



8th AIMS International Conference on
Dynamical Systems, May 25 - 28, 2010,
Technical University of Dresden, Germany,
Lecture: Successive approximation methods
for solving time-dependent problems,

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Outline of the talk

- 0) Motivation
- 1) Successive Approximation method
- 2) Analysis
- 3) Numerical Results

Introduction

We present a novel method to solve time-dependent differential equations.

The method is based on iterative splitting schemes and generally known as successive approximation methods.

Benefits:

- Iterative scheme can help to obtain higher order scheme
- Idea: Freezing time-dependent case to a temporarily inhomogeneous equation (method of Tanabe and Sobolevski)

Drawback of standard methods, e.g. Exponential splitting, Magnus expansion schemes:

- Exponential character of the schemes and problem to achieve algorithms beyond 6th order of accuracy.
- Magnus expansion is based on Bernoulli-series, so the convergence radius is also restricted in that direction (finite intervall).
- Computations for solving commutator exponentials are expensive.

Iterative Splitting Method

Methods of iterative operator splitting methods:

- Time-decomposition methods, one can split operators with respect to their time-scale (Waveform relaxation schemes, Successive approximation methods)
- Spatial-decomposition methods, one can split operators with respect to their spatial scale (Schwarz-Waveform relaxation schemes, Domain decomposition methods)

Aims and contribution with the decomposition methods

- Decoupling time-scales and spatial-scales (Reduce the stiffness in single operators).
- Decoupling the multi-physics (Reduce the unphysical behaviour with best choice of discretization and solver methods).
- Time-adaptivity, Space-adaptivity (Efficiency and local accurate computations).
- Parallelization in Time and Space (Speed up of computations)

Results : More efficient and fast algorithms with high accuracy and simple implementation.

Iterative Operator Splitting Methods (classical version)

$$\frac{\partial c_i(t)}{\partial t} = A c_i(t) + B c_{i-1}(t), \text{ with } c_i(t^n) = u^n, \quad (1)$$

$$\frac{\partial c_{i+1}(t)}{\partial t} = A c_i(t) + B c_{i+1}(t), \text{ with } c_{i+1}(t^n) = u^n, \quad (2)$$

where $c_0(t)$ is any fixed function for each iteration. (Here, as before, u^n denotes the known split approximation at the time level $t = t^n$.) The split approximation at the time-level $t = t^{n+1}$ is defined as $c_{\text{sp}}^{n+1} = c_{2m+1}(t^{n+1})$. (Clearly, functions $c_k(t)$ ($k = i - 1, i, i + 1$) depend on the interval $[t^n, t^{n+1}]$, too, but, for the sake of simplicity, in our notation we omit the dependence on n .)

Approximation Methods to timedependent operators

We deal with the following timedependent linear evolution equations

$$\partial_t u = A(t)u, \quad u(0) = u_0, \quad (3)$$

where A can be an unbounded and time-dependent operator. For solving Hamiltonian problems, it is often the case that $A(t) = T + V(t)$, where only the potential operator $V(t)$ is time-dependent.

The solution is given as

$$u(t) = \exp(\Omega(t))u(0) . \quad (4)$$

This can be expressed as:

$$u(t) = \mathcal{T} \left(\exp\left(\int_0^t A(s) ds\right) u(0) \right) , \quad (5)$$

where the time-ordering operator \mathcal{T} .

Magnus expansion to timedependent operators

The Magnus expansion is defined as:

$$\Omega(t) = \sum_{n=1}^{\infty} \Omega_n(t), \quad (6)$$

where the first few terms are [Blanes and Casas 2008]

$$\begin{aligned} \Omega_1(t) &= \int_0^t dt_1 A_1, & \Omega_2(t) &= \frac{1}{2} \int_0^t dt_1 \int_0^{t_1} dt_2 [A_1, A_2] \\ \Omega_3(t) &= \frac{1}{6} \int_0^t dt_1 \int_0^{t_1} dt_2 \int_0^{t_2} dt_3 ([A_1, [A_2, A_3]] + [[A_1, A_2], A_3]) \\ &\dots\dots & \text{etc.} & \end{aligned} \quad (7)$$

where $A_n = A(t_n)$.

