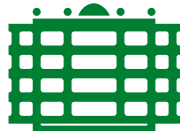


# New Hamiltonian Eigensolvers with Applications in Control

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## Outline

- (Skew-)Hamiltonian Eigenproblems
- Algorithms for (Skew-)Hamiltonian Eigenproblems
- HAPACK
- Applications
- Conclusions

## (Skew-)Hamiltonian Matrices

$$\begin{aligned} \text{Hamiltonian: } \mathcal{H} &:= \left\{ \begin{bmatrix} A & G \\ Q & -A^T \end{bmatrix} \in \mathbb{R}^{2n,2n}, \quad G, Q \text{ symmetric} \right\} \\ \text{skew-Hamiltonian: } \mathcal{SH} &:= \left\{ \begin{bmatrix} A & G \\ Q & A^T \end{bmatrix} \in \mathbb{R}^{2n,2n}, \quad G, Q \text{ skew-symmetric} \right\} \end{aligned}$$

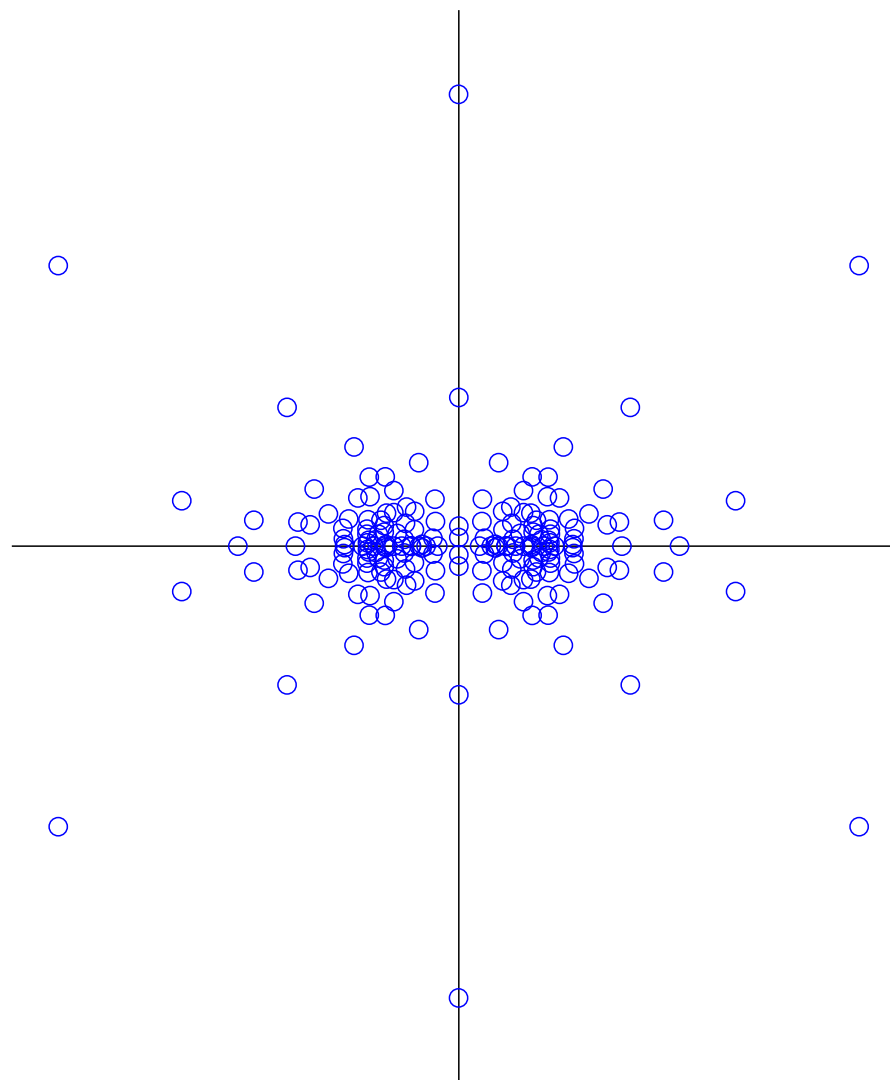
Fundamental properties:

$$J = \begin{bmatrix} 0 & I_n \\ -I_n & 0 \end{bmatrix} \Rightarrow \begin{aligned} \mathcal{H} &= \{H : (JH)^T = JH\} \\ \mathcal{SH} &= \{W : (JW)^T = -JW\} \end{aligned}$$

