

5. Eigenvectors & Eigenvalues

In this chapter we work mainly with complex vector spaces.

* Inner product on \mathbb{C}^n : $x = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}, y = \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix} \in \mathbb{C}^n$

$\Rightarrow \langle x, y \rangle = \sum_{j=1}^n x_j \bar{y}_j$

• linear in first component, "antilinear" in second comp.

$\|x\| = \left(\sum_{j=1}^n |x_j|^2 \right)^{1/2}$

Properties:

- $\forall x \in \mathbb{C}^n: \langle x, x \rangle \geq 0$
- $\langle x, x \rangle = 0 \Leftrightarrow x = 0$
- $\forall x, y \in \mathbb{C}^n: \langle x, y \rangle = \overline{\langle y, x \rangle}$

• Triangle inequality: $\|x + y\| \leq \|x\| + \|y\|$

• Cauchy-Schwarz: $|\langle x, y \rangle| \leq \|x\| \|y\|$

* $A \in M(n \times n, \mathbb{C})$ is called hermitian $\Leftrightarrow A = A^*$

where $A^* =$ transposed and conjugated matrix of A

$(A = (a_{ij})_{i,j=1, \dots, n} \Rightarrow A^* = (\bar{a}_{ji})_{i,j=1, \dots, n})$

* $U \in M(n \times n, \mathbb{C})$ is called unitary $\Leftrightarrow U^* = U^{-1}$

It can be shown: $A, B \in M(n \times n) \Rightarrow (B = A^* \Leftrightarrow \forall x, y \in \mathbb{C}^n \langle Ax, y \rangle = \langle x, By \rangle)$

Observe: $\forall A \in M(n \times n): \det(A^*) = \overline{\det(A)}$

In particular: A hermitian $\Rightarrow \det(A^*) = \det(A) \in \mathbb{R}$.

• U unitary $\Rightarrow |\det U| = 1$

Some facts on polynomials:

Let p be a polynomial of degree n with complex coefficients.

$\Rightarrow \exists \lambda_1, \dots, \lambda_k \in \mathbb{C}, r_1, \dots, r_k \in \mathbb{N}, d \in \mathbb{C}$ st.

$p(x) = a(x - \lambda_1)^{r_1} \dots (x - \lambda_k)^{r_k} + d$

If p has real coefficients and λ is a zero of p then $\bar{\lambda}$ is a zero of p too.

Proof. Assume $p(\lambda) = 0$

$\Rightarrow p(\bar{\lambda}) = \overline{p(\lambda)} = \overline{0} = 0$
coeff. of p are real.

5.1. Eigenvectors & Eigen values

Definition. Let V be a finite-dimensional \mathbb{K} -vector space ($\mathbb{K} = \mathbb{R}$ or \mathbb{C})

Let $T: V \rightarrow V$ linear. Then $\lambda \in \mathbb{K}$ is called an eigenvalue of T

if the equation $(T - \lambda \mathbb{1})v = 0$ (*)

has a non-trivial solution v .

Every non-trivial solution v of (*) is called eigenvector of T .

For fixed λ , we set $E_\lambda := E_\lambda(T) := \text{Eig}(\lambda, T) = \{v \in V \mid (T - \lambda)v = 0\}$

E_λ is called the eigenspace of T .

The set of all eigenvectors of T is called the spectrum of T , denoted by $\sigma(T)$.

Notation. Instead of $\lambda \mathbb{1}$ we write simply λ .

Lemma. For every eigenvalue λ of T , we have

$E_\lambda = \{ \text{all eigenvectors of } T \text{ for } \lambda \} \cup \{0\}$
 $= \ker(T - \lambda)$

In particular, E_λ is a subspace of V .

Remark: 0 EVal. of A
 $\Leftrightarrow A$ is not inv.
 and $\ker A = E_0(A)$.

Definition $\dim(E_\lambda(\tau)) =$ geometric multiplicity of the eigenvalue λ .