Successive Approximation Method to timedependent operators

Instead of the Magnus series (6) for solving explicit time-dependent problems, one can also directly implement successive approximation method.

The problem is given as:

$$\frac{\partial u}{\partial t} = A(t)u(t), \quad a \leq t \leq b \quad (8)$$

We rewrite:

$$\frac{\partial u}{\partial t} = A(a)u(t) + (A(t) - A(a))u(t) \quad (9)$$

The abstract integral is given as, by the so called Duhamel Principle:

$$u(t) = \exp((t - a)A(a))u_0 + \int_a^t \exp((t - s)A(a))(A(s) - A(a))u(s) ds \quad (10)$$

and $a \in [0, t]$ is a fixed time-point.

In order to construct $Y(t)$ we use Tanabe-Solevski's method with successive approximation we obtain:

$$u_1(t) = \exp((t - a)A(a))u_0, \quad (11)$$

...

$$u_{n+1}(t) = \exp((t - a)A(a))u_0 + \int_a^t \exp((t - s)A(a))(A(s) - A(a))u_n(s) ds \quad (12)$$

and formally we have:

$$u(t) = \exp((t - a)A(a))u_0 + \int_a^t \exp((t - s)A(a))R(s, a)u_0 ds \quad (13)$$

The recursive operators are given as:

$$R(t, s) = \sum_{m=1}^{\infty} R_m(t, s), \quad (14)$$

$$R_1(t, s) = \begin{cases} (A(t) - A(s)) \exp((t - s)A(a)) ds & , s < t \\ 0 & , s \geq t \end{cases} \quad (15)$$

$$R_m(t, s) = \int_s^t R_1(t, \sigma) R_{m-1}(\sigma, t) d\sigma. \quad (16)$$

Analysis

We consider the following time dependent problem,

$$\frac{\partial u}{\partial t} = Au(t) + B(t)u(t), \quad u(0) = u_0 \quad (17)$$

Our intension is to solve this problem by iterative scheme explained as follows, since this might balance the dominant terms in the equations (9),

$$\frac{\partial u_i}{\partial t} = Au_i + B(t)u_{i-1}, \quad i = 1, 2, \dots, m, \quad (18)$$

$$\frac{\partial u_{j+1}}{\partial t} = Au_j + B(t)u_{j+1}, \quad j = m, m+1, \dots, M, \quad (19)$$

where $m < M$ are integers. We apply only the one-side iteration to operator A .

The exact solutions of this system of equation then can be written by using the integration constant formula as follows:

$$\begin{aligned}\frac{\partial u_i}{\partial t} &= \exp(At)u_0 + \int_0^t \exp((t-s)A)B(s)u_{i-1}(s) ds \\ \frac{\partial u_{i+1}}{\partial t} &= \phi(t)u_0 + \int_0^t \phi(t)\phi^{-1}(s)Au_i(s) ds\end{aligned}$$

where $\phi(t)$ is the fundamental set of solution of the second equation in the system (18).

The function $\phi(t)$ can be written in terms of the Magnus series, we then have

$$\phi(t) = \exp(\Omega(t)) \quad (20)$$

$$= \exp\left(\sum_{j=1}^{\infty} \Omega_j(t)\right) \quad (21)$$

where

$$\begin{aligned} \Omega_1(t) &= \int_0^t dt_1 B_1, & \Omega_2(t) &= \frac{1}{2} \int_0^t dt_1 \int_0^{t_1} dt_2 [B_1, B_2] \\ \Omega_3(t) &= \frac{1}{6} \int_0^t dt_1 \int_0^{t_1} dt_2 \int_0^{t_2} dt_3 ([B_1, [B_2, B_3]] + [[B_1, B_2], B_3]) \\ &\dots\dots & \text{etc.} & \end{aligned} \quad (22)$$

where $B_j = B(t_j)$.