$$\mathcal{H}^2 = \mathcal{SH}$$

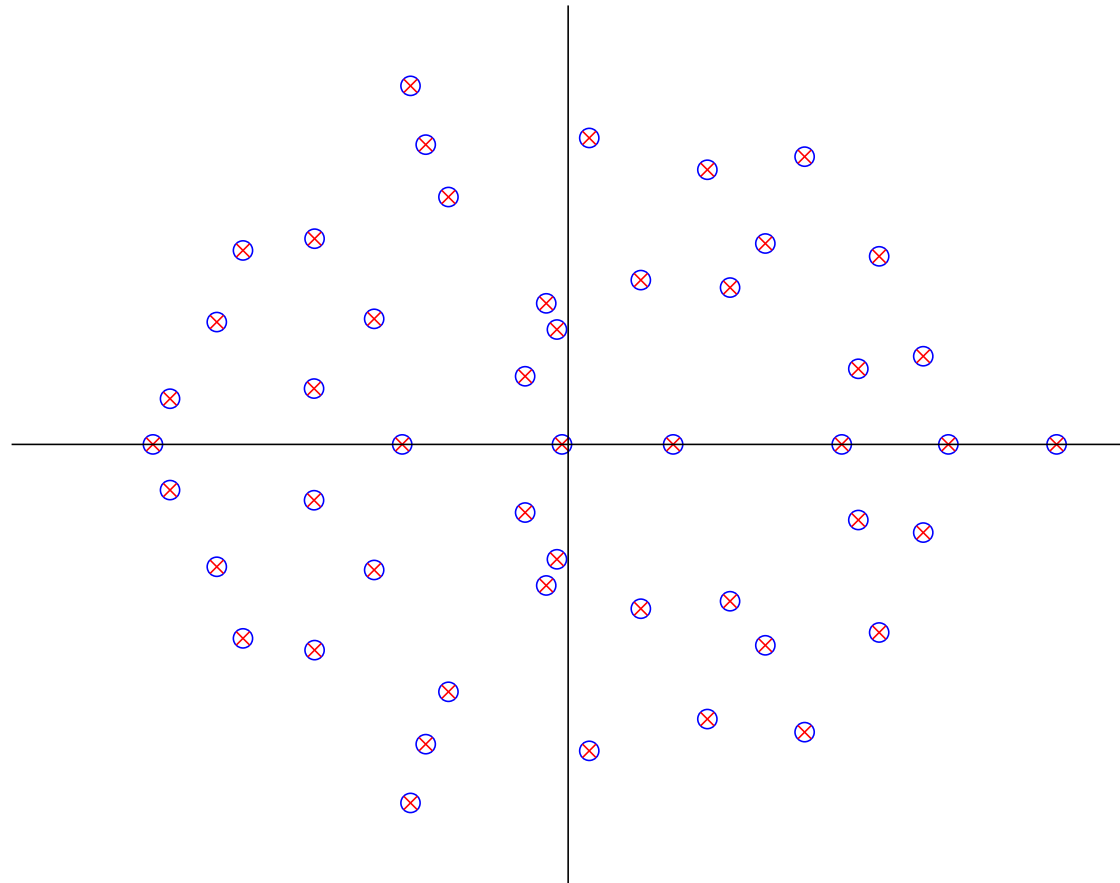
# Spectra of Hamiltonian Matrices

$$\lambda \in \text{spec}(H) \implies \bar{\lambda}, -\lambda, -\bar{\lambda} \in \text{spec}(H)$$



# Spectra of Skew-Hamiltonian Matrices

$$\lambda \in \text{spec}(W) \implies \bar{\lambda}, \lambda, \bar{\lambda} \in \text{spec}(W)$$



## Applications

- LQR/LQG control, solution of algebraic Riccati equations
- Computation of stability radius and  $H_\infty/L_\infty$  norm
- $H_\infty$  control
- Gyroscopic systems, quadratic eigenvalue problems
- Model reduction: stochastic balanced truncation, truncation based on positive-real or bounded-real balancing, LQG balancing.

Preservation of spectral symmetries is essential for all applications!

General-purpose algorithms (QR, Jacobi, Jacobi-Davidson, Arnoldi, ...) do not preserve symmetries in finite precision arithmetic.