Examples:

$$A: \mathbb{R}^6 \rightarrow \mathbb{R}^6, \quad A = \begin{pmatrix} 1 & 1 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{pmatrix}$$

$\Rightarrow A$ has three eigenvalues $\lambda_1 = 1, \lambda_2 = 2, \lambda_3 = 3$.
geometric multiplicity of λ_1 is 3
of λ_2 is 1
of λ_3 is 2.

$$A: \mathbb{R}^2 \rightarrow \mathbb{R}^2, \quad A = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}$$

$\Rightarrow A$ has exactly one eigenvalue: $\lambda = 1$, and its geometric multiplicity is 1.

$$A: \mathbb{R}^2 \rightarrow \mathbb{R}^2, \quad A = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} \quad [\text{Reflection on } x\text{-axis}]$$

$\Rightarrow A$ has two eigenvalues: $\lambda_1 = 1, \lambda_2 = -1$.

$$A: \mathbb{R}^2 \rightarrow \mathbb{R}^2, \quad A = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix} \quad [\text{Rotation by } -90^\circ].$$

Clearly, A cannot have eigenvalues.

BUT: $A: \mathbb{C}^2 \rightarrow \mathbb{C}^2, A = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}$ has the eigenvalues

$$\lambda_{\pm} = \pm i \text{ with eigenvectors } \begin{pmatrix} 1 \\ i \end{pmatrix} \text{ and } \begin{pmatrix} 1 \\ -i \end{pmatrix}.$$

• Orth. proj on $U \Rightarrow \mathcal{G}(P) \in \{0, I\}$. $E_0(P) = \ker P = \mathcal{U}_\perp$, $E_1(P) = \text{Im } P = \mathcal{U}$.

Theorem: Let $T: V \rightarrow V$ linear and let $\lambda_1, \dots, \lambda_k$ be different eigenvalues of T with eigenvectors v_1, \dots, v_k .

$\Rightarrow \{v_1, \dots, v_k\}$ are lin. independent.

Proof. By induction:

• Base case: We will show that v_1 and v_2 are lin. indep.

Assume they are lin. dep $\Rightarrow \exists c \in \mathbb{K}$ s.t. $v_1 = cv_2$ (recall: $v_1, v_2 \neq 0$).

$$\Rightarrow \lambda_1 v_1 = A v_1 = A(c v_2) = c A v_2 = c \lambda_2 v_2 = \lambda_2 v_1$$

$$\Rightarrow \lambda_1 = \lambda_2 \quad \square$$

• Induction step: Let $2 \leq j < k$ and assume that $\{v_1, \dots, v_j\}$ are lin. ind. We will show that $\{v_1, \dots, v_{j+1}\}$ are lin. ind.

Let $c_1, \dots, c_{j+1} \in \mathbb{K}$ s.t.

$$0 = c_1 v_1 + \dots + c_{j+1} v_{j+1} \quad (*)$$

Apply A to $(*)$

$$0 = A(c_1 v_1 + \dots + c_{j+1} v_{j+1})$$

$$= c_1 \lambda_1 v_1 + \dots + c_{j+1} \lambda_{j+1} v_{j+1} \quad (**)$$

and $0 = \lambda_1 c_1 v_1 + \dots + \lambda_j c_j v_j + \dots + \lambda_{j+1} c_{j+1} v_{j+1}$ (2) Multiply $(*)$ by λ_{j+1}

$$(1) - (2) \Rightarrow 0 = c_1 (\lambda_1 - \lambda_{j+1}) v_1 + \dots + c_j (\lambda_j - \lambda_{j+1}) v_j$$

$$\underbrace{c_1 (\lambda_1 - \lambda_{j+1})}_{\substack{v_1 - v_j \\ \text{lin. ind}}} = 0, \dots, c_j (\lambda_j - \lambda_{j+1}) = 0$$

$$\Rightarrow c_1 = c_2 = \dots = c_j = 0 \quad \text{because } \lambda_k \neq \lambda_l \text{ for } k \neq l. \quad \square$$

Remark: $A|_{E_\lambda(A)}$ acts as $\lambda \cdot I|_{E_\lambda(A)}$

S.2. The characteristic polynomial.

How do we find the eigenvalues of a linear map $A: \mathbb{R}^n \rightarrow \mathbb{R}^n$?

We may assume that A is given as a matrix.

