Linear Algebra

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Contents

0 Symbols and notations used

$A_{m\times n} \to m \times n$ matrix.
$A_n \to n \times n$ matrix.
$\sim \to$ the relation below
$A \sim B \implies A = P^{-1}AP.$
iff $\to \iff$

1 Basic Linear equations theory

Every $A_{m\times n} = PR_{m\times n}$ for Row reduced Echelon form R and an invertible matrix P let this relation be denoted by A rrec R

if $m < n$ then the homogeneous system $A_{m\times n}X = 0$ has a non trivial solution
i.e. if the number of equations is less than the number of variables then the Homogeneous System has a non trivial solution

Inverse Properties

- A_n has inverse A^{-1} iff $AX = 0$ has only trivial solutions.
- A is invertible iff A rrec I (identity)
- if Elementary matrices are the corresponding matrices of elementary transforms (change of rows, addition of one row to another, multiplication of a row with an non zero constant) then A is invertible iff A is product of elementary matrices.

Echelon Form

every $A_{m\times n} = P_m R Q_n$ for P, Q invertible and R is such that it has an identity in upper corner and all other entries zero i.e. $R = \begin{bmatrix} I_k & 0 \\ 0 & 0 \end{bmatrix}$ for some identity I_k.

Consistency

System of linear equations :
$A_{m\times n}X_{n\times 1} = b_{1\times m}$ for $b \neq 0$ is consistent (has a solution) iff the row reduced Echelon form of augmented matrix $[A : b]$ has same number of non zero rows as in row reduced echelon form of A.

2 Vector Spaces

Definition

$(V, \mathbb{F}, +)$ denoted by $V(\mathbb{F})$: V is vector space over Field $\mathbb{F}$ if

■ $(V, +)$ is a commutative group,

for every $\alpha, \beta \in \mathbb{F}$ and every $a, b \in V$

■ $1a = a$ where $1 \in \mathbb{F}$ is multiplicative identity of $\mathbb{F}$.

■ $(\alpha + \beta)a = \alpha a + \beta a$

■ $\alpha(a + b) = \alpha a + \alpha b$

■ $(\alpha\beta)a = \alpha(\beta a)$

The elements of V are called **vectors** and elements of $\mathbb{F}$ are called **scalars**

Span

if $K = \{v_1, v_, .., v_n\} \subseteq V(\mathbb{F})$ then span of K is the set $\{\sum \alpha_i v_i | v_i \in K, \alpha_i \in \mathbb{F}\}$ i.e. is all the formal sums from set K with $\mathbb{F}$. This is denoted by $span(K)$.

Subspace

A subset S of vector space $V(\mathbb{F})$ is a subspace if $S(\mathbb{F})$ is a vector space by same operations as in V

■ given any $K \subseteq V(F)$ $span(K)$ is a subspace of $V(\mathbb{F})$.

■ S is a subspace of V iff $\alpha a + b \in S \; \forall a, b \in S$ and $\alpha \in \mathbb{F}$ the underlying field of both spaces

■ Intersection of subspaces (arbitrary) is again a subspace i.e. if W_1, W_2 are subspaces of V then $W_1 \cap W_2$ is also a subspace of V.

■ Union of subspaces may not be a subspace

■ Union of two subspaces is a subspace iff one of them is contained in another i.e. for W_1, W_2 subspaces of V, $W_1 \cup W_2$ is a subspace iff $W_1 \subseteq W_2$ or $W_2 \subseteq W_1$.

(note: this is not the same in case of 3 subspaces : consider $Z_2 \times Z_2(Z_2)$ vector space here $Z_2 \times Z_2 = span((0, 1)) \cup span((1, 0)) \cup span((1, 1))$.)

Dependence

a set of vectors $\{v_1, v_, .., v_n\} \subseteq V(\mathbb{F})$ are called Linearly independent in V if $\alpha_1 v_1 + \alpha_2 v_2 + .. + \alpha_n v_n = 0 \implies$ all $\alpha_i' s$ are 0 and no other choice is left. Other wise the subset is called linearly dependent

Basis

a subset K of V is a spanning set of V if $span(K) = V$.

A Linearly independent spanning set of $V(\mathbb{F})$ is called a Basis of V.

Dimension

In a given vector space $V(\mathbb{F})$.

■ The number of elements in Basis is constant $n \in \mathbb{Z}^+$.

■ if a set contains more vectors than the Basis set of a vector space then it is linearly dependent.

