open science

# Parareal in time 3D numerical solver for the LWR Benchmark neutron diffusion transient model. 

Mohamed-Kamel Riahi, Anne-Marie A.-M. Baudron, Jean-Jacques Lautard,<br>Yvon Maday, Julien Salomon

## To cite this version:

Mohamed-Kamel Riahi, Anne-Marie A.-M. Baudron, Jean-Jacques Lautard, Yvon Maday, Julien Salomon. Parareal in time 3D numerical solver for the LWR Benchmark neutron diffusion transient model.. 2011. hal-00722802v1

HAL Id: hal-00722802
https://hal.science/hal-00722802v1
Preprint submitted on 4 Nov 2012 (v1), last revised 13 Jan 2015 (v4)

HAL is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers.

L'archive ouverte pluridisciplinaire HAL, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.

# Time-parareal parallel in time integrator solver for time-dependent neutron diffusion equation 

J.-J. Lautard, A.-M. Baudron<br>CEA-DRN/DMT/SERMA, CEN-Saclay, 91191 Gif sur Yvette Cedex, France<br>Y. Maday ${ }^{1 \text {, }}$<br>UPMC Univ Paris 06, UMR 7598, Laboratoire Jacques-Louis Lions, F-75005, Paris, France, and Division of Applied Mathematics, Brown University, Providence, RI, USA.<br>M.-K. Riahi ${ }^{1,}$<br>UPMC Univ Paris 06, UMR 7598, Laboratoire Jacques-Louis Lions, F-75005, Paris, France.<br>J. Salomon ${ }^{1,}$<br>CEREMADE, Univ Paris-Dauphine, Pl. du Mal. de Lattre de Tassigny, F-75016, Paris France. 1


#### Abstract

The accurate knowledge of the time-dependent spatial flux distribution in nuclear reactor is required for nuclear safety and design. The motivation behind the development of parallel methods for solving the energy-, space-, and time-dependent kinetic equations is not only the challenge of developing a method for solving a large set of coupled partial differential equations, but also a real need to predict the performance and assess the safety of large commercial reactors, both these presently operating and those being designed for the future. Numerical results show the efficiency of the parareal method on large light water reactor transient model.


Keywords: Parareal in time algorithm, time-dependent neutron diffusion equations, reduced model, parallel computation, hight performance computing. 2010 MSC: code, primary 65Y05; Secondary 35Q20.

[^0]
## 1. Introduction

In this paper, we present a time-parallel algorithm that simulate the kinetic of neutron in a nuclear reactor. We consider a large set of coupled time-dependent partial differential equations (PDEs) that predict the nuclear engine behavior and in particular its energy production.

The flux distribution on the nuclear reactor is given by a time-dependent neutron transport equation. For reason of computing time, the time-dependent neutron diffusion equation is usually used to study the transient behavior of the flux. These diffusion equations are coupled with the dynamic of some delayed neutrons, which are called precursors.

The methods for numerically advancing the space-time diffusion group diffusion equations, along with their time dependent delayed neutron precursors counterparts, through time have been developed. For each of these considered methods, the system to be solved has been reduced to the matrix form. The H -method is an accurate and efficient finite difference scheme that has been employed in the numerical integration of the time-dependent group diffusion equations since the mid-1960s. First-order finite difference schemes such as the implicit (backward) Euler, explicit (forward) Euler, and Crank-Nicholson (central difference) schemes can be derived easily from the H-method in its general form. The time approximation used for the solution of the kinetic diffusion equation with the classical method or the improved quasi-static method [3] is based on an integral $\theta$-scheme or a difference $\theta$-scheme. The kinetics diffusion equations are solved with a time discretization using an integral $\theta$-scheme both on the flux and the precursors equations (exact integration of the precursor equations with a linear expansion of the cross sections and polynomial representation of the flux). In this paper, for a sack of simplicity, we will consider difference $\theta$-scheme for both neutron flux and it's time-dependent delayed precursor groups equation. Since not yet parallelization in time was not purely investigated with large scale Kinetic equations, our main topic in this paper is the time-parallel solving of the coupled equations present in the neutron model.

The knowledge of the energy of the nuclear reactor at forward long time is very important for nuclear safety, hence very long time machine simulation. In addition for a lack of availability of read-write memory in the sequential computers nowadays, it is relevant to often propose parallel methods, which solve these large scaled system, with massively parallel computers. Many successful works, has been done around parallelization of the approximated resolution of partial differential equation arising from neutron models. For instance [2] studies the static case i.e. Eigenvalues problems with space domain decomposition methods, and a very nice strategy [3], [4] uses quasi-stationary approach to accelerate the simulation. This approach opens some directions to parallelize the implementation.

In this paper we investigate the application of the parareal in time algorithm $[5,6]$ on neutron diffusion equation that governs the time-dependent flux distribution in the nuclear reactor. The parareal in time algorithm is an iterative scheme, that enables to improve computational time with parallel simulation. In several cases, parareal in time algorithm gives an impressive rate of convergence for the linear diffusion equations or more unexpected efficiency with non-linearity [7]. This algorithm is studied and shows stability and convergence results $[8,9,10,11,12]$ particularly for diffusion system and others. Also it presents efficiency in parallel computer simulation. We find a variety [13, $14,15,9,16,17,18,19]$ of versions of this scheme that adapt $[20,21,22,23,24,25]$
the original algorithm to tackle new settings. Furthermore the parareal algorithm can be easily coupled with other iterative schemes such as domain decomposition methods e.g. basic Schwartz algorithm or more complex one [26], and optimal control based descent algorithm [27, 28, 29].

The work in this paper is presented as follows: After this introduction, we present at the first section the partial differential equation that model the kinetic of neutron inside the nuclear reactor. We discuss that model, explain the real-physics and present the maths beside. The second section is devoted to numerical tools that we use to spacetime discretize the problem. We introduce in the end of this section the parareal in time algorithm, which we adapt to the discretized problem. The parareal in time algorithm is therefore presented in a discrete way related to unification of variables that we solve all-at-once. We develop at the fourth section some approach that enables us to reduce the first model while remaining the reduced model close to the non-reduced one. That reduction is implemented in the coarse solver of the parareal in time algorithm that helps us to predict the solution for a forward time. We finally present and discuss the numerical experiments that show the speedup given when using the parareal in time algorithm.

## 2. Model

The neutron dynamics of a nuclear system (in absence of any external source) can be modeled by a time-dependent neutron diffusion equation (e.g. the equation (1)) associated with the time-dependent delayed neutron precursor equations (e.g. the equation (2)).

### 2.1. Time-dependent partial differential equation

The directional neutron flux solution of kinetic equation is a function at $\vec{r}$ about point of reactor domain, in the interval " $d E$ " about energy $E$, moving in the cone of directions $d \vec{\omega}$ about direction of $\vec{\omega}$ at time $t$. For computational reasons, we use different simplifications. Concerning the energy variable, the time-dependent equations are usually solved using its multi-group formulation which is the basis for the vast majority of computer programs. By integrating in limited energy intervals from $E_{g}$ to $E_{g+1}$ obtained by subdividing the energy $E$ from zero to infinity into $\hat{g}$ intervals, a set of $\hat{g}$ coupled differential equations. The second simplification concerns the direction variable. We look for the scalar flux which is an average flux on different directions. The scalar flux is solution of a simplified transport equation: the following diffusion equation

$$
\left\{\begin{align*}
& \frac{1}{V_{g}} \frac{\partial \phi_{g}(\vec{r}, t)}{\partial_{t}}= \operatorname{div}\left(D_{g} \vec{\nabla} \phi_{g}(\vec{r}, t)\right)-\Sigma_{t}^{g} \phi_{g}(\vec{r}, t)+\chi_{g}^{p} \sum_{g^{\prime}}\left(1-\beta_{g^{\prime}}\right) \nu \Sigma_{f}^{g^{\prime}} \phi_{g^{\prime}}(\vec{r}, t)  \tag{1}\\
&+\sum_{g^{\prime^{\prime}}} \Sigma_{s}^{g^{\prime}} \phi_{g^{\prime}}(\vec{r}, t)+\sum_{k=1}^{K} \chi_{k, g}^{d} C_{k}(\vec{r}, t), \quad \vec{r} \in \Omega \subset \mathbb{R}^{3} . \\
& \phi_{g}(\vec{r}, t)=\quad 0 \quad \text { on the boundary of the nuclear reactor } \\
& \phi_{g}(\vec{r}, 0)=\quad \phi_{g}^{0}(\vec{r}) \quad \text { a given initial value }
\end{align*}\right.
$$

The delayed neutron precursor satisfy :

$$
\begin{equation*}
\frac{\partial C_{k}}{\partial t}(\vec{r}, t)=-\lambda_{k} C_{k}(\vec{r}, t)+\sum_{g^{\prime}}^{\hat{g}} \beta_{k g^{\prime}} v \Sigma_{f}^{g^{\prime}} \phi_{g^{\prime}}(\vec{r}, t) \tag{2}
\end{equation*}
$$

In Equations (1) and (2) one reads:
$V_{g}$ is the neutron velocity in energy group $g$,
$\phi_{g}$ is the neutron scalar flux in energy group $g$,
$D_{g}$ is the diffusion coefficient in energy group $g$,
$\Sigma_{t}^{g}, v \Sigma_{f}^{g}$ are the total and production cross-section in energy group $g$,
$\Sigma_{s}^{g^{\prime} g}$ is the transfer cross-section from energy group g' to $g$,
$\chi_{g}^{p}, \chi_{k g}^{d}$ is the fission spectra of prompt and delayed neutrons, respectively $C_{k}$ is the concentration of precursor group $k$,
$\beta_{k g}, \lambda_{k}$ are the delay fraction in energy group g and decay constant of precursor group $k$, $\beta_{g}$ is the total delay fraction in energy group $g\left(\beta_{g}=\sum_{k} \beta_{k g}\right)$.
The equation (1) is derived from the transport equation by expanding into spherical harmonics and integrated over all directions $\vec{\omega}$, neglecting several terms. For each energy group $g=1, \ldots, \hat{g}$ and $\vec{r} \in \Omega \subset \mathbb{R}^{3}$, the equations are a set of the three-dimensional, multi-energy-group neutron kinetics equations, which are derived from (1).