Then:

$$\lambda \in \mathcal{G}(A) \Leftrightarrow \dim(E_\lambda(A)) \geq 1$$

$$\Leftrightarrow \dim(\ker(A - \lambda I)) \geq 1$$

$$\Leftrightarrow A - \lambda I \text{ is not injective}$$

$$\Leftrightarrow \det(A - \lambda I) = 0.$$

Observation: For $A \in M(n \times n)$ fixed,

$$p(\lambda) := \det(A - \lambda I)$$

is a polynomial of degree n , called the characteristic polynomial of A .

It is of the form $p(\lambda) = (-1)^n \lambda^n + a_{n-1} \lambda^{n-1} + \dots + a_0 \lambda + a_0$,

where $a_0 = p(0) = \det A$, and $a_{n-1} = \sum_{j=1}^n a_j = \text{tr}(A)$

$=$ trace of A .

In summary, we obtain the following Theorem:

Theorem: $A \in M(n \times n, \mathbb{C}) \Rightarrow \begin{cases} \lambda \in \mathbb{C} \text{ is an eigenvalue of } A \\ \Leftrightarrow \lambda \text{ is a zero of the charact. polynomial of } A. \end{cases}$

Definition: $A \in M(n \times n, \mathbb{C})$ with eigenvalues $\lambda_1, \dots, \lambda_k$. ($\lambda_j \neq \lambda_k$)
 $\Rightarrow \exists v_1, \dots, v_k \in \mathbb{C}^n$ s.t. $\det(A - \lambda) = (\lambda - \lambda_1)^{g_1} \dots (\lambda - \lambda_k)^{g_k}$

Then g_j is called the algebraic multiplicity of the eigenvalue λ_j .

It can be shown: For every eigenvalue: geometric mult \leq algebraic mult.

Example: $A = \begin{pmatrix} 1 & \\ & 0 \end{pmatrix}$

- $\Rightarrow \det(A - \lambda) = (1 - \lambda)^2 \Rightarrow A$ has only eigenvalue, $\lambda = 1$.
- algebraic mult. of $\lambda = 1$: 2.
- geometric mult. of $\lambda = 1$: 1 because: $A - 1 = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}$.
- Clearly: $\dim(\ker(A - 1)) = 1$.
- $E_1(A) = \ker(A - 1) = \text{span}\left\{ \begin{pmatrix} 1 \\ 0 \end{pmatrix} \right\}$.

Receipt for finding eigenvectors & eigenvalues of A:

- Calculate $p(\lambda) = \det(A - \lambda)$
- Find all zeros $\lambda_1, \dots, \lambda_k$ of p . These are the eigenvalues of A
- Find $\ker(A - \lambda_j)$ for $j = 1, \dots, k$.

Examples:

- A diagonal matrix: $A = \begin{pmatrix} \lambda_1 & & \\ & \dots & \\ & & \lambda_n \end{pmatrix}$
 \Rightarrow all eigenvalues are $\lambda_1, \dots, \lambda_n$ and for every eigenvalue: geometric mult. = algebraic mult.
- Moreover: $v_j = \vec{e}_j$ is eigenvector of A with eigenvalue λ_j

A upper triangular matrix: $A = \begin{pmatrix} \lambda_1 & & * \\ & \dots & \\ 0 & & \lambda_n \end{pmatrix}$
 \Rightarrow The eigenvalues of A are $\lambda_1, \dots, \lambda_n$.

Theorem: $A \in M(n \times n)$. Then:

- A has n lin. ind. Eigenvectors. $\Leftrightarrow \forall \lambda$ eigenvalue of A: algebraic mult. = geom. mult.
- $\Leftrightarrow \mathbb{R}^n$ has a basis consisting of E vectors of A.

Corollary: A has n different eigenvalues

$\Rightarrow \mathbb{R}^n$ has a basis of Eigenvectors of A.
 With respect to this basis, A is of the form $\begin{pmatrix} \lambda_1 & & \\ & \dots & \\ & & \lambda_n \end{pmatrix}$,
 and $\mathbb{R}^n = E_{\lambda_1}(A) \oplus \dots \oplus E_{\lambda_k}(A)$ where μ_1, \dots, μ_k are the different E values of A.