■ if a linearly independent set contains exactly the same number of elements as a Basis then it is also a Basis.

■ These above points leads us to the Definition : Number of elements n in The Basis set of $V(\mathbb{F})$ is unique and is called the Dimension of $V(\mathbb{F})$ denoted by $dim(V) = n$.

if $W_1, W_2 \subseteq V$ are subspaces then

■ $dim(W_i) \le V$.

■ let $W_1 + W_2 = span(W_1, W_2)$ then

$$dim(W_1 + W_2) = dim(W_1) + dim(W_2) - dim(W_1 \cap W_2).$$

(note: there cannot be a definite formula for $dim(\sum_{i=1}^{n} W_i)$ using dimensions of $W_i' s$ and their counterparts (union, intersections) if $n \ge 3$.)

Direct sum

Now if for two subspaces W_1, W_2 of V if $W_1 \cap W_2 = \emptyset$ we write their sum $W_1 + W_2$ as $W_1 \oplus W_2$

■ If $V = W_1 \oplus W_2$ for some non zero subspaces W_1, W_2 then for each vector $v \in V$

can be written **uniquely** as $v = w_1 + w_2$ for unique $w_1 \in W_1$ and $w_2 \in W_2$.

Matrix Representation of vectors

Fix a basis $\beta = \{b_1, b_2, .., b_n\}$ for a vector space $V(\mathbb{F})$ then as B spans V every vector $x \in V$ can be written as $x = x_1 b_1 + x_2 b_2 + .. x_n b_n$ for $x_i \in \mathbb{F}$ and $b_i \in B$ and this representation is unique so each vector can be associated with a column matrix $x_\beta = [x_1\ x_2 .. x_n]^T$

Change of Basis Matrix

Given two basis $\beta = \{b_1, b_2, .., b_n\}$, $\beta' = \{b'_1, b'_2, .., b'_n\}$ for V Then one can change the representation of $x \in V$ from $[x]_\beta$ to $[x]_{\beta'}$ by

$$[x]_{\beta'} = P[x]_\beta$$

where P_n is a invertible matrix given by if $b_j = p_{1j} b'_1 + p_{2j} b'_2 + .. + p_{nj} b'_n$ then $[p_{1j}\ p_{2j} .. p_{nj}]^T$ forms the j^{th} column of P.

3 Linear Transform

Definition

a map $T : V(\mathbb{F}) \to W(\mathbb{F})$ (between vector spaces with same underlying field) is called a linear transform if for every $v, u \in V$ and $\alpha \in F$

- $T(v + u) = T(v) + T(u)$
- $T(\alpha v) = \alpha T(V)$

Range and Null space

For a linear transform $T : V \to W$:

- Range Space of T denoted by $R(T) \subseteq W$ is $\{w | w = T(v) \text{ for some } v \in V\}$
- Null Space of T denoted by $N(T) \subseteq V$ is $\{v | T(v) = 0 \in W\}$
- Both of them are subspaces of the underlying space.
- T is one-one iff $N(T) = \{0\}$.
- T is onto if $R(T) = W$
- if $\dim(V) = \dim(W)$ and $N(T) = \{0\}$ then T is onto thus T is bijective.

if T, U are both liner transforms from $V \to W$ and if both agree on a basis of V (i.e. $T(b_i) = U(b_i)\ \forall i$ for some basis $\beta = \{.., b_i, ..\}$ of V) then both of then are same i.e. $T \equiv U$.

Rank Nullity Theorem

for a linear transform $T : V(\mathbb{F}) \to W(\mathbb{F})$ if $rank(T) = \dim(R(T))$ and $nullity(T) = \dim(N(T))$ then

$$rank(T) + nullity(T) = \dim(V)$$

(this is just an analogue of 1^{st} isomorphism theorems of Groups)

Matrix of Linear Transform

Given a linear transform $T : V \to W$, basis $\beta = \{b_1, b_2 .., b_n\}$ of V and basis $\beta' = \{b'_1, b'_2, .., b'_m\}$ of W then we can write the liner transform in the corresponding matrix representation of vectors as

$$[T(x)]_{\beta'} = [T]_\beta^{\beta'} [x]_\beta$$

where $[T]_\beta^{\beta'}$ is a $m \times n$ matrix called Matrix of linear transform of T and is given by if $T(b_j) = t_{1j} b'_1 + t_{2j} b'_1 + .. + t_{mj} b'_m$ then $[t_{1j}\ t_{2j} .. t_{mj}]^T$ forms the j^{th} column of $[T]_{\beta'}^{\beta}$.

