PERFORMANCE OPTIMIZATION OF RADIOACTIVE WASTE TRANSMUTATION IN ACCELERATOR DRIVEN SYSTEM

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Abstract
Long-lived actinides transmutation in accelerator driven system (ADS) is considered in this paper. The objective is to improve the transmutation performance from the overall radiotoxicity of nuclear waste point of view by means of an optimization approach based on control theory. Parameters subjected to determination are the initial values of actinides nuclear concentration in the loaded fuel. The change in time of actinides' atomic concentration is described by a system of ordinary differential equations with initial conditions, that can be given with some inaccuracy. For such dynamical system a gradient-based optimization algorithm taking into account robustness of initial loading will be constructed. Moreover, constraint on value of effective multiplication factor is considered during numerical optimization. Based on this approach initial atomic concentration values providing the best indicators of actinides burnup can be obtained. Calculation results of actinides volume fraction and effective multiplication factor dynamics are presented.

Key words
Optimization, optimal control, ADS, transmutation, radioactive waste.

1 Introduction
The level of ecological impact of nuclear power engineering is measured by number of radionuclides (fission products (FP) and actinides (MA)), which are formed during nuclear reactors operation. The radiotoxicity of spent nuclear fuel is firstly dominated by short-lived fission products, and later, by actinides [nuc, 1996]. A few hundred-thousand years after the discharge, the radiotoxicity of the unprocessed fuel drops to the natural toxicity level for traditional thermal reactors, i.e. the equilibrium radiotoxicity of the natural uranium ore [Mai, 2002]. The separation (and intermediate storage) of uranium and plutonium could reduce the radiotoxicity of the remaining high-level radioactive waste by an order of magnitude, but it would still take about twenty-thousand years for the radiotoxicity of this waste to reach the natural toxicity level [oec, 2006].

Ecological impact of short-lived radionuclides can be reduced till the admissible level using organizational and technical measures in storage and fuel cycle facilities. However, in case of geological disposal, long-lived nuclear waste has a potential risk over a few ten thousands years. The geological disposal site is difficult to select, mainly because it is hard to predict the changes in the structure of ground layers for the coming ten thousands years. Due to these facts, the problem of long-lived radioactive waste elimination arises, as an important technological research topic. The de-
2 Problem Statement

Evolution of fuel concentration in ADS is described by a system of ordinary differential equations (also known as Bateman equations [Bateman, 1910]):

\[
\frac{dN_i}{dt} = \sum_{j \neq i} f_{j \rightarrow i} \lambda_j N_j + \sum_{k \neq i} \sum_{x} g_{x}^{k \rightarrow i} N_x \langle \sigma_x^k \phi \rangle + \sum_{l} \gamma_{l \rightarrow i} F_l - (\lambda_i + \langle \sigma_x^i \phi \rangle) N_i,
\]

with given initial conditions:

\[
N_i(0) = N_i^0,
\]

where

\(N_i\) — atomic concentration of isotope \(i\);
the first term on the right-hand side expresses the production rate of nuclide \(i\) due to radioactive decay of another nuclide \(j\) on the burnup chain;
\(\lambda_j\) is the decay constant of nuclide \(j\) and \(f_{j \rightarrow i}\) is the probability of decay to nuclide \(i\);
the second term is the production rate of nuclide \(i\) due to the nuclear reaction \(x\) of another nuclide \(k\); the major nuclear reaction is the neutron capture reaction and other reactions such as the \((n, 2n)\) reaction can be considered if necessary;
\(g_{x}^{k \rightarrow i}\) is the probability of transmutation into nuclide \(i\) for nuclear reaction \(x\) of nuclide \(k\);
the third term is for the production of FPs;
\(F_l\) is the fission rate of heavy nuclide \(l\) and \(\gamma_{l \rightarrow i}\) is the production probability (yield fraction) of nuclide \(i\) for the fission reaction;
the last term is the loss rate of nuclide \(i\) due to radioactive decay and absorption reactions.

\(\langle \sigma_x^k \phi \rangle\) — the microscopic reaction rate of nuclide \(k\) integrated over all neutron energies — is defined by the following relation:

\[
\langle \sigma_x^k \phi \rangle = \int_{E_{min}}^{E_{max}} \sigma_x^k(E)\phi(E)dE
\]

\[
\approx \sum_{i=1}^{G} \sigma_x^k(E_i)\phi(E_i)\Delta E_i,
\]

where \(\phi(E_i)\) represents the value of neutron flux normalized to the reactor heat power in the \(i\)-th energy group, \(G\) — number of considered energy groups.

In general, neutron flux is a volume-averaged solution of stationary neutron transport equation. Its diffusion
approximation will be considered in the present paper [Okumura et al., 2014]:

\[ D \nabla^2 \phi - \Sigma_a \phi + S = 0. \quad (4) \]

with the notation

- \( D \) — diffusion coefficient,
- \( \Sigma_a = \sum_{i=1}^{n} N^i \sigma^i_a \) — macroscopic absorption cross section in the core (\( n \) — total number of considered isotopes),
- \( S \) — neutron supply from fission, scattering and external neutron source.

