^{47}Sc , ^{64}Cu , ^{67}Cu , ^{90}Y , and ^{188}Re as examples, are shown in Table 5.

Table 5 shows that when 5.00×10^{18} atom/cm³ of ^{155}Gd was dispersed into the target for producing ^{47}Sc , the yield of ^{47}Sc increased by 154%, which is an impressive result. The yields of medical isotopes increased significantly, proving that the dataset of neutron energy region values can be used for neutron spectrum regulation to enhance the production of medical isotopes. This dataset enables the rapid identification of positive- and negative-energy regions for producing these 20 medical isotopes in fast or thermal reactors. The production efficiency of medical isotopes can be effectively improved by increasing the neutron flux in positive-energy regions and reducing it in negative-energy regions. Additionally, this dataset can be used for the rapid evaluation and comparison of the economics of different irradiation schemes. As shown in Eq. (7), the total value of the neutron spectrum of an irradiation scheme can be calculated, and schemes with higher total neutron spectrum values are more economically viable.

Table 4 Correlation coefficient between ΔY and ΔT

Nuclides	Addition amount	ΔT	ΔY
^4He	1.00×10^{18}	-8.80×10^{-9}	-8.50×10^{17}
	5.00×10^{18}	6.47×10^{-10}	-3.00×10^{17}
	1.00×10^{19}	-1.10×10^{-9}	-5.10×10^{17}
^7Li	1.00×10^{18}	7.85×10^{-9}	2.66×10^{16}
	5.00×10^{18}	-1.70×10^{-9}	-5.70×10^{17}
	1.00×10^{19}	-7.30×10^{-10}	-4.20×10^{17}
^{50}Cr	1.00×10^{18}	-5.20×10^{-9}	-7.40×10^{17}
	5.00×10^{18}	-6.30×10^{-9}	-8.90×10^{17}
	1.00×10^{19}	-3.70×10^{-9}	-5.80×10^{17}
^{64}Ni	1.00×10^{18}	2.81×10^{-9}	-1.50×10^{16}
	5.00×10^{18}	2.92×10^{-9}	8.95×10^{16}
	1.00×10^{19}	7.29×10^{-9}	1.49×10^{17}
^{73}Ge	1.00×10^{18}	-4.60×10^{-10}	-2.00×10^{17}
	5.00×10^{18}	-7.40×10^{-9}	-7.70×10^{17}
	1.00×10^{19}	1.11×10^{-8}	6.27×10^{17}
^{75}As	1.00×10^{18}	-1.30×10^{-9}	-2.91×10^{17}
	5.00×10^{18}	-3.40×10^{-9}	-4.23×10^{17}
	1.00×10^{19}	-4.80×10^{-9}	-6.01×10^{17}
^{130}Ba	1.00×10^{18}	2.31×10^{-9}	-2.42×10^{17}
	5.00×10^{18}	-1.50×10^{-9}	-7.63×10^{17}
	1.00×10^{19}	5.84×10^{-9}	-2.38×10^{15}
^{145}Nd	1.00×10^{18}	-8.00×10^{-10}	-4.79×10^{17}
	5.00×10^{18}	-1.50×10^{-9}	-4.30×10^{17}
	1.00×10^{19}	1.58×10^{-11}	-3.18×10^{17}
^{165}Ho	1.00×10^{18}	3.56×10^{-9}	7.96×10^{16}
	5.00×10^{18}	7.29×10^{-10}	-2.40×10^{17}
	1.00×10^{19}	-8.40×10^{-9}	-1.10×10^{18}
^{176}Hf	1.00×10^{18}	3.61×10^{-9}	6.73×10^{16}
	5.00×10^{18}	-2.00×10^{-9}	-7.00×10^{17}
	1.00×10^{19}	7.00×10^{-9}	1.95×10^{17}
Correlation coefficients between ΔY and ΔT		0.9409	

4.3 Data availability

The process files generated during the calculation of this dataset can be accessed at <https://doi.org/10.57760/sciencedb.j00186.00468>. The calculation results of the subgroup burnup analysis method for each medical isotope can be viewed in the “Subgroup burnup” folder, which contains the raw data records and yields of this medical isotope after flux reduction in each energy region along with the corresponding neutron value curves and top-10 maximum and minimum values within the neutron value curves. The calculation results of the extreme burnup analysis method for each medical isotope can be accessed in the “Extremum burnup” folder, which records the yields of this medical isotope after flux reduction in each energy region across the 20 time steps as well as the corresponding extreme neutron value curves and top-10 maximum and minimum values within the curves.

Both the folders named “Subgroup burnup” and “Extremum burnup” contain two subfolders each, named “HFFR” and “HFIR,” which represent the computational data from these two reactors. Furthermore, both the “HFFR” and “HFIR” folders were further divided into 20 subfolders, with each named after the 20 medical isotopes covered in this dataset. These 20 subfolders were not further subdivided and contained the lowest-level data information from the computational process.

The raw data for use in technical validation and applications are stored in the “Spectrum Regulation” folder, which contains two folders named “Technical Validation” and “Application.” The “Technical Validation” folder includes 10 subfolders and an Excel sheet, which provide detailed data for technical validation. The “Application” Folder contains 5 subfolders and an Excel sheet that offer detailed data on the application process.

Table 5 Improvements of adding filter nuclides on the final yields

Medical isotopes	Filter nuclides	Addition amount (atom/cm ³)	Yield before filter (atom/cm ³)	Yield after filter (atom/cm ³)	ΔY (%)
^{47}Sc	^{153}Gd	5.00×10^{18}	3.06×10^{17}	7.19×10^{17}	135.30
	^{155}Gd	5.00×10^{18}	3.06×10^{17}	1.54×10^{17}	154.10
^{64}Cu	^{157}Gd	5.00×10^{18}	1.50×10^{19}	1.84×10^{19}	22.49
	^{149}Sm	5.00×10^{18}	1.50×10^{19}	2.58×10^{19}	72.28
^{67}Cu	^{155}Gd	5.00×10^{18}	2.83×10^{16}	3.10×10^{16}	9.43
	^{157}Gd	5.00×10^{18}	2.83×10^{16}	5.00×10^{16}	76.40
^{90}Y	^{157}Gd	5.00×10^{18}	2.81×10^{20}	3.83×10^{20}	36.23
	^{149}Sm	5.00×10^{18}	2.81×10^{20}	3.69×10^{20}	31.26
^{188}Re	^{243}Cm	5.00×10^{18}	1.15×10^{19}	1.47×10^{19}	27.60
	^{245}Cm	5.00×10^{18}	1.15×10^{19}	2.02×10^{19}	75.20

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Declarations

Conflict of interest Xiao-Jing Liu is an editorial board member for Nuclear Science and Techniques and was not involved in the editorial review, or the decision to publish this article. All authors declare that there are no competing interests.

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