Change of Basis

if $T : V \to V$ then $[T]_\beta^\beta$ is simply written as $[T]_\beta$ now if P is the change of basis matrix from basis β' to basis β of V i.e. $[x]_\beta = P[x]_{\beta'}$ then

$$[T]_{\beta'} = P^{-1}[T]_\beta P$$

(This can be treated as the origin of 'similar' equivalence matrix relationship $A \sim B \iff A = P^{-1}BP$.)

Isomorphism of Vector spaces

Two spaces V, W over same vector space $\mathbb{F}$ are said to be isomorphic to each other if there exist an invertible linear transform $T : V \to W$ (i.e. T is linear bijective map) and this is denoted by $V \cong W$.

■ if $V(\mathbb{F})$ is of dimension n then $V \cong \mathbb{F}^n = \{(\alpha_1, \alpha_2, .. \alpha_n) | \alpha_i \in \mathbb{F}\}$ i.e. set of n tuples of $\mathbb{F}$ with component wise addition.

■ clearly $V(\mathbb{F}) \cong W(\mathbb{F})$ iff $dim(W) = dim(V)$.

Space of Linear Transform

Set of linear transforms $L(V, W) = \{T | T : V \to W \text{ is linear transform}\}$ forms a commutative group under addition i.e. $(T + U)(v) = T(v) + U(v)$ (as in W) so it also forms a Vector space over $\mathbb{F}$ (same field as in V and W.)

■ if $dim(V) = n$ and $dim(W) = m$ both finite then $dim(L(V, W)) = nm$

Linear Functional

Linear transformation $f : V(\mathbb{F}) \to \mathbb{F}$ is called a Linear Functional

■ This is possible as $\mathbb{F}(\mathbb{F})$ is an one dimensional vector space.

■ $rank(f) = 1$ or 0 so $Nullity(f) = n - 1$ or n if $dim(V) = n < \infty$.

■ **Dual space** of V denoted by $V^* = L(V, \mathbb{F})$ is the set of all linear functionals on V

■ clearly $dim(V^*) = dim(V)$ if $dim(V)$ is finite

■ **Dual Basis :** for every basis $\beta = \{b_1, b_2, .., b_n\}$ of V there exist a corresponding basis $\beta^* = \{f_1, f_2, .., f_n\}$ of V^* such that $f_i(b_j) = \delta_{ij} = \begin{cases} 1 & \text{if } i = j \\ 0 & \text{if } i \neq j \end{cases}$ this β^* is called the dual basis of β

■ if $\{.., f_i, ..\}$ is the dual basis of $\{.., b_i, ..\}$ and $x \in V$ is represented as $x = x_1 b_1 + x_2 b_2 + .. + x_n b_n$ then $x_i = f_i(x)$ i.e. the coordinate functions in representation is nothing but the dual functions, i.e. $x = \sum_{i=1}^{n} f_i(x) b_i$.

■ $V \cong V^* \cong V^{**} = L(V^*, \mathbb{F})$ (note: $\cong$ in $V \cong V^{**}$ is nothing but functional evaluation at a point(vectors) only i.e. every element of V^{**} is of form $\hat{x}$ for $\hat{x}(\psi) = \psi(x)$ for some $x \in V$.)

Functional representation Theorem

if V is finite dimensional vector space, $\beta = \{b_i\}$ is its basis and $[x]_\beta = [x_1\ x_2 .. x_n]$ then every functional f is of form

$$f(x) = a_1 x_1 + a_2 x_2 + .. + a_n x_n$$

in which $a_i = f(b_i)$. are fixed but x_i varies on input representation x.

Annihilator

if $A \subset V(\mathbb{F})$ be any subset of V then annihilators of A is the set of linear functionals $A^o = \{f | f(A) = 0, f \in V^*\} \subseteq V^*$

■ clearly A^o is a subspace of V^* for any subset A of V

■ subspaces $W_1 = W_1$ iff $W_1^o = W_2^o$

■ $(W_1 + W_2)^o = W_1^o \cap W_2^o$.

■ if W is subspace of V then

$$dim(W) + dim(W^o) = dim(V).$$

■ if W is subspace of V then $W \cong W^{oo}$.

