

Invariant Subspaces And Direct Sums:

It is useful to decompose a finite dimensional vector space V into a direct sum of as many T -invariant subspaces as possible because the behaviour of T on V can be inferred from its behaviour on the direct summands. For example, T is diagonalizable if and only if V can be decomposed into a direct sum of one-dimensional T -invariant subspaces.

Theorem 5.24. Let T be a linear operator on a finite-dimensional vector space V , and suppose that $V = W_1 \oplus W_2 \oplus \dots \oplus W_k$ where W_i is a T -invariant subspace of V for each i ($1 \leq i \leq k$). Suppose that $f_i(t)$ is the characteristic polynomial of $T|_{W_i}$ ($1 \leq i \leq k$). Then $f_1(t) \cdot f_2(t) \cdot \dots \cdot f_k(t)$ is the characteristic polynomial of T .

Proof. The proof is by mathematical induction on k . Let $f(t)$ denote the characteristic polynomial of T . Suppose first that $k=2$. Let $\underline{\beta}_1$ be an ordered basis for W_1 , $\underline{\beta}_2$ an ordered basis for W_2 , and $\underline{\beta} = \underline{\beta}_1 \cup \underline{\beta}_2$. Then $\underline{\beta}$ is an ordered basis for V by Theorem 5.10 (d). Let $A = [T]_{\underline{\beta}}$, and ~~$[T]_{\underline{\beta}}$~~ and $\underline{B}_1 = [T|_{W_1}]_{\underline{\beta}_1}$ and $\underline{B}_2 = [T|_{W_2}]_{\underline{\beta}_2}$. It follows that

$$\Rightarrow A = \begin{bmatrix} \underline{B}_1 & \underline{0} \\ \underline{0}' & \underline{B}_2 \end{bmatrix}$$

where $\underline{0}$ and $\underline{0}'$ are zero matrices of the appropriate sizes.

Then $f(t) = \det(A - tI)$

$$= \det \begin{pmatrix} B_1 - tI_1 & 0 \\ 0 & B_2 - tI_2 \end{pmatrix}$$

$$= \det(B_1 - tI_1) \cdot \det(B_2 - tI_2)$$

$$\Rightarrow f(t) = f_1(t) \cdot f_2(t)$$

Here I_1 and I_2 are identity matrices of appropriate sizes. Since $f(t) = f_1(t) \cdot f_2(t)$. This proves the result for $k=2$.

Now assume that the theorem is valid for $k-1$ summands, where $k-1 \geq 2$ and suppose that V is a direct sum of k subspaces, say

$$V = W_1 \oplus W_2 \oplus \dots \oplus W_k$$

Let $W = W_1 \oplus W_2 \oplus \dots \oplus W_{k-1}$. Since W_1, W_2, \dots, W_{k-1} are T -invariant subspaces, W is also a T -invariant subspace of V and $V = W \oplus W_k$. So by the case for $k=2$,

$f(t) = g(t) \cdot f_k(t)$, where $g(t)$ is the characteristic polynomial of $T|_W$. Clearly $W = W_1 \oplus W_2 \oplus \dots \oplus W_{k-1}$ and therefore $g(t) = f_1(t) \cdot f_2(t) \cdot \dots \cdot f_{k-1}(t)$ by the induction hypothesis. We conclude

$$f(t) = g(t) \cdot f_k(t) = f_1(t) \cdot f_2(t) \cdot \dots \cdot f_{k-1}(t) \cdot f_k(t).$$

As an illustration of this result, suppose that T is a diagonalizable linear operator on a finite-dimensional vector

space V with distinct eigenvalues $\lambda_1, \lambda_2, \dots, \lambda_k$. By Theorem 5.11, V is a direct sum of the eigenspaces of T . Since each eigenspace is T -invariant, we can use Theorem 5.24. For each eigenvalue λ_i , the restriction of T to E_{λ_i} has characteristic polynomial $(\lambda_i - t)^{m_i}$, where m_i is the dimension of E_{λ_i} . By Theorem 5.24, the characteristic polynomial of T is the product

$$f(t) = (\lambda_1 - t)^{m_1} (\lambda_2 - t)^{m_2} \dots (\lambda_k - t)^{m_k}$$

It follows the multiplicity of each eigenvalue is equal to the dimension of the corresponding eigenspace.