### 2.2. Boundary condition in symmetrical physical core

In what follows, for reason of simplicity, we restrict our simulation for a quarter of the domain (nuclear reactor). Consequently, we consider Neuman boundary condition at the symmetry axes. That translate some null outcome across theses interfaces. The boundary condition is thus:

$$
\frac{\partial \phi}{\partial \vec{n}}(\vec{r}, t)=0 \quad \text { on the interior interfaces. }
$$

Such a way, the rest of the domain (nuclear rector) is simply deduced by symmetry.

### 2.3. Initial condition determination

For the initial condition, we consider the stationary version of the equation (7). We use the concept of reactivity which is related to the energetic behavior of the nuclear reactor. We focus on the distribution of the flux when the nuclear reactor is stable. This distribution is thus said stable state. In order to calculate this stable state, we introduce an eigenvalues problem on effective reactivity denoted by a $k_{\text {eff }}$ factor. This factor indicates the greatest generalized eigenvalue of the two matrices $\widehat{F}:=\boldsymbol{\Sigma}_{h}^{f}$ and $\widehat{A}$ (see Eq (9) below) of production and absorption respectively. The production matrix characterizes the production of the neutrons by fission, whereas absorption matrix caracterize the escapes/leaks. The eigenvalues problem is as follows:

$$
\widehat{A} \Phi=\frac{1}{k_{e f f}} \widehat{F} \Phi
$$

We use finally an the power algorithm to compute the $k_{\text {eff }}$ factor.
The nuclear core may have three states of energy (subcritical, critical and supercritical). These three states are related to the density of the neutrons diffusing in the core. That density is due to neutron production by the chain reaction whereas the disappearance of the neutrons by absorption and/or some leaks. The core is said critical if the neutron
outcome is null. It means: at every moment the number of produced neutrons is equal to the number of disappeared ones. The neutron outcome inside core at equilibrum obeys:

$$
\begin{equation*}
\operatorname{production}_{\text {fission }}+\text { source }_{\text {external }}=\text { absorption }+ \text { leaks } . \tag{3}
\end{equation*}
$$

Neutrons injected into the system constitute the source of the reaction, absorption cans be important, the escape/leaks represents the neutrons leaving the system.
The multiplicative factor in a medium without escape/leaks, noted by $k_{\infty}$ is defined by:

$$
\begin{equation*}
k_{\infty}:=\frac{\text { neutron account at time } t}{\text { neutron account at time } t-1} . \tag{4}
\end{equation*}
$$

In a bounded domain, The effective multiplicity factor $k_{e f f}$ is as follows:

$$
\begin{equation*}
k_{e f f}=\frac{k_{\infty}}{1+\text { leaks }} . \tag{5}
\end{equation*}
$$

This coefficient should be very close to unity in order to guarantee a state of equilibrium. Thus, if $k_{e f f}<1$ the state of the core is said subcritical, else if $k_{e f f}>1$ the state is supercritical and if $k_{e f f}=1$ the state is critical.
In order to determine such a state, it is essential to evaluate the stationary state without external source i.e. the unique source considered is fission.
The stationary Boltzmann equation (6) admits solutions only in the critical case. In this particular framework, we consider the fission source term extracted from Equation (1):

$$
\chi^{(g)} \Sigma_{m=1}^{\hat{g}} v^{\left(g^{\prime}\right)} \sigma_{f}^{\left(g^{\prime}\right)} \phi^{\left(g^{\prime}\right)}(\vec{r}, t),
$$

and absorption term (diffusion and scattering) also from Equation (1):

$$
-\operatorname{div}\left(D(\vec{r}, t) \nabla \phi^{(g)}\right)(\vec{r}, t)+\sigma_{t}^{(g)}(x) \phi^{(g)}(\vec{r}, t)-\sum_{g^{\prime}=1}^{\hat{g}} \sigma_{s}^{\left(g \rightarrow g^{\prime}\right)} \phi^{\left(g^{\prime}\right)}(\vec{r}, t) .
$$

The input data of modeling do not necessary characterizing a critical state and if necessary, we force our system becoming critical by adding the $k_{\text {eff }}$ factor in the stationary version of Equation (1). This equation becomes then:

$$
\begin{gather*}
-\operatorname{div}\left(D(\vec{r}, t) \nabla \phi^{(g)}\right)(\vec{r}, t)+\sigma_{t}^{(g)}(x) \phi^{(g)}(\vec{r}, t)-\sum_{g^{\prime}=1}^{\hat{g}} \sigma_{s}^{\left(g-g^{\prime}\right)} \phi^{\left(g^{\prime}\right)}(\vec{r}, t) \\
=  \tag{6}\\
\frac{1}{k_{e f f}} \chi^{(g)} \sum_{g^{\prime}=1}^{\hat{g}} v^{\left(g^{\prime}\right)} \sigma_{f}^{\left(g^{\prime}\right)} \phi^{\left(g^{\prime}\right)}(\vec{r}, t) .
\end{gather*}
$$

One thus bring himself back to a generalized eigenvalues problem that consists to look for $k_{e f f}$ factor for which the critical outcome equation (3) holds.
We thus use the power method to compute the largest eigenvalues. The corresponding eigenvector will be the neutron flux distribution that is the solution of the stationary problem (6).

## 3. Numerical methods and time-parallel-solver

For the convenience of presentation, we gather variables of various groups of energy both on neutron flux and delayed neutrons all at ones sought a vectorial form within the continuous framework. Then, we proceed to the time-space discretization, in order to have finally a block matrix structure representing the linear system that solves the problem at each time step.
This procedure of unification facilitates the application of the parareal in time algorithm in the resolution of the neutron model. Indeed, in a practical point of view the vectorial variable form makes building of the time-propagator easier.

### 3.1. Space discretization

We assume that the neutronic flux i.e. $\phi$ modeled by Equations (1)-(2) belongs to $C^{0}\left([0, T] ; H_{0}^{1}(\Omega)\right)$ and refere to $[30,31,32,33]$ for more theoretical results. We denote by $H_{0}^{1}(\Omega)$ the Hilbert space defined as follows:

$$
\left.H_{0}^{1}(\Omega)\right):=\left\{u \in L^{2}(\Omega), \text { such that } \nabla u \in L^{2}(\Omega), u=0 \text { on } \partial \Omega\right\},
$$

and denote by $\langle.,$.$\rangle its scalar product.$
In order to discretize the problem, we use finite elements space approximation associated with the weak formulation of the flux and the concentration of precursors. We note that the flux belongs to $H:=H_{0}^{1}(\Omega)$ and the precursors belong to $U:=L^{2}(\Omega)$ (see [35]), so by meshing the domain $\Omega \subset \mathbb{R}^{3}$ with an uniform triangulation $\mathcal{T}_{h}(\Omega)$, one can defines finite dimensional Hilbert space $H_{h} \subset H$ (repectively $U_{h} \subset U$ ), with associated standard basis functions $\left(\psi_{i}\right)_{i=1}^{q}\left(\right.$ respectively $\left.\left(\phi_{i}\right)_{i=1}^{p}\right)$. In addition one has:

$$
\begin{aligned}
H_{h} & :=\left\{u_{h} \in C^{0}(\Omega), u_{h \mid \mathbf{K}} \in \mathbb{P}_{1}, \forall \mathbf{K} \in \mathcal{T}_{h}\right\}, \\
U_{h} & :=\left\{u_{h} \in C^{0}(\Omega), u_{h \mid \mathbf{K}} \in \mathbb{P}_{0}, \forall \mathbf{K} \in \mathcal{T}_{h}\right\} .
\end{aligned}
$$

The physical problem admits a dynamic geometry because energy control rods are moving as well as the reaction proceeds. As the cross-sections are constant in each domain, it is more judicious to attach their dependence in time to the movement of those rods. These coefficients belong to

$$
L^{\infty}(\Omega):=\{u: \Omega \rightarrow \mathbb{R}, \exists c>0,|u(\vec{r})| \leq c, \forall \vec{r} \in \Omega\},
$$

of which we consider an approximation $\mathbb{P}_{0}$ for the discrete forms.
For the deterministic model of the kinetics equation, the discretized cross sections are supposed to be constant on each elements of the mesh (called media).

In order to indicate the space-dependency on the cross-sections in their discretized notations, we use the notation with the hat symbol (i.e.: "^") over each vector of nodal coefficients, which is identified to it's continuous notation.

By convenience, a block matrix belongs to $\mathbb{R}^{(a b) \times(c d)}$ means that it has $a$ block rows and $c$ block columns of matrices belonging to $\mathbb{R}^{b d}$ i.e. $b$ rows and $d$ columns.