## Structure-Preserving Algorithms

Advantages of exploiting the structure in numerical algorithms:

- Faster computations as complexity is reduced.
- Structured condition numbers (sometimes) smaller than unstructured ones.
- (Sometimes) more accurate results.
- **Physically meaningful results.** Unstructured algorithms lead to non-physical results, e.g.
  - applying eig (QR algorithm) to Hamiltonian matrices may yield  $n + k$  stable and  $k$  unstable computed eigenvalues, where  $1 \leq k \leq n$ ,
  - purely imaginary eigenvalues of Hamiltonian matrices cannot be detected in general ( $\rightarrow$  stability radius,  $H_\infty$ -norm, . . . )
- More fun.

## Structured Algorithm for Skew-Hamiltonian Matrices (Van Loan '84)

**Based on:** If  $U$  is orthogonal ( $U^T U = I$ ) and symplectic ( $U^T J U = J$ )  
 $\Rightarrow U^T W U$  is skew-Hamiltonian.

1. Compute orthogonal symplectic  $U$  s.t.

$$U^T W U = \begin{bmatrix} W_{11} & W_{12} \\ 0 & W_{11}^T \end{bmatrix} = \begin{bmatrix} \square_{\text{diag}} & \square \\ \square & \square_{\text{diag}} \end{bmatrix}.$$

2. Apply standard QR algorithm to  $W_{11}$ .



## Structured Algorithm for Hamiltonian Matrices (B./Mehrmann/Xu'97)

**Based on:**  $H^2$  is skew-Hamiltonian, but  $H^2$  is not formed explicitly.

If  $U, V$  are orthogonal symplectic, then

$$U^T H^2 U = (U^T H V)(V^T H U) = U^T H V J (U^T H V)^T J$$

$\Rightarrow U^T H V$  can be used to reveal eigenvalues of  $H$ .

1. Compute orthogonal symplectic  $U, V$  s.t.

$$U^T H V =: R = \begin{bmatrix} R_{11} & R_{12} \\ 0 & -R_{22}^T \end{bmatrix} = \begin{bmatrix} \begin{array}{|c|} \hline \triangle \\ \hline \end{array} & \begin{array}{|c|} \hline \square \\ \hline \end{array} \\ 0 & \begin{array}{|c|} \hline \square \\ \hline \triangle \\ \hline \end{array} \end{bmatrix}.$$

2. Apply periodic QR to  $R_{11} \cdot R_{22} \Rightarrow \Lambda(R_{11} R_{22}) = \Lambda(H)^2$ .

## HAPACK Routines for Skew-Hamiltonian Matrices

FORTRAN 77 routines:

DSHBAL	symplectic balancing
DSHPVB	blocked reduction to PVL form (step 1)
DSHSEN	specified invariant subspaces
DSHSNA	structured eigenvalue and eigenvector condition numbers/estimates
DSH***	various driver routines

MATLAB routines (based on MEX interfaces):

shbal	symplectic scaling and balancing
sheig	eigenvalues and eigenvectors
shschur	skew-Hamiltonian Schur form
shinv	specified invariant subspaces
shcondeig	structured eigenvalue and eigenvector condition numbers/estimates

## HAPACK Routines for Hamiltonian Matrices

FORTRAN 77 routines:

