

# Variational characterizations of eigenvalues

MATH 6610 Lecture 04

September 9, 2020

# Hermitian matrices

If  $\mathbf{A} \in \mathbb{C}^{n \times n}$  is Hermitian, recall that:

- $\mathbf{A}$  is unitarily diagonalizable,  $\mathbf{A} = \mathbf{U}\mathbf{\Lambda}\mathbf{U}^*$
- The spectrum of  $\mathbf{A}$  is real-valued
- (Hermitian) Positive-definite matrices have matrix square roots

(and unique spd square roots)

# Hermitian matrices

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Today: Variational characterizations of eigenvalues for Hermitian matrices.

# Rayleigh quotients, I

Let  $A \in \mathbb{C}^{n \times n}$  be a matrix, and let  $x \in \mathbb{C}^n \setminus \{0\}$  be a vector.

The Rayleigh Quotient (of  $A$  at  $x$ ) is the scalar,

$$R_A(x) := \frac{\langle Ax, x \rangle}{\langle x, x \rangle} = \frac{x^* Ax}{x^* x}$$

$$R_A(x) = R_A\left(\frac{x}{\|x\|_2}\right) \quad (\text{i.e. } R_A(x) \text{ is invariant under scaling of } x).$$

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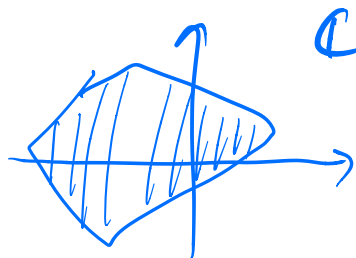
The Rayleigh Quotient (of  $\mathbf{A}$  at  $\mathbf{x}$ ) is the scalar,

$$R_{\mathbf{A}}(\mathbf{x}) := \frac{\langle \mathbf{A}\mathbf{x}, \mathbf{x} \rangle}{\langle \mathbf{x}, \mathbf{x} \rangle}$$

- The numerical range of  $\mathbf{A}$  is the set of all possible values of  $R_{\mathbf{A}}$ :

$$W_{\mathbf{A}}(\mathbb{C}^n) := R_{\mathbf{A}}(\mathbb{C}^n \setminus \{\mathbf{0}\}).$$

$$W_{\mathbf{A}}(\mathbb{C}^n) \subset \mathbb{C}$$



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$$R_A(x) := \frac{\langle Ax, x \rangle}{\langle x, x \rangle}$$

- The numerical range of  $A$  is the set of all possible values of  $R_A$ :

$$W_A(\mathbb{C}^n) := R_A(\mathbb{C}^n \setminus \{0\}).$$

- The numerical range contains the spectrum of  $A$ .

Let  $v$  be (only) eigenvector of  $A$  associated to eigenvalue  $\lambda$ . Then  $R_A(v) = \frac{v^* A v}{\|v\|_2^2} = \lambda \frac{v^* v}{\|v\|_2^2} = \lambda$

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- The numerical range contains the spectrum of  $\mathbf{A}$ .
- If  $\mathbf{A}$  is Hermitian, then  $\lambda_{\min}(\mathbf{A}) \leq R_{\mathbf{A}}(\mathbf{x}) \leq \lambda_{\max}(\mathbf{A})$ .

Upper inequality:  $A = U\Lambda U^*$ ,  $\Lambda = \text{diag}(\lambda_1, \dots, \lambda_n)$

$$\lambda_j \leq \lambda_{j+1} \quad R_{\mathbf{A}}(\mathbf{x}) = \frac{\langle \mathbf{x}, \mathbf{A}\mathbf{x} \rangle}{\langle \mathbf{x}, \mathbf{x} \rangle} = \frac{\langle U^* \mathbf{x}, \Lambda U^* \mathbf{x} \rangle}{\langle U^* \mathbf{x}, U^* \mathbf{x} \rangle}$$

Let  $y = U^* x$

$$\begin{aligned} \text{Then: } R_A(x) &= \frac{\langle y, Ay \rangle}{\|y\|_2^2} = \frac{\sum_{j=1}^n \lambda_j |y_j|^2}{\|y\|_2^2} \leq \frac{\sum_{j=1}^n \lambda_n |y_j|^2}{\|y\|_2^2} \\ &= \lambda_n (= \lambda_{\max}(A)) \end{aligned}$$



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$$W_{\mathbf{A}}(\mathbb{C}^n) := R_{\mathbf{A}}(\mathbb{C}^n \setminus \{\mathbf{0}\}).$$

- The numerical range contains the spectrum of  $\mathbf{A}$ .
- If  $\mathbf{A}$  is Hermitian, then  $\lambda_{\min}(\mathbf{A}) \leq R_{\mathbf{A}}(\mathbf{x}) \leq \lambda_{\max}(\mathbf{A})$ .