Transpose of linear transform

if $T : V \to W$ is linear transform then its transpose $T^t : W^* \to V^*$ is a linear transform defined by the evaluation $T^t(g(.)) = g(T(.))$ i.e. for $g \in W^*$, $T^t(g)$ is the functional $f = g(T(.)) \in V^*$

■ $[T^t]_{\gamma^*}^{\beta^*} = ([T]_\beta^\gamma)^t$ i.e. the corresponding matrix of T^t in dual basis of γ in W and β in V is just the Transpose of the matrix of T in β and γ.

■ if W is finite dimensional then for linear $T : V \to W$ we have $R(T^t) = (N(T))^o$ and $N(T^t) = (R(T))^o$

■ T is $1-1$ iff T^t is onto and T is onto iff T^t is $1-1$.

■ $Rank(T^t) = Rank(T)$.

if linear transform $T \in L(V) = L(V,V)$ then it is called a linear operator.

4 Determinant

Motivation

for a finite dimensional space every linear transform in $L(V)$ can be represented as a unique Matrix, but we need to 'uncover' this matrix to gain the properties of corresponding linear transform one such way is to create a Function from set of matrices to the underlying field with some properties which helps us with this 'gain'.

Some Properties needed for such a function are :

■ It must be a linear in terms of rows (or columns) of the matrix this is called n-linear.

■ It must be alternating i.e. if any 2 rows (or columns) are equal then it is zero.

■ its vale on Identity should be 1.

Say we obtain a function D with this property for $(n-1) \times (n-1)$ matrices then this can be extend to $n \times n$ by

$$E_j(A_n) = \sum_{i=1}^{n} (-1)^{i+j} a_{ij} D(A_{ij})$$

for fixed $j \in \{1, 2, .., n\}$, where $a_i j$ is the i^{th} row j^{th} column entry of A and $A_i j$ is the $n-1 \times n-1$ matrix obtained from A_n by removing i^{th} row and j^{th} column.

Definition

From above points we get determinant for a $n \times n$ matrix with entries from $\mathbb{F}$ as $D : \mathbb{F}^{n \times n} \to \mathbb{F}$ that is n-linear, Alternating and $D(I) = 1$ is Defined by recursion from the above point or if $(i_1, i_2, .., i_n)$ runs trough all the possible permutations of n i.e n- tuple with elements from $\{1, 2, .., n\}$ with out repetition then $D(A = [a_{ij}]) =$

$$\sum_{(i_1, i_2, .., i_n)} (-1)^{i_1 + i_2 + .. + i_n} a_{1i_1} a_{2i_2} .. a_{ni_n}$$

Additional Properties

■ $\det(A) = \det(B)$ if B is obtained by interchanging rows of A

■ $\det \begin{bmatrix} A & B \\ 0 & C \end{bmatrix} = \det(A)\det(C)$.

5 Diagonalizability

For linear operator $T \in L(V)$ a vector $\alpha \in V$ is called an eigenvector and λ called eigenvalue if $T(\alpha) = \lambda\alpha$. i.e. $\alpha \in N(T - \lambda I)$

■ if $A \in M_n(\mathbb{F})$ (all $n \times n$ matrices with entries from $\mathbb{F}$) then λ is an eigenvalue og A iff $\det(A - \lambda I) = 0$.

■ From above point we get all eigenvalues of $A \in M_n(\mathbb{F})$ are the solutions of **Characteristic polynomial** $f(t) = \det(A - tI)$.

for a linear operator T on finite dimensional space V

■ The polynomial $p(T)$ such that $p(T) \equiv 0$ i.e $p(T)x = 0 \ \forall x \in V$ then $p(T)$ is called the **annihilating polynomial** of T

■ the set of all annihilating polynomials of T forms an ideal in $\mathbb{F}[x]$ now as $\mathbb{F}$ is a field it is also an euclidean domain so this ideal is principle thus is generated by a monic polynomial of minimum degree in it called the **minimal polynomial** of T.

Algebraic Multiplicity of an eigenvalue λ for a linear operator T is multiplicity of λ in the characteristic polynomial of T.
Geometric multiplicity of an eigenvalue λ for a linear operator T is the dimension of the nullspace of $T - \lambda I$.

A linear operator T on V is said to be Diagonalizable if there exist a basis of V containing only eigenvectors of T.

■ T is diagonalisable iff every eigenvalue of T belongs to the underlying field and Algebraic multiplicity = Geometric multiplicity for every eigenvalue of T.

