

The benefits of changing identity

in Lagrangian subspaces and doubling algorithms

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Lagrangian subspaces

Definition

A subspace $\mathcal{U} = \text{Im } U$ of \mathbb{C}^{2n} is **Lagrangian** if it has dimension n and

$$U^* \mathcal{J}_{2n} U = 0 \qquad \mathcal{J}_{2n} = \begin{bmatrix} 0 & I_n \\ -I_n & 0 \end{bmatrix}$$

- Property of the subspace, not of the basis: can post-multiply $U \rightarrow UM$
- They arise naturally as stable invariant subspaces of **Hamiltonian** and **symplectic** problems
- Central role in control

Dealing with Lagrangian subspaces

Problem: find a (Lagrangian) invariant subspace of ...
but first: represent suitably a Lagrangian subspace, operate on it and return it to the user **preserving structure**

Subspace \mathcal{U} often represented as range of (full column rank) U ...
... but this is **not unique**: $\text{Im } U = \text{Im } UM$ for any M nonsingular

Graph bases and Riccati equations

One of the first choices [Willems, '71] : **graph basis**

$$\mathcal{U} = \text{Im} \begin{bmatrix} U_1 \\ U_2 \end{bmatrix} = \text{Im} \begin{bmatrix} I \\ X \end{bmatrix} \quad X = U_2 U_1^{-1}$$

Transforms invariant subspace problems into **algebraic Riccati equations**

$$\begin{bmatrix} A & B \\ C & D \end{bmatrix} \begin{bmatrix} I_n \\ X \end{bmatrix} = \begin{bmatrix} I_n \\ X \end{bmatrix} F \quad \iff \quad XBX + XA - DX - C = 0$$

- ☺ Easy to ensure Lagrangianity: \mathcal{U} Lagrangian $\iff X$ Hermitian
- ☹ Not all subspaces well represented: what about these?

$$\begin{bmatrix} 0 & 0 \\ 0 & 0 \\ 1 & 0 \\ 0 & 1 \end{bmatrix} \quad \begin{bmatrix} 1 & 1 \\ 1 & 1 + \varepsilon \\ 1 & 0 \\ 0 & 1 \end{bmatrix}$$

Orthogonal bases

Natural solution: **orthogonal basis**

$$\mathcal{U} = \text{Im} \begin{bmatrix} Q_1 \\ Q_2 \end{bmatrix} \quad \text{with orthogonal columns}$$

- ☺ All subspaces representable
 - ☺ Very stable, no element growth
 - ☹ Computationally more expensive
 - ☹ **Too many parameters:** Lagrangianity $\Leftrightarrow Q_1^* Q_2 = Q_2^* Q_1$
Easily lost through numerical computation, difficult to enforce explicitly
- Loss of Lagrangianity is a serious problem, e.g. in Laub Trick [Laub, '79]

Trying to save the Riccati approach

A “folklore result”: in some cases useful to switch to $\begin{bmatrix} Y \\ I \end{bmatrix}$ (so $Y = X^{-1}$)

Dual ARE

$$XBX + XA - DX - C = 0 \quad \implies \quad B + AY - YD - YCY = 0$$

But still both approaches can fail: e.g.,

$$\begin{bmatrix} 1 & 0 \\ 0 & 0 \\ 0 & 0 \\ 0 & 1 \end{bmatrix}$$

Neither $\begin{bmatrix} I \\ X \end{bmatrix}$ nor $\begin{bmatrix} Y \\ I \end{bmatrix}$ work

Idea: The identity that we are looking for is already there, in rows 1 and 4!