We can also consider the image of the Rayleigh quotient but only on a subspace  $V$ :

$$W_{\mathbf{A}}(V) := R_{\mathbf{A}}(V \setminus \{\mathbf{0}\}).$$

# Rayleigh quotients, II

Let  $\mathbf{A} \in \mathbb{C}^{n \times n}$  be Hermitian. Consider a subspace  $V \subset \mathbb{C}^n$ .

The image of the  $V$  under the Rayleigh quotient,  $W_{\mathbf{A}}(V)$ , is some subset of  $\mathbb{R}$ .

# Rayleigh quotients, II

L04-S03

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- The minimum of  $W_{\mathbf{A}}(V)$  can be  $\lambda_{\min}(\mathbf{A})$ . (if  $V$  contains minimal eigenvector of  $\mathbf{A}$ )  
What is the largest possible minimum value?

# Rayleigh quotients, II

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The image of the  $V$  under the Rayleigh quotient,  $W_{\mathbf{A}}(V)$ , is some subset of  $\mathbb{R}$ .

- The minimum of  $W_{\mathbf{A}}(V)$  can be  $\lambda_{\min}(\mathbf{A})$ .  
What is the largest possible minimum value?
- The maximum of  $W_{\mathbf{A}}(V)$  can be  $\lambda_{\max}(\mathbf{A})$ .  
What is the smallest possible maximum value?

# The “min-max” theorem

## Theorem (Courant-Fischer-Weyl “min-max”)

Let  $A \in \mathbb{C}^{n \times n}$  be Hermitian, with eigenvalues  $\lambda_1 \leq \lambda_2 \leq \dots \leq \lambda_n$ . Then for each  $1 \leq k \leq n$ ,

$$\begin{aligned}
 (\star) \quad \lambda_k &= \min_{\substack{V \subset \mathbb{C}^n \\ \dim V = k}} \max W_A(V) \\
 &\xrightarrow{\quad} \lambda_k = \max_{\substack{V \subset \mathbb{C}^n \\ \dim V = n-k+1}} \min W_A(V)
 \end{aligned}$$

$k=n \Rightarrow V = \mathbb{C}^n \rightarrow \lambda_n = \max W_A(\mathbb{C}^n)$

In addition, if  $(\mathbf{u}_j)_{j=1}^n$  are the eigenvectors associated with  $(\lambda_j)_{j=1}^n$ , then:

- $V = \text{span}\{\mathbf{u}_1, \dots, \mathbf{u}_k\}$  achieves the outer minimum
- $V = \text{span}\{\mathbf{u}_k, \dots, \mathbf{u}_n\}$  achieves the outer maximum

we'll prove first min-max equation. ( $\star$ ).

$$(\star) \lambda_k = \min_{\substack{V \subset \mathbb{C}^n \\ \dim V = k}} \max W_A(V).$$

We'll accomplish this by showing the  $\leq$  and  $\geq$  inequalities.

Let  $V = \text{span} \{u_1, \dots, u_k\}$  ( $(\lambda_j, u_j)_{j=1}^k$  are eigenpairs of  $A$ ).

$$\begin{aligned} \max W_A(V) & \\ \underline{x \in V} : R_A(x) &= \frac{\langle x, Ax \rangle}{\langle x, x \rangle} \quad x \in V \Rightarrow x = \sum_{j=1}^k c_j u_j \\ &= \frac{\langle x, \sum_{j=1}^k c_j \lambda_j u_j \rangle}{\langle x, x \rangle} = \frac{\sum_{j=1}^k |c_j|^2 \lambda_j}{\langle x, x \rangle} \\ &= \frac{\sum_{j=1}^k |c_j|^2 \lambda_j}{\sum_{j=1}^k |c_j|^2} \leq \lambda_k \frac{\sum_{j=1}^k |c_j|^2}{\sum_{j=1}^k |c_j|^2} = \lambda_k. \end{aligned}$$

$$\max W_A(V) = \lambda_k$$

$$\min_{\substack{V \subset \mathbb{C}^n \\ \dim V = k}} \max W_A(V) \leq \lambda_k. \quad (\leq)$$

(already showed  $\lambda_k \geq \min_{\substack{V \subset \mathbb{C}^n \\ \dim V = k}} \max W_A(V)$ )

Let  $V$  be any  $k$ -dim subspace of  $\mathbb{C}^n$ .