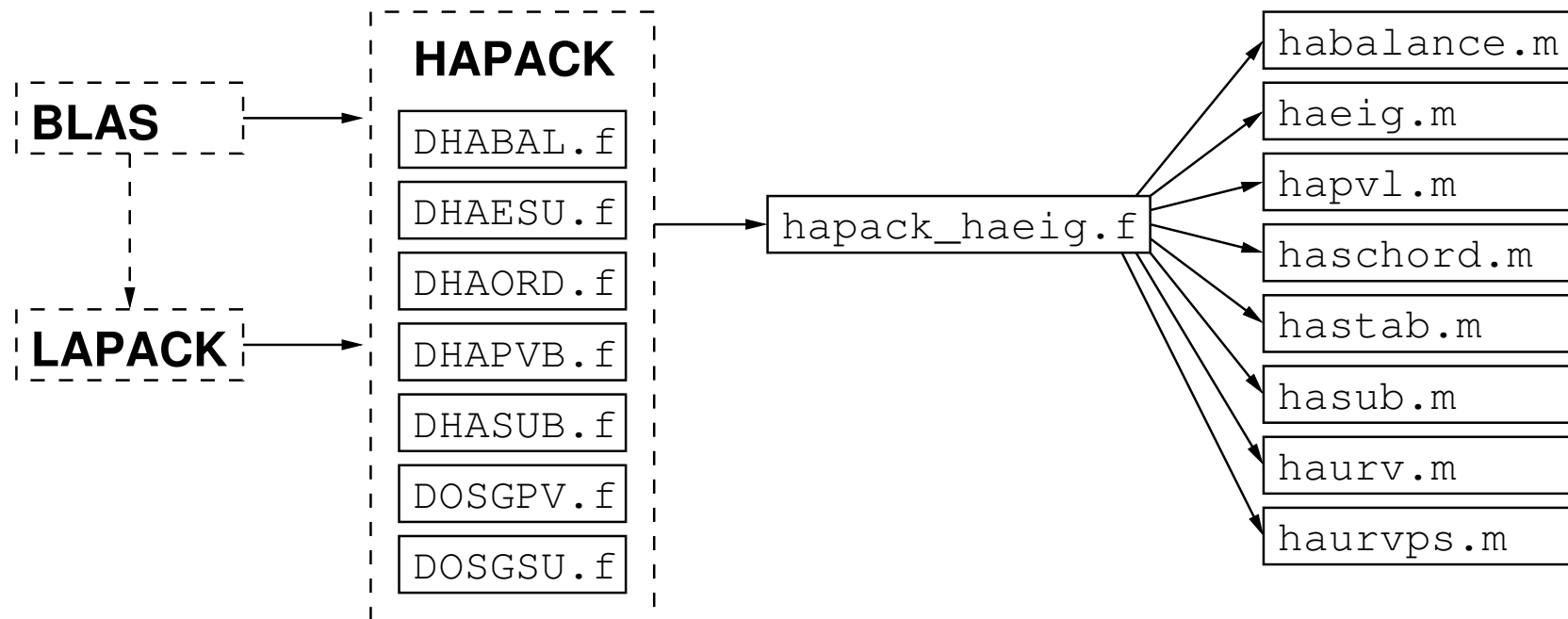
DHABAL	symplectic balancing
DHAPVB	blocked reduction to PVL form
DHAESU	blocked reduction to symplectic URV form (step 1)
DHGPQR	multi-shift periodic QR for $A \cdot B$ (step 2)
DHAORD	reorder the Hamiltonian Schur decomposition
DHASUB	specified invariant subspaces

MATLAB routines (based on MEX interfaces):

habalance	symplectic balancing
haeig	eigenvalues (supports complex Hamiltonian matrices)
hapvl	PVL decomposition
haschord	reorders Hamiltonian Schur form
hastab	stable/unstable subspaces
hasub	selected stable/unstable subspaces
haurv	symplectic URV decomposition
haurvps	symplectic URV/periodic Schur decomposition of a general matrix