Example 8. Let T be the linear operator on \mathbb{R}^4 defined by

$$T(a, b, c, d) = (2a - b, a + b, c - d, c + d)$$

and let $W_1 = \{(s, t, 0, 0) : s, t \in \mathbb{R}\}$ and

$$W_2 = \{(0, 0, s, t) : s, t \in \mathbb{R}\}.$$

for any $(a, b, 0, 0) \in W_1 \Rightarrow T(a, b, 0, 0) \in W_1$ since

$$T(a, b, 0, 0) = (2a - b, a + b, 0, 0)$$

\Rightarrow W_1 is T -invariant.

Also for any $(0, 0, c, d) \in W_2 \Rightarrow T(0, 0, c, d) \in W_2$ since

$$T(0, 0, c, d) = (0, 0, c - d, c + d)$$

\Rightarrow W_2 is also T -invariant.

Now, any $(a, b, c, d) \in \mathbb{R}^4$ can be written as

$$(a, b, c, d) = (a, b, 0, 0) + (0, 0, c, d)$$

where $(a, b, 0, 0) \in W_1$ and $(0, 0, c, d) \in W_2$

$$\Rightarrow \mathbb{R}^4 = W_1 + W_2$$

$$\text{since } W_1 \cap W_2 = \{0, 0, 0, 0\}$$

$$\Rightarrow \underline{\mathbb{R}^4 = W_1 \oplus W_2}$$

Let $\beta_1 = \{e_1, e_2\}$, $\beta_2 = \{e_3, e_4\}$, and $\beta = \beta_1 \cup \beta_2 = \{e_1, e_2, e_3, e_4\}$

Then β_1 is an ordered basis for W_1 , β_2 is an ordered basis for W_2 , and β is an ordered basis for \mathbb{R}^4 . Let $\underline{A} = [T]_{\beta}$

$\underline{B}_1 = [T_{W_1}]_{\beta_1}$, and $\underline{B}_2 = [T_{W_2}]_{\beta_2}$. Then

$$T(e_1) = T(1, 0, 0, 0) = (2, 1, 0, 0) = 2e_1 + e_2$$

$$T(e_2) = (-1, 1, 0, 0) = -e_1 + e_2$$

$$T(e_3) = (0, 0, 1, 1) = e_3 + e_4$$

$$T(e_4) = (0, 0, -1, 1) = -e_3 + e_4$$

$$\Rightarrow \underline{B}_1 = \begin{pmatrix} 2 & -1 \\ 1 & 1 \end{pmatrix}, \quad \underline{B}_2 = \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix}$$

$$\text{and } \underline{A} = \begin{pmatrix} \underline{B}_1 & 0 \\ 0 & \underline{B}_2 \end{pmatrix} = \begin{pmatrix} 2 & -1 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & -1 \\ 0 & 0 & 1 & 1 \end{pmatrix}$$

Let $f_0(t)$, $f_1(t)$, and $f_2(t)$ denote the characteristic polynomial of T , T_{W_1} , and T_{W_2} , respectively. Then

$$f(t) = \det(A - tI) \\ = \det(B_1 - tI) \cdot \det(B_2 - tI)$$

$$\Rightarrow f(t) = f_1(t) \cdot f_2(t).$$

Definition. Let $B_1 \in M_{m \times m}(F)$ and let $B_2 \in M_{n \times n}(F)$. We define the direct sum of B_1 and B_2 , denoted by $B_1 \oplus B_2$, as the $(m+n) \times (m+n)$ matrix A such that

$$A_{ij} = \begin{cases} (B_1)_{ij} & \text{for } 1 \leq i, j \leq m \\ (B_2)_{(i-m), (j-m)} & \text{for } m+1 \leq i, j \leq m+n \\ 0 & \text{otherwise} \end{cases}$$