Let us consider $\dot{\boldsymbol{\Phi}}$ and $\dot{\mathbf{c}}$ the nodal representation of first derivative in time of the neutron flux and the delayed precursors concentration respectively. The discrete multi-
group unified equation governing the neutron flux is as follows:

$$
\begin{equation*}
\mathcal{M}_{h}^{v} \dot{\mathbf{\Phi}}(t)=\mathcal{A}_{h}^{d} \boldsymbol{\Phi}(t)-\boldsymbol{\Sigma}_{h} \boldsymbol{\Phi}(t)+\mathcal{R}_{h} \mathbf{c}(t), \tag{7}
\end{equation*}
$$

In Equation (7) $\mathcal{M}_{h}^{v}:=\operatorname{Blockdiag}\left(M_{h}^{v}\right) \in \mathbb{R}^{(\hat{g} \hat{q}) \times(\hat{g} \hat{q})}$, with $M_{h}^{v} \in \mathbb{R}^{\hat{q} \hat{q}}$, and $\mathcal{A}_{h}^{d}:=\operatorname{Blockdiag}\left(A_{h}^{d}\right) \in$ $\mathbb{R}^{(\hat{g} \hat{q}) \times(\hat{g} \hat{q})}$. The matrix $\boldsymbol{\Sigma}_{h} \in \mathbb{R}^{(\hat{g} \hat{q}) \times(\hat{g} \hat{q})}$, the matrix $\mathcal{R}_{h}:=\operatorname{Blockdiag}\left(R_{h}\right) \in \mathbb{R}^{(\hat{g} \hat{q}) \times(\hat{g} \hat{q})}$. Where $\left(M_{h}^{v}\right)_{i, j}:=V M_{h}$, on the other hand $\left(M_{h}\right)_{i j}:=\left\langle\psi_{i}, \psi_{j}\right\rangle,\left(A_{h}^{d}\right)_{i, j}:=\left\langle\bar{D}_{i} \nabla \psi_{i}, \nabla \psi_{j}\right\rangle$, where $\bar{D}_{i}$ is the nodal representation of $D$ (constant by element as $\left.\mathbb{P}_{0}\right),\left(R_{h}\right)_{i, j}:=\left\langle\left(\widehat{\chi_{d} \mu}\right)_{i} \psi_{i}, \psi_{j}\right\rangle$. The matrix $\boldsymbol{\Sigma}_{h}$ of the cross section are defined as :

$$
\left.\boldsymbol{\Sigma}_{h}:=\boldsymbol{\Sigma}_{h}^{t}-\boldsymbol{\Sigma}_{h}^{s}+(I d-\widehat{\beta \beta}]\right) \boldsymbol{\Sigma}_{h}^{f},
$$

where,

$$
\begin{aligned}
& \boldsymbol{\Sigma}_{h}^{t}:=\operatorname{Blockdiag}\left(\Sigma_{h}^{t}\right) \in \mathbb{R}^{(\hat{g} \hat{q}) \times(\hat{g} \hat{q})}, \\
& \boldsymbol{\Sigma}_{h}^{s}:=\operatorname{Blockdiag}\left(\Sigma_{h}^{s}\right) \in \mathbb{R}^{(\hat{q} \hat{q}) \times(\hat{g} \hat{q})}, \\
& \mathbf{\Sigma}_{h}^{f}:=\operatorname{Blockdiag}\left(\Sigma_{h}^{f}\right) \in \mathbb{R}^{(\hat{q} \hat{q}) \times(\hat{g} \hat{q})}, \\
& \widehat{[\beta]}:=\operatorname{Blockdiag}\left(\beta^{(i)} I_{\mathbb{R}^{\hat{g}} \hat{g}}\right) \in \mathbb{R}^{(\hat{q} \hat{q}) \times(\hat{g} \hat{q})},
\end{aligned}
$$

with

$$
\begin{aligned}
\left(\Sigma_{h}^{t}\right)_{i, j} & :=\left\langle\left(\widehat{\sigma}_{t}\right)_{i} \psi_{i}, \psi_{j}\right\rangle \in \mathbb{R}^{\hat{q} \hat{q}} \\
\left(\Sigma_{h}^{s}\right)_{i, j} & :=\left\langle\left(\widehat{\bar{\sigma}_{s}}\right)_{i} \psi_{i}, \psi_{j}\right\rangle \in \mathbb{R}^{\hat{q} \hat{q}} \\
\left(\Sigma_{h}^{f}\right)_{i, j} & :=\left\langle\left(\widehat{\chi_{p} v \sigma_{f}}\right)_{i} \psi_{i}, \psi_{j}\right\rangle \in \mathbb{R}^{\hat{q} \hat{q}}
\end{aligned}
$$

where the sections $\widehat{\sigma}_{t}, \widehat{\sigma_{s}}$ et $\widehat{\chi_{p} v \sigma_{f}}$ are the finite elements nodal representations with $\mathbb{P}_{0}$ approximation, in addition $\left(\widehat{\sigma}_{t}\right)_{i},\left(\widehat{\overline{\sigma_{s}}}\right)_{i}$ and $\left(\widehat{\chi_{p} v \sigma_{f}}\right)_{i}$ referred to the tetrahedral indexed

$$
\begin{equation*}
\left.\mathcal{M}_{h} \dot{\mathbf{c}}(t)=\mathcal{N}_{h} \mathbf{c}(t)+\widetilde{\beta}\right] \mathbf{\Sigma}_{h}^{f} \boldsymbol{\Phi}(t), \tag{8}
\end{equation*}
$$

where $\mathcal{M}_{h}^{c} \in \mathbb{R}^{(\hat{k} \hat{p}) \times(\hat{k} \hat{p})}, \mathcal{M}_{h}^{c}:=\operatorname{Diag}\left(M_{h}^{c}\right)$, with $\left(M_{h}^{c}\right)_{i j}:=\left\langle\phi_{i}, \phi_{j}\right\rangle$. The matrix $\mathcal{N}_{h}:=$ $\operatorname{Blockdiag}\left(N_{h}^{k}\right) \in \mathbb{R}^{(\hat{k} \hat{p}) \times(\hat{k} \hat{p})}$ with $\left(N_{h}^{k}\right)_{i, j}:=\left(\mu_{i}^{(k)} \phi_{i}, \phi_{j}\right)$, and $\widetilde{[\beta]} \in \mathbb{R}^{(\hat{k} \hat{p}) \times(\hat{k} \hat{p})}$, such that $\left.(\widetilde{\beta \beta}]\right)_{k, g}:=$ $\left(\bar{\beta}^{(g)}\right) \in \mathbb{R}^{\hat{p} \hat{p}}$ où $\left(\bar{\beta}^{(g)}\right)_{i, j}:=\beta^{(g, k)} I_{\mathbb{R}^{\hat{p}}}$.

In the case where we do not have initial condition for Equation (7), we can build it using some techniques that we develop hereafter.
Let us now describe, with more details, implied matrices in the problem. The matrices disappearance (absorption+leaks) are described as follows:

$$
\begin{equation*}
\widehat{\mathrm{A}}:=\mathcal{A}_{h}^{d}-\boldsymbol{\Sigma}_{h}^{t}+\boldsymbol{\Sigma}_{h}^{s}, \tag{9}
\end{equation*}
$$

and the matrix related to the fission source term $\widehat{\mathrm{F}}$. Thanks to those matrices, we can
compute the generalized eigenvalue $k_{\text {eff }}$ of the couple $(\widehat{\mathrm{A}}, \widehat{\mathrm{F}})$, which satisfies:

$$
\widehat{\mathrm{A}} \boldsymbol{\Phi}=\frac{1}{k_{e f f}} \widehat{\mathrm{~F}} \boldsymbol{\Phi},
$$

where $\boldsymbol{\Phi}$ is the eigen-flux associated with $k_{\text {eff }}$. The algorithm hereafter (power algorithm) gives us the eigenvalue of the largest modulus. The space norm $\|.\|_{2}$ is associated to the scalar product $\langle., \text {. }\rangle_{\mathbb{R}^{\hat{q}} \bar{q}}$.

```
Algorithm 1: Fission source problem using power Algorithm
    Input: Initial flow vector \(\boldsymbol{\Phi}^{\mathbf{0}}\), tolerance \(\tilde{\boldsymbol{\epsilon}}\);
    Initialize the source term \(\mathcal{S}^{0}=\widehat{\mathrm{F}} \boldsymbol{\Phi}^{0} ; l \leftarrow 0\);
    repeat
        1. Solve iteratively: \(\widehat{\mathrm{A}} \boldsymbol{\Phi}^{l+1}=\mathcal{S}^{l}\);
        2. Normalize \(\tilde{\boldsymbol{\Phi}}^{l+1}:=\boldsymbol{\Phi}^{l+1} /\left\|\boldsymbol{\Phi}^{l+1}\right\|_{2}\);
        3. Evaluate the relative error \(\operatorname{Err}:=\left\|\tilde{\boldsymbol{\Phi}}^{l+1}-\tilde{\boldsymbol{\Phi}}^{1}\right\|_{2} /\left\|\tilde{\boldsymbol{\Phi}}^{l+1}\right\|_{2}\);
        4. Update the source : \(\mathcal{S}^{l+1}=\widehat{\mathrm{F}} \tilde{\boldsymbol{\Phi}}^{l+1}\);
        \(l \leftarrow l+1 ;\)
    until \(\operatorname{Err} \leq \tilde{\epsilon}\);
    Result: \(\boldsymbol{\Phi}:=\boldsymbol{\Phi}^{l} \quad \& \quad k_{\text {eff }}:=\left\|\boldsymbol{\Phi}^{l}\right\|_{2} ;\)
```

Using Algorithm 1, we can build the initial condition $\Phi^{(g, \star)}$ of the neutronic flux (including all energy groups). Then using the formula

$$
c^{(k)}(\vec{r}, 0):=\frac{1}{\mu^{(k)}} \Sigma_{g^{\prime}=1}^{\hat{g}} \beta^{\left(k, g^{\prime}\right)} v \sigma_{f}^{\left(g^{\prime}\right)} \boldsymbol{\Phi}^{\left(g^{\prime}\right)}
$$

Therefore, we are able to build the initial condition related to the concentrations of precursors groups.