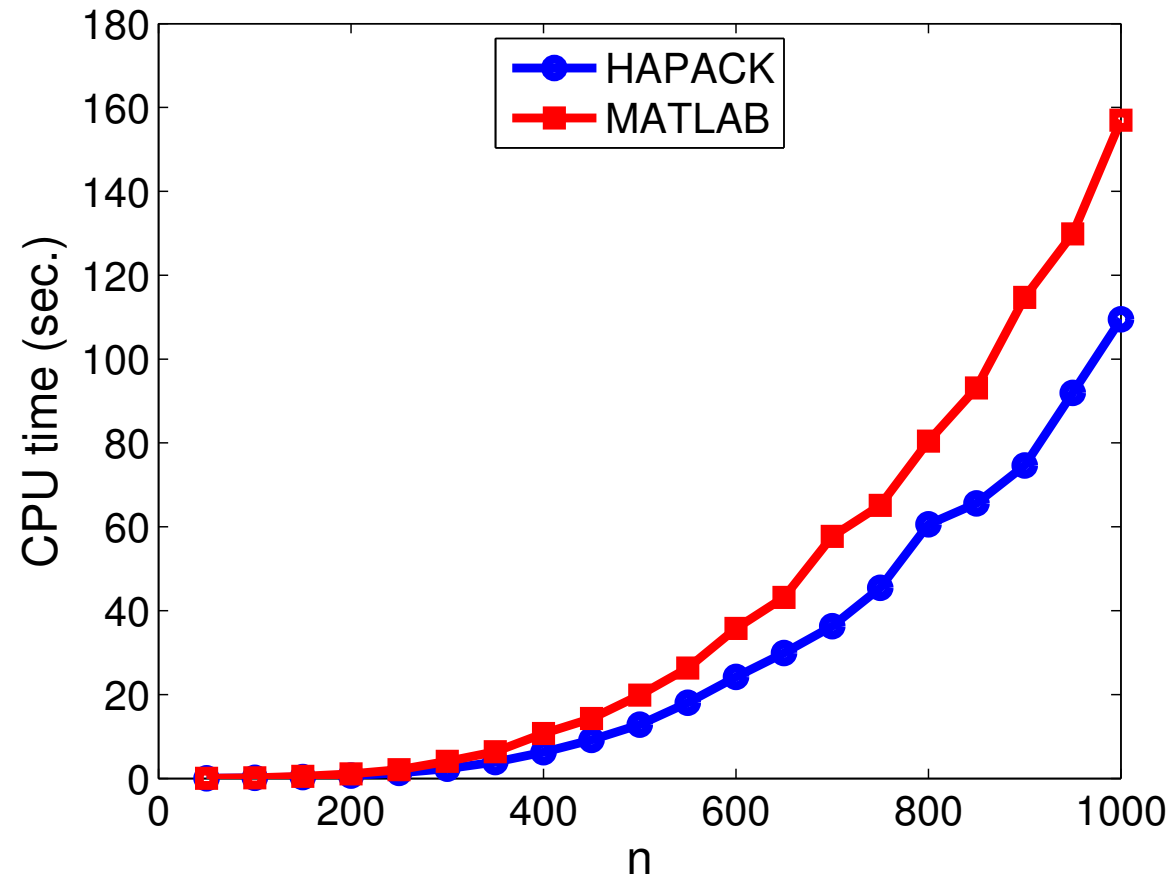
## Structure of HAPACK

HAPACK MATLAB functions are called via the MEX gateway `hapack_haeig.f`.



## Performance

Random Hamiltonian matrices, CPU times for `haeig` (HAPACK) and `eig` (MATLAB).



# Applications

Linear continuous-time systems with constant coefficients in state-space form,

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t), \\ y(t) &= Cx(t) + Du(t).\end{aligned}$$

Related issues:

1. algebraic Riccati equations
2. LQ(R,G) control (see paper)
3. stability radius computation
4. model reduction

## Stability Radius Computation

Stability radius of  $A$ :  $\gamma(A) := \min\{\|E\|_2 : \lambda(A + E) \cap i\mathbb{R} \neq \emptyset\}$ .

Bisection method for computing  $\gamma(A)$  can be based on the following observation:  
[Byers '88, Hinrichsen/Pritchard '86]

if  $\alpha \geq 0$ , then the Hamiltonian matrix

$$H(\alpha) = \begin{bmatrix} A & -\alpha I_n \\ \alpha I_n & -A^T \end{bmatrix}$$

has an eigenvalue on the imaginary axis if and only if  $\alpha \geq \gamma(A)$ .

## HAPACK Implementation

For lower and upper bounds  $\beta \geq 0$  and  $\delta > \gamma(A)$ , obtain HAPACK implementation:

```
nA = norm(A+A', 'fro') / 2;
beta = 0;
delta = nA;
e = ones(n,1);
QG = full(spdiags([e -e], 0:1, n, n+1));
while (delta-beta) > 100*eps*nA,
    e = haeig(A, (beta + delta)/2 * QG);
    if isempty( find(real(e)==0) ),
        beta = (beta + delta)/2;
    else
        delta = (beta + delta)/2;
    end
end
end
```

Quadratically convergent variants [Bruinsma/Steinbuch '90, Boyd/Balakrishnan '90] can also be based on HAPACK, will be available in SLICOT!