In should be noted that flux value in (1) is considered to be constant during the step of integration. Equation (4) is solved again with new atomic concentrations obtained after one step of (1) integration. Neutron flux from (4) is calculated in multi-group diffusion approximation using the program code MDSR [Golovkina and Kudinovich, 2017].

System (1) can be rewritten in the compact matrix form:

\[ \dot{x} = f(t, x) = A(t) x, \quad (5) \]

where we assume that \( x = [N^1, N^2, \ldots, N^n] \), \( A(t) \) — matrix of \( n \times n \) dimensions, constructed on the basis of burnup chain (see Fig. 1 [Okumura et al., 2014]) for considered actinides and \( n \) is the number of considered isotopes [Golovkina et al., 2018b].

Initial conditions for system (5), determining the initial loading of fuel to the reactor core are given

\[ x(0) = \bar{x}_0. \quad (6) \]

The aim of optimization process is to find \( \bar{x}_0 \) values providing the best performance of transmutation: minimal radiotoxicity of discharged fuel. The initial values determining dynamics of system (5) can be effectively realized in the process with some error that can lead to the different performance of waste transmutation. In order to take into account such possibility it is supposed that initial values of nuclear concentrations \( x_i(0) \) can be taken from some neighborhood of point \( \bar{x}_0 \). Then the initial values of nuclear concentrations can be taken from the interior part of the ellipsoid

\[ (x_0 - \bar{x}_0)^* G_0 (x_0 - \bar{x}_0) = 1, \quad (7) \]

where \( G_0 \) is a positive definite symmetric matrix, \( \bar{x}_0 \) — the ellipsoid center. Ellipsoid shift along the solutions of a linear system is also ellipsoid with center in \( \bar{x}(t) \), bounding nuclear concentration initial values set in time moment \( t \):

\[ (x - \bar{x}(t))^* G(t) (x - \bar{x}(t)) = 1. \quad (8) \]

Here \( \bar{x}(t) = x(t, \bar{x}_0) \) — the solution of system (5) with initial conditions (6), matrix \( G(t) \) — determined by the solution of the following equation [Ovsyannikov, 1980]

\[ \frac{dG}{dt} = -A^*(t) G(t) - G(t) A(t) \quad (9) \]

with initial conditions \( G(0) = G_0 \).

The inverse matrix \( G^{-1}(t) \) satisfies the similar to (9) equation [Ovsyannikov, 1980], [Ovsyannikov, 1990]

\[ \frac{dG^{-1}}{dt} = AG^{-1} + G^{-1} A^*. \quad (10) \]

It’s clear from the matrix equation (9) that initial condition (6) (considered here as a control actions) doesn’t influence the matrix \( G(t) \), determining the ellipsoid shape at some time moment \( t \).

Because ADS is considered in the paper as transmutation facility, the requirement of its reactor to stay subcritical throughout the operation time should be taken
3 Optimization Approach

The aim of optimization is to provide the minimal output waste radioactivity after ADS operation ends in time \( T \). The corresponding performance criterion in this case is

\[
J(p) = \sum_{i=1}^{n} \omega(\bar{x}_i^2(T)),
\]

where \( \omega(\bar{x}_i^2(T)) = \bar{x}_i^2(T) \lambda_i \) — function, translating isotopes nuclear concentrations [1/cm\(^3\)] in their specific radioactivity [Bq/cm\(^3\)], \( \lambda_i \) — radioactive decay constant [1/s], \( p \) — control parameter subjected to optimization (in our case the initial values \( \bar{x}_0 \)).

In this formulation the similar problem for advanced fuel cycle was solved in [Golovkina et al., 2018a]. In this paper it’s proposed to consider the problem of robust optimization, stipulatory to the possible errors in fuel loading to the ADS reactor core. In order to provide the robustness to the optimal solution besides minimization of (12) the problem of ellipsoid (8) radius constriction in moment \( T \) should be taken into account.

Let us denote \( S = G^{-1}, z = x - \hat{x}(t) \), then ellipsoid with matrix \( S \) is described by the equation

\[
z^* S(t) z = 1.
\]

The objective is then to find the maximal value of \( z \) on the ellipsoid (8) surface. Using the standard approach with Lagrangian multipliers we will get the following estimation [Ovsyannikov, 1990]

\[
\max_{z^*G(t)z = 1} z_i = \sqrt{s_{ii}(t)},
\]

where \( s_{ii} \) — diagonal elements of matrix \( S \), \( z_i \) — the \( i \)-th component of vector \( z \). Thus functions \( \sqrt{s_{ii}(t)} \) are the envelopes on the variables \( x - \hat{x}(t) \) of system (5) [Ovsyannikov, 2012].

