

MTH 102A - Linear Algebra - 2015-16-II Semester

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P. Eigenvalues and eigenvectors

- Matrix $A = \begin{bmatrix} 1 & 2 \\ 2 & 1 \end{bmatrix}$ behaves like the scalar 3 when multiplied with $\mathbf{x} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$. That is, $\begin{bmatrix} 1 & 2 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = 3 \begin{bmatrix} 1 \\ 1 \end{bmatrix}$.
- Physically: The LT $f(\mathbf{x}) = A\mathbf{x}$ magnifies the nonzero vector $\begin{bmatrix} 1 \\ 1 \end{bmatrix} \in \mathbb{C}^2$ three (3) times.
- Similarly, $A \begin{bmatrix} 1 \\ -1 \end{bmatrix} = -1 \begin{bmatrix} 1 \\ -1 \end{bmatrix}$. So, behaves by changing the direction of the vector $\begin{bmatrix} 1 \\ -1 \end{bmatrix}$
- Take $A = \begin{bmatrix} 1 & 2 \\ 1 & 3 \end{bmatrix}$. Do I have a nonzero $\mathbf{x} \in \mathbb{C}^2$ which gets magnified by A ?
- So I am looking for $\mathbf{x} \neq 0$ and α s.t. $A\mathbf{x} = \alpha\mathbf{x}$. Using $\mathbf{x} \neq 0$, we have $A\mathbf{x} = \alpha\mathbf{x}$ if and only if $[\alpha I - A]\mathbf{x} = \mathbf{0}$ if and only if $\text{DET}[\alpha I - A] = 0$.
- $\text{DET}[\alpha I - A] = \text{DET} \begin{bmatrix} \alpha - 1 & -2 \\ -1 & \alpha - 3 \end{bmatrix} = \alpha^2 - 4\alpha + 1$. So $\alpha = 2 \pm \sqrt{3}$.
- Take $\alpha = 2 + \sqrt{3}$. To find \mathbf{x} , solve $\begin{bmatrix} 1 + \sqrt{3} & -2 \\ -1 & \sqrt{3} - 1 \end{bmatrix} \mathbf{x} = \mathbf{0}$: using GJE, for instance.

We get $\mathbf{x} = \begin{bmatrix} \sqrt{3} - 1 \\ 1 \end{bmatrix}$. Moreover

$$A\mathbf{x} = \begin{bmatrix} 1 & 2 \\ 1 & 3 \end{bmatrix} \begin{bmatrix} \sqrt{3} - 1 \\ 1 \end{bmatrix} = \begin{bmatrix} \sqrt{3} + 1 \\ 2 + \sqrt{3} \end{bmatrix} = (2 + \sqrt{3}) \begin{bmatrix} \sqrt{3} - 1 \\ 1 \end{bmatrix}.$$

- We call $\lambda \in \mathbb{C}$ an **eigenvalue** of $A_{n \times n}$ if there exists $\mathbf{x} \in \mathbb{C}^n$, $\mathbf{x} \neq 0$ s.t. $A\mathbf{x} = \lambda\mathbf{x}$. We call \mathbf{x} an **eigenvector** of A for the eigenvalue λ . We call (λ, \mathbf{x}) an **eigenpair**.
- If (λ, \mathbf{x}) is an eigenpair of A , then so is $(\lambda, c\mathbf{x})$, for each $c \neq 0$, $c \in \mathbb{C}$.

Th. The number λ is an eigenvalue of $A_{n \times n}$ if and only if $\text{DET}(\lambda I - A) = 0$.

Pr. Follows from: there exists a nontrivial solution of $[\lambda I - A]\mathbf{x} = \mathbf{0}$ if and only if $\text{DET}[\lambda I - A] = 0$. ■

- Find eigenpairs of $A = \begin{bmatrix} 1 & 3 \\ -2 & 5 \end{bmatrix}$.
- To find the eigenvalues, solve $\text{DET}(\lambda I - A) = 0$. So $\lambda = 3 \pm \sqrt{2}i$.
- To find an eigenvector $\begin{bmatrix} x \\ y \end{bmatrix}$ for the eigenvalue $3 + \sqrt{2}i$:
$$\left[\begin{array}{cc|c} 2 + \sqrt{2}i & -3 & 0 \\ 2 & \sqrt{2}i - 2 & 0 \end{array} \right] \xrightarrow{E_{21}(\frac{2}{2 + \sqrt{2}i})} \left[\begin{array}{cc|c} 2 + \sqrt{2}i & -3 & 0 \\ 0 & 0 & 0 \end{array} \right]. \text{ So } \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} \frac{3}{2 + \sqrt{2}i} \\ 1 \end{bmatrix}.$$

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$$\left(3 + \sqrt{2}i, \begin{bmatrix} \frac{3}{2+\sqrt{2}i} \\ 1 \end{bmatrix}\right).$$

- Let D be a diagonal matrix. Then eigenpairs of D are (d_{ii}, \mathbf{e}_i) .

