Unconditionally stable scheme for Riccati equation

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Abstract

In this communication we present a numerical scheme for the resolution of matrix Riccati equation used in control problems. The scheme is unconditionally stable and the solution is definite positive at each time step of the resolution. We prove the convergence in the scalar case and present several numerical experiments for classical test cases.

1 Introduction.

We study the optimal control of a differential linear system

\[
\frac{dy}{dt} = Ay + Bv
\]

where the state variable \( y(t) \) belongs to \( \mathbb{R}^n \) and the control function \( v(\bullet) \) takes its values in \( \mathbb{R}^m \), with \( n \) and \( m \) being given integers. Matrix \( A \) is composed by \( n \) lines and \( n \) columns and matrix \( B \) contains \( n \) lines and \( m \) columns. Both matrices \( A \) and \( B \) are independent of time. With the ordinary differential equation (1) is associated an initial condition

\[
y(0) = y_0
\]

with \( y_0 \) given in \( \mathbb{R}^n \) and the solution of system (1) (2) is parametrized by the function \( v(\bullet) \). The control problem consists of finding the minimum \( u(\bullet) \) of some quadratic functional \( J(\bullet) \):

\[
J(u(\bullet)) \leq J(v(\bullet)), \quad \forall v(\bullet).
\]

The functional \( J(\bullet) \) depends on the control variable function \( v(\bullet) \), is defined by the horizon \( T > 0 \), the symmetric semi-definite positive \( n \) by \( n \) constant matrix \( Q \) and the symmetric definite positive \( m \) by \( m \) constant matrix \( R \). We set classically :

\[
J(v(\bullet)) = \frac{1}{2} \int_0^T (Qy(t), y(t)) \, dt + \frac{1}{2} \int_0^T (Rv(t), v(t)) \, dt.
\]

Problem (1) (2) (3) (4) is a classical linear quadratic mathematical modelling of dynamical systems in automatics (see e.g. Lewis [Le86]). When the control function

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v(\bullet) is supposed to be square integrable \((v(\bullet) \in L^2([0, T], IR^n))\) then the control problem (1) (2) (3) (4) has a unique solution \(u(\bullet) \in L^2([0, T], IR^n)\) (see for instance Lions [Li68]). When there is no constraint on the control variable the minimum \(u(\bullet)\) of the functional \(J(v)\) is characterized by the condition:

\[
(5) \quad dJ(u) \cdot w = 0, \quad \forall w \in L^2([0, T], IR^n),
\]

which is not obvious to compute directly.

• When we introduce the differential equation (1) as a constraint between \(y(\bullet)\) and \(v(\bullet)\), the associated Lagrange multiplier \(p(\bullet)\) is a function of time and is classically named the adjoint variable. Research of a minimum for \(J(\bullet)\) (condition (5)) can be rewritten in the form of research of a saddle point and the evolution equation for the adjoint variable is classical (see e.g. Lewis [Le86]):

\[
(6) \quad \frac{dp}{dt} + A^t p + Q y = 0, \\
\text{with a final condition at } t = T, \\
(7) \quad p(T) = 0
\]

and the optimal control in terms of the adjoint state \(p(\bullet)\) takes the form:

\[
(8) \quad Ru(t) + B^t p(t) = 0.
\]

• We observe that the differential system (1) (6) together with the initial condition (2) and the final condition (7) is coupled through the optimality condition (8). In practice, we need a linear feedback function of the state variable \(y(t)\) instead of the adjoint variable \(p(t)\). Because adjoint state \(p(\bullet)\) depends linearly on state variable \(y(\bullet)\) we can set: \(p(t) = X(T - t) y(t), \quad 0 \leq t \leq T, \) with a symmetric \(n\) by \(n\) matrix \(X(\bullet)\) which is positive definite. The final condition (7) is realized for each value \(y(T)\), then we have the following condition:

\[
(9) \quad X(0) = 0.
\]

We set \(K = BR^{-1} B^t\), we remark that matrix \(K\) is symmetric positive definite, we replace the control \(u(t)\) by its value obtained in relation (8) and we deduce after elementary algebra the evolution equation for the transition matrix \(X(\bullet)\):

\[
(10) \quad \frac{dX}{dt} - (XA + A^t X) + KK X - Q = 0, \\
\]

which defines the Riccati equation associated with the control problem (1) (2) (3) (4).

