

4) Suppose a symmetric matrix $A \in \mathbb{R}^{n \times n}$, U is an A -invariant subspace. Show that U^\perp must also be A -invariant.

$$U^\perp = \{v \in \mathbb{R}^n \mid \langle u, v \rangle = 0 \text{ for } u \in U\}.$$

$$\text{g } A = A^T$$

Once U is invariant under A :

if $\forall u_i \in U$ then $\exists u'_i \in U$

such that

$$A u'_i = u_i$$

consider $w_1 \in U^\perp$ then we know that

$$\langle u_i | w_1 \rangle = 0$$

~~xx~~

~~xx~~

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$$\Rightarrow w_1^T u_i = 0 \quad [u_i = A u'_i]$$

$$\Rightarrow w_1^T (A u'_i) = 0 \quad [\text{since } A = A^T]$$

$$\Rightarrow (A w_1)^T u'_i = 0$$

$\exists y \in V$ such that $A w_1 = y$.

$$y^T u'_i = 0 \Rightarrow y \in U^\perp$$

$$A w_1 \in U^\perp \text{ g } w_1 \in U^\perp$$

$\therefore U^\perp$ is also invariant under A .

Q6

a) Suppose $v \in \text{Im} [B \ AB \ A^2B \ \dots \ A^nB]$

$$\Rightarrow v = [B \ AB \ \dots \ A^nB]x.$$

Consider $Av = [AB \ A^2B \ \dots \ A^nB]x.$

Now the characteristic polynomial of A is of degree ' n ' &
 A satisfies its own characteristic equation.

Hence if $\chi_A(s)$ is the characteristic polynomial of A ,

then $\chi_A(A) = 0$. But $\deg(\chi_A(s)) = n$ & it is non-zero.

Hence any power of A greater than $n-1$, can be written

as a linear combination of $\{I, A, A^2, \dots, A^{n-1}\}$

$$\Rightarrow A^nB = \left(\sum_{i=0}^{n-1} d_i A^i \right) B \quad [\because A^n = I]$$

$$\Rightarrow Av = [AB \ A^2B \ \dots \ A^nB]x = [AB \ A^2B \ \dots \ \sum_{i=0}^{n-1} d_i A^i B]x \quad x \in \mathbb{R}^n$$

∴ Av is expressible as a linear combination

of the columns of $B, AB, \dots, A^n B$.

Hence, $Av \in \text{im}[B, AB, \dots, A^n B]$. Thus, C is invariant.

b) let $v \in \Theta$ ie $\begin{bmatrix} v \\ CA \\ \vdots \\ CA^{n-1} \end{bmatrix}_{v=0} \quad \text{--- } \textcircled{*}$

Consider $Av = \begin{bmatrix} v \\ CA \\ \vdots \\ CA^{n-1} \end{bmatrix}_{AV} = \begin{bmatrix} CA \\ CA^2 \\ \vdots \\ CA^n \end{bmatrix}_v$

from $\textcircled{*}$ we know that
 $CV = 0$
 $CAv = 0$
 \vdots
 $CA^{n-1}v = 0$

$\begin{bmatrix} CA \\ CA^2 \\ \vdots \\ CA^n \end{bmatrix}_v = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ CA^n v \end{bmatrix}_0$

Now, $A^2 = \sum_{i=0}^{n-1} d_i A^i \Rightarrow CA^n v = C \sum_{i=0}^{n-1} d_i A^i v = 0$.

$\therefore \begin{bmatrix} CA \\ CA^2 \\ \vdots \\ CA^n \end{bmatrix}_v = 0 \dots \text{Hence, } \Theta \text{ is also } A\text{-invariant.}$

Q7

Since W is A -invariant & $W = \langle w_1, w_2, \dots, w_k \rangle$

$$\Rightarrow Aw_i \in W \Rightarrow Aw_i = \sum_{j=1}^k a_{ij} w_j$$

$$\text{Now, } W = [w_1, w_2, \dots, w_k]$$

$$\Rightarrow Aw = [Aw_1, Aw_2, \dots, Aw_k] = \left[\sum_{j=1}^k a_{1j} w_j, \sum_{j=1}^k a_{2j} w_j, \dots, \sum_{j=1}^k a_{kj} w_j \right]$$

$$\Rightarrow Aw = [w_1, w_2, \dots, w_k] \begin{bmatrix} a_{11} & \cdots & a_{1k} \\ a_{21} & \cdots & a_{2k} \\ \vdots & \ddots & \vdots \\ a_{k1} & \cdots & a_{kk} \end{bmatrix} = WS$$

$$\text{where } S = \begin{bmatrix} a_{11} & \cdots & a_{1k} \\ a_{21} & \cdots & a_{2k} \\ \vdots & \ddots & \vdots \\ a_{k1} & \cdots & a_{kk} \end{bmatrix}$$

Let v be an eigenvector of S with eigenvalue (λ) .

$$\Rightarrow Sv = \lambda v \Rightarrow WSW = \lambda Whv \quad (\text{premultiplied by } W)$$

$$\Rightarrow Awv = \lambda Whv \quad [\because WSW = Aw]$$

$$\Rightarrow Ay = \lambda y \quad \text{where } Whv = y$$

So, if λ is an eigenvalue of S with eigenvector v ,

Then λ is also an eigenvalue of A with eigenvector
 Wv , unless $v \in \text{Ker}(W)$.