Q. Should $A_{n \times n}$ have n eigenvalues (including multiplicities)? Yes.

- While discussing eigenvalues etc, we consider square matrices only.

Q. What are the eigenvalues of an upper triangular matrix? Diagonal entries.

- The multiset of eigenvalues of A is called the **spectrum** of A . Notation: $\sigma(A)$.

Q. Let $\lambda \in \sigma(A)$. Must we have at least one eigenvector for λ ? Yes, by GJE.

- The multiplicity of λ in $\sigma(A)$ is called the **algebraic multiplicity** of A .

Q. Suppose that algebraic multiplicity of λ in $\sigma(A)$ is 2. Must we have 2 lin.ind eigenvectors for λ ? No. For example, take $A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$. Then $\sigma(A) = \{0, 0\}$ and \mathbf{e}_1 is an eigenvector. That is, $A\mathbf{e}_1 = \mathbf{0}$. If \mathbf{x} is another eigenvector, then $A\mathbf{x} = \mathbf{0}$. That is, $\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} x_2 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$. So $x_2 = 0$ and $\mathbf{x} = x_1\mathbf{e}_1$.

[If $(\lambda_1, \mathbf{x}_1), \dots, (\lambda_k, \mathbf{x}_k)$ are eigenpairs, then $\mathbf{x}_1, \dots, \mathbf{x}_k$ may not be linearly independent unless λ_i 's are distinct!!]

- If $\lambda \in \sigma(A)$, then the maximum number of linearly independent eigenvectors for λ is the **geometric multiplicity** of λ . [Geometry: maximum k such that the LT magnifies a k -dimensional subspace by λ . Take LT: $\mathbf{e}_1 \rightarrow \mathbf{e}_1, \mathbf{e}_2 \rightarrow \mathbf{e}_2, \mathbf{e}_3 \rightarrow \mathbf{e}_1 + \mathbf{e}_3$.

Th. Let $(\lambda_i, \mathbf{v}_i)$ be some eigenpairs of $A_{n \times n}$, λ_i 's are distinct. Then $\mathbf{v}_1, \dots, \mathbf{v}_k$ are linearly independent. [Contrast it with the earlier comment.]

Pr. Suppose that $\mathbf{v}_1, \dots, \mathbf{v}_k$ are lin.dep. Then there exists ℓ smallest, and β s.t. $\mathbf{v}_{\ell+1} = \beta_1\mathbf{v}_1 + \dots + \beta_\ell\mathbf{v}_\ell$. Note that $\beta \neq 0$. So

$$\begin{aligned} A\mathbf{v}_{\ell+1} &= A[\beta_1\mathbf{v}_1 + \dots + \beta_\ell\mathbf{v}_\ell] = && \lambda_1\beta_1\mathbf{v}_1 + \dots + \lambda_\ell\beta_\ell\mathbf{v}_\ell \\ \lambda_{\ell+1}\mathbf{v}_{\ell+1} &= && \lambda_{\ell+1}\beta_1\mathbf{v}_1 + \dots + \lambda_{\ell+1}\beta_\ell\mathbf{v}_\ell \\ 0 &= && [\lambda_{\ell+1} - \lambda_1]\beta_1\mathbf{v}_1 + \dots + [\lambda_{\ell+1} - \lambda_\ell]\beta_\ell\mathbf{v}_\ell. \end{aligned}$$

So $\mathbf{v}_\ell \in \text{LS}(\mathbf{v}_1, \dots, \mathbf{v}_{\ell-1})$, a contradiction. ■

- The converse of the above is not true. Think of I .
- $0 \in \sigma(A)$ if and only if $\text{NULL } A \neq \{\mathbf{0}\}$ if and only if A is singular.
- Let A be invertible. Then $\lambda \in \sigma(A)$ if and only if $\lambda^{-1} \in \sigma(A^{-1})$.
- Polynomial $p_A(x) = \text{DET}[xI - A]$ is called the **characteristic polynomial** of A .

The equation $p_A(x) = 0$ is called the **characteristic equation** of A . So eigenvalues are roots of the characteristic equation.

- Let $\lambda \in \sigma(A)$. The **eigenspace** for λ is the span of all eigenvectors for λ .

[The dimension of the eigenspace is the geometric multiplicity.]