Permuted graph bases

$$\begin{bmatrix} 1 & 0 \\ * & * \\ * & * \\ 0 & 1 \end{bmatrix}$$

We look for bases with an identity submatrix spread along different rows

$$= II \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ * & * \\ * & * \end{bmatrix}$$

Equivalently: keep identity on top, but premultiply with a **permutation matrix**

But permutations are not the right tool here: want to preserve Lagrangianity

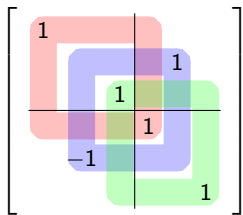
The right thing: symplectic swap matrices

Symplectic row swap matrices: those that act (separately) as either

$$I = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \text{ or } \mathcal{J}_2 = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}$$

on every pair of indices $(i, n + i)$

Examples:



The diagram shows a 4x4 matrix with a vertical line between columns 2 and 3, and a horizontal line between rows 2 and 3. The matrix is partitioned into four 2x2 blocks. The top-left block is red and contains a 1 in the top-left corner. The top-right block is purple and contains a 1 in the top-right corner. The bottom-left block is blue and contains a -1 in the bottom-left corner. The bottom-right block is green and contains a 1 in the bottom-right corner.

$$I_{2n} = \begin{bmatrix} I_n & 0 \\ 0 & I_n \end{bmatrix}$$

$$\mathcal{J}_{2n} = \begin{bmatrix} 0 & I_n \\ -I_n & 0 \end{bmatrix}$$

- Can only swap row i with $n + i$
- There's 2^n of them
- All preserve Lagrangianity, so $\Pi \left[\begin{smallmatrix} I \\ X \end{smallmatrix} \right]$ Lagrangian $\Leftrightarrow X$ Hermitian

2×2 case

$$U = \begin{pmatrix} 1 \\ 2 \\ 3 \\ 4 \end{pmatrix}$$

$$\text{If } \begin{pmatrix} 1 \\ 2 \end{pmatrix} \text{ well-conditioned} \Rightarrow U \cdot \text{inv} \left(\begin{pmatrix} 1 \\ 2 \end{pmatrix} \right) = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ * & * \\ * & * \end{bmatrix}$$

$$\text{If } \begin{pmatrix} 3 \\ 4 \end{pmatrix} \text{ well-conditioned} \Rightarrow U \cdot \text{inv} \left(\begin{pmatrix} 3 \\ 4 \end{pmatrix} \right) = \begin{bmatrix} * & * \\ * & * \\ 1 & 0 \\ 0 & 1 \end{bmatrix}$$

2×2 case

$$\text{If } \begin{bmatrix} 1 \\ 4 \end{bmatrix} \text{ well-conditioned} \Rightarrow U \cdot \text{inv} \left(\begin{bmatrix} 1 \\ 4 \end{bmatrix} \right) = \begin{bmatrix} 1 & 0 \\ * & * \\ * & * \\ 0 & 1 \end{bmatrix} = H \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ * & * \\ * & * \end{bmatrix}$$

$$\text{If } \begin{bmatrix} 3 \\ 2 \end{bmatrix} \text{ well-conditioned} \Rightarrow U \cdot \text{inv} \left(\begin{bmatrix} 3 \\ 2 \end{bmatrix} \right) = \begin{bmatrix} * & * \\ 0 & 1 \\ 1 & 0 \\ * & * \end{bmatrix} = H \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ * & * \\ * & * \end{bmatrix}$$

What if none of them works? E.g., $\begin{bmatrix} 1 & 0 \\ 0 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}$? **Not Lagrangian!**

How good are row swaps?

$\mathcal{U} = \text{Im } U$ has a **permuted graph basis** with a prescribed Π iff some subset of n rows of U is linearly independent

$$U = \Pi \begin{bmatrix} Y \\ Z \end{bmatrix} \sim \Pi \begin{bmatrix} I \\ ZY^{-1} \end{bmatrix} \stackrel{\text{def}}{=} \Pi \begin{bmatrix} I \\ X \end{bmatrix}$$

Theorem

For each **Lagrangian** \mathcal{U} there's a Π such that Y is invertible. . .

