

# An Algorithm for the Complete Solution of Quadratic Eigenvalue Problems

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Joint work with **Sven Hammarling** and  
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# Quadratic Eigenvalue Problem (QEP)

Consider  $Q(\lambda) = \lambda^2 M + \lambda D + K \in \mathbb{C}[\lambda]^{n \times n}$ .

Assume  $Q(\lambda)$  regular ( $\det(Q(\lambda)) \neq 0$ ).

QEP: find nonzero vectors  $x, y$  and scalars  $\lambda \in \mathbb{C}$  s.t.

$$Q(\lambda)x = 0, \quad y^* Q(\lambda) = 0.$$

- ▶  $Q(\lambda)$  has  **$2n$  e'vals**. Finite e'vals are roots of  $\det(Q(\lambda)) = 0$ .
- ▶ E'val at 0 when  $K$  is singular and e'val at  $\infty$  when  $M$  is singular.

# Example 1

$$Q(\lambda) = \lambda^2 \begin{bmatrix} 0 & 8 & 0 \\ 0 & 6 & 0 \\ 0 & 0 & 1 \end{bmatrix} + \lambda \begin{bmatrix} 1 & -6 & 0 \\ 2 & -7 & 0 \\ 0 & 0 & 0 \end{bmatrix} + \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}.$$

**Regular:**  $\det Q(\lambda) = -6\lambda^5 + 11\lambda^4 - 12\lambda^3 + 12\lambda^2 - 6\lambda + 1 \neq 0$ .

**Six eigenpairs**  $(\lambda_k, x_k)$ ,  $k = 1 : 6$ , given by

$k$	1	2	3	4	5	6
$\lambda_k$	1/3	1/2	1	$i$	$-i$	$\infty$
$x_k$	$\begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}$	$\begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}$	$\begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$	$\begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$	$\begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$	$\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$

Collection of Nonlinear Eigenvalue Problems : T. Betcke, N. J. Higham, V. Mehrmann, C. Schröder, F. T., 2011.

- ▶ Quadratic, polynomial, rational and other nonlinear eigenproblems.
- ▶ Provided in the form of a MATLAB Toolbox. Compatible with GNU Octave.
- ▶ Problems from real-life applications + specifically constructed problems.

<http://www.mims.manchester.ac.uk/research/numerical-analysis/nlevp.html>

# Sample of Quadratic Problems

$n \times m$  quadratic  $Q(\lambda) = \lambda^2 M + \lambda D + K$ .

**Speaker box** (pep, qep, real, symmetric).

$n = m = 107$ . Finite element model of a speaker box.

$\|M\|_2 = 1$ ,  $\|D\|_2 = 5.7 \times 10^{-2}$ ,  $\|K\|_2 = 1.0 \times 10^7$ .

**Railtrack** (pep, qep, t-palindromic, sparse).

$n = m = 1005$ . Model of vibration of rail tracks under the excitation of high speed trains.  $M = K^T$ ,  $D = D^T$ .

**Surveillance** (pep, qep, real, nonsquare,

nonregular).  $n = 21$ ,  $m = 16$ . From calibration of surveillance camera using human body as calibration target.

# Standard Solution Process

Find all  $\lambda$  and  $x$  satisfying  $Q(\lambda)x = (\lambda^2 M + \lambda D + K)x = 0$ .

▶ Commonly solved by linearization:

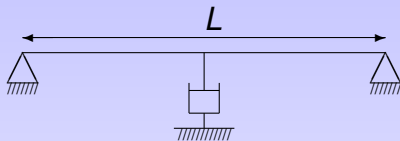
- **Convert**  $Q(\lambda)x = 0$  into  $(\mathcal{A} - \lambda\mathcal{B})\xi = 0$ , e.g.,

$$\mathcal{A} - \lambda\mathcal{B} = \begin{bmatrix} K & 0 \\ 0 & I \end{bmatrix} - \lambda \begin{bmatrix} -D & -M \\ I & 0 \end{bmatrix}, \quad \xi = \begin{bmatrix} x \\ \lambda x \end{bmatrix}.$$

- **Solve**  $(\mathcal{A} - \lambda\mathcal{B})\xi = 0$  with an eigensolver for generalized eigenproblem (e.g., QZ algorithm).
- **Recover** eigenvectors of  $Q(\lambda)$  from those of  $\mathcal{A} - \lambda\mathcal{B}$ .