Thereby the optimized criterion, taking into account the system robustness is

\[
J(p) = \sum_{i=1}^{n} \omega \left( \bar{x}_i(T) + \sqrt{s_{ii}(T)} \right)^2 .
\]

Gradient based optimization methods can be used to minimize functional (14). In this case the explicit expression of its gradient should be obtained

\[
\nabla J(p) = \frac{d J}{d p} = \left( \frac{\partial J}{\partial p_1}, \ldots, \frac{\partial J}{\partial p_n} \right)^* .
\]

Let us denote \( y_i = \bar{x}_i(T) + \sqrt{s_{ii}(T)}, \varphi() = \sum \omega(\cdot^2) \), then the expression (14) can be rewritten in the form

\[
J(p) = \sum_{i=1}^{n} \omega(y_i^2(T)) = \varphi(y(T)).
\]

\[
\frac{d J}{d p} = \frac{d \omega}{d y} \cdot \frac{d y}{d p} = \left( \frac{\partial \omega}{\partial y_1}, \ldots, \frac{\partial \omega}{\partial y_n} \right) \cdot \left( \frac{\partial y}{\partial p_1}, \ldots, \frac{\partial y}{\partial p_n} \right)^* .
\]

\[
\frac{d y}{d p_i} = \frac{\partial \bar{x}_i(T)}{\partial p_i} + \frac{\partial \sqrt{s_{ii}(T)}}{\partial p_i} .
\]

In the expression (16) the first term can be obtained using variation equation [Golovkina et al., 2018a]

\[
\frac{d}{d t} \left( \frac{\partial x(t, p)}{\partial p_i} \right) = A(t) \frac{\partial x(t, p)}{\partial p_i},
\]

\[
\frac{\partial x(0, p)}{\partial p_i} = \mathbb{E}.
\]

The second term in (16) is equal to zero, because as shown above in the case of dynamical process described by a linear system (5) the elements of matrix \( S \) do not depend on the parameter \( p \). Their values at each moment of time \( t \) are completely determined by the initial values from \( S_0 \) and matrix of the system (5) \( A \). Thus the robustness of initial loading (uncertainty limits of the parameter \( p \)) can be estimated by matrix \( S_0 \).

Taking into account the previous considerations gradient components can be calculated using the following formula

\[
\frac{\partial J}{\partial p_i} = 2 \left( \bar{x}_i(T) + \sqrt{s_{ii}(T)} \right) \frac{\partial \bar{x}_i(T)}{\partial p_i} .
\]
Thus optimization problem is narrowed down to the solution of equation (17) and the following use of one of the gradient-based methods constructed with the obtained expression for gradient components (18). It should be noted that taking into account the constraint (11) now is carried out in the following way. At each optimization step the value of (11) is computed, if it exceeds 1 then the step size in optimization method should be reduced. More comprehensive way of considering this condition is being developed.

4 Calculation Results

Program code for calculation of fuel isotopes composition dynamics and optimization of initial fuel loading is developed in order to implement the described techniques. It is shown that the burnup capability of ADS is better in comparison with conventional nuclear reactors. Transmutation performance criterion (14), taking into account robustness of the system is used as an objective function. As was described earlier control of constraints values (11) is carried out on each of the step during numerical optimization. Fig. 2 shows the evolution of the neutron multiplication factor during the operation time. In table 1 volume fractions of transmutated actinides in initial ADS loading and their fractions by the end of operation cycle are presented. Calculation also showed that overall radioactivity of the charged fuel was reduce in approximately 5 times.

![Figure 2. $k_{eff}$ time evolution during transmutation cycle (dotted line is an approximation curve)](image)

| Isotope | vol % before | vol % after | Isotope | vol % before | vol % after |
|---------|--------------|-------------|---------|--------------|-------------|
| Pu$^{239}$ | 52.0 | 17.0 | Am$^{241}$ | 12.0 | 6.0 |
| Pu$^{239}$ | 21.0 | 37.0 | Cm$^{242}$ | 5.0 | 13.0 |
| Pu$^{241}$ | 2.0 | 6.0 | Cm$^{243}$ | 0.2 | 0.5 |
| Pu$^{242}$ | 6.0 | 2.0 | Cm$^{244}$ | 0.1 | 1.0 |

Table 1. Comparison of actinides volume fraction in charged and discharged fuel in ADS

5 Conclusion

Optimization approach based on control theory applied to the problem of long-lived actinides transmutation in accelerator driven system is considered in the paper. The transmutation performance criterion taking into account the system robustness is constructed. Gradient-based optimization algorithm is designed and implemented as a program tool. Constraint on value of effective multiplication factor is taken into account during numerical optimization. Based on this approach some calculations results were obtained. After optimization radioactivity of fuel charged in the core was reduced in approximately 5 times.

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