Follows easily from a result on symplectic matrices [Dopico, Johnson '06]

Theorem [Mehrmann, P., preprint]

. . . moreover, there's one with X **entrywise small**:

$$|(X)_{ij}| \leq \begin{cases} 1 & \text{if } i = j \\ \sqrt{2} & \text{otherwise} \end{cases}$$

Geometrical interpretation

In geometrical terms: the 2^n maps

$$f_{II} : X \text{ Hermitian and bounded} \mapsto \text{Im } II \begin{bmatrix} I \\ X \end{bmatrix}$$

are an atlas for the manifold of Lagrangian subspaces

For each subspace, we can find II giving “tame” structure-preserving basis

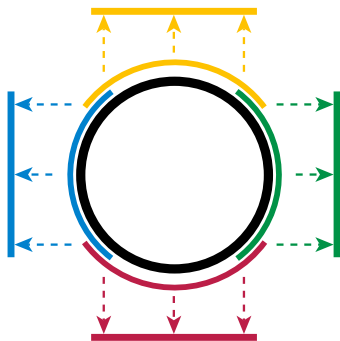


Image: ©Wikimedia

Unstructured case

Similar, **unstructured version** known for a generic subspace:

Theorem [Knuth, ≈'84??]

For every n -dimensional subspace $\mathcal{U} \subseteq \mathbb{C}^{n+m}$, there are a **permutation matrix** Π and an $X \in \mathbb{C}^{m \times n}$ with $|x_{i,j}| \leq 1$ for all i, j such that

$$\mathcal{U} = \text{Im } \Pi \begin{bmatrix} I_n \\ X \end{bmatrix}$$

Connected to rank revealing QR: existing work by Knuth, C.-T. Pan, Gu–Eisenstat, Goreinov *et al.* . .

Key word: **Plücker coordinates**

Sketch of the proof

$$U = \Pi \begin{bmatrix} Y_{\Pi} \\ Z_{\Pi} \end{bmatrix} \sim \Pi \begin{bmatrix} I \\ Z_{\Pi} Y_{\Pi}^{-1} \end{bmatrix} \stackrel{\text{def}}{=} \Pi \begin{bmatrix} I \\ X_{\Pi} \end{bmatrix}$$

Different Π give different Y_{Π} ; take R so that $|\det Y_R|$ maximal

Cramer's rule on $X_R = Z_R Y_R^{-1}$ gives

$$|x_{ii}| = \left| \frac{\det Y_Q}{\det Y_R} \right| \leq 1$$

Can only swap i with $n + i \Rightarrow$ this works only for **diagonal** elements x_{ii}

But similarly

$$\left| \det \begin{bmatrix} x_{i,i} & x_{i,j} \\ x_{i,j} & x_{j,j} \end{bmatrix} \right| = \left| \frac{\det Y_Q}{\det Y_R} \right| \leq 1 \quad \Rightarrow \quad |x_{i,j}| \leq \sqrt{2}$$

Computing a good Π

The proof can be turned into a “greedy” algorithm: given U s.t. $\text{Im } U = \mathcal{U}$,

- 1 choose an admissible Π
- 2 compute basis $\Pi \begin{bmatrix} I \\ X \end{bmatrix}$
- 3 if $|x_{i,i}| > 1$, update Π with a row swap to enlarge $\det Y_{\Pi}$, goto 2
- 4 if $|x_{i,j}| > \sqrt{2}$, two row swaps, goto 2

- Best to work with **thresholds** $S > 1$, $T > \sqrt{2}$
- Ends with a matrix X with $|(X)_{ij}| \leq \begin{cases} S & \text{if } i = j \\ T & \text{otherwise} \end{cases}$
- Every update of X is essentially a rank-1 update, $O(n^2)$
- Can be made very robust

Similar (less robust) algorithm for the unstructured case: [Goreinov et al., '08]

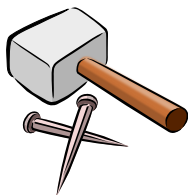
Applications?

Applications

Several different problems can be “reshaped” in order to use this theorem

Maslow's law

“If all you have is a hammer, everything looks like a nail”

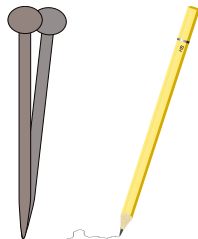
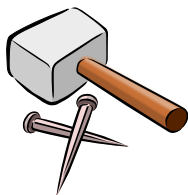


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First, a **pencil** and a **nail** are very similar objects. . .