▶ Eigensolver often **absent from numerical libraries**.

# Beam Problem



- ▶ Transverse displacement  $u(x, t)$  governed by

$$\rho A \frac{\partial^2 u}{\partial t^2} + c(x) \frac{\partial u}{\partial t} + EI \frac{\partial^4 u}{\partial x^4} = 0.$$

$$u(0, t) = u''(0, t) = u(L, t) = u''(L, t) = 0.$$

- ▶ Separation of variables  $u(x, t) = e^{\lambda t} v(x, \lambda)$  yields the **eigenvalue problem for the free vibrations:**

$$\lambda^2 \rho A v(x, \lambda) + \lambda c(x) v(x, \lambda) + EI \frac{\partial^4 v}{\partial x^4}(x, \lambda) = 0.$$

# Discretized Beam Problem

Finite element method leads to

$$Q(\lambda) = \lambda^2 M + \lambda D + K$$

with **symmetric**  $M, D, K \in \mathbb{R}^{n \times n}$ .  $M > 0$ ,  $K > 0$ ,  $D \geq 0$ .

Roots of  $x^* Q(\lambda) x = 0$ ,  $x \in \mathbb{C}^n \setminus \{0\}$ ,

$$\lambda = \frac{-(x^* D x) \pm \sqrt{(x^* D x)^2 - 4(x^* M x)(x^* K x)}}{2(x^* M x)}.$$

- ▶  $M > 0$ ,  $K > 0$ ,  $D \geq 0 \Rightarrow$  all e'vals have  $\operatorname{Re}(\lambda) \leq 0$ .
- ▶  $D$  is rank 1. Can show  $n$  pure imaginary e'vals.



# Eigenvalues of $Q(\lambda) = \lambda^2 M + \lambda D + K$

When  $M, K$  are nonsingular then theoretically

$$C_1(\lambda) = \lambda \begin{bmatrix} M & 0 \\ 0 & I \end{bmatrix} + \begin{bmatrix} D & K \\ -I & 0 \end{bmatrix},$$

$$L_1(\lambda) = \lambda \begin{bmatrix} M & 0 \\ 0 & -K \end{bmatrix} + \begin{bmatrix} D & K \\ K & 0 \end{bmatrix}, \quad L_2(\lambda) = \lambda \begin{bmatrix} 0 & M \\ M & D \end{bmatrix} + \begin{bmatrix} -M & 0 \\ 0 & K \end{bmatrix}$$

have the same eigenvalues as  $Q(\lambda)$ .