In the following Lemma is used to bound the perturbation:

Lemma

If A generates an analytical semigroup $\exp(At)$ and $B(t)$ is a bounded function. We have $\text{dom}(A) \subset \mathbf{X}$, $\text{dom}(B(t)) \subset \mathbf{X}$ and $\text{dom}(A) = \text{dom}(B) = \text{dom}(A + B)$, where \mathbf{X} is a Banach space with the operator norm $\|\cdot\|$.

With the assumption: $\|B(s + \tau)u_{i-1}(s + \tau) - B(s)u_{i-1}(s)\| \leq \gamma \|u_{i-1}(s + \tau) - u_{i-1}(s)\|$, $\forall s, s + \tau \in [0, t]$, and γ is a positive constant independent of t .

We obtain:

a.) $A + B(t)$

b.) and the perturbed semigroup $y(t) = \exp(At + \Omega(t))$ can be given as

$$u_i(t) = \exp(tA)u(0) + \int_0^t \exp((t-s)A)B(s)u_{i-1}(s)ds. \quad (23)$$

Proof.

The proof idea is perturbation in semigroups, see [Engel and Nagel 2000]. The operator B can be estimated with A -boundedness ($\|\cdot\|_A$ norm).



Time-dependent Case:

We deal with the perturbation theory [Engel, Nagel 2000]. The same proof methods are used for the time-dependent case. The variation of constants can be extended to a time-dependent case, see [Pazy 1983].

Theorem

Let us consider the abstract Cauchy problem in a Banach space \mathbf{X}

$$\begin{aligned}\partial_t c(t) &= Ac(t) + B(t)c(t), \quad 0 < t \leq T \text{ and } x \in \Omega, \\ c(0) &= c_0, \quad t \in [0, T],\end{aligned}\tag{24}$$

where $A, B(t) : D(A) \subset \mathbf{X} \rightarrow \mathbf{X}$, $B(t) : D(B(t)) \subset \mathbf{X} \rightarrow \mathbf{X}$ are given linear bounded operators which are generators of the C_0 -semigroup and $c_0 \in \mathbf{X}$ is a given element.

Further, we assume the estimations of the bounded time-dependent operator:

$$\|B(t) \exp(At)x\| \leq \beta \|x\|, \quad (25)$$

$$\tau_n = (t^{n+1} - t^n).$$

The error of the first time-step is of accuracy $\mathcal{O}(\tau_n^m)$, where $\tau_n = t^{n+1} - t^n$ and we have equidistant time-steps, with $n = 1, \dots, N$. Then the iteration process (18) for $i = 1, 2, \dots, m$ is consistent with the order of the consistency $\mathcal{O}(\tau_n^m)$.

Proof.

For $i = 1$, we have:

$$c_1(t^{n+1}) = \exp(A\tau_n)c(t^n), \quad (26)$$

and the solution is given as

$$c(t^{n+1}) = \exp(A\tau_n)c(t^n) + \int_{t^n}^{t^{n+1}} \exp(A(t^{n+1} - s))B(s) \exp(sA) + \int_0^s B(\tilde{s})d\tilde{s} c(t^n) ds,$$



We obtain:

$$\|e_1\| \leq \left\| \exp\left(A\tau + \int_{t^n}^{t^{n+1}} B(s) ds\right) c(t^n) - c_1 \right\| \quad (27)$$

$$\begin{aligned} &= \left\| \int_{t^n}^{t^{n+1}} \exp(A(t^{n+1} - s)) B(s) \exp(sA + \int_0^s B(\tilde{s}) d\tilde{s}) c(t^n) ds \right\| \\ &\leq C_\tau \|B(\xi)\| \|c(t^n)\| \end{aligned} \quad (28)$$

where $\xi \in [t^n, t^{n+1}]$ and the exp functions can be estimated by C .