Now if atmost $K-r$ linearly independent vectors v_i s.t.
 $Wv_i = 0$. So, $\text{Ker}(W)$ can almost accommodate $(K-r)$
eigenvectors of S . The remaining $K - (K-r) = r = \text{rank}(W)$
eigenvalues of S must have their eigenvectors s.t.
 $Wv \neq 0$ & $Sw = \lambda v$. Thus atleast r eigenvalues
must be shared between A & S .

13)

(a) Consider A & A^H together with their characteristic polynomials given by.

$$\chi_A(x) = \det(A - xI)$$

$$\text{And } \chi_{A^H}(y) = \det(A^H - yI)$$

Now

$$\begin{aligned}\chi_{A^H}(y) &= \overline{\det(A^H - yI)} \\ &= \overline{\det(\bar{A}^H - y^* \bar{I})} \\ &= \overline{\det(A^T - y^* I)} \\ &= \overline{\det(A^T - y^* I^T)}\end{aligned}$$

$$= \chi_{A^T}(y^*)$$

But $\chi_A(z) = \chi_{A^T}(z)$, since $I^T = I$
 & determinant is invariant under transposition.

$$\Rightarrow \chi_{A^H}(y) = \chi_A(y^*).$$

Hence for $z \in \mathbb{C}^n$: $\chi_A(z) = 0$,
 we have $\chi_{A^H}(z^*) = 0$.

(b) Suppose

$$Av = \lambda v \rightarrow ①$$

$$\& A^H w = \mu w, \rightarrow ②$$

when $v \neq 0, w \neq 0$ &
 $\lambda \neq \mu^*$.

from ①

$$A^H v = \lambda^* v$$

$$\Rightarrow v^* A = \lambda v^* - \textcircled{3}$$

$$\text{Now, } \langle w/v \rangle = v^* w - \textcircled{4}$$

Also, from \textcircled{3}, we get

$$v^* \underbrace{A^H w}_T = \lambda^* v^* w. \quad [\text{by post mult. with } w]$$

$$\Rightarrow v^* \cancel{A^H w} = \lambda^* v^* w \quad [\text{using } \textcircled{2}]$$

$$\Rightarrow (v - \lambda^*) v^* w = 0$$

$$\Rightarrow (v - \lambda^*) \langle w/v \rangle = 0 \quad [\text{from } \textcircled{4}]$$

$$\Rightarrow \langle w/v \rangle = 0 \quad [\because \lambda^* \neq \lambda]$$

$\Rightarrow w$ is \perp to v .

(c) Consider $J \in \mathbb{C}^{n \times n}$ as the Jordan canonical form of A (even if $A \in \mathbb{R}^{n \times n}$), we can always consider it to be

matrix in $\mathbb{C}^{n \times n}$ & consider its
complex JCF).

Since $a.m.(\lambda) = 1.$,

we have

$$J_1 = \begin{bmatrix} \lambda & \\ 0 & * \end{bmatrix}$$

for some. $V = [v_1 \dots v_n]$

where $A v_i = \lambda v_i$.

V is non-singular, with its
columns containing the
(generalized) eigenvectors of A .

v_1 is clearly the right
eigenvector of A corresp. to λ .

For left eigenvectors we have.
similarly.

$$\underbrace{\begin{bmatrix} w_1^* \\ w_2^* \\ \vdots \\ w_n^* \end{bmatrix}}_w A = \begin{bmatrix} \lambda & & \\ & Q & \\ & & \lambda \end{bmatrix} \begin{bmatrix} w_1^* \\ \vdots \\ w_n^* \end{bmatrix},$$

where w_i^* is the left eigenvector of A w.r.t. λ_i .

Suppose w_i & v_i are orthogonal,

i.e. $\langle v_i | w_i \rangle = 0$

$$\Rightarrow w_i^* v_i = 0.$$

Hence,

$$WA = \begin{bmatrix} \lambda & & \\ & Q & \\ & & \lambda \end{bmatrix} w$$

$$WA v_i = \begin{bmatrix} \lambda & & \\ & Q & \\ & & \lambda \end{bmatrix} w v_i$$

$$\Rightarrow \lambda \begin{bmatrix} 0 \\ w_2^* v_1 \\ \vdots \\ w_n^* v_1 \end{bmatrix} = \underbrace{\begin{bmatrix} 0 & & & \\ & 0 & & \\ & & 0 & \\ & & & 0 \end{bmatrix}}_{\mathcal{Q}} \begin{bmatrix} 0 \\ w_2^* v_1 \\ \vdots \\ w_n^* v_1 \end{bmatrix}$$

Comparing the last $(n-1)$ terms in the vectors on both sides, we get

$$\lambda \underbrace{\begin{bmatrix} w_2^* v_1 \\ \vdots \\ w_n^* v_1 \end{bmatrix}}_P = \mathcal{Q} \underbrace{\begin{bmatrix} w_2^* v_1 \\ \vdots \\ w_n^* v_1 \end{bmatrix}}_P$$

$\Rightarrow \lambda$ is an eigenvalue of \mathcal{Q} . Note that v_i cannot be orthogonal to w_i for

all $i = 2, 3, \dots, n$, since
then we would have

$v_i = 0$, a
contradiction. Hence.