Useful theorems:

Theorem: (Cayley-Hamilton) $A \in M(n \times n)$ and $p(\lambda) = \det(A - \lambda)$.
 $\Rightarrow p(A) = 0$.

Definition. $A, B \in M(n \times n)$. A, B are called similar if there exists a $C \in M(n \times n)$, invertible, such that

$$A = C^{-1} B C \quad (*)$$

C is called a similarity transformation.

If A and B are similar, we write $A \sim B$.

Observe: $(*) \Leftrightarrow CA = BC$.

Proposition.

(i) \sim is an equivalence relation; that is:

- $R \sim R$
- $R \sim S, S \sim T \Rightarrow R \sim T$
- $R \sim S \Rightarrow S \sim R$

(ii) For $C \in M(n \times n)$, C invertible; fixed, the map

$$T: M(n \times n) \rightarrow M(n \times n), A \mapsto C^{-1} A C$$
 is a linear map.

Observation. Let $A, B, C \in M(n \times n)$, C invertible.

$\Rightarrow C$ can be seen as the matrix of a change of bases and

" $A = C^{-1} B C$ " says that A and B describe the same linear map, but with respect to two different bases.

Theorem. Let $A, B \in M(n \times n)$ s.t. $A \sim B$.

$\Rightarrow A$ and B have the same characteristic polynomial. In particular, they have the same eigenvalues (but in general not the same eigenvectors).

Proof. Let $C \in M(n \times n)$ s.t. $A = C^{-1} B C$

$$\begin{aligned} \Rightarrow P_A(\lambda) &= \det(A - \lambda I) = \det(C^{-1} B C - \lambda C^{-1} C) = \det(C^{-1} (B - \lambda I) C) \\ &= (\det C^{-1}) \det(B - \lambda I) \det C = \det(B - \lambda I) = P_B(\lambda). \\ &= (\det C)^{-1} \det C \end{aligned}$$

□

Relation between eigenvectors: Let $v \in \mathbb{R}^n$ s.t. $(A - \lambda) v = 0$

$$\Rightarrow (A - \lambda) v = 0 \Leftrightarrow (A - \lambda) C C^{-1} v = 0 \Leftrightarrow C^{-1} (A - \lambda) C C^{-1} v = 0$$

$$\Leftrightarrow (C^{-1} A C - \lambda C^{-1} C) C^{-1} v = 0$$

$$\Leftrightarrow (B - \lambda) C^{-1} v = 0$$

$\Rightarrow (v$ Eigenvector of A with Eigenvalue $\lambda \Leftrightarrow C^{-1} v$ eigenvector of B with Eigenvalue λ)

Definition. $A \in M(n \times n)$ is called diagonalizable if and only if A is similar to a diagonal matrix.

Observation: A diagonalizable $\Rightarrow \det A = \lambda_1 \cdot \dots \cdot \lambda_n$ (λ_j : eigenvalues of A)

Theorem. A is diagonalizable $\Leftrightarrow \mathbb{R}^n$ has a basis consisting of eigenvectors of A .

In particular: A has n different eigenvalues $\Rightarrow A$ is diagonalizable

Proof. \Rightarrow Suppose that A is diagonalizable.

Let $D = \text{diag}(\lambda_1, \dots, \lambda_n)$ and $C \in M(n \times n)$ s.t. $D = C^{-1} A C$

$$\Rightarrow C D = A C$$

Let $\vec{c}_1, \dots, \vec{c}_n$ be the columns of C .

$$\Rightarrow A \vec{c}_j = j\text{-th column of } A C$$

$$= j\text{-th column of } C D \text{ by } (*)$$

$$= \lambda_j \cdot (j\text{-th column of } C)$$

$$= \lambda_j \vec{c}_j$$

$\Rightarrow \vec{c}_j$ is an eigenvector of A with eigenvalue λ_j .

Since C is invertible, the $\vec{c}_1, \dots, \vec{c}_n$ form a basis of \mathbb{R}^n and " \Leftarrow " is proved.

" \Leftarrow " Let $\vec{v}_1, \dots, \vec{v}_n$ be a basis of \mathbb{R}^n consisting of eigenvectors of A with eigenvalues $\lambda_1, \dots, \lambda_n$ respectively.