## Example: The Demmel Matrix

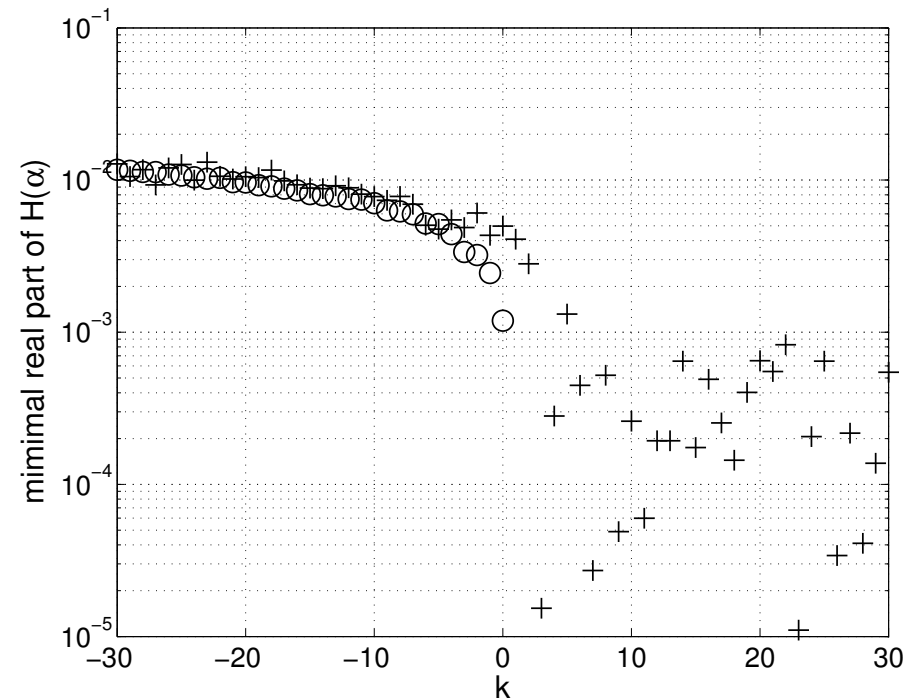
$$A = Q^T \begin{bmatrix} -0.05 & -10 & -10^2 & -10^3 & -10^4 \\ 0 & -0.05 & -10 & -10^2 & -10^3 \\ 0 & 0 & -0.05 & -10 & -10^2 \\ 0 & 0 & 0 & -0.05 & -10 \\ 0 & 0 & 0 & 0 & -0.05 \end{bmatrix} Q, \quad Q \text{ orthogonal, randomly generated.}$$

$$\gamma(A) \approx 3.6 \times 10^{-11}!$$

Right: minimum absolute values for real parts of  $\text{spec}(H(\alpha))$ , computed by `eig` ('+') and `haeig` ('o').

$$\alpha = 1.005^k \times \gamma(A), \quad k \in [-30, 30].$$

For  $k > 0$ , `haeig` computes exact zeroes which are not shown in the figure.



## Model Reduction

Passivity-preserving model reduction for linear systems can be based on truncating a positive real balanced realization, obtained from balancing the Gramians given as solutions of the dual algebraic Riccati equations

$$\begin{aligned}0 &= \hat{A}P + P\hat{A}^T + PC^T R^{-1}CP + BR^{-1}B^T, \\0 &= \hat{A}^T Q + Q\hat{A} + QBR^{-1}B^T Q + C^T R^{-1}C,\end{aligned}$$

where

$$\begin{aligned}R &= D + D^T, \\ \hat{A} &= A - BR^{-1}C.\end{aligned}$$

## HAPACK Implementation

Solving positive real AREs with HAPACK requires only **one** call to `hastab` as  $P, Q$  can be obtained from stable and anti-stable invariant subspaces:

```
R = chol(D+D');  
B = B / R;  
C = R' \ C;  
A = A - B*C;  
H = [ A   -B*B'; C'*C  -A'];  
[X,Y,e] = hastab(H);  
P = -Y(1:n,:) / Y(n+1:2*n,:);  
Q = -X(n+1:2*n,:) / X(1:n,:);
```

## Conclusions

- HAPACK provides efficient Fortran 77 and MATLAB routines for Hamiltonian eigenproblems and related control problems.
- In contrast to other MATLAB software for the considered control problems, HAPACK provides reliable and physically meaningful results.
- Software is freely available at

[www.tu-chemnitz.de/mathematik/hapack/](http://www.tu-chemnitz.de/mathematik/hapack/)

and will (partially) be included in SLICOT ([www.slicot.net](http://www.slicot.net)).

- Most HAPACK routines are described in [Algorithm 8xx: Fortran 77 Subroutines for Computing the Eigenvalues of Hamiltonian Matrices II](#)  
PETER BENNER, DANIEL KRESSNER, *ACM Transactions on Mathematical Software*, to appear.

... Thanks for your attention!