Cayley-Hamilton Theorem

if T is a linear operator on finite dimensional space V then characteristic polynomial of T divides minimal polynomial of T i.e. if f is characteristic polynomial of T then $f(T) \equiv 0$.

for a given eigenvalue λ of $T \in L(V)$ the set of all eigenvectors corresponding to λ form a subspace of V this is called eigenspace of λ.

Invariant subspace

W is an invariant subspace of T over V if $T(W) \subseteq W$.

Eigenspaces are invariant subspaces.

Diagonalizability test

T is diagonalizable iff minimal polynomial of T ($m_T(x)$) splits into distinct linear factors in the underlying field $\mathbb{F}$ i.e.
T is diagonalizable $\iff$ $m_T(x) = (x - c_1)(x - c_2)..(x - c_n)$ for distinct $c_i \in \mathbb{F}$

matrix representation

T is diagonalizable iff their exist a representation of T in matrix form which is diagonal matrix i.e. if A is matrix of T in some basis then T is diagonalizable iff there exist an invertible matrix P such that $P^{-1}AP = D$ where D is diagonal i.e. iff

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

$$\sim D = \begin{bmatrix} d_{11} & 0 & .. & 0 \\ 0 & d_{22} & .. & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & .. & d_{nn} \end{bmatrix}$$

6 Projections or Idempotent Operators

Projections

$E : V(\mathbb{F}) \to V(\mathbb{F})$(is a projection if $E^2 = E$
■ if E is a projection then $\alpha \in R(T)$ iff $E(\alpha) = \alpha$.

if V is a finite dimensional vector space, say $\{b_1, b_2, .. b_n\}$ is a given ordered basis then we can define projection operators E_i ($i = 1, 2, .. n-1$) as follows: for $x \in V$, $x = \sum_{j=1}^{n} a_j b_j$ we have $E_i(x) = \sum_{j=1}^{i} a_j b_j$ i.e. restriction of the element to a particular subspace. Here we get $R(E_i) = span(\{b_1, .. b_i\})$ and $N(E_i) = span(\{b_i, .. b_n\})$ (note : 0 and I are also projection operator so we can extend these definitions to include 0-space and whole space.)

By intuition of above point we get
if vector space $V = W_1 \oplus W_2 \oplus .. \oplus W_n$ then there exists linear operators $E_1, E_2 .. E_n$ such that
■ Range of $E_i = W_i$
■ each E_i is a projection.
■ $E_i E_j = 0$ for $i \neq j$.
■ $I = E_1 + E_2 + .. + E_n$
Conversely if above 4 points are satisfied for some set of linear operators $\{E_i\}$ on finite dimensional vector space V then for $W_i = R(E_i)$ we have $V = W_1 \oplus W_2 \oplus .. \oplus W_n$.

if a linear operator T on V (finite dimensional) and if E the projection operator of subspace $W \subseteq V$ (defining it can be done by using basis definition of the projections) then T commutes with E iff W is invariant on T i.e.

$$\text{for } E^2 = E \text{ and } R(E) = W$$
$$TE = ET \iff T(W) \subseteq W$$

If vector space $V = U \oplus W$ for some non zero subspaces U, W and if P is the projection operator on V such that $R(P) = U$ then $I - P$ is also a projection operator on V such that $R(I - P) = W$.

Diagonalizability and Projections

if a linear operator T on V is diagonalizable on V then for distinct eigenvalues $\lambda_1, \lambda_2, .., \lambda_n$ of T $\exists$ projections $E_1, E_2, .. E_n$ on V such that

- ■ range of E_i = eigenspace of λ_i in V.
- ■ $T = \lambda_1 E_1 + \lambda_2 E_2 + .. + \lambda_n E_n$.
- ■ $E_i E_j = 0$ for $i \neq j$.
- ■ $I = E_1 + E_2 + .. + E_n$

Conversely if last 3 points are satisfied for any linear operator T and some set of projections $\{E_i\}$ on finite dimensional vector space V then T is Diagonalisable.

Primary Decomposition Theorem

for a Linear operator T on finite dimensional vector space V and if minimal polynomial of $T = m_T(x) = P_1^{r_1}(x) P_2^{r_2}(x) .. P_n^{r_n}(x)$ where P_i are distinct **primes** $\mathbb{F}[x]$ then for W_i = Nullspace of $P_i^{r_i}(T)$ we have

- ■ $V = V = W_1 \oplus W_2 \oplus .. \oplus W_n$.
- ■ W_i is T invariant i.e. $T(W_i) \subseteq W_i$.
- ■ for T_i restriction of T on subspace W_i has minimal polynomial $P_i^{r_i}$.