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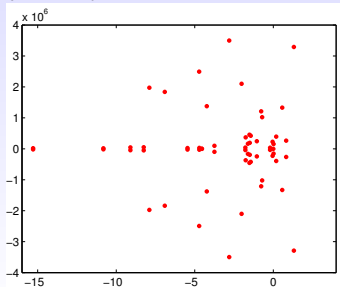
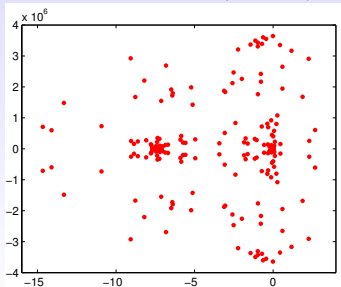
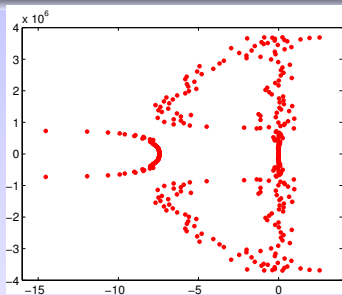
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```
coeffs = nlevp('damped_beam', 100);  
K = coeffs{1}; D = coeffs{2}; M = coeffs{3};  
I = eye(2*nele); O = zeros(2*nele);  
eval = eig([D K; -I O], -[M O; O I]; %  $C_1$   
%eval = eig([D K; K O], -[M O; O -K]; %  $L_1$   
%eval = eig([-M O; O K], -[O M; M D]; %  $L_2$   
plot(eval, '.r');
```

# Computed Spectra of $C_1$ , $L_1$ and $L_2$



# Sensitivity and Stability of Linearizations

- **Condition number** measures sensitivity of the solution of a problem to perturbations in the data.
- **Backward error** measures how well the problem has been solved.

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For a given  $Q(\lambda)$ , infinitely many linearizations exist:

- ▶ can have **widely varying eigenvalue condition numbers**,
- ▶ computed eigenpairs can have **widely varying backward errors**.

# Objectives

- ▶ To design a general purpose eigensolver for dense QEPs—**quadeig**.
- ▶ Incorporate:
  - Appropriate choice of linearization.
  - Deflation of 0 and  $\infty$  eigenvalues.
  - Eigenvalue parameter scaling.
  - Advantageous use of block structure.
  - Careful recovery of the eigenvectors.
- ▶ A MATLAB and Fortran implementation.

# Linearization for $Q(\lambda) = \lambda^2 M + \lambda D + K$

Opt for  $C(\lambda) = \begin{bmatrix} D & -I \\ K & 0 \end{bmatrix} + \lambda \begin{bmatrix} M & 0 \\ 0 & I \end{bmatrix}$  **(companion form, Fiedler pencil)**

- ▶ Simple block structure.
- ▶ Always a linearization, i.e., there exist  $E(\lambda)$  and  $F(\lambda)$  with constant, nonzero determinants s.t.

$$\begin{bmatrix} Q(\lambda) & 0 \\ 0 & I \end{bmatrix} = E(\lambda)C(\lambda)F(\lambda).$$

- ▶ Left/right e'vecs of  $Q(\lambda)$  are easily recovered from those of companion linearizations.
- ▶ Deflation of 0 and  $\infty$  e'vals easy to implement.
- ▶ "Good" backward error and conditioning properties.

# Eigenvalue Condition Numbers $\kappa_Q(\lambda)$

$$Q(\lambda)x = 0, \quad y^*Q(\lambda) = 0, \quad \Delta Q(\lambda) = \lambda^2\Delta M + \lambda\Delta D + \Delta K.$$

For  $\lambda$  **simple, nonzero and finite**,