$p \neq 0]$

But $X_A(x) = X_{J_2}(x)$

$$= (x - \lambda) \cdot X_\varnothing(x).$$

Now if λ is an eigenvalue of
 \varnothing , then $X_\varnothing(x) = (x - \lambda)g(x)$

for some $g(x) \in R(x)$.

Hence $X_A(x) = (x - \lambda)^2 g(x)$.

\Rightarrow a.m. (λ) = $2 > 1$, a

contradiction. Thus

contradiction. Thus

$$\langle v_1/w_1 \rangle \neq 0.$$

20) Which of the following are Ideals and Why/Why not?

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a) {all polynomials in $\mathbb{C}[x]$ having constant term equal to zero}.

$$a := \{ p(x) \in \mathbb{C}[x], p(x) = a_n x^n + a_{n-1} x^{n-1} + \dots + a_1 x \}$$

Let $p, q \in a$

$$\text{clearly } p+q = (a_n+b_n)x^n + \dots + (a_1+b_1)x \in a$$

now,

$$\text{Consider } g(x) = c_m x^m + \dots + c_1 x + c_0 \in \mathbb{C}[x]$$

$$pg = c_m x^m (p(x)) + \dots + c_0 (p(x))$$

notice $p(x)g(x)$ is also a polynomial with zero constant co-efficient.

$$\therefore p(x)g(x) \in a$$

$\therefore a$ is an ideal.

b) $a := \{ p(x) \mid p(x) \text{ has only even degree terms} \}$

if $p, q \in a$, it is clear

that $p+q \in a$

now consider:

$$g = a_0 + a_2 x^2 + \dots + a_k x^k \in \mathbb{C}[x]$$

$$g \quad p(x) = b_0 + b_2 x^2 + b_4 x^4 + \dots + b_{2k} x^{2k}$$

$$p(x)g(x) = b_2 x^2 (g(x)) + \dots + b_{2k} x^{2k} (g(x))$$

$p(x)g(x)$ clearly has terms with odd degree. i.e. $p(x)g(x) \notin a$

\therefore not an ideal.

cy $\{0, 2, 4\} \subseteq \mathbb{Z}_6$

$$0+2=2 \bmod 6 = 2 \quad ; \quad p, q \in \alpha$$

$$2+4=6 \bmod 6 = 0 \quad \& \quad p+q \in \alpha$$

Similarly when $p \in \{0, 1, 2, 3, 4, 5\}$

it can be verified that

every $2p \in \{0, 2, 4\}$ & $4p \in \{0, 2, 0\}$ &

$$0p=0 \in \{0, 2, 4\}$$

$\therefore \{0, 2, 4\}$ is an ideal.

dy $\alpha := \{p(x) \in \mathbb{Z}[x] \mid p(x) \text{ has even co-efficients}\}$

Let $p(x) = \sum_{i=0}^k a_i x^i \in \alpha$ & $q(x) = \sum_{i=0}^m b_i x^i$

a_i & b_i are even $\forall i$. we observe that

$\therefore p+q \in \alpha$ since a_i+b_i is even $\forall i$

Let $g(x) = c_n x^n + \dots + c_1 x + c_0 \in \mathbb{Z}[x]$

c_i & i are not necessarily all odd or even

$p(x) g(x) = \cancel{c_n x^n} (p(x)) \dots + c_0 (p(x))$

every co-efficient of $g(x)$ is multiplied by an even number resulting in all even co-efficients

& their addition (as far the terms with same degree) results in an even number

$\therefore p \in \alpha \Rightarrow \alpha$ is an ideal.

~~ex~~ $R \subseteq R[x]$

not an ideal, consider $x \in R[x]$

e.g. $p \in R$, $px \notin R$.

?

Q22)

Given

$$\langle v, Av, \dots, A^{n-1}v \rangle = \mathbb{C}^n$$

$$\Rightarrow B = \{v, Av, \dots, A^{n-1}v\}$$

is a basis for \mathbb{C}^n

(set of n vectors spanning

\mathbb{C}^n must be lin. ind.

Now, hence, a basis.)

$$[A]_B = \begin{bmatrix} [Av]_B & [AAv]_B & \dots & [A \cdot A^{n-1}v]_B \end{bmatrix}$$

since we can have an
ordered basis

$$\{v_1, v_2, \dots, v_n\}$$

$$\text{where } v_i^o = A^{i-1}v.$$

Thus, $A v_i^{\circ} = v_{i+1}$

$$\Rightarrow [A v_i]_B = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \\ 0 \end{bmatrix} \leftarrow (i+1)^{\text{th}} \text{ position.}$$

for $i = 1, 2, \dots, n-1$,

from Cayley Hamilton theorem,
we know that

$$\chi_A(A) = 0$$

$$\Rightarrow A^n = - \sum_{i=0}^{n-1} \alpha_i A^i$$

[since $\chi_A(x)$ is

$\Rightarrow A \cdot v_n$ monic, $\alpha_n = 1$]

$$\therefore = A \cdot A^{n-1} v_n$$

$$= A^n v_n$$

$$= - \sum_{i=0}^{n-1} \alpha_i A^i v_n$$

$$= - \sum_{i=0}^{n-1} \alpha_i v_i$$

$$\Rightarrow [Av_n]_B = \begin{bmatrix} -\alpha_0 \\ -\alpha_1 \\ \vdots \\ -\alpha_{n-1} \end{bmatrix}$$

$$\Rightarrow [A]_B = \begin{bmatrix} 0 & 0 & 0 & -\alpha_0 \\ 1 & 0 & 0 & -\alpha_1 \\ 0 & 1 & \ddots & \vdots \\ 0 & 0 & \ddots & \vdots \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 1 & -\alpha_{n-1} \end{bmatrix}$$

Since. $\{v, Av, \dots, A^{n-1}v\}$

is a linearly ind-set,
there does not exist

$$\{\alpha_0, \alpha_1, \dots, \alpha_{n-1}\};$$

$$\sum \alpha_i A^i v = 0, \text{ unless}$$

$$\alpha_i = 0 \quad \forall i = 0, 1, \dots, n-1.$$

minimal polynomial of A w.r.t. v must be of degree at least equal to n.