Pencils and subspaces

Eigenvalues and right invariant subspaces of a pencil: well defined up to

$$sE - A \sim s(ME) - MA \quad \text{“right-handed equivalence”}$$

Or: they depend on the subspace $\text{Im} \begin{bmatrix} E^* \\ A^* \end{bmatrix}$, not on the matrix $\begin{bmatrix} E^* \\ A^* \end{bmatrix}$

Results

Structured pencils

Symplectic pencils

$$E \mathcal{J}_{2n} E^* = A \mathcal{J}_{2n} A^* \quad \iff \quad \begin{bmatrix} E & A \end{bmatrix} \begin{bmatrix} \mathcal{J}_{2n} & 0 \\ 0 & -\mathcal{J}_{2n} \end{bmatrix} \begin{bmatrix} E^* \\ A^* \end{bmatrix} = 0$$

Hamiltonian pencils

$$E \mathcal{J}_{2n} A^* = -A \mathcal{J}_{2n} E^* \quad \iff \quad \begin{bmatrix} E & A \end{bmatrix} \begin{bmatrix} 0 & \mathcal{J}_{2n} \\ \mathcal{J}_{2n} & 0 \end{bmatrix} \begin{bmatrix} E^* \\ A^* \end{bmatrix} = 0$$

By exchanging blocks, we can transform the two matrices in red into \mathcal{J}_{4n}

Up to some reordering, $\text{Im} \begin{bmatrix} E^* \\ A^* \end{bmatrix}$ is **Lagrangian!**

Our theory can be used to give tame, structure-preserving representations. . .

Bounded representations of symplectic pencils

Theorem

Every symplectic pencil is (right-handed-)equivalent to one in the form

$$s \begin{bmatrix} I_n & X_{21} \\ 0 & X_{22} \end{bmatrix} \Pi_1 - \begin{bmatrix} X_{11} & 0 \\ X_{21} & I_n \end{bmatrix} \Pi_2$$

with Π_i symplectic swap matrices, $X = \begin{bmatrix} X_{11} & X_{12} \\ X_{21} & X_{22} \end{bmatrix}$ **bounded** Hermitian

(Without the parts in red, well known)

Bounded representations of Hamiltonian pencils

Theorem

Every Hamiltonian pencil is (right-handed-)equivalent to one in the form

$$s \begin{bmatrix} E_1 & E_2 \end{bmatrix} - \begin{bmatrix} A_1 & A_2 \end{bmatrix}$$

with

$$\begin{bmatrix} E_1 & A_1 \end{bmatrix} = \begin{bmatrix} I & X_{11} \\ 0 & X_{21} \end{bmatrix} \Pi_1, \quad \begin{bmatrix} -A_1 & E_2 \end{bmatrix} = \begin{bmatrix} X_{12} & 0 \\ X_{22} & I \end{bmatrix} \Pi_2,$$

and $X = \begin{bmatrix} X_{11} & X_{12} \\ X_{21} & X_{22} \end{bmatrix}$ **bounded** Hermitian

Idea: start from $sI - H$, you can swap vectors between the “outer” blocks and between the “inner” ones

Deflating R in general-form control problems

In control problems, originally an “extended” $(2n + m) \times (2n + m)$ pencil
 Need deflation to get a Hamiltonian matrix/pencil

$$s \begin{bmatrix} 0 & I & 0 \\ -I & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} - \begin{bmatrix} 0 & A & B \\ A^* & Q & S \\ B^* & S^* & R \end{bmatrix} \implies s \begin{bmatrix} I & 0 & 0 \\ 0 & I & 0 \\ 0 & 0 & 0 \end{bmatrix} - \begin{bmatrix} \mathcal{H} & 0 \\ * & * \\ * & * & I \end{bmatrix}$$

When R ill-conditioned or singular, **trouble**

Solution: allow identities to move!

$$s \begin{bmatrix} 0 & I & 0 \\ -I & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} - \begin{bmatrix} 0 & A & B \\ A^* & Q & S \\ B^* & S^* & R \end{bmatrix} \implies s \begin{bmatrix} \mathcal{H}_1 & 0 \\ * & * \\ * & * & 0 \end{bmatrix} - \begin{bmatrix} \mathcal{H}_2 & 0 \\ * & * \\ * & * & I \end{bmatrix}$$