• In this paper we study the numerical approximation of differential system (9) (10). Recall that the given matrices \(Q\) and \(K\) are \(n \times n\) symmetric matrices, with \(Q\) semi-definite positive and \(K\) positive definite ; the matrix \(A\) is an \(n\) by \(n\) matrix without any other condition and the unknown matrix \(X(t)\) is symmetric. We have the following property (see e.g. [Le86]).

**Proposition 1.1** Let \(K, Q, A\) be given \(n \times n\) matrices with \(K, Q,\) symmetric, \(Q\) positive and \(K\) definite positive. Let \(X(\bullet)\) be the solution of the Riccati differential equation (10) with initial condition (9). Then \(X(t)\) is well defined for any \(t \geq 0\),
is symmetric and for each $t > 0$, $X(t)$ is definite positive and tends to a definite positive matrix $X_\infty$ as $t$ tends to infinity: $X(t) \rightarrow X_\infty$ if $t \rightarrow \infty$. Matrix $X_\infty$ is the unique positive symmetric matrix which is solution of the so-called algebraic Riccati equation:

$$-(X_\infty A + A^t X_\infty) + X_\infty K X_\infty - Q = 0.$$ 

- As a consequence of this proposition it is useful to simplify the feedback command law (8) by the associated limit command obtained by taking $t \rightarrow \infty$, that is:

$$(11) \quad v(t) = -R^{-1} B^t X_\infty y(t),$$

and the differential system (1) (11) is stable (see e.g. [Le86]). The practical computation of matrix $X_\infty$ by direct methods is not obvious and we refer e.g. to Laub [La79]. If we wish to compute directly a numerical solution of instationary Riccati equation (10), classical methods for ordinary differential equations like e.g. the forward Euler method

$$\frac{1}{\Delta t} (X_{j+1} - X_j) + X_j K X_j - (A^t X_j + X_j A) - Q = 0,$$

or Runge Kutta method fail to maintain positivity of the iterate $X_{j+1}$ at the order $(j + 1)$:

$$(12) \quad (X_{j+1} x, x) > 0, \quad \forall x \in \mathbb{R}^n, \quad x \neq 0,$$

if $X_j$ is positive definite and if time step $\Delta t > 0$ is not small enough (see e.g. Dieci and Eirola [DE96]). Moreover, there is to our best knowledge no simple way to determine a priori if time step $\Delta t$ is compatible or not with condition (12).

- In the following, we propose a method for numerical integration of Riccati equation (10) which maintains condition (12) for each time step $\Delta t > 0$. We present in second section the simple case of scalar Riccati equation and present the numerical scheme and its principal properties of the general case in section 3. We describe several numerical experiments in section 4.

## 2 Scalar Riccati equation.

When the unknown is a scalar variable, we write Riccati equation in the following form:

$$(13) \quad \frac{dx}{dt} + kx^2 - 2ax - q = 0,$$

with

$$(14) \quad k > 0, \quad q \geq 0, \quad a^2 + q^2 > 0,$$

and an initial condition:

$$(15) \quad x(0) = d, \quad d \geq 0.$$
We approach the ordinary differential equation (13) with a finite difference scheme of the type proposed by Baraille [Ba91] for hypersonic chemical kinetics and independently with the "family method" proposed by Cariolle [Ca79] and studied by Miellou [Mi84]. We suppose that time step $\Delta t$ is strictly positive. The idea is to write the approximation $x_{j+1}$ at time step $(j+1)\Delta t$ as a rational fraction of $x_j$ with positive coefficients. We decompose first the real number $a$ into positive and negative parts: $a = a^+ - a^-$, $a^+ = \max(0, a) \geq 0$, $a^- = \max(0, -a) \geq 0$, $a^+ a^- = 0$ and factorize the product $x^2$ into the very simple form: $(x^2)_{j+1/2} = x_j x_{j+1}$.