Also, $M_A(x)$ must be a multiple of this polynomial.

Thus $\deg(M_A(x)) = n$.

and we know that

$$M_A(x) \mid \chi_n(x).$$

$$\text{so } M_A(x) = \chi_n(x)$$

Q.12) Given $A \in \mathbb{R}^{m \times n}$ and SVD of A is

$$\text{given by } A = U \begin{bmatrix} \Sigma_{n \times n} & 0 \\ 0 & 0 \end{bmatrix} V^T.$$

$$A^+ = V \begin{bmatrix} \Sigma_{n \times n}^{-1} & 0 \\ 0 & 0 \end{bmatrix} U^T$$

(a) $\det A$ is invertible $\Rightarrow A \in \mathbb{R}^{n \times n}$

$$\text{and } A^+ = V \Sigma_{n \times n}^{-1} U^T$$

with $V \in \mathbb{R}^{n \times n}$ and $U \in \mathbb{R}^{n \times n}$.

$$\text{Now } A^{-1} = \left(U \Sigma_{n \times n} V^T \right)^{-1}$$

$$= (V^T)^{-1} \Sigma_{n \times n}^{-1} U^{-1}$$

$$= V \Sigma_{n \times n}^{-1} U^T = A^+$$

$$(b) (A^+)^+ = \left(V \begin{bmatrix} \Sigma_{n \times n}^{-1} & 0 \\ 0 & 0 \end{bmatrix} U^T \right)^+$$

$$= U \begin{bmatrix} (\Sigma_{n \times n}^{-1})^{-1} & 0 \\ 0 & 0 \end{bmatrix} V^T$$

$$= U \begin{bmatrix} \Sigma_{n \times n} & 0 \\ 0 & 0 \end{bmatrix} V^T = A$$

$$\begin{aligned}
 (c) \quad (A^+)^T &= \left(V \begin{bmatrix} \Sigma_{n \times n}^{-1} & 0 \\ 0 & 0 \end{bmatrix} U^T \right)^T \\
 &= U \begin{bmatrix} (\Sigma_{n \times n}^{-1})^T & 0 \\ 0 & 0 \end{bmatrix} V^T \\
 (A^T)^+ &= \left(\left(U \begin{bmatrix} (\Sigma_{n \times n}^{-1})^T & 0 \\ 0 & 0 \end{bmatrix} V^T \right)^T \right)^+ \\
 &= \left(V \begin{bmatrix} (\Sigma_{n \times n}^{-1})^T & 0 \\ 0 & 0 \end{bmatrix} U^T \right)^+ \\
 &= U \begin{bmatrix} \Sigma_{n \times n}^{-1} & 0 \\ 0 & 0 \end{bmatrix} V^T = (A^+)^T
 \end{aligned}$$

(d) A^+ has full column rank.

$$\text{Then } A^T A = V \sum_{n \times n}^{-2} V^T \in \mathbb{R}^{n \times n}$$

$$\begin{aligned}
 \text{Now, } (A^T A)^{-1} A &= V^T \sum_{n \times n}^{-2} V V^T \begin{bmatrix} \Sigma_{n \times n}^{-1} & 0 \\ 0 & 0 \end{bmatrix} U^T \\
 &= V^T \begin{bmatrix} \Sigma_{n \times n}^{-2} & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} \Sigma_{n \times n}^{-1} & 0 \\ 0 & 0 \end{bmatrix} U^T \\
 &= V^T \begin{bmatrix} \Sigma_{n \times n}^{-1} & 0 \\ 0 & 0 \end{bmatrix} U^T \\
 &= A^+
 \end{aligned}$$

Similarly when A has full row rank,

$$AA^T = U \Sigma_{m \times m}^2 U^T \in \mathbb{R}^{m \times m}$$

$$\begin{aligned} & A^T (AA^T)^+ = V \begin{bmatrix} \Sigma_{m \times m}^{-1} & 0 \\ 0 & 0 \end{bmatrix} U^T U \Sigma_{m \times m}^{-2} U^T \\ &= V \begin{bmatrix} \Sigma_{m \times m}^{-1} & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} \Sigma_{m \times m}^{-2} & 0 \\ 0 & 0 \end{bmatrix} U^T \\ &\approx V \begin{bmatrix} \Sigma_{m \times m}^{-1} & 0 \\ 0 & 0 \end{bmatrix} U^T = A^+ \end{aligned}$$

$$\begin{aligned} & (A^T A)^+ = (V \begin{bmatrix} \Sigma_{r \times r}^{-2} & 0 \\ 0 & 0 \end{bmatrix} V^T)^+ \\ &= V \begin{bmatrix} \Sigma_{r \times r}^{-2} & 0 \\ 0 & 0 \end{bmatrix} V^T \end{aligned}$$