$s\mathcal{H}_1 - \mathcal{H}_2$ Hamiltonian pencil (bounded as in the theorem)

Doubling algorithms – I

Doubling algorithms compute invariant subspaces of Hamiltonian / symplectic pencils

Key operation

Given $A - sE$, find full-rank $\begin{bmatrix} C & S \end{bmatrix}$ such that $\begin{bmatrix} C & S \end{bmatrix} \begin{bmatrix} A \\ E \end{bmatrix} = 0$

Two main strategies:

- QR-factorize $\begin{bmatrix} A \\ E \end{bmatrix}$ and construct C, S from Q
- permute and invert a block to reduce to $\begin{bmatrix} A \\ E \end{bmatrix} = \begin{bmatrix} I \\ X \end{bmatrix}$; then use

$$\begin{bmatrix} -X & I \end{bmatrix} \begin{bmatrix} I \\ X \end{bmatrix} = 0$$

Doubling algorithms – II

- QR-factorize $\begin{bmatrix} A \\ E \end{bmatrix}$: **inverse-free matrix sign/disc method** [Benner, '96] , [Benner, Byers '06] , [Bai et al., '97]
- enforce identity: **structure-preserving doubling algorithm** (SDA) [Anderson, '78] , [Chu et al., '04]

	QR-based	SDA
$O(n^3)$?	Yes	Yes
Structure-preserving?	No way!	Yes
Stable?	Yes	No way!

This looks familiar... again, it's hammer time!

An attempt at a new doubling algorithm

Doubling + permuted graph bases

- 1 Compute bounded permuted graph basis $\begin{bmatrix} E \\ A \end{bmatrix} = \tilde{\Pi} \begin{bmatrix} I \\ \tilde{X} \end{bmatrix}$
- 2 $\begin{bmatrix} C & S \end{bmatrix} = \begin{bmatrix} -\tilde{X} & I \end{bmatrix} \tilde{\Pi}^{-1}$
- 3 Use C, S to perform a doubling step
- 4 Compute bounded permuted graph basis of $\begin{bmatrix} E^* \\ A^* \end{bmatrix} = \Pi \begin{bmatrix} I \\ X \end{bmatrix}$
- 6 Repeat until convergence

An attempt at a new doubling algorithm

Doubling + permuted graph bases

- 1 Compute bounded permuted graph basis $\begin{bmatrix} E \\ A \end{bmatrix} = \tilde{\Pi} \begin{bmatrix} I \\ \tilde{X} \end{bmatrix}$
- 2 $[C \ S] = [-\tilde{X} \ I] \tilde{\Pi}^{-1}$ (**unstructured** version)
- 3 Use C, S to perform a doubling step
- 4 Compute bounded permuted graph basis of $\begin{bmatrix} E^* \\ A^* \end{bmatrix} = \Pi \begin{bmatrix} I \\ X \end{bmatrix}$
- 5 **Enforce Lagrangianity:** $X \leftarrow \frac{1}{2}(X + X^*)$
- 6 Repeat until convergence

Still **not 100% satisfactory**: $\begin{bmatrix} E \\ A \end{bmatrix}$ is not Lagrangian: switch to unstructured arithmetic and then project back

Looking for a compact version of 1–4 using only Hermitian arithmetic
Possible in the known special cases (SDA, Cyclic Reduction)

But still, great numerical results. . .

Figure: Relative subspace residual for the 33 CAREX problems in [Chu et al., '07]