Q. Is $\sigma(A) = \sigma(A^t)$? Yes, same characteristic polynomial: $p_A(x) = p_{A^t}(x)$.

• Let $A \in \mathcal{M}_n(\mathbb{R})$. Then $p_A(\lambda) = 0$ if and only if $p_A(\bar{\lambda}) = 0$. So $\lambda \in \sigma(A)$ if and only if $\bar{\lambda} \in \sigma(A)$.

[This is not true for complex matrices.]

Similar: A and B are said to be matrices if there exists an invertible matrix S s.t. $A = SBS^{-1}$.

Th. Similar matrices have the same characteristic polynomial.

Pr. Let A be similar to B . So there exists S s.t. $B = S^{-1}AS$. So $\text{DET}[xI - B] = \text{DET}[xI - S^{-1}AS] = \text{DET}[S^{-1}(xI - A)S] = \text{DET}[xI - A]$. ■

• Let B be invertible. Then AB is similar to BA , as $BA = B(AB)B^{-1}$. So $p_{AB}(x) = p_{BA}(x)$ and $\sigma(AB) = \sigma(BA)$.

• Let $B = S^{-1}AS$, $\lambda \in \sigma(A)$ and $\mathbf{v}_1, \dots, \mathbf{v}_k$ be lin.ind eigenvectors of A for λ .

Are $S^{-1}\mathbf{v}_1, \dots, S^{-1}\mathbf{v}_k$ lin.ind? **Yes, as S is invertible.** Is $S^{-1}\mathbf{v}_i$ an eigenvector of B ?

Yes, $B(S^{-1}\mathbf{v}_i) = S^{-1}ASS^{-1}\mathbf{v}_i = S^{-1}A\mathbf{v}_i = \lambda(S^{-1}\mathbf{v}_i)$. So the geometric multiplicity of λ in A and B are the same. [Geometry: if A has eigenvector \mathbf{v} and $T\mathbf{x} = A\mathbf{x}$, then $[T]$ has eigenvector $[\mathbf{v}]$.]

P. Geometric and algebraic multiplicity

Th. Let $\lambda \in \sigma(A)$. Then geom. multipl of $\lambda \leq$ algeb multipl of λ .

Pr. Let geometric multiplicity of λ be k and $\mathbf{v}_1, \dots, \mathbf{v}_k$ be lin.ind eigenvectors for λ . Apply GS: let $\mathbf{w}_1, \dots, \mathbf{w}_k$ be the orthonormal eigenvectors for λ . Extend it to an orthonormal basis:

$$\{\mathbf{w}_1, \dots, \mathbf{w}_k, \mathbf{w}_{k+1}, \dots, \mathbf{w}_n\}. \text{ Put } P(:, i) = \mathbf{w}_i.$$

$$\text{Then } P^*AP = P^*[A\mathbf{w}_1 \ \dots \ A\mathbf{w}_k \ A\mathbf{w}_{k+1} \ \dots \ A\mathbf{w}_n] =$$

$$\begin{bmatrix} \mathbf{w}_1^* \\ \vdots \\ \mathbf{w}_k^* \\ \mathbf{w}_{k+1}^* \\ \vdots \\ \mathbf{w}_n^* \end{bmatrix} \begin{bmatrix} \lambda \mathbf{w}_1 & \dots & \lambda \mathbf{w}_k & * & \dots & * \end{bmatrix} = \begin{bmatrix} \lambda & \dots & 0 & * & \dots & * \\ 0 & \ddots & 0 & * & \dots & * \\ 0 & \dots & \lambda & * & \dots & * \\ \hline 0 & \dots & 0 & * & \dots & * \\ \vdots & & & \mathbf{D} & & \\ 0 & \dots & 0 & * & \dots & * \end{bmatrix}.$$

Now $p_A(x) = p_{P^*AP}(x) = \text{DET}(xI - P^*AP) = (x - \lambda)^k \text{DET}(xI - D)$.

So algeb. multipl of λ in $A =$ algeb. multipl of λ in $P^*AP \geq k$. ■

P. Schur unitary triangularization

Th. Let $A \in \mathcal{M}_n(\mathbb{C})$. Then there exists U unitary, s.t. $T = U^*AU$ is upper triangular. Further, if A and $\sigma(A)$ are real, then U can be real orthogonal.

Pr. Use induction on n . Case $n = 1$ is trivial. Let $n > 1$. Let $(\lambda_1, \mathbf{w}_1)$ be an eigenpair

of A , $\|\mathbf{w}_1\| = 1$. Take an orthonormal basis $\{\mathbf{w}_1, \dots, \mathbf{w}_n\}$ and form $W = [\mathbf{w}_1 \ \dots \ \mathbf{w}_n]$.