$$\kappa_Q(\lambda) = \limsup_{\epsilon \rightarrow 0} \left\{ \frac{|\Delta\lambda|}{\epsilon|\lambda|} : \left[ (Q + \Delta Q)(\lambda + \Delta\lambda) \right] (x + \Delta x) = 0, \right. \\ \left. \|\Delta M\|_2 \leq \epsilon\|M\|_2, \|\Delta D\|_2 \leq \epsilon\|D\|_2, \|\Delta K\|_2 \leq \epsilon\|K\|_2 \right\}.$$



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Can show that [T. 01]

$$\kappa_Q(\lambda) = \frac{(|\lambda|^2\|M\|_2 + |\lambda|\|D\|_2 + \|K\|_2)\|y\|_2\|x\|_2}{|\lambda|y^*(2\lambda M + D)x}.$$

When  $\lambda = 0$  or  $\infty$ , use **homogeneous form** of  $Q$ .

# Eigenvalue Condition Numbers $\kappa_Q(\alpha, \beta)$

**Homogeneous form:**  $Q(\alpha, \beta) = \alpha^2 M + \alpha\beta D + \beta^2 K$ .

E'vals are pairs  $(\alpha, \beta) \neq (0, 0)$  s.t.  $\det Q(\alpha, \beta) = 0$ ,  $\lambda = \alpha/\beta$ .

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For  $(\alpha, \beta)$  simple with right/left e'vecs  $x, y$ , [Dedieu, T., 03]

$$\kappa_Q(\alpha, \beta) = \frac{(|\alpha|^4 \|M\|_2^2 + |\alpha|^2 |\beta|^2 \|D\|_2^2 + |\beta|^4 \|K\|_2^2)^{1/2} \|y\|_2 \|x\|_2}{|y^*(\bar{\beta} \mathcal{D}_\alpha Q - \bar{\alpha} \mathcal{D}_\beta Q)|_{(\alpha, \beta) x}}.$$

Angle between original and perturbed eigenvalues satisfies

$$|\theta((\alpha, \beta), (\tilde{\alpha}, \tilde{\beta}))| \leq \kappa_Q(\alpha, \beta) \|\Delta Q\| + o(\|\Delta Q\|).$$

Here  $\Delta Q = (\Delta M, \Delta D, \Delta K)$ .

Similar expression for  $\kappa_C(\alpha, \beta)$ ,  $C(\lambda) = \begin{bmatrix} D & -I \\ K & 0 \end{bmatrix} + \lambda \begin{bmatrix} M & 0 \\ 0 & I \end{bmatrix}$ .

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Want  $\kappa_Q(\alpha, \beta) \approx \kappa_C(\alpha, \beta)$  for all  $(\alpha, \beta)$ .

# Backward Error $\eta_Q(x, \alpha, \beta)$

For an approximate (right) eigenpair  $(x, \alpha, \beta)$  of  $Q(\alpha, \beta)$ ,

$$\eta_Q(x, \alpha, \beta) = \min \{ \epsilon : (Q(\alpha, \beta) + \Delta Q(\alpha, \beta))x = 0, \\ \|\Delta M\|_2 \leq \epsilon \|M\|_2, \|\Delta D\|_2 \leq \epsilon \|D\|_2, \|\Delta K\|_2 \leq \epsilon \|K\|_2 \},$$

where  $\Delta Q(\alpha, \beta) = \alpha^2 \Delta M + \alpha \beta \Delta D + \beta^2 \Delta K$ .

Can show that [T.01]

$$\eta_Q(x, \alpha, \beta) = \frac{\|Q(\alpha, \beta)x\|_2}{(|\alpha|^2 \|M\|_2 + |\alpha| |\beta| \|D\|_2 + |\beta|^2 \|K\|_2) \|x\|_2}.$$

Similar expression for  $\eta_C(z, \alpha, \beta)$ ,  $C(\lambda) = \begin{bmatrix} D & -I \\ K & 0 \end{bmatrix} + \lambda \begin{bmatrix} M & 0 \\ 0 & I \end{bmatrix}$ .