### 3.2. Time discretization

We denote by $\boldsymbol{\Phi}_{i+1}$ the approximation of $\boldsymbol{\Phi}\left(t_{i+1}\right)$ obtained with the time-discretization $\theta$-scheme.

$$
\begin{align*}
\mathcal{M}_{h}^{v} \boldsymbol{\Phi}_{i+1}-\mathcal{M}_{h}^{v} \boldsymbol{\Phi}_{i}= & \tau \theta\left(\mathcal{A}_{h}^{d} \boldsymbol{\Phi}_{i+1}-\boldsymbol{\Sigma}_{h} \boldsymbol{\Phi}_{i+1}+\mathcal{R}_{h} \mathbf{c}_{i+1}\right) \\
& +\tau(1-\theta)\left(\mathcal{A}_{h}^{d} \mathbf{\Phi}_{i}-\boldsymbol{\Sigma}_{h} \mathbf{\Phi}_{i}+\mathcal{R}_{h} \mathbf{c}_{i}\right), \tag{10}
\end{align*}
$$

and,

$$
\begin{equation*}
\mathcal{M}_{h}^{k} \mathbf{c}_{i+1}-\mathcal{M}_{h}^{k} \mathbf{c}_{i}=\tau \theta\left(\mathcal{N}_{h} \mathbf{c}_{i+1}+\widetilde{[\beta]} \mathbf{\Sigma}_{h}^{f} \boldsymbol{\Phi}_{i+1}\right)+\tau(1-\theta)\left(\mathcal{N}_{h} \mathbf{c}_{i}+\widetilde{[\beta]} \mathbf{\Sigma}_{h}^{f} \boldsymbol{\Phi}_{i}\right) \tag{11}
\end{equation*}
$$

where $\theta \in[0,1]$ and $\tau$ is the time step.
We denote by $X_{i}:=\left[\boldsymbol{\Phi}_{i}^{T}, \mathbf{c}_{i}^{T}\right]^{T}$ the vector that represents the solution at time $t_{i}$. The
previous iterative formulas (10)-(11) can be written in the following matrix form:

$$
\left[\begin{array}{c}
\boldsymbol{\Phi}_{i+1}  \tag{12}\\
\mathbf{c}_{i+1}
\end{array}\right]=\left[\begin{array}{ll}
D_{I I} & D_{I J} \\
D_{J I} & D_{J J}
\end{array}\right]^{-1}\left[\begin{array}{cc}
B_{I I} & B_{I J} \\
B_{J I} & B_{J J}
\end{array}\right]\left[\begin{array}{c}
\boldsymbol{\Phi}_{i} \\
\mathbf{c}_{i}
\end{array}\right],
$$

where

$$
\begin{array}{lr}
D_{I I}:=\mathcal{M}_{h}^{v}+\tau \theta\left(\boldsymbol{\Sigma}_{h}-\mathcal{A}_{h}^{d}\right), & B_{I I}:=\mathcal{M}_{h}^{v}+\tau(1-\theta)\left(\mathcal{A}_{h}^{d}-\boldsymbol{\Sigma}_{h}\right), \\
D_{I J}:=-\tau \theta \mathcal{R}_{h}, & B_{I J}:=\tau(1-\theta) \mathcal{R}_{h}, \\
D_{J I}:=-\tau \theta[\widetilde{\beta}] \mathbf{\Sigma}_{h}^{f}, & B_{J I}:=\tau(1-\theta)[\widetilde{\beta}] \mathbf{\Sigma}_{h}^{f}, \\
D_{J J}:=\mathcal{M}_{h}^{k}-\tau \theta \mathcal{N}_{h}, & B_{J J}:=\mathcal{M}_{h}^{k}+\tau(1-\theta) \mathcal{N}_{h},
\end{array}
$$

which we simply denote by:

$$
\begin{equation*}
X_{i+1}=E X_{i}, \tag{13}
\end{equation*}
$$

where the block matrix $E$ is :

$$
E:=\left[\begin{array}{cc}
D_{I I} & D_{I J} \\
D_{J I} & D_{J J}
\end{array}\right]^{-1}\left[\begin{array}{cc}
B_{I I} & B_{I J} \\
B_{J I} & B_{J J}
\end{array}\right]
$$

In what follows, we describe the construction of the sequential $3 D$ solver. Since control energy rods are moving as physical time progress, the space domain properties also vary with respect to the positions of rods and fuels in the nuclear reactor. Hence the matrix $E$ is time-dependent, sequential and requires an update/assembly at each time step. The construction of the serial solver can so be achieved with following three steps:

Step 1: localize the absorber rods according to the chronology of the corresponding rod group position, then calculate the cross sections.
Step 2 : construct and assemble the block matrices.
Step 3 : solve the system e.g. with the GMRES algorithm [34].
Let us describe now in more details some characteristics of our solver and the implementation we define in the context of an international standard Benchmarck [35] . In particular we will detail adaptation of cross-sections to the mesh generation and managing dynamical geometry. Let us now discuss two procedures: the first one uses an adaptation of the number of the layers (on meshes), on the "z" direction, according to chosen small times steps. The second one, what we investigate in this paper is based on the interpolation of the Heaviside functions according to the celerity of control rods. We use tetrahedron elements generated by the scientific computation software FreeFem $++[1]$ and we carry out the rods movements by interpolating the domain indicators, which characterizes the parts of the finite elements (tetrahedron) occupied by rods and/or fuel. These indicators are used to distinguish constant physical coefficients in each medium/domaine (see tables (3)-(4)) that are approximated by $\mathbb{P}_{0}$ finite elements.

In the setting of our solver, the construction of block matrices requires a construction of Heaviside functions, which identify the mediums, by the intuitive rule:

$$
\begin{equation*}
\text { Presence of rods } \Longrightarrow \text { Absence of fuel. } \tag{14}
\end{equation*}
$$

The $3 D$-mesh (Figure on the side) represents $\frac{1}{4}$ nuclear core (front view). The rest of the core is deduced by symetry.


Figure 1: $3 D$ mesh representation of $\frac{1}{4}$ of the nuclear core.

Heaviside function


Figure 2: Heaviside function indicates the rate of sinking of the rods (red) and of the Fuel (blue) on a tetrahedron.
with a an adaptive linear system such that:

$$
\begin{equation*}
X_{i+1}=E_{i} X_{i}, \tag{15}
\end{equation*}
$$

which motivate parallel computation to increase serial simulation cost in ordinary machine, and thus improve computational time.


Figure 3: Heaviside function for the rod and the fuel on one tetrahedron.

### 3.3. Acceleration using parareal in time Algorithm

In order to accelerate the resolution of Equation (13) we apply the parareal in time algorithm [5] which is based on prediction correction scheme. We present in what follows the basic tools for applaying such scheme. Let us divide the time interval $[0, T]$ into $\hat{n}$ small ones;

$$
\left[T_{n}, T_{n+1}\right]:=\cup_{i=0}^{i=l-1}\left[T_{n, i}, T_{n, i+1}\right]
$$

where $l=\frac{\Delta T}{\tau}$ with $\Delta T$ is the size of the small time interval we have $T_{n, 0}=T_{n}$ et $T_{n, l}=T_{n+1}$. Let $\left\{\mathbf{X}_{n}\right\}_{n \geq 0}$ be the sequence of solution of the system (13) at time $t=T_{0}, \ldots, T_{n}, \ldots, T_{\hat{n}}$, so that $\mathbf{X}_{n+1}:=X_{n, l}$.
In the case of sequential propagation, those elements are solutions of:

$$
\left[\begin{array}{ccccc}
\mathbf{I d} & & & &  \tag{16}\\
-\mathbf{F}_{\Delta} & \mathbf{I d} & & & \\
& \ddots & \ddots & & \\
& & & -\mathbf{F}_{\Delta} & \mathbf{I d}
\end{array}\right]\left[\begin{array}{c}
\mathbf{X}_{0} \\
\mathbf{X}_{1} \\
\vdots \\
\mathbf{X}_{\hat{n}-1}
\end{array}\right]=\left[\begin{array}{c}
\mathbf{X}_{0} \\
\mathbf{0} \\
\vdots \\
\mathbf{0}
\end{array}\right],
$$

where $\mathbf{F}_{\Delta}=\prod_{i=1}^{l} E_{i}$ could be obtained with a "for"-loop as a routine and no need to memorize such results of matrix-product.

As well known The parareal in time algorithm [5, 6] ensures an exact convergence towards the fine approximated solution after $\hat{n}$ iterations, where $\hat{n}$ is the number of subintervals of the decomposition. And more accurate results with a time shorter when applied on diffusive systems. The updating formula for the series $\left(\mathbf{X}_{n}^{p}\right)_{n \geq 0}^{p \geq 0}$ of initial conditions values is given as follows:

$$
\begin{equation*}
\mathbf{X}_{n+1}^{p+1}=\mathbf{G}_{\Delta} \mathbf{X}_{n}^{p+1}+\mathbf{F}_{\Delta} \mathbf{X}_{n}^{p}-\mathbf{G}_{\Delta} \mathbf{X}_{n}^{p} \tag{17}
\end{equation*}
$$

where $\mathbf{G}_{\Delta}$ is a coarse propagator taken with some reduction of the parallel propagator
$\mathbf{F}_{\Delta}$ that approximate the right solution with the $\theta$-scheme using fine time step $\tau_{F}$. This solvers reduction may produce some instability, one cans find in $[8,10,12]$ some stability results of such scheme. As the coarse part is solved sequentially, it is preferable to reduce model such that the coarse propagator $\mathbf{G}_{\Delta}$ is the least expensive as possible while remaining stable and always taking into account the physic problem. In what follows, we present a reduced neutron model that makes possible use of larger time step serial propagator.