**Definition 2.1** For resolution of the scalar differential equation (13), we define our numerical scheme by the following relation:

$$
(16) \quad \frac{x_{j+1} - x_j}{\Delta t} + k x_j x_{j+1} - 2a^+ x_j + 2a^- x_{j+1} - q = 0.
$$

- The scheme (16) is implicit because some linear equation has to be solved to compute $x_{j+1}$ when $x_j$ is supposed to be given. In the case of our scheme this equation is linear and the solution $x_{j+1}$ is obtained from scheme (16) by the homographic relation:

$$
(17) \quad x_{j+1} = \frac{(1 + 2a^+ \Delta t) x_j + q \Delta t}{k \Delta t x_j + (1 + 2a^- \Delta t)}.
$$

**Proposition 2.2** Let $(x_j)_{j \in \mathbb{N}}$ be the sequence defined by initial condition: $x_0 = x(0) = d$

and recurrence relation (17). Then sequence $(x_j)_{j \in \mathbb{N}}$ is globally defined and remains positive for each time step: $x_j \geq 0$, $\forall j \in \mathbb{N}$, $\forall \Delta t > 0$. If $\Delta t > 0$ is chosen such that:

$$
(18) \quad 1 + 2 |a| \Delta t - k q \Delta t^2 \neq 0,
$$

then $(x_j)_{j \in \mathbb{N}}$ converges towards the positive solution $x^*$ of the "algebraic Riccati equation": $k x^2 - 2a x - q = 0$ and

$$
(19) \quad x^* = \frac{1}{k} (a + \sqrt{a^2 + kq}).
$$

- In the exceptional case where $\Delta t > 0$ is chosen such that (18) is not satisfied, then the sequence $(x_j)_{j \in \mathbb{N}}$ is equal to the constant $\frac{1 + 2a^+ \Delta t}{k \Delta t}$ for $j \geq 1$ and the scheme (17) cannot be used for the approximation of Riccati equation (13).

**Theorem 2.3** We suppose that the data $k, a, q$ of Riccati equation satisfy (14) and (18) and that the datum $d$ of condition (15) is relatively close to $x^*$, i.e.:

$$
(20) \quad -\frac{1}{k \tau} + \eta \leq d - x^* \leq C,
$$

where $C$ is some given strictly positive constant $(C > 0)$, $x^*$ given by relation (19) is the limit in time of the Riccati equation, $\tau$ is defined from data $k, a, q$ by:

$$
\tau = \frac{1}{2 \sqrt{a^2 + kq}}, \quad \text{and } \eta \text{ is some constant chosen such that}
$$
(21) \( 0 < \eta < \frac{1}{k\tau} \).

- We denote by \( x(t; d) \) the solution of differential equation (13) with initial condition (15). Let \((x_j(t; d))_{j \in \mathbb{N}}\) be the numerical scheme defined by relation (17) and let \( d_\Delta \) be the initial condition: \( x_0(\Delta t; d_\Delta) = d_\Delta \). We suppose that the numerical initial condition \( d_\Delta > 0 \) satisfies a condition analogous to (20):

\[
- \frac{1}{k\tau} + \eta \leq d_\Delta - x^* \leq C, \text{ with } C \text{ and } \eta > 0 \text{ equal to the constant introduced in (20) and satisfying (21)}.
\]

- Then the approximated value \((x_j(\Delta t; d_\Delta))_{j \in \mathbb{N}}\) is arbitrarily close to the exact value \( x(j\Delta t; d) \) for each \( j \) as \( \Delta t \to 0 \) and \( d_\Delta \to d \). More precisely, if \( a \neq 0 \) we have the following estimate for the error at time \( t \in [0, \Delta t) \):

\[
|x(j\Delta t; d) - x(x_j(\Delta t; d_\Delta))| \leq A(\Delta t + |d - d_\Delta|), \forall j \in \mathbb{N}, 0 < \Delta t \leq B
\]

with constants \( A > 0, B > 0, \) depending on data \( k, a, q, \eta \) and \( C \) but independent on time step \( \Delta t > 0 \) and iteration \( j \).

- If \( a = 0 \), the scheme is second order accurate in the following sense:

\[
|x(j\Delta t; d) - x_j(\Delta t; d_\Delta)| \leq A(\Delta t^2 + |d - d_\Delta|), \forall j \in \mathbb{N}, 0 < \Delta t \leq B
\]

with constants \( A \) and \( B \) independent on time step \( \Delta t \) and iteration \( j \).

A direct application of the Lax theorem for numerical scheme associated to ordinary differential equations is not straightforward because both Riccati equation and the numerical scheme are nonlinear. Our proof is detailed in [DS99].

