## **Eigenvalue Problems**

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Problem:

#### Find $\lambda \in \mathbb{C}$ , $u : \Omega \to \mathbb{C}$ :

$$\begin{cases} -\Delta u(z) = \lambda u(z), & z \in \Omega, \\ u(z) = 0, & z \in \partial \Omega \end{cases}$$

discretization  $\downarrow$ 

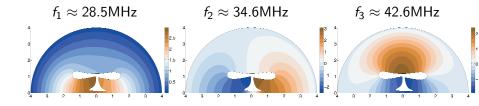
Find  $\lambda \in \mathbb{C}$ ,  $u \in \mathbb{C}^n$ :

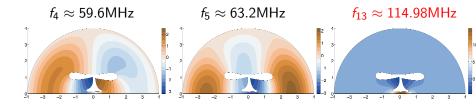
 $Au = \lambda u$ 

Eigenvalue problem!

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### Motivation: antennas





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## Motivation: mechanics

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## Motivation: SVD

$$A = U\Sigma V^{*}$$

$$\downarrow$$

$$AA^{*} = U\Sigma\Sigma^{T}U^{*}$$

$$A^{*}A = V\Sigma^{T}\Sigma V^{*}$$

 $\sigma$  - SVD of A iff  $\sigma^2$ -eigenvalue of AA\* (or A\*A)

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Straightforward algorithm for finding eigenvalues:

1. Compute coefficients of

$$p_A(\lambda) = \det(A - \lambda I)$$

2. Find roots

$$p_A(\lambda) = 0$$

3. Optionally/if needed: find eigenvectors

#### Expensive; numerically unstable!

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## Similar matrices

**Definition:** A and B-similar if  $\exists X$ , det $(X) \neq 0$ :

$$A = XBX^{-1}$$

Same eigenvalues (incl. algebraic multiplicity):

$$p_A(\lambda) = \det(A - \lambda I) = \det[X(B - \lambda I)X^{-1}]$$
  
= det(X) det(B - \lambda I) det(X^{-1}) = det(B - \lambda I) = p\_B(\lambda)

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### Similar matrices

**Definition:** A and B-similar if  $\exists X$ , det $(X) \neq 0$ :

$$A = XBX^{-1}$$

Same number of lin. indep. eigenvectors (geometric multiplicity):

 $Av = \lambda v$  $XBX^{-1}v = \lambda v$  $B[X^{-1}v] = \lambda[X^{-1}v]$ 

Eigenvalue-revealing factorizations:

If  $A = X\Lambda X^{-1}$ ,  $\Lambda$ -diagonal

$$AX = X\Lambda$$
$$AX_{*,j} = \Lambda_{j,j}X_{*,j}$$

Columns of X-eigenvectors, diagonal of  $\Lambda$ -eigenvalues!

Does not exist for defective matrices!

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## Eigenvalue-revealing factorizations:

Schur factorization:  $A = QRQ^*$ ,  $Q^*Q = I$ , R-(upper) triangular (Remember:  $A^*A = AA^* \implies R$ -diagonal.)

Exists for all matrices!

A similar to  $R \implies$ 

 $\lambda_i(A) = \lambda_i(R) = R_{i,i}$ 

### Is existence proof constructive?

Not in the sense of numerical analysis...

Start with  $Aq_1 = \lambda_1 q_1$ , complete q to ON basis  $\tilde{Q} = [q_1, \dots, q_n]$ 

$$\begin{split} \tilde{Q}^* A \tilde{Q} &= \tilde{Q}^* [Aq_1, Aq_2, \dots, Aq_n] \\ &= \tilde{Q}^* [\lambda_1 q_1, Aq_2, \dots, Aq_n] \\ &= [\lambda_1 e_1, \tilde{Q}^* Aq_2, \dots, \tilde{Q}^* Aq_n] \\ &= \begin{bmatrix} \lambda_1 & w \\ 0 & \tilde{A} \end{bmatrix} = \begin{bmatrix} \lambda_1 & w \\ 0 & \hat{Q} \hat{R} \hat{Q}^* \end{bmatrix} \end{split}$$

-used induction hypothesis on a smaller submatrix  $\tilde{A}$ .

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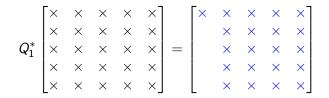
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-used induction hypothesis on a smaller submatrix  $\tilde{A}$ .

$$A = \underbrace{\tilde{Q} \begin{bmatrix} 1 & 0 \\ 0 & \hat{Q} \end{bmatrix}}_{=:Q} \underbrace{\begin{bmatrix} \lambda_1 & w\hat{Q} \\ 0 & \hat{R} \end{bmatrix}}_{=:R} \underbrace{\begin{bmatrix} 1 & 0 \\ 0 & \hat{Q}^* \end{bmatrix}}_{=:Q^*} \tilde{Q}^*$$

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Recall Householder QR-factorization algorithm



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Recall Householder QR-factorization algorithm

Could we do something similar for Schur?

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Not possible!