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Want  $\eta_Q(z_1, \alpha, \beta) \approx \eta_C(z, \alpha, \beta)$ ,  $z = \begin{bmatrix} z_1 \\ z_2 \end{bmatrix}$ .

# Eigenvalue Parameter Scaling

Let  $\lambda = \mu\gamma$ ,  $\gamma \neq 0$  and convert  $Q(\lambda) = \lambda^2 M + \lambda D + K$  to

$$\delta Q(\mu\gamma) = \mu^2(\gamma^2 \delta M) + \mu(\gamma \delta D) + \delta K = \mu^2 \tilde{M} + \mu \tilde{D} + \tilde{K} =: \tilde{Q}(\mu).$$

Choose  $\gamma$  such that

- the linearization process does not affect the eigenvalue condition numbers, i.e.,  $\kappa_Q(\alpha, \beta) \approx \kappa_C(\alpha, \beta)$  for all e'vals  $(\alpha, \beta)$ .
- the standard solution process is numerically stable, i.e.,  $\eta_Q(z_1, \alpha, \beta) \approx \eta_C(z, \alpha, \beta)$  for all e'vals  $(\alpha, \beta)$ , where  $z = \begin{bmatrix} z_1 \\ z_2 \end{bmatrix}$ .

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Try  $\gamma = \exp(r)$ , where  $r$  is a **tropical root** of a **tropical scalar quadratic** (proposed by **Gaubert & Sharify, 09**).



# Tropical Scalar Polynomials

- Let  $(\mathbb{R} \cup \{-\infty\}, \oplus, \otimes)$  be the **tropical semiring** with
$$a \oplus b = \max(a, b), \quad a \otimes b = a + b \quad \text{for all } a, b \in \mathbb{R} \cup \{-\infty\}.$$
- The piecewise affine function

$$p(x) = \bigoplus_{k=0}^d p_k \otimes x^{\otimes k} = \max_{0 \leq k \leq d} (p_k + kx), \quad p_k \in \mathbb{R} \cup \{-\infty\}$$

is a **tropical polynomial** of degree  $d$ .

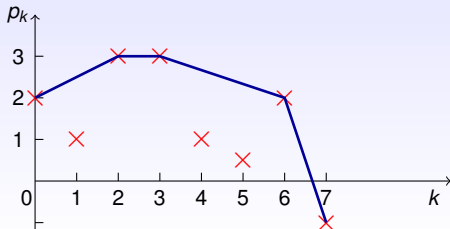
- The **tropical roots** of  $p(x)$  are the points of nondifferentiability of  $p(x)$ .

# Computation of Tropical Roots

Let  $p(x) = \bigoplus_{k=0}^d p_k \otimes x^{\otimes k} = \max(p_0, p_1 + x, \dots, p_d + dx)$ .

- ▶ Upper boundary of **convex hull** of  $(k, p_k)$ ,  $k = 0: d$ .
- ▶ **Tropical roots** are minus the **slopes** of the segments (Legendre-Fenchel duality).
- ▶ Horizontal **width** of segment gives **multiplicity**.

$p(x) = \max(2, 1 + x, 3 + 2x, 3 + 3x, 1 + 4x, \frac{1}{2} + 5x, 2 + 6x, -1 + 7x)$   
has roots  $-\frac{1}{2}, -\frac{1}{2}, 0, \frac{1}{3}, \frac{1}{3}, \frac{1}{3}, 3$ .



# Tropical Roots (cont.)

- Tropical roots can be computed in **linear time**.
- **Classical roots** of  $p(x) = a_0 + a_1x + \dots + a_nx^n$  can be **bounded in terms of tropical roots** of  $p_{\text{trop}}(x) = \bigoplus_{k=0}^d p_k \otimes x^{\otimes k}$  (Sharify, 11).