Then, $W^*AW = \begin{bmatrix} \lambda_1 & * \\ 0 & A' \end{bmatrix}$, $A' \in \mathcal{M}_{n-1}$.

$W^*AW = \begin{bmatrix} \lambda_1 & * \\ 0 & A' \end{bmatrix}$, $A' \in \mathcal{M}_{n-1}$. By induction hypothesis, there exists U' unitary, s.t. $T' = U'^*A'U'$ is upper triangular.

Put $U = W \begin{bmatrix} 1 & 0 \\ 0 & U' \end{bmatrix}$. Note: U is unitary. Now

$$U^*AU = \begin{bmatrix} 1 & 0 \\ 0 & U'^* \end{bmatrix} W^*AW \begin{bmatrix} 1 & 0 \\ 0 & U' \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & U'^* \end{bmatrix} \begin{bmatrix} \lambda_1 & * \\ 0 & A' \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & U' \end{bmatrix} = \begin{bmatrix} \lambda_1 & * \\ 0 & U'^*A'U' \end{bmatrix} = \begin{bmatrix} \lambda_1 & * \\ 0 & T' \end{bmatrix}.$$

Further: 'if A, λ are real then there exists $\mathbf{x} \neq \mathbf{0}$ real, s.t. $A\mathbf{x} = \lambda\mathbf{x}$ '. ■

P. Applications: Schur unitary triangularization (SUT)

Remark. In SUT, as $U^*AU = T$, we have

$$\{\lambda_1, \dots, \lambda_n\} = \sigma(A) = \sigma(T) = \{t_{11}, \dots, t_{nn}\}.$$

Further, we can get the λ_i 's in the diagonal of T in any prescribed order.

Cor. Let $\sigma(A) = \{\lambda_1, \dots, \lambda_n\}$. Then $\text{DET } A = \prod \lambda_i$ and $\text{TR } A = \sum \lambda_i$.

Pr. By SUT, there exists U unitary, s.t. $U^*AU = T$ is upper triangular. So $\sigma(T) = \sigma(A)$.

So $\sum \lambda_i = \sum t_{ii} = \text{TR}(T) = \text{TR}(U^*AU) = \text{TR}(AUU^*) = \text{TR}(A)$.

And $\prod \lambda_i = \prod t_{ii} = \text{DET } T = \text{DET } U^*AU = \text{DET } A$. ■

• By SUT, $U^*AU = T$ and

$$\{\lambda_1, \dots, \lambda_n\} = \sigma(A) = \sigma(T) = \{t_{11}, \dots, t_{nn}\}.$$

• By **DIAG A** we denote the diagonal $\{a_{11}, \dots, a_{nn}\}$ of A . By **DIAG v** we denote the diagonal matrix A with $a_{11} = \mathbf{v}_1, \dots, a_{nn} = \mathbf{v}_n$.

Th[spectral Theorem]. Let A be normal ($AA^* = A^*A$). Then there exists U unitary, s.t. $U^*AU = D$ is diagonal. Further, $\text{DIAG } D = \sigma(A)$.

Pr. By SUT, there exists U unitary, s.t. $U^*AU = T$ is upper triangular. As $A^*A = AA^*$, we get $T^*T = TT^*$. Note: $\sum |t_{1i}|^2 = (TT^*)_{11} = (T^*T)_{11} = |t_{11}|^2$. So $t_{12} = \dots = t_{1n} = 0$. Repeating the process, T is diagonal. ■

P. Applications: Schur unitary triangularization

Cor. Let A be Hermitian. Then there exists U unitary, s.t. $U^*AU = D$ is real diagonal. Further, if A is real, then U can be chosen as real orthogonal.

Pr. By SUT, we can find U unitary, s.t. $U^*AU = T$ is upper triangular. As $A^* = A$, we get $T^* = T$. So T is real diagonal. ■