## 4. Reduced neutron model and input experiments data

In practice, we consider some reduction of the model. This reduction will relate mainly to the dynamic aspect of rods in the nuclear reactor.

Let us describe the physics. At time $t=0 \mathrm{sec}$ the first group of rods has an initial position at $z=100 \mathrm{~cm}$, while the second group has a higher position at $z=180 \mathrm{~cm}$.

The linear movement of rods is as follows:
Group 1: $(t=0 \mid z=100 \mathrm{~cm}) \nearrow(t=26,5 \mid z=180 \mathrm{~cm})$, with velocity of $3 \mathrm{~cm} / \mathrm{sec}$.
Group 2: $(t=7,5 \mid z=180 \mathrm{~cm}) \searrow(t=47,5 \mid z=60 \mathrm{~cm})$, with velocity of $3 \mathrm{~cm} / \mathrm{sec}$.
We present now both dynamics and statics scenarios which definitions are strongly related to the rods movement, and especially at their levels of sinking in the nuclear reactor. We show that a simplification of the model can be an advantage to accelerate the resolution by the parareal in time algorithm.

We call complete-data model and reduced-data model the neutron model i.e. Equation (7) with and whithout precursors concentration groups respectively.

A thorough study of the reduction of the model is fundamental in order to allow a degradation of the fine solver and to determine the characteristics of a stable propagator. Degradation can be carried out either by some de-refinement of the mesh-grid (spacetime), or by a reduction of the number of the iterations of inversion (when using an iterative solver) or by increasing the tolerance at the time of the resolution of the linear system or by simplification of the physics of the model. In what follows, we basically use this last way to build a coarse propagator.

In the framework of neutron kinetics, several parameters allow a simplification of the model while keeping the simulation results rather close to those obtained with a complete model. For this purpose, one can for example decrease the number of the groups of energies for the neutron flux, or even the number of the groups of the concentrations of the precursors. We can also approach the productivity of the reactor by a simple change of the reactivity factor.
Hereafter, we detail the dynamic aspect of the physical model and we propose a reduction of its dynamics. This reduction enables us to make computation cheaper while keeping accurate results as we show it in the experimental part later on (see section 5.3).

### 4.1. Dynamic rods scenario

The kinetics in this framework is characterized by a constant velocity (i.e. of $3 \mathrm{~cm} / \mathrm{sec}$ ) of control rods. At the initial configuration, rods sinking level provides a critical state of the nuclear reactor (i.e. $k_{e f f}=1$ ). In this configuration rods of group-1 are at $z=100 \mathrm{~cm}$ and group- 2 are at $z=180 \mathrm{~cm}$. During time the nuclear reactor is gradually switched

```
Algorithm 2: Algorithme pararéel de la cinétique neutronique
    Input: \(\hat{n}:=\#\) slave proc, \(\tau_{F}, \tau_{G}\)
    Input: \(\mathbf{X}_{0}^{0}=\left[\left(\Phi^{\star}\right)^{T},\left(\mathbf{c}^{\star}\right)^{T}\right]^{T}\) as initial conditions, \(\boldsymbol{\epsilon}\) a tolerance of the algorithm ;
    Input: a solver \(\mathcal{A}\), a data vector \(\mathbf{X}\);
    Routine \((\mathcal{A}, \mathbf{X})\)
    1) Positioning the absorber rods with respect to the dynamic chronology;
    2) Constructing matrices related to equations (10)-(11)-(12);
    3) Serial propagation of \(\mathbf{X}\) using \(\mathcal{A}\) with respect to its rods' scenario (the result is
        denoted by \(\mathcal{A} \mathbf{X}\) );
    end Routine;
    \(p \longleftarrow 0\);
    repeat
        if master processor then
            foreach \(n \in\{0, . ., \hat{n}-1\}\) do
                            1) Call: Routine \(\left(\mathbf{G}_{\Delta}, \mathbf{X}_{n}^{p}\right)\) (i.e. coarse-serial propagation);
            2)
                if \(p=0\) then
                    repeat return to 1 with \(\mathbf{X}_{n+1}^{p}\) until \(n=\hat{n}-1\)
                    else
                            Construct \(\left(\mathbf{X}_{n}^{p}\right)_{n \geq 1}\) with respect to relationship (17);
                    end
                    3) Send ( \(\mathbf{X}_{n}^{p}\), processor \(\left.(n)\right)\);
            end
        else
            forall slave processor( \(n\) )/ \(n \in\{0, \ldots, \hat{n}-1\}\) do
                    Recv ( \(\mathbf{X}_{n}^{p}\), master processor);
            Call: Routine( \(\mathbf{F}_{\Delta}, \mathbf{X}_{n}^{p}\) ) (i.e. fine-parallel propagation);
            Send ( \(\mathbf{F}_{\Delta} \mathbf{X}_{n}^{p}\), processor \(\left.(n)\right)\);
            end
        end
        if master processor then
            foreach \(n \in\{0, \ldots, \hat{n}-1\}\) do
                    \(\operatorname{Recv}\left(\mathbf{F}_{\Delta} \mathbf{X}_{n}^{p}\right.\), processor \(\left.(n)\right)\);
                    Evaluate \(\epsilon_{n}^{p}=\left\|\mathbf{F}_{\Delta} \mathbf{X}_{n}^{p}-\mathbf{X}_{n+1}^{p}\right\|_{\left.\mathcal{L}_{\left(\mathbb{R}^{8} \delta q\right)}\right)} ;\)
        end
        end
        \(p \leftarrow p+1\);
        Broadcast (master processor, \(\epsilon_{n}^{p}\) );
    until \(\max _{n} \epsilon_{n}^{p} \leq \epsilon\);
```

on to produce neutrons by fission. This production is less important if the rods sinking level on the engine is important. By this way, one controls the fission productivity on a nuclear reactor.

### 4.2. Static rods scenario

In this scenario rods are fixed. Then we approach the last scenario by choosing the rods positions correctly.

In the framework of the dynamic scenario, there are two principal phases of the core: subcritical and supercritical. A reproduction of these phases (related to the states of the core) is completely possible when we can arbitrarily control the core neutrons production.

In what follows we propose two approaches that ensure those states with a fixed position of control rods.

Special rods positioning. In order to approximate the latest scenario, we choose only two levels for the two rods groups.

|  | $t \in[0,20]$ | $t \in] 20,80]$ |
| :---: | :---: | :---: |
| Group 1 | $\mathrm{z}=160 \mathrm{~cm}$ | $\mathrm{z}=180 \mathrm{~cm}$ |
| Group 2 | $\mathrm{z}=180 \mathrm{~cm}$ | $\mathrm{z}=60 \mathrm{~cm}$ |

Table 1: Control of the state of the nuclear core by a change of the positions of the rods.

External variation of reactivity. This approach does not change the initial configuration of the rods. Nevertheless we are able to reproduce the phenomenon ensured by rods positioning. Indeed, on a physical point of view: these positions correspond to a certain reactivity, it is thus sufficient to reproduce this reactivity (corresponding to a given factor of $k_{e f f}$ ) to reproduce the same scenario.

|  | $t \in[0,20]$ | $t \in] 20,80]$ |
| :---: | :---: | :---: |
| reactivity | $10008 . \mathrm{e}-5$ | $9998 . \mathrm{e}-5$ |

Table 2: Control of the state of the nuclear core by a change of its reactivity.

Remark 1. In the case of static scenario, it is easy and practice to implement the second method. Indeed, one does not require several updates of the coefficients, contrary to the first method. Moreover, the method 4.2 facilitates the development and accelerates the execution of the algorithm resulting without any change with respect to the results of the first approach 4.1.

Remark 2. With the static scenario it is difficult to reproduce the results of the real scenario. This is clear, because we already lose the continuity of the dynamics. In fact, it has a direct impact on energy resulting from this model. Nevertheless, with the static scenario of the second method 4.2, we can exploit the reactivity to reach the same peak as that of the real scenario. The static version 4.1 related to the rods position does not allow this possibility.

## 5. Numerical results

To validate the method and to test its performance in terms of numerical accuracy and computational effort, we consider the $3 D$ variation of the LMW reference test case, which can be characterized as an operation transient with large spatial variation involving a LWR type reactor [35]. This test case is a three dimensional quarter core benchmark with two energy groups and six precursor groups. Figure 4 show the geometry of the model reactor. The withdrawal of the rods of group-1 initiates a transient, which is terminated by the insertion of the rods group-2. The average power increases and reaches maximum values between 20 and 21 seconds, then decreases slightly under the influence of the insertion of the rod group-2, and crosses its initial value after 37 seconds.

The physical data are presented at Table 3 and supplemented by the Tables 4. The cross sections are presented in Table 3. Each Table presents the section in a specific medium. The neutron data relating to the section of scattering are presented in Tables 4.

For the implementation and the numerical experiments of the preceding methods, we used data of various cross sections given from: [35]. The numerical simulations were carried out on a parallel SGI shared-memories-machine, which has 64 processor given rhythm with $2.0 \mathrm{GHz}, 256$ Go of shared memory and a communication network Numalink ( $15 \mathrm{~GB} / \mathrm{s}$ ). We thus exploited this machine massively in parallel with the use of the MPI library and the scientific computation software FreeFem $++[1]$.

