

if there exists a scalar $\lambda \in F$ such that $AX = \lambda X$ holds. ($V_n(F)$ is a vector space of dimension n over the field F)
 Let X be an eigenvector of the matrix A . Then for some suitable scalar λ , $AX = \lambda X$ holds. That is,

$(A - \lambda I_n)X = 0$. This is a homogeneous system of n equations in n unknowns. Since there exists a non-null solution of the system, $\det(A - \lambda I_n) = 0$. This implies that λ is an eigenvalue of A . Thus for an eigenvector, if it exists, there corresponds an eigenvalue of A .

Theorem 3.5.11 Let A be an $n \times n$ matrix over a field F . To an eigenvector of A there corresponds a unique eigenvalue of A .

Proof: Let there be two distinct eigenvalues λ_1, λ_2 of A corresponding to an eigenvector X . Then $AX = \lambda_1 X$ and $AX = \lambda_2 X$. So, $(\lambda_1 - \lambda_2)X = 0$. But this is a contradiction, since X is a non-null vector and $\lambda_1 \neq \lambda_2$. Hence the theorem.

Theorem 3.5.12 Let A be an $n \times n$ matrix over a field F and λ be an eigenvalue belonging to F .

To each such eigenvalue of A there corresponds at least one eigenvector.

Proof: Since λ is an eigenvalue, $\det(A - \lambda I) = 0$. So, the homogeneous system of equations $(A - \lambda I_n)X = 0$ has a non-null solution,

say $X = X_1$ where $X_1 \in \mathbb{P}^n(F)$. Then $(A - \lambda I_n) X_1 = 0$
 or, $A X_1 = \lambda X_1$. This shows that X_1 is an eigen vector
 of A corresponding to λ . This proves the theorem.

Note: In fact, there are many eigen vectors of A
 corresponding to an eigen value $\lambda \in F$, because
 $\det(A - \lambda I_n) = 0$ implies that there are infinite
 number of non-null solutions of the homogeneous system
 $(A - \lambda I_n) X = 0$ and each such non-null solution gives an
 eigen vector of A corresponding to λ .

Example Let $A = \begin{pmatrix} 1 & 3 \\ 4 & 5 \end{pmatrix}$; The eigen values of A are
 $-1, 7$. Let $X = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$ be an eigen vector corresponding
 to -1 . Then $A X = -X$ and this gives $2x_1 + 3x_2 = 0$
 $4x_1 + 6x_2 = 0$

The equivalent system is $x_1 + \frac{3}{2}x_2 = 0$

The solution of the system is $k(-\frac{3}{2}, 1)$ where $k \in \mathbb{R}$.

The eigen vectors are $k \begin{pmatrix} -3/2 \\ 1 \end{pmatrix}$ or equivalently, $c \begin{pmatrix} 3 \\ -2 \end{pmatrix}$, where

~~c~~ c is a non-zero real number.

Let $X = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$ be an eigen vector corresponding to

7 . Then $A X = 7X$ and this gives $-6x_1 + 3x_2 = 0$
 $4x_1 - 2x_2 = 0$

The system is equivalent to $x_1 - \frac{1}{2}x_2 = 0$

The solution of the system is $c(1, 2)$, where $c \in \mathbb{R}$

Therefore the eigen vectors are $c \begin{pmatrix} 1 \\ 2 \end{pmatrix}$, where c is a non-zero
 real number.

Theorem 3.5.13 If X_i be the eigen vector corresponding to eigen value λ_i , $i=1, 2, \dots, k$ (where $\lambda_1, \lambda_2, \dots, \lambda_k$ are all distinct) of the square matrix A over a field F , then $\{X_1, X_2, \dots, X_k\}$ is linearly independent.

Proof: $AX_i = \lambda_i X_i$, $i=1, 2, \dots, k$

Let us consider the relation $\sum_{i=1}^k c_i X_i = 0$ — (1)

$$\text{So, } \sum_{i=1}^k c_i (AX_i) = 0$$

$$\text{or, } \sum_{i=1}^k c_i \lambda_i X_i = 0 \quad \text{--- (2)}$$

$$\text{Similarly, } \sum_{i=1}^k c_i \lambda_i^2 X_i = 0$$

$$\text{or, } \sum_{i=1}^k c_i \lambda_i^2 X_i = 0 \quad \text{--- (3)}$$

$$\text{Similarly, } \sum_{i=1}^k c_i \lambda_i^3 X_i = 0 \quad \text{--- (4)}$$

$$\dots \dots \dots \sum_{i=1}^k c_i \lambda_i^{k-1} X_i = 0 \quad \text{--- (k)}$$

So, we can write above equations (1) to (k),

$$(c_1 X_1, c_2 X_2, \dots, c_k X_k) \begin{pmatrix} 1 & \lambda_1 & \dots & \lambda_1^{k-1} \\ 1 & \lambda_2 & \dots & \lambda_2^{k-1} \\ \dots & \dots & \dots & \dots \\ 1 & \lambda_k & \dots & \lambda_k^{k-1} \end{pmatrix} = (0, 0, \dots, 0) \quad \text{--- (A)}$$

$$\text{Let } P = \begin{pmatrix} 1 & \lambda_1 & \dots & \lambda_1^{k-1} \\ 1 & \lambda_2 & \dots & \lambda_2^{k-1} \\ \vdots & \vdots & \dots & \vdots \\ 1 & \lambda_k & \dots & \lambda_k^{k-1} \end{pmatrix} \text{ Now } \det P = \prod_{i < j} (\lambda_i - \lambda_j) \neq 0$$

as $\lambda_1, \lambda_2, \dots, \lambda_k$ are all distinct, So, P^{-1} exists

Pre-multiplying both sides of (A) by P^{-1} , we have

$$(c_1 X_1, c_2 X_2, \dots, c_k X_k) = (0, 0, \dots, 0)$$

So, $c_i = 0$, $i=1, 2, \dots, k$ since x_1, x_2, \dots, x_k are all non-zero.

This proves that $\{x_1, x_2, \dots, x_k\}$ is linearly dependent.

Theorem 3.5.14 The eigenvectors of an $n \times n$ matrix A over a field F corresponding to an eigen value $\lambda \in F$, together with the null vector, form a vector space, a subspace of $V_n(F)$.

Proof: To an eigen value $\lambda \in F$, there corresponds an eigen vector of A . Let S be the set of all eigen vectors of A corresponding to λ . Then S is the set of all non-null solutions of the homogeneous system of equations $(A - \lambda I_n)x = 0$. The null vector $0 \in V_n(F)$ is also a solution of the system.

As the solutions of a homogeneous system with an $n \times n$ matrix over F as the coefficient matrix form a subspace of $V_n(F)$, it follows that $S \cup \{0\}$ is a subspace of $V_n(F)$.

In other words, the eigenvectors corresponding to λ , together with the null vector, form a non-null subspace of $V_n(F)$. This completes the proof.

Definition 3.5.15 The subspace $E_\lambda = \{x \in V_n(F) : Ax = \lambda x\}$ in Theorem 3.5.14 is called the eigenspace corresponding to an eigen value λ .

Theorem 3.5.16 If λ be an r -fold eigen value of an $n \times n$ matrix A , then rank of $(A - \lambda I_n) \geq n - r$.

Proof: Since λ is an r -fold eigen value of A ,