If B_1, B_2, \dots, B_k are square matrices with entries from F , then we define the direct sum of B_1, B_2, \dots, B_k recursively

$$\text{by } B_1 \oplus B_2 \oplus \dots \oplus B_k = (B_1 \oplus B_2 \oplus \dots \oplus B_{k-1}) \oplus B_k.$$

If $A = B_1 \oplus B_2 \oplus \dots \oplus B_k$, then we often write

$$A = \begin{bmatrix} B_1 & 0 & \dots & 0 \\ 0 & B_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & B_n \end{bmatrix}$$

Recursion \rightarrow An expression such that each term is generated by repeating a particular mathematical operation.

Example 9. Let

$$B_1 = \begin{pmatrix} 1 & 2 \\ 1 & 1 \end{pmatrix}$$

$$B_2 = (3)$$

$$\text{and } B_3 = \begin{pmatrix} 1 & 2 & 1 \\ 1 & 2 & 3 \\ 1 & 1 & 1 \end{pmatrix}$$

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Then

$$B_1 \oplus B_2 \oplus B_3 = \begin{pmatrix} 1 & 2 & 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 & 0 \\ \hline 0 & 0 & 3 & 0 & 0 & 0 \\ \hline 0 & 0 & 0 & 1 & 2 & 1 \\ 0 & 0 & 0 & 1 & 2 & 3 \\ 0 & 0 & 0 & 1 & 1 & 1 \end{pmatrix}$$

Theorem 5.25. Let T be a linear operator on a finite-dimensional vector space V , and let W_1, W_2, \dots, W_k be T -invariant subspaces of V such that $V = W_1 \oplus W_2 \oplus \dots \oplus W_k$. For each i , let β_i be an ordered basis for W_i , and let $\beta = \beta_1 \cup \beta_2 \cup \dots \cup \beta_k$. Let $A = [T]_\beta$ and $B_i = [T_{W_i}]_{\beta_i}$ for $i = 1, 2, \dots, k$. Then $A = B_1 \oplus B_2 \oplus \dots \oplus B_k$.

Matrix Polynomial \rightarrow

An expression of the form

$$F(x) = A_0 + A_1 x + A_2 x^2 + \dots + A_m x^m$$

where A_0, A_1, \dots, A_m are matrices each of order $n \times n$ over a field F , is called a matrix polynomial of degree m , provided $A_m \neq 0$.

Example. Let

$$A = \begin{bmatrix} 1 + 2x + 3x^2 & x^2 & 6 - 4x \\ 1 + x^3 & 3 - 4x^2 & 1 - 2x + 4x^3 \\ 2 - 3x + 2x^3 & 5 & 6x^2 + 7x \end{bmatrix}$$

then, A can be written as

$$A = \begin{bmatrix} 1 & 0 & 6 \\ 1 & 3 & 1 \\ 2 & 5 & 0 \end{bmatrix} + \begin{bmatrix} 2 & 0 & -4 \\ 0 & 0 & -2 \\ -3 & 0 & 7 \end{bmatrix} x + \begin{bmatrix} 3 & 1 & 0 \\ 0 & -4 & 0 \\ 0 & 0 & 6 \end{bmatrix} x^2 + \begin{bmatrix} 0 & 0 & 0 \\ 1 & 0 & 4 \\ 2 & 0 & 0 \end{bmatrix} x^3$$

which is a matrix polynomial of degree 3.