$\Rightarrow V_j$: $A \vec{v}_j = \lambda_j \vec{v}_j$

$$\text{Let } C = (\vec{v}_1 | \dots | \vec{v}_n)$$

$$\Rightarrow C^{-1} A C = \text{diag}(\lambda_1, \dots, \lambda_n) = A \text{ represented in the basis } \vec{v}_1, \dots, \vec{v}_n.$$

Eigenvectors & Eigenvalues for linear maps (not necessarily matrices):

Let V be a vector space with $\dim V = n < \infty$ and

$T: V \rightarrow V$ linear.

Choose a basis $B := \{\vec{v}_1, \dots, \vec{v}_n\}$ of V and let A be the matrix repr of T with respect to B .

Then: Characteristic polynomial of $T := P_T(\lambda) := \det(A - \lambda I)$.

Observe: P_T does not depend on the chosen basis.

Because: Let $B' := \{\vec{w}_1, \dots, \vec{w}_n\}$ basis of V and A' the matrix repr of T wrt. B' .

Let $C =$ transition matrix from B to B' .

$\Rightarrow A = C^{-1}A'C$ $\Rightarrow A$ and A' are similar.

$\Rightarrow A$ and A' have the same char. poly.

Clearly: λ Eigenvalue of $T \Leftrightarrow T - \lambda I$ not invertible

$\Leftrightarrow A - \lambda I$ not injective

$\Leftrightarrow \det(A - \lambda I) = 0$

$\Leftrightarrow P_T(\lambda) = 0$.

□

Example

$T: \mathbb{R}^3 \rightarrow \mathbb{R}^3, T \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} -y+z \\ -3x-2y+3z \\ -2x-2y+3z \end{pmatrix}$

Find Eigenvalues & Eigenvectors of T .

Solution. Matrix repr of $T: A = \begin{pmatrix} 0 & -1 & 1 \\ -3 & -2 & 3 \\ -2 & -2 & 3 \end{pmatrix}$

Char. polynomial:

$p(\lambda) = \det(A - \lambda I) = \det \begin{pmatrix} -\lambda & -1 & 1 \\ -3 & -2-\lambda & 3 \\ -2 & -2 & 3-\lambda \end{pmatrix}$

$= \lambda(2+\lambda)(3-\lambda) + 6 + 6 - [2(2+\lambda) + 6\lambda + 3(3-\lambda)]$

$= -\lambda^3 + \lambda^2 + 6\lambda + 12 - 5\lambda - 13$

$= -\lambda^3 + \lambda^2 + \lambda - 1 = (\lambda-1)[- \lambda^2 + 1] = -(\lambda-1)^2(\lambda+1)$.

\Rightarrow Eigenvalues of $T: \lambda_1 = 1, \lambda_2 = -1$.

Eigenspaces:

$\bullet \lambda_1: A - \lambda_1 I = \begin{pmatrix} -1 & -1 & 1 \\ -3 & -3 & 3 \\ -2 & -2 & 2 \end{pmatrix} \rightarrow \begin{pmatrix} -1 & -1 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$

$\Rightarrow \dim E_{\lambda_1} = 2, E_{\lambda_1} = \text{span} \left\{ \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \right\}$.

$\bullet \lambda_2: A - \lambda_2 I = \begin{pmatrix} 1 & -1 & 1 \\ -3 & -1 & 3 \\ -2 & -2 & 4 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & -1 & 1 \\ 0 & -4 & 6 \\ 0 & 2 & -3 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & -1 & 1 \\ 0 & 2 & -3 \\ 0 & 0 & 0 \end{pmatrix}$

$\Rightarrow \dim E_{\lambda_2} = 1, E_{\lambda_2} = \text{span} \left\{ \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} \right\}$.

$\Rightarrow \vec{v}_1, \vec{v}_2, \vec{v}_3$ is a basis of $\mathbb{R}^3 \Rightarrow A$ is diagonalizable.

Let $C = (\vec{v}_1 | \vec{v}_2 | \vec{v}_3)$

$\Rightarrow C^{-1}AC = \begin{pmatrix} 1 & & \\ & -1 & \\ & & -1 \end{pmatrix}$

S.S. Symmetric and Hermitian matrices.

