

Interval arithmetic methods to verify the stabilizing solution of an algebraic Riccati equation

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Overview

Goal: compute a set \mathbf{X} which contains (for sure, not “up to small computational errors”) the stabilizing solution X_s of

$$0 = F(X) = A^\top X + XA + Q - XGX.$$

Do not use more than $O(n^3)$ flops.

Plan

- Convince you that interval arithmetic is a good idea.
- Show you what people did to verify Riccati equations.
- Show you the improvements we introduced.
- Competitors, experiments, and other ideas.

Interval arithmetic [Rump, '10 review]

Basic idea if $a \in [1, 2]$ and $b \in [3, 4]$, then $a + b \in [4, 6]$ and $ab \in [3, 8]$.
Store (min, max) (or $(center, radius)$) and operate on them. \mathbb{IR}, \mathbb{IC} .

With IEEE arithmetic + **rounding** in the correct direction, the inclusions work irrespective of machine errors.

Machine numbers can be embedded in \mathbb{IR} as radius-0 intervals.

Wrapping effect

$$\begin{bmatrix} x \\ y \end{bmatrix} \mapsto \begin{bmatrix} -1 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

$$v = \begin{bmatrix} [1, 2] \\ [3, 4] \end{bmatrix}$$

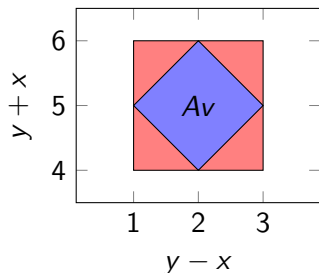
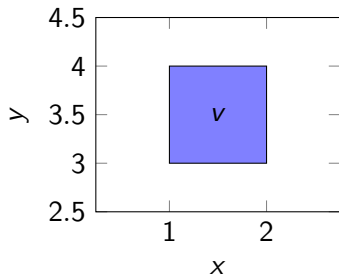


Image of v : blue square. Interval result: red+blue.
This happens even though $\kappa(A) = 1$!

Verify, don't solve

The first rule of interval arithmetic

You don't solve your problem with interval arithmetic.

Things like back-substitution would create huge intervals.
E.g., solving $AX + XB = C$ with Bartels-Stewart: hopeless.

Instead:

$g(\mathbf{x}) \subset \mathbf{x}$ implies $\mathbf{x} = g(\mathbf{x})$ for some $\mathbf{x} \in \mathbf{x}$

- compute (with usual methods) an approximate solution $\tilde{\mathbf{x}}$.
- reformulate as $\mathbf{x} = g(\mathbf{x})$, e.g., $\mathbf{x} = \mathbf{x} - Rf(\mathbf{x})$.
- choose an interval $\mathbf{x} \ni \tilde{\mathbf{x}}$, e.g., $\tilde{\mathbf{x}} - [0.9, 1.1]f(\tilde{\mathbf{x}})$
- check (hopefully) $g(\mathbf{x}) \subset \mathbf{x}$.
- if not, enlarge \mathbf{x} , e.g., $\mathbf{x} \leftarrow [0.9, 1.1]g(\mathbf{x})$ and **retry**.

Details omitted; e.g.: need care with computing $\mathbf{x} - Rf(\mathbf{x})$.

The Krawczyk method

Ingredients:

- **approximate solution** \tilde{x} .
- **slope** \mathbf{S}_x : set such that there is $S \in \mathbf{S}_x$ satisfying

$$f(x) - f(\tilde{x}) = S(x - \tilde{x}) \quad \text{for all } x \in \mathbf{x}. \quad (*)$$

Often related to an interval evaluation of $f'(\mathbf{x})$.

- **preconditioner** R : approximate inverse of some matrix in \mathbf{S}_x .

Theorem [Krawczyk '69, Rump '83]

If, for some interval δ ,

$$\text{int}(\delta) \supseteq -Rf(\tilde{x}) + (I - R\mathbf{S}_{\tilde{x}+\delta})\delta,$$

then $\tilde{x} + \delta$ contains a solution of $f(x) = 0$.

If (*) holds replacing \tilde{x} with every $y \in \mathbf{x}$, then the solution is unique.

Verifying Riccati equations

$$F(X) = A^T X + XA + Q - XGX$$

$O(n^3)$ algorithm: [Hashemi '12]

- \tilde{X} from your favorite method.
- \mathbf{S}_x : $F'(\mathbf{X}) = (A - G\mathbf{X})^T \otimes I + I \otimes (A - G\mathbf{X})^T$ works.
- R : **can't use Bartels-Stewart**. Instead: explicit eigendecomposition $(A - GX) \approx VDV^{-1}$ and

$$R = (V^{-T} \otimes V^{-T})(D^T \otimes I + I \otimes D^T)^{-1}(V^T \otimes V^T)$$

Additional manipulations: $(I - R\mathbf{S}_x) = (V^{-T} \otimes V^{-T})(\dots)(V^T \otimes V^T)$

Again, many details omitted; for instance, dealing properly with $W \approx V^{-1}$.