Proof of Cayley-Hamilton Theorem For Matrices. Let A be an $n \times n$ matrix, and let $f(t)$ be the characteristic polynomial of A . Then $f(A) = 0$, the $n \times n$ zero matrix.

let $A = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \dots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{bmatrix}$

then the characteristic polynomial of A is

$$f(t) = \det(A - tI) = \det \begin{bmatrix} a_{11} - t & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} - t & \dots & a_{2n} \\ \vdots & \vdots & \dots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} - t \end{bmatrix}$$

Therefore, $f(t)$ can be written as

$$f(t) = (-1)^n [t^n + a_1 t^{n-1} + a_2 t^{n-2} + \dots + a_n]$$

where $a_1, a_2, \dots, a_n \in F$.

now to prove that $f(A) = 0$, we must show that

$$\Rightarrow A^n + a_1 A^{n-1} + a_2 A^{n-2} + \dots + a_n I = 0$$

Since all the elements of $A - tI$ are at most of first degree in t , all the elements of $\text{adj}(A - tI)$ are polynomials in t of degree $(n-1)$ or less. (since elements of $\text{adj}(A - tI)$ are cofactors of the elements of $(A - tI)$).

\therefore Hence, $\text{adj}(A - tI)$ can be written as a matrix polynomial in t of degree $(n-1)$

$$\text{let } \text{adj}(A - tI) = B_0 t^{n-1} + B_1 t^{n-2} + \dots + B_{n-2} t + B_{n-1}$$

where $B_0, B_1, B_2, \dots, B_{n-1}$ are square matrices of order $n \times n$.

Now we have

$$(A - tI) \text{adj}(A - tI) = \det(A - tI) I$$

(since $A \cdot \text{adj} A = \det(A) I$)

$$\Rightarrow (A - tI) (B_0 t^{n-1} + B_1 t^{n-2} + \dots + B_{n-2} t + B_{n-1}) = \det(A - tI) \cdot I$$

$$\Rightarrow (A - tI) (B_0 t^{n-1} + B_1 t^{n-2} + \dots + B_{n-2} t + B_{n-1}) = (-1)^n [t^n + a_1 t^{n-1} + \dots + a_n] \cdot I$$

comparing coefficients of like powers of t , we obtain

$$-B_0 = (-1)^n I$$

$$AB_0 - B_1 = (-1)^n a_1 I$$

$$AB_1 - B_2 = (-1)^n a_2 I$$

$$\dots$$
$$- AB_{n-1} = (-1)^n a_n I$$

pre-multiplying the above equations ~~the~~ successively by A^n, A^{n-1}, \dots, I and adding, we obtain

$$0 = (-1)^n [A^n + a_1 A^{n-1} + a_2 A^{n-2} + \dots + a_n I]$$

$$\Rightarrow A^n + a_1 A^{n-1} + a_2 A^{n-2} + \dots + a_n I = 0 \Rightarrow f(A) = 0$$

where 0 is the zero matrix of order $n \times n$

Hence, A satisfies its characteristic equation.

Exercise 15. Use the Cayley-Hamilton theorem (Theorem 5.23) to prove its corollary for matrices.

Theorem 5.23. Let T be a linear operator on a finite-dimensional vector space V , and let $f(t)$ be the characteristic polynomial of T . Then ~~f(T)~~ $f(T) = T_0$, the zero transformation. That is, T "satisfies" its characteristic equation.

Solution. To prove this theorem for matrices, consider the linear transformation $L_A: F^n \rightarrow F^n$ such that $L_A(v) = Av$.

Let $B = \{e_1, e_2, \dots, e_n\}$ ~~be the~~

be the standard basis for F^n . Then

$$\boxed{[L_A]_B = A}$$

Hence, A and L_A have the same characteristic

polynomial f(t). Let

$$f(t) = a_0 + a_1 t + a_2 t^2 + \dots + a_n t^n \quad \text{--- (1)}$$

Since L_A is a linear transformation, hence, from Theorem 5.23, it will satisfy its characteristic equations

$$\Rightarrow \underline{f(L_A) = (L_A)_0}$$

$$\Rightarrow f(L_A)(v) = 0 \quad \text{for all } v \in F^n$$

$$\Rightarrow \underline{f(L_A)(v) = 0(v)}$$

$$\Rightarrow a_0 v + a_1 L_A(v) + a_2 L_A^2(v) + \dots + a_n L_A^n(v) = 0(v) \quad \text{--- (2)}$$