3 Matrix Riccati equation.

In order to define a numerical scheme to solve the Riccati differential equation (10) with initial condition (9) we first introduce a strictly positive real number \( \mu \), which is chosen positive in such a way that the real matrix \( [\mu I - (A + A^\dagger)] \) is definite positive:

(22) \( \frac{1}{2}(\mu x, x) - (A x, x) > 0, \forall x \neq 0. \)

Then we introduce the definite positive matrix \( M \) which depends on \( \mu \) and matrix \( A : M = \frac{1}{2} \mu I - A \). The numerical scheme is then defined by analogy with relation (16). We have the following decomposition:

(23) \( A = A^+ - A^- \),

with \( A^+ = \frac{1}{2} \mu I, A^- = M, \mu > 0, M \) positive definite. Taking as an explicit part the positive contribution \( A^+ \) of the decomposition (23) of matrix \( A \) and in the implicit part the negative contribution \( A^- = M \) of the decomposition (23), we get

(24) \[
\begin{aligned}
\frac{1}{\Delta t}(X_{j+1} - X_j) + \frac{1}{2}(X_jKX_{j+1} + X_{j+1}KX_j) + \\
\quad + (M^tX_{j+1} + X_{j+1}M) = \mu X_j + Q.
\end{aligned}
\]
The numerical solution given by the scheme $X_{j+1}$ at time step $j+1$ is then defined as a solution of Lyapunov matrix equation with matrix $X$ as unknown:

$$S_j^t X + X S_j = Y_j \quad \text{with}$$

$$S_j = \frac{1}{2} I + \frac{\Delta t}{2} K X_j + \Delta t M$$

and

$$Y_j = X_j + \mu \Delta t X_j + \Delta t Q.$$  

We notice that $S_j$ is a (non necessarily symmetric) positive matrix and that $Y_j$ is a symmetric definite positive matrix if it is the case for $X_j$.

**Definition 3.1** Let $n$ be an integer greater or equal to 1. We define by $\mathcal{S}_n(\mathbb{R})$, (respectively $\mathcal{S}_n^+(\mathbb{R})$, $\mathcal{S}_n^{++}(\mathbb{R})$) the linear space (respectively the closed cone, the open cone) of symmetric-matrices (respectively symmetric positive and symmetric definite positive matrices). The following inclusions $\mathcal{S}_n^+(\mathbb{R}) \subseteq \mathcal{S}_n^{++}(\mathbb{R}) \subseteq \mathcal{S}_n(\mathbb{R})$ are natural.

**Proposition 3.2** Let $S$ be a matrix which is not necessary symmetric, such that the associated quadratic form $\mathbb{R}^n \ni x \mapsto (x, S x) \in \mathbb{R}$, is strictly positive, i.e. $S + S^t \in \mathcal{S}_n^{++}(\mathbb{R})$. Then the application $\varphi$ defined by:

$$\varphi(X) = S^t X + X S \in \mathcal{S}_n(\mathbb{R}),$$

is a one to one bijective application on the space $\mathcal{S}_n(\mathbb{R})$ of real symmetric matrices of order $n$. Moreover, if matrix $\varphi(X)$ is definite positive then the matrix $X$ is also definite positive: if $\varphi(X) \in \mathcal{S}_n^{++}(\mathbb{R})$ then $X \in \mathcal{S}_n^{++}(\mathbb{R})$.

- The numerical scheme has been written as an equation with unknown $X = X_{j+1}$ which takes the form: $\varphi_j(X) = Y_j$ with $\varphi_j$ given by a relation of the type (27) with the help of matrix $S_j$ defined in (25) and a datum matrix $Y_j$ defined by relation (26). Then we have the following propositions.

**Proposition 3.3** The matrix $X_j$ defined by numerical scheme (24) with the initial condition $X_0 = 0$ is positive for each time step $\Delta t > 0$: $X_j \in \mathcal{S}_n^{++}(\mathbb{R})$, $\forall j \geq 0$.

- If there exists some integer $m$ such that $X_m$ belongs to open cone $\mathcal{S}_n^{++}(\mathbb{R})$, then matrix $X_{m+j}$ belongs to open cone $\mathcal{S}_n^{++}(\mathbb{R})$ for each index $j$.

**Proposition 3.4** Under the condition: $\frac{1}{2} (K X_{\infty} + X_{\infty} K) < (\mu + \frac{1}{\Delta t}) I$, the scheme (24) is monotone and we have more precisely:

$$0 \leq X_j \leq X_\infty \implies 0 \leq X_j \leq X_{j+1} \leq X_\infty.$$
4 First numerical experiments.