### Connexion between polynomial roots and eigenvalues

Arbitrary polynomial:

$$p(x) = x^n + a_{n-1}x^{n-1} + \dots + a_1x + a_0$$

Define:

$$A = \begin{bmatrix} 0 & & & -a_0 \\ 1 & 0 & & -a_1 \\ & 1 & 0 & & -a_2 \\ & & 1 & \ddots & & \vdots \\ & & \ddots & 0 & -a_{n-2} \\ & & & 1 & -a_{n-1} \end{bmatrix}$$

Then  $p(\lambda) = \det[\lambda I - A]!$ 

No formula for polynomial roots (Abel, 1842)  $\implies$  no finite algorithm for eigenvalues!

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Remember: <u>all</u> eigenvalue algorithms are inherently iterative! (But in practice, only a few iterations are needed for good algorithms.)

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## Phase 1/phase 2 approach

Phase 1: finite factorization

$$A = Q \tilde{A} Q^*$$

 $Q^*Q = I$ ,  $ilde{A}$ -significantly simpler than A

### Phase 2: iterative computation of eigenvalues of $\tilde{A}$ (hence also A)

$$A = QHQ^*$$

 $Q^*Q = I$ , *H*-(upper) Hessenberg Algorithm: similar to Householder-QR

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## Full algorithm: Householder Hessenberg

1: for 
$$k = 1$$
 to  $n - 2$  do  
2:  $x := A_{k+1:n,k}$   
3:  $v_k := \text{sign}(x_1) ||x||_2 e_1 + x$   
4:  $v_k := v_k / ||v_k||_2$   
5:  $A_{k+1:n,k:n} = A_{k+1:n,k:n} - 2v_k (v_k^* A_{k+1:n,k:n})$   
6:  $A_{1:n,k+1:n} = A_{1:n,k+1:n} - 2(A_{1:n,k+1:n}v_k) v_k^*$   
7: end for

Complexity:

$$\sim \sum_{k=1}^{n-2} \left[ \underbrace{4(n-k)^2}_{\text{QR,line 5}} + \underbrace{4n(n-k)}_{\text{line 6}} \right] \approx \underbrace{4n^3/3}_{\text{QR}} + 4n^3/2 \\ = 10n^3/3 = 2.5 \times \text{QR}$$

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Phase 1, 
$$A^* = A$$

$$A = QHQ^*$$
$$A^* = QH^*Q^*$$
$$\Downarrow$$
$$H - \text{tri-diagonal}$$

Householder Hessenberg complexity: same as QR

Given: Hessenberg (or tri-diagonal) matrix Construct iterative process for computing eigenvalues

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# Rayleigh quotient

### Suppose $x \in \mathbb{C}^n \setminus \{0\}$ is a given approximation of an eigenvector. Task: find eigenvalue! I.e., find $\lambda \in \mathbb{C}$ :

 $x\lambda \approx Ax$ 

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## Rayleigh quotient

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Solve as least-squares problem! Normal equations:

$$(x^*x)\lambda = x^*Ax$$
  
 $\lambda = \frac{x^*Ax}{x^*x} =: r(x)$ 

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### **Properties:**

- 1.  $r(\alpha x) = r(x), \forall \alpha \neq 0, x \neq 0$
- r: smooth function of x [for x ∈ C<sup>n</sup>: consider real and complex components]
- 3. ||Ax r(x)x|| = 0 iff x-eigenvector; then r(x) is eigenvalue.

Possible algorithm: solve a system of non-linear equations Ax - r(x)x = 0, ||x|| = 1.

### Derivatives:

Assume  $x \in \mathbb{R}^n \setminus \{0\}$ ,  $A \in \mathbb{R}^{n \times n}$ :

$$\partial_i r(x) = \frac{\left[\partial_i (x^{\mathrm{T}} A x)\right] x^{\mathrm{T}} x - x^{\mathrm{T}} A x \partial_i (x^{\mathrm{T}} x)}{(x^{\mathrm{T}} x)^2}$$
$$= \frac{\left([Ax]_i + [A^{\mathrm{T}} x]_i\right) x^{\mathrm{T}} x - 2(x^{\mathrm{T}} A x) x_i}{(x^{\mathrm{T}} x)^2}$$
$$= \frac{2}{x^{\mathrm{T}} x} \left[\frac{[Ax]_i + [A^{\mathrm{T}} x]_i}{2} - r(x) x_i\right]$$
$$\nabla r(x) = \frac{2}{x^{\mathrm{T}} x} \left[\frac{A + A^{\mathrm{T}}}{2} x - r(x) x\right]$$

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### Derivatives:

Additionally assume  $A = A^{T}$ . Then

$$\nabla r(x) = \frac{2}{x^{\mathrm{T}}x} [Ax - r(x)x]$$

Therefore  $\nabla r(x) = 0$  iff  $x \neq 0$  – eigenvector.

In particular, if  $A\bar{x} = \lambda \bar{x}$  then

$$r(x) - \lambda = O(||x - \bar{x}||^2),$$

in the vicinity of  $\bar{x}$ .

(Note: in non-Hermitian case  $r(x) - \lambda = O(||x - \bar{x}||)$  only.)