Multiply

$$\begin{bmatrix} 0 & * & * & * \\ 0 & * & * & * \\ 0 & 0 & * & * \\ 0 & 0 & 0 & * \end{bmatrix} \begin{bmatrix} * & * & * & * \\ 0 & \mathbf{0} & * & * \\ 0 & 0 & * & * \\ 0 & 0 & 0 & * \end{bmatrix} = \begin{bmatrix} 0 & 0 & * & * \\ 0 & 0 & * & * \\ 0 & 0 & * & * \\ 0 & 0 & 0 & * \end{bmatrix}$$

$$\begin{bmatrix} 0 & 0 & * & * \\ 0 & 0 & * & * \\ 0 & 0 & * & * \\ 0 & 0 & 0 & * \end{bmatrix} \begin{bmatrix} * & * & * & * \\ 0 & * & * & * \\ 0 & 0 & \mathbf{0} & * \\ 0 & 0 & 0 & * \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 & * \\ 0 & 0 & 0 & * \\ 0 & 0 & 0 & * \\ 0 & 0 & 0 & * \end{bmatrix}$$

$$\begin{bmatrix} 0 & 0 & 0 & * \\ 0 & 0 & 0 & * \\ 0 & 0 & 0 & * \\ 0 & 0 & 0 & * \end{bmatrix} \begin{bmatrix} * & * & * & * \\ 0 & * & * & * \\ 0 & 0 & * & * \\ 0 & 0 & 0 & \mathbf{0} \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

So $\begin{bmatrix} \mathbf{0} & * & * & * \\ 0 & * & * & * \\ 0 & 0 & * & * \\ 0 & 0 & 0 & * \end{bmatrix} \begin{bmatrix} * & * & * & * \\ 0 & \mathbf{0} & * & * \\ 0 & 0 & * & * \\ 0 & 0 & 0 & * \end{bmatrix} \begin{bmatrix} * & * & * & * \\ 0 & * & * & * \\ 0 & 0 & \mathbf{0} & * \\ 0 & 0 & 0 & * \end{bmatrix} \begin{bmatrix} * & * & * & * \\ 0 & * & * & * \\ 0 & 0 & * & * \\ 0 & 0 & 0 & \mathbf{0} \end{bmatrix} = \mathbf{0}.$

Cor[Cayley-Hamilton]. Every matrix satisfies its characteristic equation.

Pr. Let $\sigma(A) = \{\lambda_1, \dots, \lambda_n\}$. By SUT, there exists U unitary, s.t. $U^*AU = T$ is upper triangular; $t_{ii} = \lambda_i$ for each i . Let $p(x)$ be the characteristic polynomial of A . So $p(x) = \prod(x - \lambda_i)$. So $p(A) = \prod(A - \lambda_i I) = \prod(UTU^* - \lambda_i UIU^*) = U \left[(T - \lambda_1 I)(T - \lambda_2 I) \cdots (T - \lambda_n I) \right] U^* = U \mathbf{0} U^* = \mathbf{0}$. ■

P. Diagonalizability

• We say ' A is **diagonalizable**' if there exists S invertible, s.t. $S^{-1}AS = D$ is diagonal. Obviously, $\lambda_i = d_{ii}$ are the eigenvalues of A .

Th. $A_{n \times n}$ is diagonalizable if and only if it has n linearly independent eigenvectors.

Pr. \Rightarrow : Let A be diagonalizable. So there exists S invertible, s.t. $S^{-1}AS = D$ is diagonal. Put $\mathbf{s}_i = S(:, i)$ and $\lambda_i = d_{ii}$. Then

$$[A\mathbf{s}_1 \cdots A\mathbf{s}_n] = A[\mathbf{s}_1 \cdots \mathbf{s}_n] = AS = SD = [\mathbf{s}_1 \cdots \mathbf{s}_n]D = [\lambda_1 \mathbf{s}_1 \cdots \lambda_n \mathbf{s}_n].$$

That is, $A\mathbf{s}_i = \lambda_i \mathbf{s}_i$. So $\mathbf{s}_1, \dots, \mathbf{s}_n$ are n linearly independent eigenvectors.

\Leftarrow : let $\mathbf{s}_1, \dots, \mathbf{s}_n$ be n linearly independent eigenvectors, $A\mathbf{s}_i = \lambda_i \mathbf{s}_i$.

Put $S = [\mathbf{s}_1 \cdots \mathbf{s}_n]$ and $D = \text{DIAG}(\lambda_1, \dots, \lambda_n)$. Then

$$AS = [A\mathbf{s}_1 \cdots A\mathbf{s}_n] = [\lambda_1 \mathbf{s}_1 \cdots \lambda_n \mathbf{s}_n] = SD. \text{ So } S^{-1}AS = D. \quad \blacksquare$$

Cor. Let $A_{n \times n}$ have n distinct eigenvalues. Then A is diagonalizable.

Pr. Follows as the corresponding eigenvectors are linearly independent. ■

P. Non-diagonalizability-Example

Q. Is $A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$ diagonalizable? No. Assume it is. Then there exists S invertible, s.t. $S^{-1}AS = D$ is diagonal. As $\sigma(A) = \sigma(S^{-1}AS) = \sigma(D)$, we must have $D = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$. But

then $A = SDS^{-1} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$, a contradiction.