now $L_A(v) = Av \Rightarrow \underline{L_A^2(v) = L_A(L_A(v)) = L_A(Av)}$
 $= A \cdot Av = \underline{A^2 v}$

similarly $\underline{L_A^3(v) = L_A(L_A^2(v)) = L_A(A^2(v)) = A \cdot A^2 v = \underline{A^3 v}}$

and so on $\Rightarrow \underline{L_A^n(v) = \underline{A^n(v)}}$

substituting these values in equation (2), we get

$$\Rightarrow a_0 v + a_1 Av + a_2 A^2(v) + \dots + a_n A^n v = 0(v) = 0$$

$$\Rightarrow \underline{(a_0 + a_1 A + a_2 A^2 + \dots + a_n A^n)(v) = 0}$$

$$\Rightarrow \underline{f(A)(v) = 0} \quad \text{for all } v \in F^n$$

Thus f(A) = 0, the zero matrix of order n x n.

Exercise 36. Let T be a linear operator on a finite-dimensional vector space V. Prove that T is diagonalizable if and only if V is the direct sum of one-dimensional T-invariant subspaces.

Solution. Let $\dim V = n$.

Suppose that V is the direct sum of one dimensional T -invariant subspaces, that is

$$V = W_1 \oplus W_2 \oplus \dots \oplus W_k$$

since W_i ($1 \leq i \leq k$) is one-dimensional T -invariant subspace

$$W_i = \text{span}\{v_i\} \quad \text{and} \quad T(W_i) \subseteq W_i$$

since $v_i \in W_i \Rightarrow T(v_i) \in W_i$

$$\Rightarrow T(v_i) = \lambda v_i \quad (\text{linear combination of element of basis})$$

$\Rightarrow v_i$ is an eigenvector (for $1 \leq i \leq k$)

Thus $B_1 = \{v_1\}$, $B_2 = \{v_2\}$, ..., $B_k = \{v_k\}$ are ^{ordered} bases of the subspaces W_1, W_2, \dots, W_k respectively. Hence, from #

Theorem 5.10(d) $B_1 \cup B_2 \cup \dots \cup B_k = \{v_1, v_2, \dots, v_k\}$ is an ordered basis for V . But $\dim V = n \Rightarrow$ Thus $k = n$

$\Rightarrow \beta = \{v_1, v_2, \dots, v_n\}$ is an ordered basis for V

where all v_i ($1 \leq i \leq n$) are eigenvectors of T .

\Rightarrow Thus there exists an ordered basis β for V consisting of eigenvectors of T . Hence, using Theorem 5.1, T is diagonalizable.

Now, suppose that T is diagonalizable, then there will exist an ordered basis for V consisting of eigenvectors of T .

Let $\beta = \{v_1, v_2, \dots, v_n\}$

be an ordered basis for V consisting of eigenvectors of T .

Now consider $W_i = \text{span}\{v_i\} = \{\lambda v_i, \lambda \in F\}$

now for all $\lambda v_i \in W_i$

$T(\lambda v_i) = \lambda T(v_i) = \lambda \cdot \lambda_i v_i$ (since v_i is an eigenvector) $= (\lambda \lambda_i) v_i \in W_i$ (since $\lambda \lambda_i \in F$)

$\Rightarrow T(\lambda v_i) \in W_i$ for all i ($1 \leq i \leq n$)

$\Rightarrow W_i$ is a T -invariant subspace of V and $\dim(W_i) = 1$

Let $\beta_1 = \{v_1\}, \beta_2 = \{v_2\}, \dots, \beta_n = \{v_n\}$ are ordered bases

for the ~~sub~~ subspaces W_1, W_2, \dots, W_n respectively and

$\beta_1 \cup \beta_2 \cup \dots \cup \beta_n = \{v_1, v_2, \dots, v_n\}$ is an ordered basis

for V . Then, from Theorem 5.10 (a) and (d), we can say that

$V = W_1 \oplus W_2 \oplus \dots \oplus W_n$

$\Rightarrow V$ is the direct sum of one-dimensional T -invariant subspaces of V .