## Invariant subspace methods for CAREs

$X$ solves CARE $A^{*} X+X A+Q=X G X$ iff

$$
\left[\begin{array}{cc}
A & -G \\
-Q & -A^{T}
\end{array}\right]\left[\begin{array}{c}
I \\
X
\end{array}\right]=\left[\begin{array}{c}
I \\
X
\end{array}\right] \mathcal{R}, \quad \mathcal{R}=A-G X
$$

One can find $X$ through an invariant subspace of the Hamiltonian.
>> [A,G,Q] = carex(4) \%if test suite is installed
>> n = length(A);
>> $H=[A-G ;-Q-A)] ;$
>> [U, T] = schur(H);
>> [U, T] =ordschur(U, T, 'lhp');
>> $X=U(n+1: 2 * n, 1: n) / U(1: n, 1: n) ;$
People are not satisfied with this method though - it is not structured backward stable.

Eigenvalues close to the imaginary axis can be 'mixed up' - try carex (14) for instance.

## Symplectic transformations

Ideal setting: make transformations at each step that are orthogonal and symplectic, i.e., orthogonal w.r.t the scalar product $J=\left[\begin{array}{cc}0 & I \\ -I & 0\end{array}\right]$ : they satisfy $S^{T} J S=J$.
For instance:

- If $Q \in \mathbb{R}^{n \times n}$ is any orthogonal matrix, then $\operatorname{blkdiag}(Q, Q)$ is orthogonal and symplectic.
- A Givens matrix that acts on entries $k$ and $n+k$ (i.e., $\mathrm{G}=\operatorname{eye}(2 * \mathrm{n}) ; \mathrm{G}([\mathrm{k}, \mathrm{n}+\mathrm{k}],[\mathrm{k}, \mathrm{n}+\mathrm{k}])=[\mathrm{c} \mathrm{s} ;-\mathrm{s} \mathrm{c}] ;)$ is orthogonal and symplectic.

Laub trick: let $U=\left[\begin{array}{ll}U_{1} & U_{3} \\ U_{2} & U_{4}\end{array}\right]$ the unitary matrix produced by schur (H). Then, $\left[\begin{array}{l}U_{1} \\ U_{2}\end{array}\right]$ is an orthogonal matrix that spans the stable subspace. We know that $-J\left[\begin{array}{l}U_{1} \\ U_{2}\end{array}\right]=\left[\begin{array}{c}U_{2} \\ -U_{1}\end{array}\right]$ is orthogonal to it (and spans the left unstable invariant subspace).
It turns out that $V=\left[\begin{array}{cc}U_{1} & U_{2} \\ U_{2} & -U_{1}\end{array}\right]$ is orthogonal and symplectic,
and $V^{T} \mathcal{H} V=\left[\begin{array}{cc}R & S \\ 0 & -R^{*}\end{array}\right]$, with $R$ upper triangular and $S$ symmetric (Hamiltonian Schur form).

## An orthogonal symplectic algorithm

This produces the same subspace as the previous method, so it is not really a 'structured' method. Can one do a 'symplectic QR' and compute the Hamiltonian Schur form using a sequence of orthosymplectic transformations?

Open problem for a while; it turns out that that Schur form does not exist for all Hamiltonian matrices (there are counterexamples with eigenvalues on the unit circle). $\Longrightarrow$ algorithms must be unstable 'nearby'.
(This problem was known as Van Loan's curse.)

## Chu-Liu-Mehrmann algorithm

Closest thing to a solution: Chu-Liu-Mehrmann algorithm. Based on a different decomposition: $\mathcal{H}=U R V^{\top}$, with $U, V$ orthosymplectic and

$$
R=\left[\begin{array}{cc}
R_{11} & R_{12} \\
0 & R_{22}
\end{array}\right]
$$

with $R_{11}, R_{22}^{*}$ upper triangular. Can be computed 'almost' directly in $O\left(n^{3}\right)$ (it's an LU-like decomposition).

## URV — simpler version (produces Hessenberg $R_{22}$ )

- Left-multiply by blkdiag(Q, Q) to get
- Left-multiply by a Givens on (1, $n+1$ ) to get $* * * * * *$
$0 * * * * *$
$0 * * * * *$ $0 * * * * *$
- Left-multiply by blkdiag( $\mathrm{Q}, \mathrm{Q}$ ) to get





## Using URV

Note that $\mathcal{H}=U R V+$ symplecticity implies

$$
\mathcal{H}=V\left[\begin{array}{cc}
-R_{22}^{T} & R_{12}^{T} \\
0 & -R_{11}^{T}
\end{array}\right] U^{T} .
$$

Hence

$$
\mathcal{H}^{2}=V\left[\begin{array}{cc}
-R_{11} R_{22}^{T} & * \\
0 & -R_{22} R_{11}^{T}
\end{array}\right] V^{T}
$$

can be used to compute eigenvalues (easily) and eigenvectors of $\mathcal{H}$ (for instance: the columns of $V$ cause breakdown at step 2 in Arnoldi).