The same argumentation is used for $i = 2$:

$$\|e_2\| \leq \tag{29}$$

$$= \int_{t^n}^{t^{n+1}} \|\exp(A(t-s))B(s)\| \tag{30}$$

$$\int_{t^n}^s \|\exp(A(s-\rho))B(\rho) \exp((\rho-t^n)(A+B))c(t^n) d\rho\| ds$$

$$= \tilde{C}_T^2 \|B(\xi)\|^2 \|c(t^n)\|$$

where $\xi \in [t^n, t^{n+1}]$

Based on the bounded operators we can apply the recursive argument.

For the iterations till $i = m$, we obtain for c_i and c :

By shifting $0 \rightarrow t^n$ and $\tau_n \rightarrow t^{n+1}$, we obtain our result:

$$\|e_i\| \leq \tilde{C} \|B(\xi)\|^m \tau_n^m \|c(t^n)\|,$$

where $\xi \in [t^n, t^{n+1}]$ and \tilde{C} is a non-timedependent constant.
The same proof idea can be applied to the even iterative scheme.

Algorithm

The algorithm is given as, for one time step, h , in the interval $[t_n, t_n + h]$,

$$u(t_n + h) = e^{hA_a} u(t_n) + \int_{t_n}^{t_n+h} e^{(t_n+h-s)A_a} (A(s) - A_a) u(s) ds \quad (31)$$

where $A_a = A(a)$ is $n \times n$ constant matrix.

Successive approximation steps then can be read as

$$u_1(t_n + h) = e^{hA_a} u(t_n),$$

$$u_2(t_n + h) = e^{hA_a} u(t_n) + \int_{t_n}^{t_n+h} e^{(t_n+h-s)A_a} (A(s) - A_a) u_1(s) ds$$

.....

$$u_k(t_n + h) = e^{hA_a} y(t_n) + \int_{t_n}^{t_n+h} e^{(t_n+h-s)A_a} (A(s) - A_a) u_{k-1}(s) ds.$$

After approximating the integrals in each iterations by quadrature formulas, we rewrite the solutions as

$$u_k(t_n + h) = e^{hA_a} u(t_n) + \sum_{j=1}^s w_j F(u_j^*), \quad k = 2, \dots, m \quad (32)$$

where $F(s) = e^{(t_n+h-s)A_a}$, w_j are weights and $u_j^* \in [t_n, t_n + h]$ are nodes.

We simply use the trapezoidal rule for approximating the integrals, we then have following iterative solving scheme,

$$\begin{aligned}
 u_k(t_n + h) = & e^{hA_a} \left(I + \frac{h}{2}(A(t_n) - A_a) \right) u(t_n) \\
 & + \frac{h}{2}(A(t_n + h) - A_a) u_{k-1}(t_n + h) \quad (33)
 \end{aligned}$$

for $k = 2, \dots, m$. Here $u(t_0) = u_0$ (initial condition), $u(t_n) = u_k(t_{n-1} + h)$, $n = 1, \dots, N$ and $N = \frac{b-a}{h}$. The algorithm will continue until the following condition is fulfilled,

$$|u_k - u_{k-1}| \leq Tol.$$

It can be easily seen in Equation (34), the scheme involves only one approximation of exponential of a constant matrix.

Experiments

Consider the following scalar equation,

$$u'(t) = 2u + tu, \quad u(0) = 1, \quad (34)$$

the exact solution is

$$u(t) = e^{-2} e^{\frac{(t+2)^2}{2}}. \quad (35)$$

The comparison of the numerical solution obtained by successive approximation and the exact solution of the scalar equation is shown in Figures (1) and (2) for different time intervals.