This paragraph presents the various sections. The neutron flux domain is the sum of the two mediums of combustible (group-A and group-B), control rods and reflectors. Two energy groups are used for the discretization in each medium. The corresponding

| Physical data | medium |  |  |  |
| :---: | :---: | :---: | :---: | :---: |
| Cross sections | fuel A |  | fuel B |  |
|  | group-1 | group-2 | group-1 | group-2 |
| $\sigma_{t}$ | 0.23409670 | 0.93552546 | 0.23381787 | 0.95082160 |
| $\sigma_{f}$ | 0.006477691 | 0.1127328 | 0.007503284 | 0.1378004 |
| rods |  | reflecteur |  |  |
| $\sigma_{t}$ | group-1 | group-2 | group-1 | group-2 |
| $\sigma_{f}$ | 0.23409670 | 0.93552546 | 0.20397003 | 1.26261670 |
| celerity | 0.006477691 | 0.1127328 | .0 | .0 |

Table 3: Neutron data for the core of benchmark 3D.
neutrons are in permanent interaction. These neutrons can change from an energy group to another by a deceleration (shocks) or by a simple change of direction (diffusion).

### 5.1. The sequential solver

Experiment are carried-out with various neutron models and shows that the explicit Euler scheme is unconditionally unstable, while the $\theta$-scheme related to a value of $0<$ $\theta<1$ presents some oscillations with model when precursors concentration groups are present. It appears that the single value for which the $\theta$-scheme is unconditionally stable and does not present oscillations is the Implicit Euler scheme i.e. $\theta=1$. We present in
fuel $\mathbf{A}$

| $\sigma_{s}^{\left(g-g^{\prime}\right)}$ | group-1 | group-2 |
| :---: | :---: | :---: |
| group-1 | .20613914 | .01755550 |
| group-2 | .0 | .84786329 |

fuel B

| $\sigma_{s}^{\left(g-g^{\prime}\right)}$ | group-1 | group-2 |
| :---: | :---: | :---: |
| groupe-1 | .20564756 | .01717768 |
| groupe-2 | .85156526 | .0 |

rods of control

| $\sigma_{s}^{\left(g \rightarrow g^{\prime}\right)}$ | groupe-1 | groupe-2 |
| :---: | :---: | :---: |
| groupe-1 | .20558914 | .01755550 |
| groupe-2 | .84406329 | .0 |


| reflector <br> $\sigma_{s}^{\left(g \rightarrow g^{\prime}\right)}$ <br> $n n$ <br> groupe-1 | groupe-2 |  |
| :---: | :---: | :---: |
| groupe-1 | .17371253 | .02759693 |
| groupe-2 | 1.21325319 | .0 |

Table 4: "scattering" cross section data.


Figure 4: Initial configuration of rods as well as fuel volumes inside nuclear core .
what follows some graphs showing the behavior of the standard of the neutron flux in the nuclear reactor.

First of all, we simulate a complete neutron model, with six groups of precursors and two groups of neutron energy of flux. We eliminate then precursors group and consider the impact of their presence on the behavior of the reactor power.
5.2. Parareal in time algorithm behavior with respect to the standard of the average flux on nuclear core
In this subsection, we present a series of curves (see Figures 5 and 6) describing the state of the standard of the neutron flux in the engine in real time of the reaction.


Figure 5: Parareal in time algorithm behavior applied on neutron model (1-4 iteration). Complet model with dynamic rods scenario $\tau_{F}=10^{-1}, \tau_{G}=4$ et $\hat{n}=10$.

### 5.3. Numerical study of the convergence of the parareal in time algorithm

In this subsection we show the numerical convergence of the solutions calculated with the parareal in time algorithm to the sequential solution (see Tables 5-6-7-8) of which we vary the physics we treat. The error considered in these curves is the maximum one among $\epsilon_{n}^{p}$, where :

$$
\epsilon_{n}^{p}:=\frac{\left\|\mathbf{X}_{n}^{p}-X_{n}\right\|_{L^{2}\left(\mathbb{R}^{8}(q+p)\right.}}{\left\|X_{n}\right\|_{L^{2}\left(\mathbb{R}^{8}(q+p)\right.}^{17}} .
$$



Figure 6: Parareal in time algorithm behavior applied on neutron model (1-4 iteration). Complet model with static rods scenario. $\tau_{F}=10^{0}, \tau_{G}=4$ et $\hat{n}=10$.

We first present curves relating to the dynamic scenario with the complete model. In the second time, we present error curves obtained with the static scenario of a reduced-data model.
Taking into account the accuracy of the time-discretization scheme against referred to solution (reproduced with a very small time step e.g. $10^{-2}$ ). We were thus interested in errors of the same order as this one. It is thus necessary to consider a convergence of the parareal in time algorithm to an error of order $10^{-2}$ or of $10^{-3}$. Nevertheless, we have more thorough results of convergence (see Figure 7-10-11) to show convergence towards the sequential numerical solution.

Remark 3. We note that the error curves in Figure 7 (and in Figure 10 respectively) correspond to the experiment results of the behavior of the flux norm presented in Figure 5 (respectively in Figure 6)

### 5.4. Reduced coarse solver using static rods scenario

We present in this sub-section a degradation of the coarse model allowing an acceleration of the resolution without harming the convergence of the algorithm. This degradation is carried out for the coarse solver at which we use static scenario. This


Figure 7: Convergence of the algorithm (iteration 00-09), complete model with dynamic rods scenario for both serial and parallel computation; $\tau_{F}=10^{-1}, \tau_{G}=4$ et $\hat{n}=10$.
scenario as explained before uses a variation of the reactivity of the system on two intervals of physical time of the reaction. The employed procedure enables us to accelerate calculation by no computing the matrix of production at all time steps. Indeed, only one multiplication per a real coefficient of the matrix (already in memory) is enough.

To have an order of idea, the sequential simulation (execution with only one processor) lasts 03h:06mn:57s.

We present in Figure 11 the acceleration of the resolution according to the number of subdivisions by temporal sub steps. The resolution on each sub steps is entrusted to a processor of the machine massively parallel. As we are interested in the errors of convergence about $10^{-2}$ or about $10^{-3}$ we note a saving of parallel time CPU of treatment of the neutron kinetics compared to a sequential calculation. Moreover, the fact of passing from 8 processors to 16 processors makes it possible to pass from $2 h: 45 \mathrm{mn}$ with $1 h: 15 \mathrm{mn}$ of treatment. We notice a division by two (approximate) of the time CPU, which is in adequacy with a local effectiveness of a machine with 8 proc the USSR by processor.
Notes that if the number of the subdivisions $\hat{n}$ increases then the number of the processors agents in parallel also increases. The communication thus becomes increasingly important. This explains the fact that the curve of error in time wallclock (CPU) relating to the use of 40 processors is worse than that for 20 processors. The machine on which we submitted these numerical tests thus reached its saturation of speedup.

Table 5: Iteration $1 \& 2$ of the algorithm, complet model with dynamic rods scenario.

| $\tau_{F}$ | $\tau_{G}$ | $\max _{n \geq 0} \epsilon_{n}^{1}$ |
| :---: | :---: | :---: |
| 0.01 | 2 | $1.42 \mathrm{e}-02$ |
| 0.01 | 4 | $2.93 \mathrm{e}-02$ |
| 0.01 | 8 | $6.09 \mathrm{e}-02$ |
| 0.1 | 0.5 | $0.29 \mathrm{e}-02$ |
| 0.1 | 1 | $0.65 \mathrm{e}-02$ |
| 0.1 | 2 | $1.35 \mathrm{e}-02$ |
| 0.1 | 4 | $2.88 \mathrm{e}-02$ |
| 0.1 | 8 | $6.04 \mathrm{e}-02$ |
| 0.5 | 2 | $1.13 \mathrm{e}-02$ |
| 0.5 | 4 | $2.68 \mathrm{e}-02$ |
| 0.5 | 8 | $5.83 \mathrm{e}-02$ |
| 1 | 2 | $0.75 \mathrm{e}-02$ |
| 1 | 4 | $2.29 \mathrm{e}-02$ |
| 1 | 8 | $5.43 \mathrm{e}-02$ |


| $\tau_{F}$ | $\tau_{G}$ | $\max _{n \geq 0} \epsilon_{n}^{2}$ |
| :---: | :---: | :---: |
| 0.01 | 2 | $3.04 \mathrm{e}-04$ |
| 0.01 | 4 | $1.16 \mathrm{e}-03$ |
| 0.01 | 8 | $4.11 \mathrm{e}-03$ |
| 0.1 | 0.5 | $1.31 \mathrm{e}-05$ |
| 0.1 | 1 | $6.48 \mathrm{e}-05$ |
| 0.1 | 2 | $0.28 \mathrm{e}-04$ |
| 0.1 | 4 | $1.11-03$ |
| 0.1 | 8 | $4.02 \mathrm{e}-03$ |
| 0.5 | 2 | $1.71 \mathrm{e}-04$ |
| 0.5 | 4 | $0.88 \mathrm{e}-03$ |
| 0.5 | 8 | $3.55 \mathrm{e}-03$ |
| 1 | 2 | $7.42 \mathrm{e}-05$ |
| 1 | 4 | $0.63 \mathrm{e}-04$ |
| 1 | 8 | $3.01 \mathrm{e}-03$ |

Table 6: Iteration $3 \& 4$ of the algorithm, complet model with dynamic rods scenario.