Improving Hashemi's method

Our goal: construct an enclosure \mathbf{X} for the **stabilizing** solution X_S .

Plan:

- Compute an enclosure \mathbf{X} starting from $\tilde{X} \approx X_S$.
- Verify that each matrix in $A - G\mathbf{X}$ is stabilizing.
- Uniqueness follows from classical Riccati theory. [Brockett, '70]

Letting go of uniqueness allows some improvements:

- 1 Tighter slope \mathbf{S}_X .
- 2 Defer the change of basis as in [Frommer Hashemi '09].
- 3 Verify a different equation using tricks from [Mehrmann P. '12].

Improvements

1 Tighter slope \mathbf{S}_X

We can use $\mathbf{S}_X = (A - G\mathbf{X})^\top \otimes I + I \otimes (A - G\tilde{\mathbf{X}})^\top$.

2 Defer the change of basis

Find \mathbf{Y} that encloses a solution of $\hat{F}(\mathbf{Y}) = V^\top f(V^{-\top} \mathbf{Y} V^{-1}) V$:
Easier, because \hat{F}' is diagonal.

Then, compute $\mathbf{X} = V^{-\top} \mathbf{Y} V^{-1}$.

Even if $\mathbf{Y} \in \mathbf{Y}$ unique solution, other solutions may end up in \mathbf{X} due to wrapping effects.

[Frommer Hashemi '09] introduced this trick for `sqrtm`.

Improvements

3 Verify a different equation

$$CARE \iff \begin{bmatrix} A & -G \\ -Q & -A^\top \end{bmatrix} \begin{bmatrix} I \\ X \end{bmatrix} = \begin{bmatrix} I \\ X \end{bmatrix} (A - GX)$$

[Mehrman P. '12]: one can find a basis for $\text{im} \begin{bmatrix} I \\ X \end{bmatrix}$ with an identity in different position (i.e., $\text{im} \begin{bmatrix} I \\ X \end{bmatrix} = \text{im} \Pi \begin{bmatrix} I \\ Y \end{bmatrix}$, Π permutation matrix) so that $|Y|_{ij} \leq \sqrt{2}$.

As above, we can **verify a Riccati equation for Y** rather than one for X .

Smaller / more balanced entries \implies easier verification.

Verify a different equation

Algorithm

- Compute approximate CARE solution \tilde{X}
- Compute Π so that $\text{im} \begin{bmatrix} I \\ \tilde{X} \end{bmatrix} = \text{im} \Pi \begin{bmatrix} I \\ \tilde{Y} \end{bmatrix}$, with \tilde{Y} bounded.
- Form the CARE associated with $\Pi^{-1} \begin{bmatrix} A & -G \\ -Q & -A^\top \end{bmatrix} \Pi$ instead of $\begin{bmatrix} A & -G \\ -Q & -A^\top \end{bmatrix}$.
- Compute an inclusion $\mathbf{Y} \supseteq Y_s$ of its stable solution.
- $\mathbf{X} = \mathbf{U}_2 \mathbf{U}_1^{-1}$, where $\Pi \begin{bmatrix} I \\ \mathbf{Y} \end{bmatrix} = \begin{bmatrix} \mathbf{U}_1 \\ \mathbf{U}_2 \end{bmatrix}$. Other solutions may enter \mathbf{X} .

Summing up

- Start from an approximate stabilizing solution \tilde{X}
- Use the above methods to construct $\mathbf{X} \ni \tilde{X}$ containing a solution
- If all the matrices in \mathbf{X} are stabilizing, bingo!

Alternative approach (main competitor): [Miyajima '15].

Mix between the above methods and explicit normwise bounds. **Idea:**

- Newton-like iteration $X = g(X)$, $g(X) = X - (F'_X)^{-1}(F(X))$.
- Formula for F'_X using an eigendecomposition of $A - GX$, as earlier.
- Expand $g(\mathbf{X})$, where $\mathbf{X} = (\tilde{X} - \eta R, \tilde{X} + \eta R)$ (for a specific choice of R), as a function of η .
- Using inequalities, determine η such $\mathbf{X} \supseteq g(\mathbf{X})$ (if possible).
- Compute η using interval arithmetic and rounding.
- Uniqueness and stabilizing-ness verified *a posteriori*.

Diagonalizability

Verification methods tested on the benchmarks in CAREX [Benner et al '95]

OK on many of them, but we are still not satisfied:

CAREX Example 1 [Benner et al '95]

$$A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, \quad G = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}, \quad Q = \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix}, \quad X_s = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}$$

This example can be used for a first verification of any solver for [CAREs] since the solution may be computed by hand.

$A - GX_s$ is not diagonalizable \implies **all methods fail** on this 'warm-up' example.

Non-diagonalizable problems

New algorithm: not as effective as the others, but it works in $O(n^3)$ even if $A - GX_s$ is (almost) not diagonalizable.