- Let  $r_1, r_2$  be the **tropical roots** of

$$p_{\text{trop}}(r) = \max(\log(\|K\|), \log(\|D\|) + r, \log(\|M\|) + 2r).$$

When  $r_1 \gg r_2$  and  $M, D, K$  are well conditioned,  $e^{r_1} \approx |\lambda_{\max}(Q)|$  and  $e^{r_2} \approx |\lambda_{\min}(Q)|$ , where  $Q(\lambda) = \lambda^2 M + \lambda D + K$  (Gaubert & Sharify, 09).

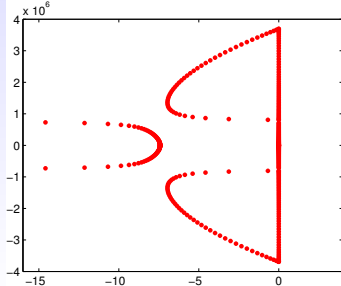
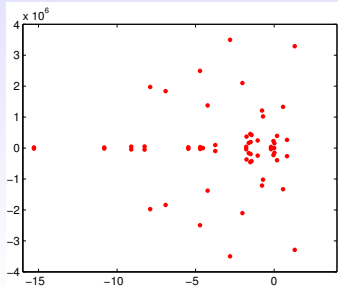
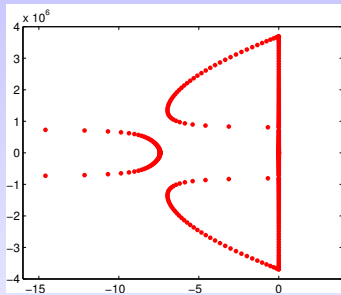
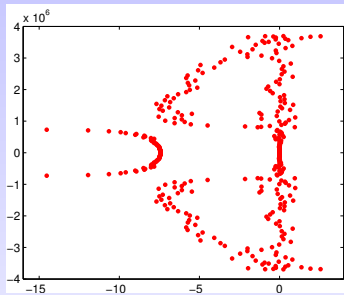
# Tropical Scaling

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Convert  $Q(\lambda)$  to  $\delta Q(\mu\gamma)$ , where

- $\gamma = e^r$ ,  $r$  is a tropical root of  $p_{\text{trop}}$ ,
  - $\delta = e^{p_{\text{trop}}(r)}$ .
- If  $\|D\|_2^2 \leq \|M\|_2 \|K\|_2$  then  $r_1 = r_2$  (one double root).  
Can show that  $\eta_{\tilde{Q}} \approx \eta_{\tilde{C}}$ ,  $\kappa_{\tilde{Q}} \approx \kappa_{\tilde{C}}$  for all e'vals.
- Otherwise, two distinct tropical roots  $r_1 > r_2$ . Can show  
that  $\eta_{\tilde{Q}} \approx \eta_{\tilde{C}}$ ,  $\kappa_{\tilde{Q}} \approx \kappa_{\tilde{C}}$  if  $\begin{cases} r = r_1 \text{ and } |\lambda| \geq \gamma, \\ r = r_2 \text{ and } |\lambda| \leq \gamma. \end{cases}$

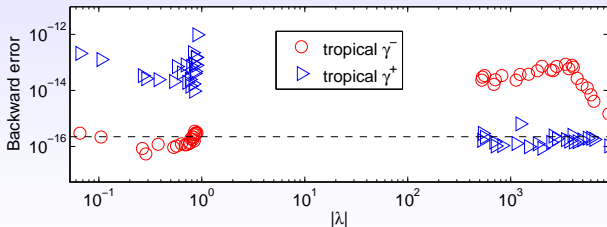
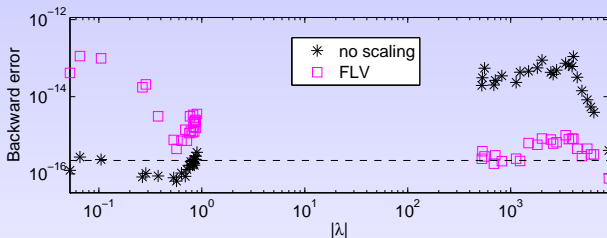
# Beam Pb: $\Lambda(C)$ , $\Lambda(L_2)$ before/after Scaling



# Modified Hospital Problem

Damping added s.t.  $\|D\|_2^2 \gg \|M\|_2 \|K\|_2$ .  $\gamma_{FLV} = \left( \frac{\|K\|_2}{\|M\|_2} \right)^{1/2} \approx 56$ .

Two tropical roots:  $\gamma^+ \approx 3.7 \times 10^3$  and  $\gamma^- \approx 0.8$ .



# Deflation of 0 and $\infty$ eigenvalues

Transform  $\tilde{C}(\lambda) = \begin{bmatrix} \gamma\delta D & -I \\ \delta K & 0 \end{bmatrix} - \lambda \begin{bmatrix} -\gamma^2\delta M & 0 \\ 0 & -I \end{bmatrix}$  into

$$\begin{bmatrix} \mathbf{S}_{11} & \mathbf{S}_{12} & \mathbf{S}_{13} \\ 0 & \mathbf{S}_{22} & \mathbf{S}_{23} \\ 0 & 0 & \mathbf{0}_{n-r_K} \end{bmatrix} - \lambda \begin{bmatrix} \mathbf{T}_{11} & \mathbf{T}_{12} & \mathbf{T}_{13} \\ 0 & \mathbf{0}_{n-r_M} & \mathbf{T}_{23} \\ 0 & 0 & \mathbf{I}_{n-r_K} \end{bmatrix},$$

where  $r_M = \text{rank}(M)$  and  $r_K = \text{rank}(K)$ .

- ▶ 2 QR fact with col piv and 1 COD if  $r_M < n$  and  $r_K < n$ .