| $\tau_{F}$ | $\tau_{G}$ | $\max _{n \geq 0} \epsilon_{n}^{3}$ |
| :---: | :---: | :---: |
| 0.01 | 2 | $3.04 \mathrm{e}-04$ |
| 0.01 | 4 | $1.16 \mathrm{e}-03$ |
| 0.01 | 8 | $4.11 \mathrm{e}-03$ |
| 0.1 | 0.5 | $1.31 \mathrm{e}-05$ |
| 0.1 | 1 | $6.48 \mathrm{e}-05$ |
| 0.1 | 2 | $2.81 \mathrm{e}-04$ |
| 0.1 | 4 | $1.11 \mathrm{e}-03$ |
| 0.1 | 8 | $4.02 \mathrm{e}-03$ |
| 0.5 | 2 | $1.71 \mathrm{e}-04$ |
| 0.5 | 4 | $8.80 \mathrm{e}-04$ |
| 0.5 | 8 | $3.55 \mathrm{e}-03$ |
| 1 | 2 | $7.42 \mathrm{e}-05$ |
| 1 | 4 | $6.30 \mathrm{e}-04$ |
| 1 | 8 | $3.01 \mathrm{e}-03$ |


| $\tau_{F}$ | $\tau_{G}$ | $\max _{n \geq 0} \epsilon_{n}^{4}$ |
| :---: | :---: | :---: |
| 0.01 | 2 | $1.73 \mathrm{e}-05$ |
| 0.01 | 4 | $1.22 \mathrm{e}-04$ |
| 0.01 | 8 | $7.85 \mathrm{e}-04$ |
| 0.1 | 0.5 | $1.41 \mathrm{e}-07$ |
| 0.1 | 1 | $1.67-06$ |
| 0.1 | 2 | $1.51 \mathrm{e}-05$ |
| 0.1 | 4 | $1.13 \mathrm{e}-04$ |
| 0.1 | 8 | $7.53 \mathrm{e}-04$ |
| 0.5 | 2 | $7.02 \mathrm{e}-06$ |
| 0.5 | 4 | $7.80 \mathrm{e}-05$ |
| 0.5 | 8 | $6.17 \mathrm{e}-04$ |
| 1 | 2 | $1.94 \mathrm{e}-06$ |
| 1 | 4 | $4.64 \mathrm{e}-05$ |
| 1 | 8 | $4.77 \mathrm{e}-04$ |

Table 7: Iteration $1 \& 2$ of the algorithm, reduced model with dynamic rods scenario.

| $\tau_{F}$ | $\tau_{G}$ | $\max _{n \geq 0} \epsilon_{n}^{1}$ |
| :---: | :---: | :---: |
| 0.1 | 2 | $1.39 \mathrm{e}-02$ |
| 0.1 | 4 | $2.78 \mathrm{e}-02$ |
| 0.1 | 8 | $5.34 \mathrm{e}-02$ |
| 0.5 | 2 | $1.09 \mathrm{e}-02$ |
| 0.5 | 4 | $2.48 \mathrm{e}-02$ |
| 0.5 | 8 | $5.03 \mathrm{e}-02$ |
| 1 | 2 | $7.22 \mathrm{e}-03$ |
| 1 | 4 | $2.10 \mathrm{e}-02$ |
| 1 | 8 | $4.65 \mathrm{e}-02$ |


| $\tau_{F}$ | $\tau_{G}$ | $\max _{n \geq 0} \epsilon_{n}^{2}$ |
| :---: | :---: | :---: |
| 0.1 | 2 | $5.19 \mathrm{e}-05$ |
| 0.1 | 4 | $1.88 \mathrm{e}-04$ |
| 0.1 | 8 | $5.64 \mathrm{e}-04$ |
| 0.5 | 2 | $3.17 \mathrm{e}-05$ |
| 0.5 | 4 | $1.47 \mathrm{e}-04$ |
| 0.5 | 8 | $4.91 \mathrm{e}-04$ |
| 1 | 2 | $1.33 \mathrm{e}-05$ |
| 1 | 4 | $1.03 \mathrm{e}-04$ |
| 1 | 8 | $4.06 \mathrm{e}-04$ |



Figure 8: Norm of flux, behavior against (iteration 1-2-3-4) of the algorithm with complete model using static rods scenario for the coarse solver and dynamic rods scenario for the fine solver; $\tau_{F}=10^{-1}, \tau_{G}=2$ and $\hat{n}=10$.

Remark 4. The parareal in time algorithm is implemented with a master-slave configuration for which we have two types of communications: a distribution and a collection communication. In the distribution communication, the main processor sends information towards all its processors agents. On the other hand in the collection communication; the Master himself receives and collects information since his agents. In both cases, it is about the same quantity of informations which passes in the two directions. The second type of communication is devoted to the correction of the coarse error, which requires fine information to be communicated by agent processors toward master. Those communications are important and influence directly the period of information's passing. This is clear, in particular, if one increases the number of agent processors or the number of subintervals.

## 6. Conclusion

In this paper we present a time parallelization of the simulation of the neutron model, which is governed by the Boltzmann PDEs. The parallelization in time direction is


Figure 9: Convergence of the algorithm (iteration 00-09), complete model with dynamic rods scenario for the parallel computation and static rods scenario with reduced model for the coarse serial computation; $\tau_{F}=10^{-1}, \tau_{G}=2$ and $\hat{n}=10$.
achieved with parareal in time algorithm. We showed that the neutron model could be reduced, in number of concentration of delayed neutrons also in the dynamic of the geometry, in such way the coarse propagator of the parareal in time algorithm is lightweight and can predict solution i.e. spatial flux distribution with a higher speedup. Furthermore, in the numerical point of view we can consider space domain decomposition (e.g. multi grid) in order to more reduce the model. This direction was inversigated in the static case (see [2]) and is under study with more complicated neutron model. In the experiments point of view, we showed convergence rate and performance of the parallelization using supercomputer with MPI implementation of the parareal in time algorithm. The efficiency of the parallelization could be much more enhanced with some particular implementation using MPI non-blocking communications and managing agent processor's topology in a supercomputer.

Table 8: Iteration $3 \& 4$ of the algorithm, reduced model with dynamic rods scenario..

| $\tau_{F}$ | $\tau_{G}$ | $\max _{n \geq 0} \epsilon_{n}^{3}$ |
| :---: | :---: | :---: |
| 0.1 | 2 | $1.91 \mathrm{e}-07$ |
| 0.1 | 4 | $1.74 \mathrm{e}-06$ |
| 0.1 | 8 | $1.63 \mathrm{e}-05$ |
| 0.5 | 2 | $1.01 \mathrm{e}-07$ |
| 0.5 | 4 | $1.30 \mathrm{e}-06$ |
| 0.5 | 8 | $1.21 \mathrm{e}-05$ |
| 1 | 2 | $3.18 \mathrm{e}-08$ |
| 1 | 4 | $8.35 \mathrm{e}-07$ |
| 1 | 8 | $9.42 \mathrm{e}-06$ |


| $\tau_{F}$ | $\tau_{G}$ | $\max _{n \geq 0} \epsilon_{n}^{4}$ |
| :---: | :---: | :---: |
| 0.1 | 2 | $5.71 \mathrm{e}-09$ |
| 0.1 | 4 | $1.52 \mathrm{e}-07$ |
| 0.1 | 8 | $2.53 \mathrm{e}-06$ |
| 0.5 | 2 | $2.69 \mathrm{e}-09$ |
| 0.5 | 4 | $7.39 \mathrm{e}-08$ |
| 0.5 | 8 | $2.17 \mathrm{e}-06$ |
| 1 | 2 | $6.96 \mathrm{e}-10$ |
| 1 | 4 | $5.50 \mathrm{e}-08$ |
| 1 | 8 | $1.86 \mathrm{e}-06$ |



Figure 10: Convergence of the algorithm (iteration 00-09), complete model with dynamic rods scenario for the parallel computation and static rods scenario with reduced model for the coarse serial computation; $\tau_{F}=10^{-1}, \tau_{G}=4$ and $\hat{n}=10$.


Figure 11: Real time machine simulation CPU (Wallclock format: h:m). Relative errors between solution computed with the parareal in time algorithm and the serial one, we vary the number of used processor per subinterval; $\tau_{F}=10^{-1}, \tau_{G}=2$ and $\hat{n} \in\{1,4,8,10,16,20,40\}$.