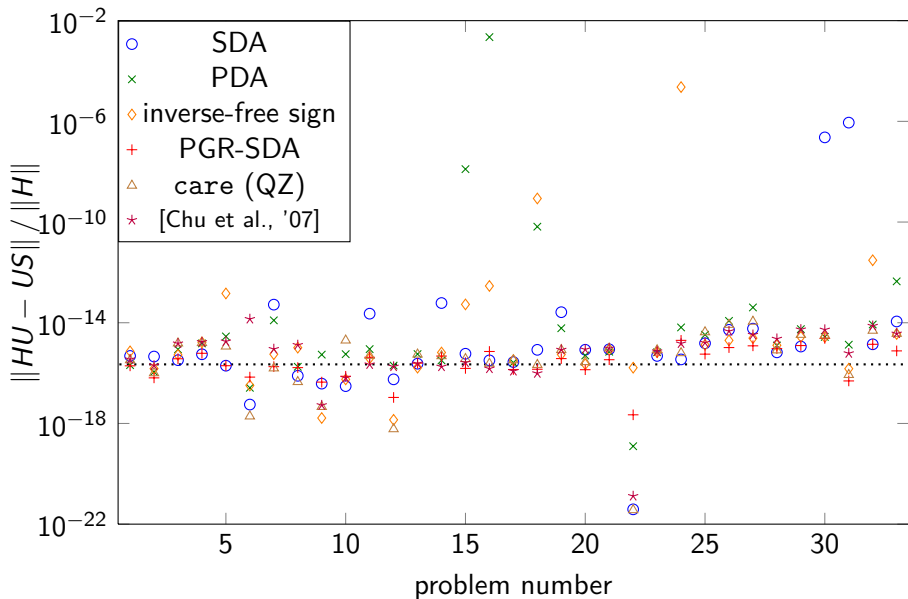
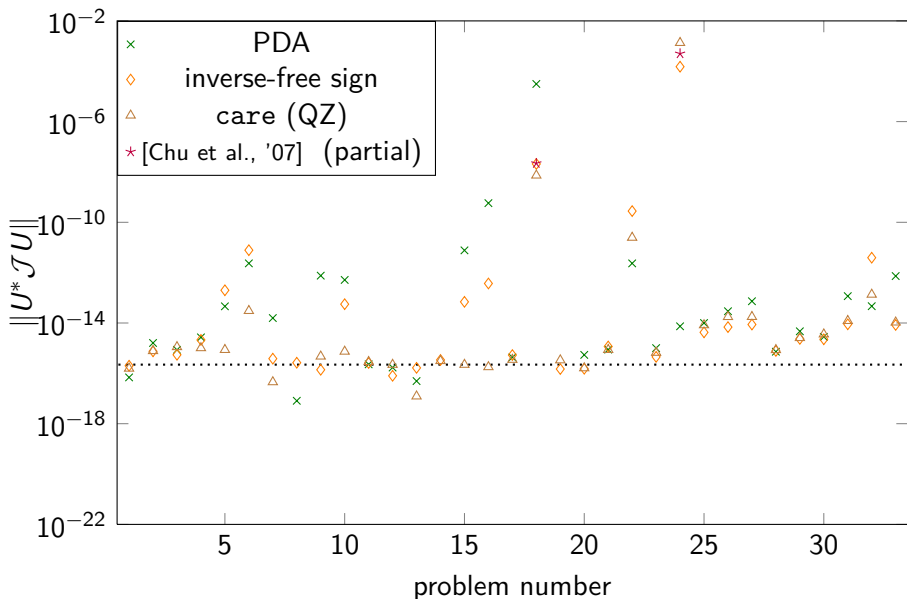


Figure: Lagrangianity residual for the 33 CAREX problems in [Chu et al., '07]



Now, stability analysis. . .

Stability results

- $\kappa(\Pi \begin{bmatrix} I \\ X \end{bmatrix}) \leq Cn$, with $\kappa(Z) = \sigma_{\max}(Z)/\sigma_{\min}(Z)$
- Given an initial basis U , can construct $\Pi \begin{bmatrix} I \\ ZY^{-1} \end{bmatrix}$ with $\kappa(Y) \leq Cn\kappa(U)$

Unfortunately, textbook backward stability analysis not well suited to doubling (or, more generally, matrix squaring):

$$\begin{array}{ccc} \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} & \xrightarrow{\text{squaring}} & \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} \\ & & \downarrow \text{perturb} \\ \color{red}{???) & \xrightarrow{\text{squaring}} & \begin{bmatrix} 0 & \varepsilon \\ 0 & 0 \end{bmatrix} \end{array}$$

Note: 0 **not** a critical eigenvalue here

How good is this in theory?

	QR-based	SDA	Permuted SDA
$O(n^3)$?	Yes	Yes	
Structure-preserving?	No way!	Yes	
Stable?	Yes	No way!	