- ▶ Make use of block structure of  $\tilde{C}(\lambda)$ .
- ▶ Call QZ on  $\mathbf{S}_{11} - \lambda\mathbf{T}_{11}$ , with  $\mathbf{T}_{11}$  usually upper triang.
- ▶  $Q(\lambda)$  nonregular if  $\mathbf{S}_{22}$  singular.
- ▶ Cost of deflation negligible compared with overall cost.

# Some QEPs from NLEVP Collection

$\eta_Q(x, \lambda)$ : backward error of computed eigenpair  $(x, \lambda)$ .

Problem	$n$	$\tau_Q$	<b>polyeig</b>	<b>quadeig</b>	
			$\eta_Q(x, \lambda)$	$\eta_Q(x, \lambda)$	$\eta_Q(y^*, \lambda)$
<b>power_plant</b>	8	7e-1	1e-8	4e-16	5e-17
<b>cd_player</b>	60	9e+3	2e-10	7e-16	2e-15
<b>speaker_box</b>	107	2e-5	2e-11	2e-16	4e-16
<b>damped_beam</b>	200	2e-4	3e-9	1e-15	1e-15
<b>shaft</b>	400	1e-6	9e-8	9e-16	6e-16
<b>railtrack</b>	1005	2e1	2e-8	2e-15	8e-15

**quadeig** is **backward stable** for  $\tau_Q \lesssim 1$ , where

$$\tau_Q = \|A_1\| / \sqrt{\|A_2\| \|A_0\|}.$$



# Some Timings

Problem	$n$	<b>polyeig</b>		<b>quadeig</b>	
		$\Lambda$	$(\Lambda, X)$	$\Lambda$	$(\Lambda, X)$
<code>acoustic_wave_2D</code>	870	118s	202s	114s	191s
<code>damped_beam</code>	1000	92s	156s	97s	163s
<code>spring</code>	1000	182s	272s	94s	170s
<code>railtrack2</code>	1410	129s	306s	79s	113s

`railtrack2`: QEPs with singular  $M, K$ .

# Extension to Higher Degree Polynomials

$$P(\lambda)x = (\lambda^\ell A_\ell + \cdots + \lambda A_1 + A_0)x = 0.$$

- ▶ Use **Fiedler linearization**:

$$F(\lambda) = \lambda \begin{bmatrix} A_\ell & & & & \\ & I & & & \\ & & \ddots & & \\ & & & \ddots & \\ & & & & I \end{bmatrix} - \begin{bmatrix} -A_{\ell-1} & -A_{\ell-2} & \cdots & -A_1 & I \\ & I & & & \\ & & \ddots & & \\ & & & I & \\ & & & & -A_0 \end{bmatrix}.$$

Stability/conditioning similar to companion forms.

- ▶ Deflation of 0 and  $\infty$  e'vals easy to implement.
- ▶ Work in progress: scaling strategy (Betcke's scaling, tropical roots, tropical eigenvalues, ...).


# Summary

- ▶ **quadeig** is backward stable for not too heavily damped problems.
- ▶ Tropical mathematics useful for e'val scaling.
- ▶ Deflation strategy can produce significant speedups.
- ▶ Returns e'vals, right & left e'vecs, e'val condition numbers, b'errs of right/left approx. eigenpairs.
- ▶ MATLAB function available. Fortran implementation available soon.
- ▶ Extension to polynomial eigenproblem of degree  $\ell > 2$ .

For papers, eprints and codes,

<http://www.ma.man.ac.uk/~ftisseur/>

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