## References

[1] O. Pironneau, F. Hecht, K. Ohtsuka, FreeFem++-mpi, third-edition:2012.
[2] P. Guérin, A.-M. Baudron, J.-J. Lautard, Domain decomposition methods for the neutron diffusion problem, Math. Comput. Simulation 80 (11) (2010) 2159-2167, ISSN 0378-4754, doi: 10.1016/j.matcom.2010.04.009, URL http://dx.doi.org/10.1016/j.matcom.2010.04.009.
[3] M. Dahmani, A. Baudron, J. Lautard, L. Erradi, A 3D nodal mixed dual method for nuclear reactor kinetics with improved quasistatic model and a semiimplicit scheme to solve the precursor equations, Annals of Nuclear Energy 28 (8) (2001) 805-824, ISSN 0306-4549, doi:10.1016/S0306-4549(00)00089-X, URL http://www.sciencedirect.com/science/article/pii/S030645490000089X.
[4] S. Chauvet, Multi-scale method for the resolution of the neutronic kinetics equations, Ph.D. thesis, URL oai:tel.archives-ouvertes.fr:tel-00348435, 2008.
[5] J.-L. Lions, Y. Maday, G. Turinici, Resolution d'EDP par un schema en temps "parareel", C. R. Acad. Sci. Paris Ser. I Math. 332 (7) (2001) 661-668, ISSN 0764-4442, doi:10.1016/S0764-4442(00)01793-6, URL http://dx.doi.org/10.1016/S0764-4442(00)01793-6.
[6] Y. Maday, G. Turinici, The parareal in time iterative solver: a further direction to parallel implementation, in: Domain decomposition methods in science and engineering, vol. 40 of Lect. Notes Comput. Sci. Eng., Springer, Berlin, 441-448, doi:10.1007/3-540-26825-1_45, URL http://dx.doi.org/10.1007/3-540-26825-1_45, 2005.
[7] G. Bal, Y. Maday, A "parareal" time discretization for non-linear PDE's with application to the pricing of an American put, in: Recent developments in domain decomposition methods (Zürich, 2001), vol. 23 of Lect. Notes Comput. Sci. Eng., Springer, Berlin, 189-202, 2002.
[8] G. Bal, On the convergence and the stability of the parareal algorithm to solve partial differential equations, in: Domain decomposition methods in science and engineering, vol. 40 of Lect. Notes Comput. Sci. Eng., Springer, Berlin, 425-432, doi:10.1007/3-540-26825-1_43, URL http://dx.doi.org/10.1007/3-540-26825-1_43, 2005.
[9] M. J. Gander, E. Hairer, Nonlinear convergence analysis for the parareal algorithm, in: Domain decomposition methods in science and engineering XVII, vol. 60 of Lect. Notes Comput. Sci. Eng., Springer, Berlin, 45-56, doi:10.1007/978-3-540-75199-1_4, URL http://dx.doi.org/10.1007/978-3-540-75199-1_4, 2008.
[10] D. S. Daoud, Stability of the parareal time discretization for parabolic inverse problems, in: Domain decomposition methods in science and engineering XVI, vol. 55 of Lect. Notes Comput. Sci. Eng., Springer, Berlin, 275-282, doi:10.1007/978-3-540-34469-8_32, URL http://dx.doi.org/10.1007/978-3-540-34469-8_32, 2007.
[11] M. J. Gander, S. Vandewalle, Analysis of the parareal time-parallel time-integration method, SIAM J. Sci. Comput. 29 (2) (2007) 556-578 (electronic), ISSN 1064-8275, doi:10.1137/05064607X, URL http://dx.doi.org/10.1137/05064607X.
[12] G. A. Staff, E. M. Rønquist, Stability of the parareal algorithm, in: Domain decomposition methods in science and engineering, vol. 40 of Lect. Notes Comput. Sci. Eng., Springer, Berlin, 449-456, doi: 10.1007/3-540-26825-1_46, URL http://dx.doi.org/10.1007/3-540-26825-1_46, 2005.
[13] M. Sarkis, C. E. Schaerer, T. Mathew, Block diagonal parareal preconditioner for parabolic optimal control problems, in: Domain decomposition methods in science and engineering XVII, vol. 60 of Lect. Notes Comput. Sci. Eng., Springer, Berlin, 409-416, doi:10.1007/978-3-540-75199-1_52, URL http://dx.doi.org/10.1007/978-3-540-75199-1_52, 2008.
[14] C. Farhat, J. Cortial, C. Dastillung, H. Bavestrello, Time-parallel implicit integrators for the near-real-time prediction of linear structural dynamic responses, Internat. J. Numer. Methods Engrg. 67 (5) (2006) 697-724, ISSN 0029-5981, doi:10.1002/nme.1653, URL http://dx.doi.org/10.1002/nme. 1653.
[15] G. Bal, Q. Wu, Symplectic parareal, in: Domain decomposition methods in science and engineering XVII, vol. 60 of Lect. Notes Comput. Sci. Eng., Springer, Berlin, 401-408, doi:10.1007/978-3-540-75199-1_51, URL http://dx.doi.org/10.1007/978-3-540-75199-1_51, 2008.
[16] S. Ulbrich, Generalized SQP methods with "parareal" time-domain decomposition for timedependent PDE-constrained optimization, in: Real-time PDE-constrained optimization, vol. 3 of Comput. Sci. Eng., SIAM, Philadelphia, PA, 145-168, 2007.
[17] M. J. Gander, S. Vandewalle, On the superlinear and linear convergence of the parareal algorithm, in: Domain decomposition methods in science and engineering XVI, vol. 55 of Lect. Notes Comput. Sci. Eng., Springer, Berlin, 291-298, doi:10.1007/978-3-540-34469-8_34, URL http://dx.doi.org/10.1007/978-3-540-34469-8_34, 2007.
[18] L. He, The reduced basis technique as a coarse solver for parareal in time simulations, J. Comput. Math. 28 (5) (2010) 676-692, ISSN 0254-9409.
[19] M. Gander, M. Petcu, Analysis of a Krylov subspace enhanced parareal algorithm for linear problems, in: Paris-Sud Working Group on Modelling and Scientific Computing 2007-2008, vol. 25 of ESAIM Proc., EDP Sci., Les Ulis, 114-129, 2008.
[20] M. J. Gander, Analysis of the parareal algorithm applied to hyperbolic problems using characteristics, Bol. Soc. Esp. Mat. Apl. SëMA (42) (2008) 21-35, ISSN 1575-9822.
[21] Y. Maday, Parareal in time algorithm for kinetic systems based on model reduction, in: Highdimensional partial differential equations in science and engineering, vol. 41 of CRM Proc. Lecture Notes, Amer. Math. Soc., Providence, RI, 183-194, 2007.
[22] D. Guibert, D. Tromeur-Dervout, Adaptive parareal for systems of ODEs, in: Domain decomposition methods in science and engineering XVI, vol. 55 of Lect. Notes Comput. Sci. Eng., Springer, Berlin, 587-594, doi:10.1007/978-3-540-34469-8_73, URL http://dx.doi.org/10.1007/978-3-540-34469-8_73, 2007.
[23] P. F. Fischer, F. Hecht, Y. Maday, A parareal in time semi-implicit approximation of the NavierStokes equations, in: Domain decomposition methods in science and engineering, vol. 40 of Lect. Notes Comput. Sci. Eng., Springer, Berlin, 433-440, doi:10.1007/3-540-26825-1_44, URL http://dx.doi.org/10.1007/3-540-26825-1_44, 2005.
[24] M. L. Minion, A hybrid parareal spectral deferred corrections method, Commun. Appl. Math. Comput. Sci. 5 (2) (2010) 265-301.
[25] L. Baffico, S. Bernard, Y. Maday, G. Turinici, G. Zérah, Parallel-in-time moleculardynamics simulations, Phys. Rev. E 66 (2002) 057701, doi:10.1103/PhysRevE.66.057701, URL http://link.aps.org/doi/10.1103/PhysRevE.66.057701.
[26] Y. Maday, G. Rim, Couplage de l'algorithme pararéel avec quelques méthodes de décomposition de domaine, Ph.D. thesis, Paris, 2011.
[27] Y. Maday, G. Turinici, A parareal in time procedure for the control of partial differential equations, C. R. Math. Acad. Sci. Paris 335 (4) (2002) 387-392, ISSN 1631-073X, doi:10.1016/S1631-073X(02)02467-6, URL http://dx.doi.org/10.1016/S1631-073X (02) 02467-6.
[28] Y. Maday, J. Salomon, G. Turinici, Monotonic parareal control for quantum systems, SIAM J. Numer. Anal. 45 (6) (2007) 2468-2482 (electronic), ISSN 0036-1429, doi:10.1137/050647086, URL http://dx.doi.org/10.1137/050647086.
[29] T. P. Mathew, M. Sarkis, C. E. Schaerer, Analysis of block parareal preconditioners for parabolic optimal control problems, SIAM J. Sci. Comput. 32 (3) (2010) 1180-1200, ISSN 1064-8275, doi: 10.1137/080717481, URL http://dx.doi.org/10.1137/080717481.
[30] R. Dautray, J.-L. Lions, Analyse mathématique et calcul numérique pour les sciences et les techniques. Tome 1, Collection du Commissariat à l'Énergie Atomique: Série Scientifique. [Collection of the Atomic Energy Commission: Science Series], Masson, Paris, ISBN 2-225-80406-9, with the collaboration of Michel Artola, Marc Authier, Philippe Bénilan, Michel Cessenat, Jean-Michel Combes, André Gervat, Hélène Lanchon, Bertrand Mercier, Claude Wild and Claude Zuily, 1984.
[31] R. Dautray, J.-L. Lions, Analyse mathématique et calcul numérique pour les sciences et les techniques. Vol. 9, INSTN: Collection Enseignement. [INSTN: Teaching Collection], Masson, Paris, ISBN 2-225-81303-5, Évolution: numérique, transport. [Evolution: numerical methods, transport], Reprint of the 1985 edition, 1988.
[32] F. Brezzi, M. Fortin, Mixed and hybrid finite element methods, vol. 15 of Springer Series in Computational Mathematics, Springer-Verlag, New York, ISBN 0-387-97582-9, 1991.
[33] J. E. Roberts, J.-M. Thomas, Mixed and hybrid methods, in: Handbook of numerical analysis, Vol. II, Handb. Numer. Anal., II, North-Holland, Amsterdam, 523-639, 1991.
[34] Y. Saad, M. H. Schultz, GMRES: a generalized minimal residual algorithm for solving nonsymmetric linear systems, SIAM J. Sci. Statist. Comput. 7 (3) (1986) 856-869, ISSN 0196-5204, doi: 10.1137/0907058, URL http://dx.doi.org/10.1137/0907058.
[35] W. W. Langenbuch S., Maurer W., Coarse-Mesh Flux-Expansion Method for the Analysis of SpaceTime Effects in Large Light Water Reactor Cores, Nucl. Sci. Eng. 63 (1977) 437-456.


[^0]:    A
    Email addresses: jean-jacques.lautard@cea.fr (J.-J. Lautard, A.-M. Baudron), anne-marie.baudron@cea.fr (J.-J. Lautard, A.-M. Baudron), maday@ann.jussieu.fr (Y. Maday), riahi@ann.jussieu.fr -/- riahi.mk@gmail.com (M.-K. Riahi), salomon@ceremade.dauphine.fr (J. Salomon)

    URL: http://www.ljll.math.upmc.fr/maday (Y. Maday), http://www.ljll.math.upmc.fr/riahi (M.-K. Riahi), http://www.ceremade.dauphine.fr/~salomon (J. Salomon)