7 Jordan Form

Generalised eigenvectors

For a linear operator T on V, if λ is an eigenvalue of T then a vector v is such that $(T - \lambda I)^k v = 0$ for some positive integer k is generalised eigenvector.

■ The Subspace $K_\lambda = \{v | (T - \lambda I)^k v = 0 \text{ for some +ve integer } k\}$ is called generalised eigenspace.

properties of generalised eigenspaces

For a given linear operator let K_λ denote generalised eigenspace of T w.r.t (with respect to) eigenvalue λ of T then

- ■ K_λ is T invariant.
- ■ for eigenvalue $\mu \neq \lambda$ of T: $T - \mu I$ is one-one on K_λ.
- ■ $\dim(K_\lambda) = m_\lambda$ where m_λ = Algebraic multiplicity of λ.
- ■ $K_\lambda = N((T - \lambda I)^{m_\lambda})$where m_λ = Algebraic multiplicity of λ.
- ■ if all of the eigenvalues of T belong to the underlying field then $V = K_{\lambda_1} \oplus K_{\lambda_2} \oplus .. \oplus K_{\lambda_n}$. where $\lambda_1, \lambda_2, .. \lambda_n$ are distinct eigenvalues of T.

Cycle of generalised eigenvector : if $v \in K_\lambda$ then the set $\gamma = \{(T - \lambda I)^{k-1} v, (T - \lambda I)^{k-2} v, .. (T - \lambda I) v, v\}$, where $(T - \lambda I)^k v = 0$ and $(T - \lambda I)^{k-1} v$ called as initial vector, forms a linearly independent set in K_λ

■ if $\gamma_1, \gamma_2, .., \gamma_l$ are cycle of generalised eigenvectors for a given eigenvalue λ such that for each γ_i initial vectors are distinct and linearly independent in K_λ then $\gamma = \cup \gamma_i$ is a linearly independent set in K_λ.

existence Jordan canonical form

for any linear operator $T \in L(V(\mathbb{F}))$

- ■ every K_λ (generalised eigenspace) has a ordered basis constituting of cycle of generalised eigenvectors.
- ■ if characteristic polynomial of T completely splits into linear factors in $\mathbb{F}$ then

there exist a basis of $\mathbf{V}$ containing only Cycle of generalised eigenvectors of $\mathbf{T}$, this basis gives a unique characteristic to $\mathbf{T}$ which when viewed in matrix form of $\mathbf{T}$ gives raise to Jordan canonical form.

Consequences of Jordan Form

■ Two linear operators or square matrices (whose characteristics polynomial completely splits into linear factors in their under lying filed) are similar iff they have the same Jordan form representation.

■ $\mathbf{T} \sim \mathbf{T}^t$.

■ if characteristic polynomial of $\mathbf{T}$ completely splits into linear factors in $\mathbb{F}$ then

$$\mathbf{T} \sim \mathbf{D} + \mathbf{N}.$$

where $\mathbf{D}$ is diagonal and $\mathbf{N}$ is nilpotent such that $\mathbf{TN} = \mathbf{NT}$.

matrix representation

if if characteristic polynomial of $\mathbf{T}$ completely splits into linear factors in $\mathbb{F}$ then matrix of $\mathbf{T}$: $\mathbf{A}$ is similar to $\mathbf{J}$ where $\mathbf{J}$ is represented as blocks with diagonal entries as eigenvalues and super diagonal entries 1 and rest entries 0 i.e.

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

$$\sim \mathbf{D} = \begin{bmatrix} [J_1] & 0 & \cdots & 0 \\ 0 & [J_2] & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & [J_k] \end{bmatrix}$$

where $[J_i] = \begin{bmatrix} \lambda_i & 1 & 0 & \cdots & 0 \\ 0 & \lambda_i & 1 & \cdots & 0 \\ \vdots & \vdots & \ddots & \ddots & \vdots \\ 0 & \cdots & \cdots & \lambda_i & 1 \\ 0 & \cdots & \cdots & .. & \lambda_i \end{bmatrix}$, λ_i an eigenvalue of $\mathbf{T}$.

8 Rational Form

9 Inner Product Spaces

10 Forms

11 Bilinear Forms