Idea

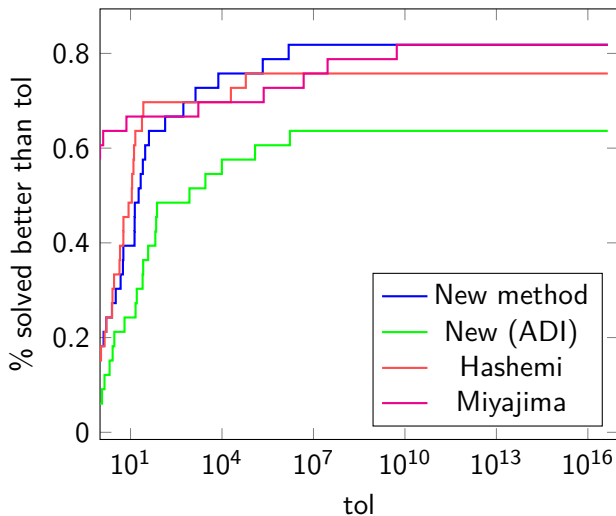
- Rewrite as a CARE in Δ , where $X = \tilde{X} + \Delta$:
 $\hat{A}^* \Delta + \Delta \hat{A} + \hat{Q} - \Delta G \Delta = 0$.
- **Mimic ADI:** fixed-point eqn $\Delta = (\hat{A} - sI)^{-\top} (\Delta G \Delta - \hat{Q} - \Delta(A + sI))$.
- Are there parameters that we can tune? Choice of s , and then **change of basis:**

$$\Delta_V = V^* \Delta V, \quad A_V = V^{-1} \tilde{A} V, \quad Q_V = V^* \tilde{Q} V, \quad G_V = V^{-1} G V^{-*}.$$

No need to diagonalize this time.

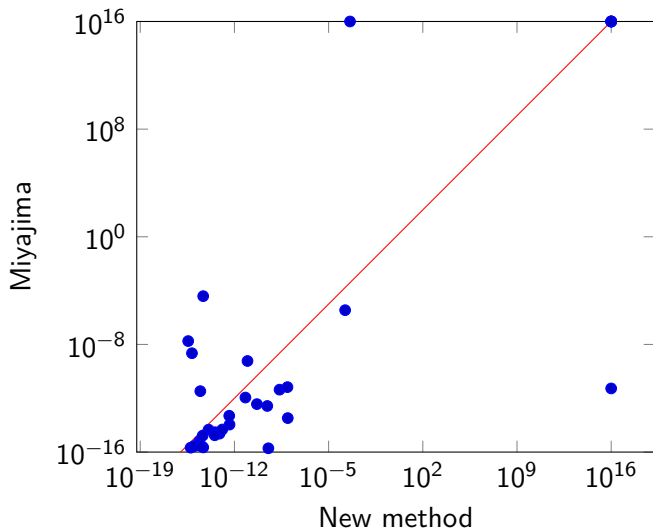
In practice, we choose $V =$ orthogonal Schur factor of \hat{A} ,
 $s = -\lambda_{\max}(\hat{A})$

Performance profile on CAREX suite in [Chu et al '07]

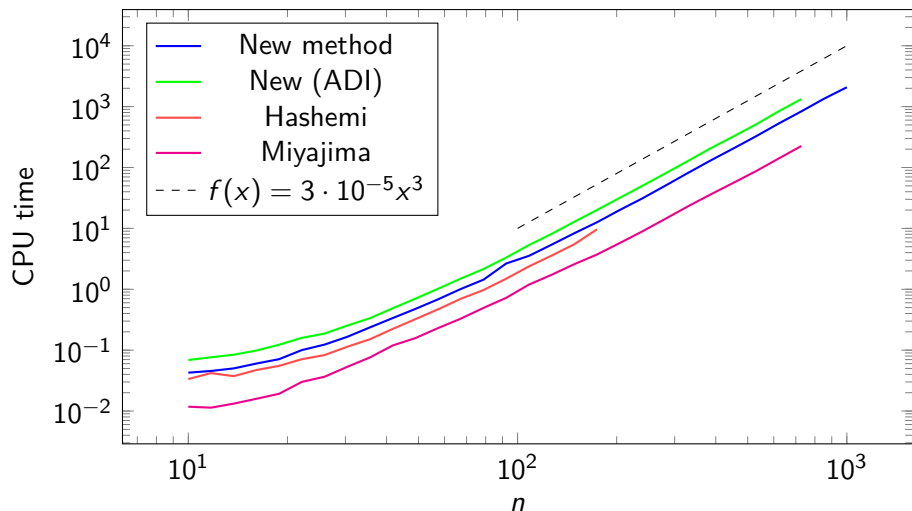


Top left = better.

(Norm-2) width of found interval



CPU time on CAREX 15



Lower = better. New = only method to reach $n = 1000$.

Conclusions

- Technical improvements and ideas from Riccati theory take Krawczyk-based method to state-of-the-art level.
- No method always better than the others, so it is useful to have more choice.
- In almost all cases, the first solution guess $\tilde{x} - [0.9, 1.1]f(\tilde{x})$ already works — so there is still room to optimize.
- Up next: transfer some of these improvements to Miyajima's method.

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[Thanks for your attention!]