$$\begin{aligned} & A^+ (A^T)^+ = V \begin{bmatrix} \Sigma_{r \times r}^{-1} & 0 \\ 0 & 0 \end{bmatrix} U^T (V \begin{bmatrix} \Sigma_{n-r}^{-1} & 0 \\ 0 & 0 \end{bmatrix} V^T)^+ \\ &= V \begin{bmatrix} \Sigma_{r \times r}^{-1} & 0 \\ 0 & 0 \end{bmatrix} U^T U \begin{bmatrix} \Sigma_{r \times r}^{-1} & 0 \\ 0 & 0 \end{bmatrix} V \\ &\approx V \begin{bmatrix} \Sigma_{r \times r}^{-2} & 0 \\ 0 & 0 \end{bmatrix} V = (A^T A)^+ \end{aligned}$$

Q24

Is $\mathbb{Z}[x]$ a P.I.D? Justify

for $\mathbb{Z}[x]$ to be a P.I.D, every ideal generated must be a principle ideal (ideal generated by a single element).

Consider the ideal $\alpha = \langle 3, x^2 \rangle$

i.e. $f(x) = 3p(x) + x^2q(x)$, $f(x) \in \alpha : p(x), q(x) \in \mathbb{Z}[x]$.

If $\mathbb{Z}[x]$ is a P.I.D then $\alpha = \langle 3, x^2 \rangle = \langle h(x) \rangle$

i.e.

$$3 = h(x) f_1(x) \quad \boxed{\textcircled{1}} \quad f_1(x) \in \mathbb{Z}[x]$$

$$x^2 = h(x) f_2(x) \quad \boxed{\textcircled{2}} \quad f_2(x) \in \mathbb{Z}[x]$$

~~from~~ ~~1~~ ~~2~~

from $\textcircled{1}$ it is clear that $h(x) f_1(x)$ has to be a constant polynomial.

$$\therefore h(x) = \{ \pm 1, \pm 3 \}$$

Suppose $\alpha = \langle 3, x^2 \rangle = \langle 1 \rangle$

exists $y_1(x) \& y_2(x)$, such that $[y_1, y_2 \in \mathbb{Z}[x]]$

$$3y_1(x) + x^2y_2(x) = 1$$

The term $x^2y_2(x)$ has no ~~the~~ constant term thus the only constant term available is that $y_1(x)$

Suppose $y_1(0)$ is the constant term then by comparison of Poly Coefficient, we have

$3r_1(0) = 1$. this only possible only if $r_1(0) = \frac{1}{3}$

$r_1(n) \notin \mathbb{Z}[n]$

$$\langle 3, x^2 \rangle \neq \langle 1 \rangle$$

Suppose $\langle 3, x^2 \rangle = \langle 3 \rangle$ [i.e. $h(n) = \pm 3$]

then $\exists r(n)$ such that

$$x^2 = 3r(n)$$

which again implies $r(n) \notin \mathbb{Z}[n]$

$\therefore \langle 3, x^2 \rangle$ is not a principle ideal

$\Rightarrow \mathbb{Z}[n]$ is not PFD.

Q27

Given two diagonalizable matrices A, B .

$$\langle 1 \rangle AB = BA$$

$$\langle 2 \rangle \exists S \in \mathbb{R}^{n \times n} \text{ s.t. } S^T AS \text{ & } S^T BS \text{ are both diagonal.}$$

$\langle 2 \rangle \Rightarrow \langle 1 \rangle$ let $S^T AS = D_1$ & $S^T BS = D_2$ are diagonal matrices.

$$S^T A S S^T B S = D_1 D_2 \quad \left\{ \text{diagonal matrices commute} \right\}.$$

$$S^T A B S = D_2 D_1 = S^T B S S^T A S.$$

$$S^T A B S = S^T B A S \Rightarrow AB = BA.$$

$\langle 1 \rangle \Rightarrow \langle 2 \rangle$ Given $AB = BA$, A & B are diagonalizable

Q.s.t. $S^T A S = D_A = \begin{bmatrix} \lambda_1 I_{n_1} & & \\ & \ddots & \\ & & \lambda_r I_{n_r} \end{bmatrix}$

is given by $\sum_{k=0}^n C_k$.

Hence, the total number of invariant subspaces is given by

$$\sum_{k=0}^n \sum_{j_1=0}^{n-k} \dots \sum_{j_n=0}^{n-k} = (1+1)^n = 2^n.$$

for V

THEOREM Consider matrix $L = \begin{bmatrix} L_1 & & \\ & L_2 & \\ & & L_K \end{bmatrix}$ L_i 's are block matrices.

Then L is diagonalizable iff each L_i is diagonalizable.

(\Leftarrow) Let L_1, L_2, \dots, L_K be diagonalizable then \exists invertible matrix

Ω_i s.t. $\Omega_i^{-1}L_i\Omega_i$ is diagonal matrix. let $\Omega = \Omega_1 \oplus \dots \oplus \Omega_K$.

Then clearly, we have $D = \Omega^{-1}L\Omega$ is a diagonal matrix.

Hence, L is diagonalizable.

(\Rightarrow) Let L is diagonalizable then \exists Ω s.t. $D = \Omega^{-1}L\Omega$

where $D = \text{diag}(\lambda_1, \lambda_2, \dots, \lambda_K)$ is a diagonal matrix.