4.1 Square root function.

- The first example studied is the resolution of the equation:

\[
\frac{dX}{dt} + X^2 - Q = 0, \quad X(0) = 0
\]

(29) with \(n = 2, \ A = 0, \ K = I\) and matrix \(Q\) equal to

\[
Q = \frac{1}{2} \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 100 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ -1 & 1 \end{pmatrix}.
\]

- We have tested our numerical scheme for fixed value \(\Delta t = 1/100\) and different values of parameter \(\mu: \mu = 0.1, 10^{-6}, 10^{+6}\). For small values of parameter \(\mu\), the behaviour of the scheme does not change between \(\mu = 0.1\) and \(\mu = 10^{-6}\). Figures 1 to 4 show the evolution with time of the eigenvalues of matrix \(X_j\) and the convergence is achieved to the square root of matrix \(Q\). For large value of parameter \(\mu (\mu = 10^{+6})\), we loose completely consistency of the scheme (see figures 5 and 6).

4.2 Harmonic oscillator.

- The second example if the classical harmonic oscillator. Dynamical system \(y(t)\) is governed by the second order differential equation with command \(v(t)\):

\[
\frac{d^2y(t)}{dt^2} + 2\delta \frac{dy(t)}{dt} + \omega^2 y(t) = b v(t).
\]

(31) This equation is written as a first order system of differential equations:

\[
Y = \begin{pmatrix} y(t) \\ \frac{dy(t)}{dt} \end{pmatrix}, \quad \frac{dY}{dt} = \begin{pmatrix} 0 & 1 \\ -\omega^2 & -2\delta \end{pmatrix} Y(t) + \begin{pmatrix} 0 \\ b v(t) \end{pmatrix}.
\]

(32) In this case, we have tested the stability of the scheme for fixed value of parameter \(\mu (\mu = 0.1)\) and different values of time step \(\Delta t\) and coefficients of matrix \(R\) inside the cost function of relation (4):

\[
R = \begin{pmatrix} \alpha & 0 \\ 0 & \alpha \end{pmatrix}.
\]

- We have chosen three sets of parameters: \(\alpha = \Delta t = 1/100\) (reference experiment, figures 7 and 8), \(\alpha = 10^{-6}, \Delta t = 1/100\) (very small value for \(\alpha\), figures 9 and 10) and \(\alpha = 1/100, \Delta t = 100\) (too large value for time step, figures 11 and 12). Note that for the last set of parameters, classical explicit schemes fail to give any answer. As in previous test case, we have represented the two eigenvalues of discrete matrix solution \(X_j\) as time is increasing. On reference experiment (figures 7 and 8), we have convergence of the solution to the solution of algebraic Riccati equation. If control parameter \(\alpha\) is chosen too small, the eigenvalues of Riccati equation oscillate during the first time steps but reach finally the correct values of limit matrix, the solution of algebraic Riccati equation. If time step is too large, we still have stability but we loose also monotonicity. Nevertheless, convergence is achieved as in previous case.
5 Conclusion.

We have proposed a numerical scheme for the resolution of the matrix Riccati equation. The scheme is implicit, unconditionally stable, needs to use only one scalar parameter and to solve a linear system of equations for each time step. This scheme is convergent in the scalar case and has good monotonicity properties in the matrix case. Our first numerical experiments show stability and robustness when various parameters have large variations. Situations where classical explicit schemes fail to give a solution compatible with the property that solution of Riccati equation is a definite positive matrix have been computed. We expect to prove convergence in the matrix case and we will present in [DS99] experiments on realistic test models such as a string of vehicles and the discretized wave equation.

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Figures 1 and 2: Square root function test. Two first eigenvalues of numerical solution for \( \mu = 0.1 \).

Figures 3 and 4: Square root function test. Two first eigenvalues of numerical solution for \( \mu = 10^{-6} \).
Figures 5 and 6: Square root function test.  
Two first eigenvalues of numerical solution for ($\mu = 10^6$)
Figures 7 and 8: Harmonic oscillator.
Two first eigenvalues of numerical solution ($\mu = 0.1, \alpha = 0.01, \Delta t = 0.01$).
Figures 9 and 10: Harmonic oscillator.
Two first eigenvalues of numerical solution ($\mu = 0.1, \alpha = 10^{-6}, \Delta t = 0.01$).
Figures 11 and 12: Harmonic oscillator.
Two first eigenvalues of numerical solution ($\mu = 0.1, \alpha = 0.01, \Delta t = 100$).

References.