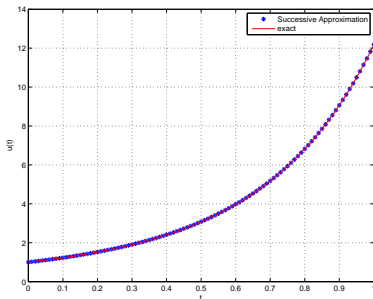


Figure: Comparison of approximate solution by successive approximation and exact solution of scalar equation for shorter time scale.

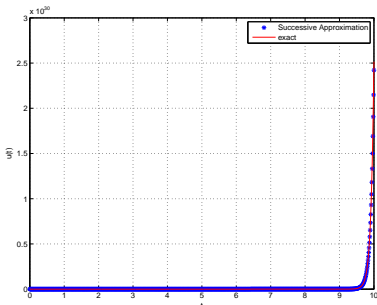


Figure: Comparison of approximate solution by successive approximation and exact solution of scalar equation for longer time scale.

Second example: Radial Schrödinger equation

We first consider the radial Schrödinger equation

$$\frac{\partial^2 u}{\partial r^2} = f(r, e)u(r) \quad (36)$$

where

$$f(r, E) = 2V(r) - 2E + \frac{l(l+1)}{r^2} \quad (37)$$

The equation (37) can be transformed as a harmonic oscillator with a time dependent spring constant after relabeling $r \rightarrow t$ and $u(r) \rightarrow q(t)$ and defining

$$k(t, E) = -f(t, E). \quad (38)$$

By redefining the variables as $u(t) = q(t)$ and $\dot{u}(t) = p(t)$, and $Y(t) = (q(t), p(t))$, the Equation (37) can be put into the system of equation as

$$\dot{Y}(t) = A(t)Y(t) \quad (39)$$

and Hamiltonian of the system is written by

$$H = \frac{1}{2}p^2 + \frac{1}{2}k(t, E)q^2. \quad (40)$$

For specific example, the ground state of hydrogen atom can be modeled as Schrodinger equation with the parameters $l = 0, E = -1/2, V(t) = -1/(t - a), a$ is arbitrary constant. Now the time dependent oscillator corresponds to

$$A(t) = \begin{pmatrix} 0 & 1 \\ f(t) & 0 \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} + \begin{pmatrix} 0 & 0 \\ f(t) & 0 \end{pmatrix} \equiv T + V(t),$$

with

$$f(t) = \left(1 - \frac{2}{t - a}\right). \quad (41)$$

The exact solution for this model with the initial conditions $q(0) = -a, p(0) = 1 + a, a = -0.001$ is

$$q(t) = (t - a)e^{-t}. \quad (42)$$

The comparison of exact and approximation of the hydrogen

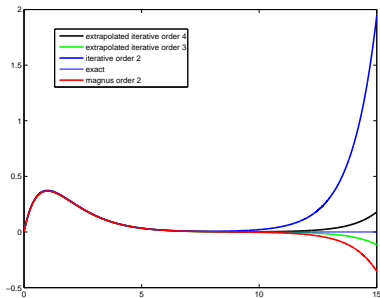


Figure: Comparison of exact and approximation of the hydrogen ground state wave function for various schemes in shorter time scale.

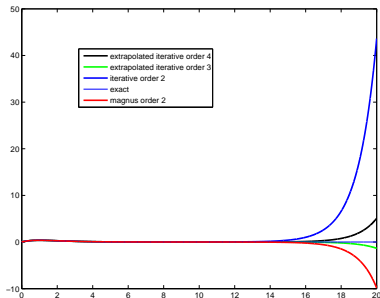


Figure: Comparison of exact and approximation of the hydrogen ground state wave function for various schemes in longer time scale.

Remark

The numerical results show that our higher order treatment gives a better long time behavior than the standard magnus expansion.

Future Works

Outview

- 1) Convergence Analysis.
- 2) Nonlinear Problems.
- 3) Application to real-life problems, e.g. Monte Carlo schemes.