Observe $A \in \mathbb{R}^{n \times n}$ is symmetric \Rightarrow A is Hermitian

Theorem. Let $A \in M(n \times n, \mathbb{C})$ be a Hermitian matrix.

\Rightarrow All eigenvalues of A are real.

Proof. Let λ be an eigenvalue of A . We have to show: $\lambda = \bar{\lambda}$.

$$\text{Let } x \in \mathbb{C}^n \setminus \{0\} \text{ s.t. } (A - \lambda)x = 0$$

$$\Rightarrow \lambda \|x\|^2 = \lambda \langle x, x \rangle = \langle \lambda x, x \rangle = \langle Ax, x \rangle = \langle x, A^* x \rangle$$

$$= \langle x, Ax \rangle = \bar{\lambda} \|x\|^2.$$

$$\Rightarrow \lambda \|x\|^2 = \bar{\lambda} \|x\|^2 \Rightarrow \lambda = \bar{\lambda} \quad (\text{because } \|x\| \neq 0).$$

□

Theorem. Let $A \in M(n \times n, \mathbb{C})$ be Hermitian. Let $\lambda \neq \mu$ be eigenvalues of A and \vec{v}, \vec{w} eigenvectors respectively. Then: $\vec{v} \perp \vec{w}$.

Proof. Assume $\vec{v}, \vec{w} \in \mathbb{C}^n \setminus \{0\}$, $(A - \lambda)\vec{v} = 0$, $(A - \mu)\vec{w} = 0$

$$\Rightarrow \lambda \langle \vec{v}, \vec{w} \rangle = \langle \lambda \vec{v}, \vec{w} \rangle = \langle A\vec{v}, \vec{w} \rangle = \langle A\vec{v}, \vec{w} \rangle = \langle \vec{v}, A^* \vec{w} \rangle = \langle \vec{v}, A\vec{w} \rangle = \langle \vec{v}, \mu \vec{w} \rangle = \mu \langle \vec{v}, \vec{w} \rangle$$

$$\Rightarrow (\lambda - \mu) \langle \vec{v}, \vec{w} \rangle = 0 \stackrel{\lambda \neq \mu}{\Rightarrow} \langle \vec{v}, \vec{w} \rangle = 0 \Rightarrow \vec{v} \perp \vec{w}.$$

□

This shows: If A is Hermitian and diagonalizable, then \mathbb{C}^n has

an ONB of eigenvectors of A (For every $\lambda \in \sigma(A)$, choose

an ONB in $\mathbb{E}_\lambda(A)$). The union of these bases gives then a basis of \mathbb{C}^n)

It can be shown:

Thm. A Hermitian \Rightarrow A diagonalizable.

With the above, it follows that every Hermitian Matrix A has an ONB of eigenvectors.

Definition. $A \in M(n \times n, \mathbb{C})$ admits an orthogonal diagonalization if there is a diagonal matrix D and an orthogonal matrix Q such that $D = Q^t A Q$

Clearly, this is the case if and only if A has an ONB of eigenvectors of A

Theorem. $A \in M(n \times n, \mathbb{C})$. Then:

A Hermitian $\Leftrightarrow A$ has an orthogonal diagonalization.

□

Other useful theorems:

Let $A \in M(n \times n, \mathbb{K})$. A is called triangularizable if it is similar to an upper triangular matrix.

Theorem. $A \in M(n \times n, \mathbb{K})$. Then the following is equivalent:

(i) A is triangularizable

(ii) The characteristic polynomial of A is product of linear factors:

$$P_A(\lambda) = (\lambda_1 - \lambda) \cdots (\lambda_n - \lambda).$$

Corollary. Every $A \in M(n \times n, \mathbb{C})$ is triangularizable.

This is false for $\mathbb{K} = \mathbb{R}$. For example, $A = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}$ is not triangularizable over \mathbb{R} . [Suppose it were. $\Rightarrow A$ must have eigenvalues; but?

$$P_A(\lambda) = \det \begin{pmatrix} -\lambda & 1 \\ 1 & -\lambda \end{pmatrix} = \lambda^2 + 1 \text{ has no zeros in } \mathbb{R}.$$