Then $L\Omega = \Omega D$, we have

$$\begin{bmatrix} L_1 & & \\ & L_2 & \\ & & L_K \end{bmatrix} \begin{bmatrix} x_1^j \\ \vdots \\ x_K^j \end{bmatrix} = \lambda_j \begin{bmatrix} x_1^j \\ \vdots \\ x_K^j \end{bmatrix}$$

where $x_j^i = [x_{1j}^j \ \dots \ x_{Kj}^j]^T$ j^{th} column of Ω & x_i^j is a vector
 s.t. $L_i x_i^j = \lambda_j x_i^j$ for $i=1, \dots, K$ & L_i is $n_i \times n_i$ matrix.

Since Ω is invertible, defining invertible submatrix

$$Y := \begin{bmatrix} x_1^{m_1} & \dots & x_i^{m_{n_i}} \end{bmatrix} \therefore Y \in \mathbb{R}^{n_i \times n_i}$$

So, $L_i Y = Y A$ where $A = \text{diag}(\lambda_{m_1}, \dots, \lambda_{m_{n_i}})$

which is diagonal matrix. Hence $L_i = \begin{pmatrix} A \\ I \end{pmatrix}^{-1}$. Proved.

Using above results, if nonsingular $X_i \in M_{n_i}$ s.t. $X_i^T D_B X_i$ for $i=1, \dots, r$

is a diagonal matrix.

Consider $X := \text{diag}(X_1, X_2, \dots, X_r)$. (a block diagonal matrix)

$\Rightarrow A = X^T D_B X$ & $\Gamma = X^T D_A X$ are diagonal.

Therefore if invertible $S := QX$ s.t. $A = S^T B_S$ & $\Gamma = S^T A S$

are diagonal matrices. Hence proved.

Q28

$$A = \begin{bmatrix} -\gamma_2 & \gamma_2 \\ \gamma_2 & -\gamma_2 \end{bmatrix},$$

$$\cos(A) = P^T \cos(D) P$$

$$sI - A = \begin{bmatrix} s + \gamma_2 & -\gamma_2 \\ -\gamma_2 & s + \gamma_2 \end{bmatrix}$$

$$\det(sI - A) \Rightarrow (s + \gamma_2)^2 - (\gamma_2)^2 = s(s + 2\gamma_2) = 0$$

$s = 0, -2\gamma_2$ (eigen values)

eig vector corresponding to $\lambda = 0$

$$\begin{bmatrix} \gamma_2 & -\gamma_2 \\ -\gamma_2 & \gamma_2 \end{bmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \Rightarrow x_1 = x_2, v_1 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

eig vector corresponding to $\lambda = -2\gamma_2$

$$\begin{bmatrix} -\gamma_2 & -\gamma_2 \\ -\gamma_2 & -\gamma_2 \end{bmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \Rightarrow x_1 = -x_2, v_2 = \begin{pmatrix} 1 \\ -1 \end{pmatrix}$$

$$\cos A = \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} \cos 0 & 0 \\ 0 & \cos(-2\gamma_2) \end{pmatrix} \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix}^{-1} = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$$

$$\boxed{\cos(A) = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}}$$

Q29

Given $A = -A^T$

Let $\Omega = e^A$

$$\begin{aligned} \text{Now } \Omega\Omega^T &= e^A e^{A^T} \\ &= e^{-A^T} e^{A^T} \\ &= e^{-A^T+A^T} \\ &= e^0 \end{aligned}$$

$\stackrel{\text{Open}}{=} I$

$= I$

where I is the Identity matrix

hence Ω is orthogonal.

(32)

Consider the function,

$$F(t) = e^{At+Bt} - e^{At} e^{Bt}$$

Differentiating both sides wrt to t , we get

$$\begin{aligned} F'(t) &= (A+B)e^{(A+B)t} - Ae^{At}e^{Bt} - e^{At}e^{Bt} \cdot B \\ &= A e^{(A+B)t} - Ae^{At}e^{Bt} + Be^{(A+B)t} - B \underbrace{e^{At}e^{Bt}}_L \\ &= A F(t) + B F(t) \end{aligned}$$

$\left[\begin{array}{l} \therefore BA = AB \\ A^k B = B A^k \\ \forall k \end{array} \right]$

$$\Rightarrow F'(t) = (A+B) F(t)$$

write $F(t) = [f_1(t), \dots, f_n(t)]$, where $f_i(t) \in \mathbb{R}^n$

$$\Rightarrow \overset{\circ}{f}_i(t) = (A+B) f_i(t) \quad \forall i = 1, 2, \dots, n$$

Consider $f_i^*(t) = e^{(A+B)t} f_i^*(0)$

and observe that $\overset{\circ}{f}_i^*(t) = (A+B) f_i^*(t)$ is satisfied for this choice of $f_i^*(t)$

$$\Rightarrow F(t) = e^{(A+B)t} F(0)$$

$$\text{Now, } F(0) = e^{A \cdot 0 + B \cdot 0} - e^{A \cdot 0} e^{B \cdot 0}$$

$$= I - I$$

$$= 0$$

$$\Rightarrow F(t) = 0, \forall t \Rightarrow e^{(A+B)t} = e^{At} e^{Bt}$$

$$\Rightarrow e^{(A+B)t} = e^{At} e^{Bt}$$

Solution of other part is not unique. Take any A, B : $AB \neq BA$ and verify

(34) Suppose we have

$$A\mathbf{v} = \lambda_M \mathbf{v} \rightarrow \textcircled{P}$$

where λ_M is the largest eigenvalue of A

Equating component-wise on both sides of \textcircled{P} , we get

$$\sum_{j=1}^n a_{pj} v_j = \lambda_M v_p \quad \text{for } p=1, \dots, n$$

Suppose $\max_i v_i = v_K$

Note that v_K is always positive, unless $v_i < 0$ for all i .
In case $v_i < 0 \forall i$, namely we may just choose $-\mathbf{v}$ as the eigenvector.