How good is this in theory?

	QR-based	SDA	Permuted SDA
$O(n^3)$?	Yes	Yes	Kind of
Structure-preserving?	No way!	Yes	Kind of
Stable?	Yes	No way!	Kind of

- $O(n^3)$: Need to bound total number of row swaps
Well-studied in the unstructured case
In practice $2n$ row swaps (overall) suffice on all experiments
- **Structure-preserving**: Would like “fully Hermitian” update formula
We haven’t nailed it down yet. . .
- **Stable**: can mimic [Bai *et al.*, '97], but large worst-case constants

Turning those “kind of” into **yes** looks **possible** for the first time

Conclusions

- Doubling now competitive with state-of-the-art algorithms for dense control problems
(Matlab code soon to be released — contact me for info)
- Recipe to add stability to existing structure-preserving algorithms

Other possible “nail” applications:

- Large scale Riccati and ADI:
 X (approx) low rank \Rightarrow many determinants (approx) 0
- \mathcal{H}_∞ control: Riccati with unbounded solutions show up
- “Butterfly” SR/SZ algorithms
- Are you working with symplectic matrices?
Maybe your problem looks like a nail, too

Conclusions

- Doubling now competitive with state-of-the-art algorithms for dense control problems (Matlab)
- Recipe to

Thanks for your attention! Questions?



Other possibilities

- Large scale X (approx) low rank \Rightarrow many determinants (approx) 0
- \mathcal{H}_∞ control: Riccati with unbounded solutions show up
- “Butterfly” SR/SZ algorithms
- Are you working with symplectic matrices?
Maybe your problem looks like a nail, too

How does it compare to [Mehrmann, Schröder, Watkins '09] ?

	QR-based	SDA	Perm-SDA	MehSW
$O(n^3)$?	Yes	Yes	Kind of	Yes
Structure-preserving?	No way!	Yes	Kind of	Yes?*
Stable?	Yes	No way!	Kind of	Yes?*
BLAS3/Parallel/ Communication optimal?	Yes	Yes	Kind of	No way!

- *MehSW (essentially: block Schur + Laub trick on every block) uses orthogonal bases and can have the same problems as Laub trick
- Schur-type algorithms not suited for large communication optimal linear algebra — for instance, [Demmel et al., '06] use doubling instead of QR for eigenvalues

Figure: Riccati residual for the 33 CAREX problems in [Chu et al., '07]

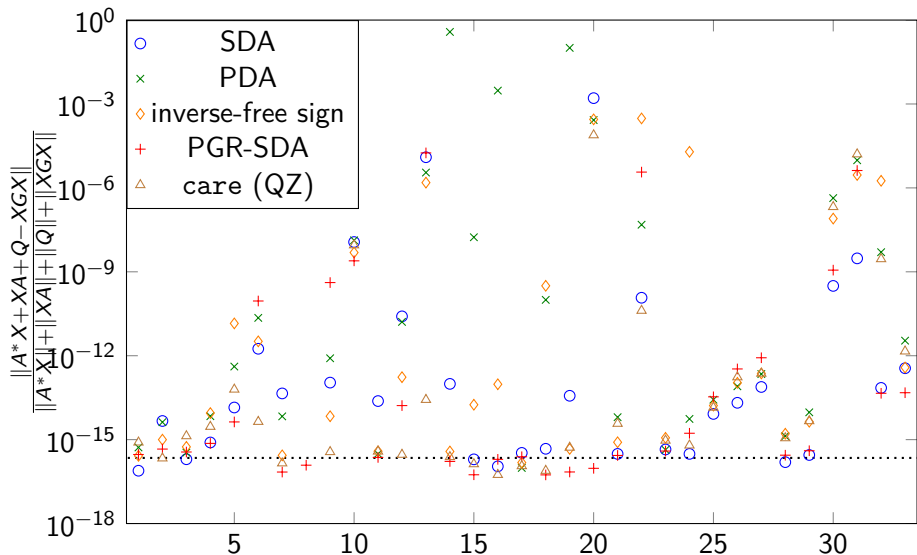


Figure: Unstructured pencil backward error for the problems in [Chu et al., '07]