$\text{span}\{u_k, u_{k+1}, \dots, u_n\}$  is a dimension  $(n-k+1)$  space.

Then  $V \cap \text{span}\{u_k, \dots, u_n\} \neq \emptyset$ .

$\Rightarrow \exists v \neq 0$  s.t.  $v = \sum_{j=k}^n c_j u_j, v \in V$ .

$$R_A(v) = \frac{\langle v, Av \rangle}{\langle v, v \rangle} = \frac{\sum_{j=k}^n |c_j|^2 \lambda_j}{\sum_{j=k}^n |c_j|^2} \geq \lambda_k$$

$\Rightarrow \max_{v \in V} R_A(v) = \max W_A(V) \geq \lambda_k \quad (\forall k\text{-dim } V)$ .

$\Rightarrow \min_{\substack{V \subset \mathbb{C}^n \\ \dim V = k}} \max W_A(V) \geq \lambda_k$ .

$\Rightarrow \lambda_k = \min_{\substack{V \subset \mathbb{C}^n \\ \dim V = k}} \max W_A(V) \quad \checkmark \quad (\star)$

# Cauchy interlacing theorem

A matrix  $B$  is a compression of  $A$  if  $B = Q^* A Q$  for some  $Q \in \mathbb{C}^{n \times r}$  with orthonormal columns.

orth

$$A \in \mathbb{C}^{n \times n}$$

$$B = Q^* A Q \in \mathbb{C}^{r \times r}$$



# Cauchy interlacing theorem

A matrix  $B$  is a compression of  $A$  if  $B = Q^* A Q$  for some  $Q \in \mathbb{C}^{n \times r}$  with orthonormal columns.

Just one consequence of the min-max theorem:

## Theorem (Cauchy interlacing)

Let  $B \in \mathbb{C}^{(n-1) \times (n-1)}$  be a compression of a Hermitian matrix  $A \in \mathbb{C}^{n \times n}$ . If  $A$  has eigenvalues  $\lambda_1 \leq \lambda_2 \leq \dots \leq \lambda_n$ , and  $B$  has eigenvalues  $\mu_1, \dots, \mu_{n-1}$ , then

$$\lambda_j \leq \mu_j \leq \lambda_{j+1},$$

for all  $j = 1, \dots, n-1$ .

$B$  is Hermitian  $\Rightarrow \mu_j$  are  $\mathbb{R}$ -valued

Assume  $\mu_1 \leq \mu_2 \leq \dots \leq \mu_{n-1}$ .

Proof:

Recall: if  $U = \text{span}\{u_1, \dots, u_j\}$ , where

$(\underbrace{\mu_j}_{k}, \underbrace{u_j}_{k})_{k=1}^{j-1}$  are eigen pairs of  $B$ .

$$\mu_j = \max_{x \in U \setminus \{0\}} R_B(x) = \max_{x \in U \setminus \{0\}} \frac{\langle Bx, x \rangle}{\langle x, x \rangle}$$

defn of  $B$ , and  $Q^*Q = I$ .

$$\text{Then: } \mu_j = \max_{x \in U \setminus \{0\}} \frac{\langle Bx, x \rangle}{\langle x, x \rangle} = \max_{x \in U \setminus \{0\}} \frac{\langle AQx, Qx \rangle}{\langle Qx, Qx \rangle}$$

$$= \max_{y \in QU \setminus \{0\}} \frac{\langle Ay, y \rangle}{\langle y, y \rangle}, \quad \text{where } QU := \{Qy \mid y \in U\}$$

is  $j$ -dim subspace  $\dim QU = j$

$$\geq \min_{\dim V = j} \max_{y \in V} \frac{\langle Ay, y \rangle}{\langle y, y \rangle} = \lambda_j$$

Courant-Fischer-Weyl.

$$\Rightarrow \mu_j \geq \lambda_j.$$

Other inequality:  $U := \text{span}\{u_j, u_{j+1}, \dots, u_{n-1}\}$

$$\text{Consider } \mu_j = \min_{x \in U \setminus \{0\}} R_B(x).$$

and perform similar computations.

.....  $\square$