Further, $\sum_{j=1}^n a_{pj} = \text{RowSum}_p(A)$

$$\text{Now, } \lambda_M v_K = \sum_{j=1}^n a_{kj} v_j \leq \sum_{j=1}^n a_{kj} v_K \quad \left[\begin{array}{l} \because a_{kj} v_j < a_{kj} v_K \\ v_j \end{array} \right]$$

$$\Rightarrow \lambda_M v_K \leq v_K \sum_{j=1}^n a_{kj}$$

as $a_{kj} > 0$ by definition]

$$\Rightarrow \lambda_M \leq \sum_{j=1}^n a_{kj} = \text{RowSum}_K(A)$$

$$\leq \max_i \text{RowSum}_i(A)$$

Since $A^T = A$, we have an orthonormal set of eigenvectors that diagonalize A .

Consider any $\mathbf{x} \in \mathbb{R}^n$, represented as

$$\mathbf{x} = \sum_{p=1}^n c_p \mathbf{v}_p$$

where $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$ is the orthonormal set of eigenvectors of A .

Observe that,

$$\begin{aligned}\langle x | Ax \rangle &= x^T A x = \left(\sum c_i v_i \right)^T A \left(\sum c_i v_i \right) \\&= \left(\sum c_i v_i \right)^T \sum c_i A v_i \\&= \sum c_i v_i^T \left[\sum c_i \lambda_i v_i \right] \\&= \sum c_i^2 \lambda_i v_i^T v_i \\&= \sum c_i^2 \lambda_i\end{aligned}$$

Notice that $\|x\|^2 = \langle x | x \rangle = \langle \sum c_i v_i | \sum c_i v_i \rangle$

$$= \sum_{i=1}^n c_i^2$$

Now, $\max_x \frac{\langle x | Ax \rangle}{\langle x | x \rangle}$

$$= \max_{\{c_1, c_2, \dots, c_n\}} \frac{\sum c_i^2 \lambda_i}{\sum c_i^2}$$
$$\leq \lambda_M \frac{\sum c_i^2}{\sum c_i^2} = \lambda_M$$

$$\therefore \lambda_M \geq \max_{x \neq 0} \frac{\langle x | Ax \rangle}{\langle x | x \rangle}$$

$\therefore \lambda_M \geq \frac{\langle x | Ax \rangle}{\langle x | x \rangle}$ for any choice of non-trivial vector, we choose $x = [1 1 \dots 1]^T$

$$\Rightarrow \lambda_M \geq \frac{\sum_{i=1}^n \text{RowSum}_i(A)}{n}$$

$$\text{Now, } \sum_{i=1}^n \text{RowSum}_i^o(A) \geq n \left[\min_i \text{RowSum}_i^o(A) \right]$$

$$\Rightarrow \frac{\sum_{i=1}^n \text{RowSum}_i^o(A)}{n} \geq \min_i \text{RowSum}_i^o(A)$$

$$\Rightarrow \lambda_M \geq \min_i \text{RowSum}_i^o(A)$$

$$\Rightarrow \min_i \text{RowSum}_i^o(A) \leq \lambda_M \leq \max_i \text{RowSum}_i^o(A)$$

(35) $A = \begin{bmatrix} 1 & 3 & 5 \\ 2 & 4 & 7 \\ 3 & 6 & 8 \end{bmatrix}$

Consider a basis for \mathbb{R}^3 as $\{e_1, e_2, e_3\}$; $e_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$

$$Ae_1 = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \quad A^2e_1 = \begin{bmatrix} 22 \\ 31 \\ 39 \end{bmatrix} \quad A^3e_1 = \begin{bmatrix} 310 \\ 441 \\ 564 \end{bmatrix}$$

$$\therefore \alpha_0 \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + \alpha_1 \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} + \alpha_2 \begin{bmatrix} 22 \\ 31 \\ 39 \end{bmatrix} = \begin{bmatrix} 310 \\ 441 \\ 564 \end{bmatrix}$$

$$\alpha_0 = 5, \alpha_1 = 19, \alpha_2 = 13$$

$$\boxed{\mu_{e_1} = s^3 - 13s^2 - 19s - 5}$$

$$e_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \quad Ae_2 = \begin{bmatrix} 3 \\ 4 \\ 6 \end{bmatrix} \quad A^2e_2 = \begin{bmatrix} 45 \\ 64 \\ 81 \end{bmatrix} \quad A^3e_2 = \begin{bmatrix} 642 \\ 913 \\ 1167 \end{bmatrix}$$

$$\alpha_0 \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} + \alpha_1 \begin{bmatrix} 3 \\ 4 \\ 6 \end{bmatrix} + \alpha_2 \begin{bmatrix} 45 \\ 64 \\ 81 \end{bmatrix} = \begin{bmatrix} 642 \\ 913 \\ 1167 \end{bmatrix}$$

$$\Rightarrow \alpha_0 = 5, \alpha_1 = 19, \alpha_2 = 13$$

$$\boxed{\mu_{e_2} = s^3 - 13s^2 - 19s - 5}$$

$$e_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \quad Ae_3 = \begin{bmatrix} 5 \\ 7 \\ 8 \end{bmatrix} \quad A^2 e_3 = \begin{bmatrix} 66 \\ 94 \\ 121 \end{bmatrix} \quad A^3 e_3 = \begin{bmatrix} 953 \\ 1355 \\ 1730 \end{bmatrix}$$

$$\alpha_0 \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} + \alpha_1 \begin{bmatrix} 5 \\ 7 \\ 8 \end{bmatrix} + \alpha_2 \begin{bmatrix} 66 \\ 94 \\ 121 \end{bmatrix} = \begin{bmatrix} 953 \\ 1355 \\ 1730 \end{bmatrix}$$

$$\Rightarrow \alpha_0 = 5, \alpha_1 = 19, \alpha_2 = 13$$

$$\Rightarrow H e_3 = s^3 - 18s^2 - 19s - 5$$

$$\mu = \text{lcm}(H e_1, H e_2, H e_3)$$

$$\boxed{\mu = s^3 - 13s^2 - 19s - 5}$$

Now, eigenvalues are roots of both $X(s)$ and $H(s)$.

Solving for $H(s) = 0$, we get by inspection that

$s = -1$ is a root

Now,

$$\begin{array}{r} s^2 - 14s - 5 \\ \hline s+1 \left[\begin{array}{r} s^3 - 13s^2 - 19s - 5 \\ s^3 + s^2 \\ \hline -14s^2 - 19s - 5 \\ -14s^2 - 14s \\ \hline -5s - 5 \\ -5s - 5 \\ \hline 0 \end{array} \right] \end{array}$$

The other two eigenvalues are roots of $s^2 - 14s - 5 = 0$.

$$\therefore \lambda = \{-1, 7 \pm \sqrt{54}\}$$

(4D) Given $\phi: V \rightarrow V$ with $\dim(V) = n < \infty$

T.S.T. $\ker(\phi^n) \cap \text{Im}(\phi^n) = \{0\}$

Suppose $v \in \ker(\phi^n) \cap \text{Im}(\phi^n)$. Then $\phi^n v = 0$

Also, $\exists u \in V$ s.t. $v = \phi^n u$,

so, $\phi^n v = \phi^{2n} u = 0$

But $\ker(\phi^n) = \ker(\phi^{n+1}) = \dots$ Null spaces stop growing

$$\Rightarrow \phi^n u = 0$$

$$\Rightarrow v = \phi^n u = 0$$

$$\Rightarrow \ker(\phi^n) \cap \text{Im}(\phi^n) = \{0\}$$

thus $\ker(\phi^n) + \text{Im}(\phi^n)$ is direct sum.

Also, $\dim \left[(\ker(\phi^n)) \oplus (\text{Im}(\phi^n)) \right] = \dim \ker(\phi^n) + \dim \text{Im}(\phi^n)$

Now by fundamental theorem of linear maps,

$$\dim \ker(\phi^n) + \dim \text{Im}(\phi^n) = \dim V$$

thus,

$$\ker(\phi^n) \oplus \text{Im}(\phi^n) = V.$$

Assignment 5

Q.7. We already have.
 $AW = WS$.

$$\Rightarrow \tilde{A}^T W = \underbrace{AWS}_T = \underbrace{WS \cdot S}_T = WS^2$$

$$\Rightarrow A^l W = WS^l \quad \forall l$$

(identically)

$$\therefore \cancel{\chi_A(A)} W = W \cancel{\chi_A(S)}$$

where $\chi_A(\lambda)$ is
 the char poly of A .

$$\Rightarrow W \cdot \chi_A(S) = 0$$

$$[\chi_A(S)]^T W^T = 0$$

$$\Rightarrow S_A(S^T) \cdot W^T = 0$$

[since
 $f(S)^T = f(S^T)$
 for any $S \in \mathbb{R}^{K \times K}$
 & $f(\cdot)$ being a polynomial
 (Check this).]

$\Rightarrow S_A(S^T)$ maps at least.
 the 'g' independent
 columns of W^T to zero.
 so the entire W^T is mapped to 0

$$\Rightarrow (S^T - \lambda_1 I) (S^T - \lambda_2 I) \dots (S^T - \lambda_n I).$$

$$W^T = 0$$

where $\lambda_1, \lambda_2, \dots, \lambda_n$ are

the (not necessarily distinct) eigenvalues of A .

\Rightarrow The s lin. ind. columns of W must belong to

$$\ker(S^T - \lambda_i I)$$
 for

some value of i or the other.

These λ_i 's are therefore eigenvalues of S^T with the corresp. columns of W^T being

the eigenvectors. Thus, distinct or not, we have at least s set of eqns of the form

$$(S^T - \lambda_i I) \bar{w}_i = 0$$

where w_i is some column of W^T .

such λ_i 's ; being r in number, are common eigenvalues of both A & S^T .

$S^T S^T$ have same eigenvalues, so these λ_i 's are also eigenvalues of S .