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Question  
Sub-Question No.

$$\boxed{\quad} + \boxed{\quad} + \boxed{\quad} + \boxed{\quad} = \boxed{\quad}$$

Q. \_\_\_\_\_

# COL726 Assignment 2

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Ans 1:

⇒ We have  $(C = AB)$  with all of the  $m \times n$  matrices.

⇒ Now, we can look at the Invertibility of matn A, B

⇒ Both A & B are invertible :

$$\rightarrow \text{We can use: } \|C\|_2 = \|AB\|_2 \leq \|A\|_2 \cdot \|B\|_2$$

$$\Rightarrow \boxed{c_m \leq a_m b_m} \leftarrow ①$$

$$\rightarrow \text{Also taking inverse on both sides, we have: } C^{-1} = B^{-1}A^{-1} \Rightarrow \|C^{-1}\|_2 \leq \|B^{-1}\|_2 \|A^{-1}\|_2$$

$$\Rightarrow \boxed{\frac{1}{c_m} \leq \frac{1}{b_m} \cdot \frac{1}{a_m}}$$

$\therefore A, B \xrightarrow{\text{invertible}} C \text{ invertible. So } c_m, a_m, b_m \neq 0$   
 And  $c_m$  smallest singular value for C.  
 So  $\frac{1}{c_m}$  largest singular value for  $C^{-1}$

$$\Rightarrow \boxed{c_m \geq a_m b_m} \leftarrow ②$$

$$\boxed{\quad} + \boxed{\quad} + \boxed{\quad} + \boxed{\quad} = \boxed{\quad}$$

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\* ~~Addition of  $A$  &  $B$  is not invertible~~

~~So here we can say that  $C$  won't be invertible~~

→ We can also write :  $\boxed{CB^{-1} = A} \Rightarrow \|A\|_2 \leq \|C\|_2 \|B^{-1}\|_2$   
(From  $C = AB$ )

$$\Rightarrow c_i \geq a_i b_m \quad \leftarrow (3)$$

→ Similarly, from  $\boxed{(A^{-1})C = B} \Rightarrow c_i \geq b_i a_m \quad \leftarrow (4)$

→ Combining the eq ①, ③, ④ :  $a_i b_m \geq c_i \geq \max(a_i b_m, b_i a_m)$

→ Also, if we consider :  $\boxed{C^{-1} = B^{-1} A^{-1}}$

$$\boxed{c_m \leq b_i a_m} \Leftrightarrow \boxed{B C^{-1} = A^{-1}} \quad \boxed{C^{-1} A = B^{-1}} \Rightarrow \boxed{c_m \leq a_i b_m}$$

→ The above ~~two~~ inequalities we obtained in a similar manner as done for the Perron case & so by combining them along with eq ② :

a\_i b\_m \leq c\_i \leq \min(a\_i b\_m, b\_i a\_m)

$$\boxed{\quad} + \boxed{\quad} + \boxed{\quad} + \boxed{\quad} = \boxed{\quad}$$

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\* If at least one of A & B are not invertible

→ We can say that C won't be invertible. So  $c_m = 0$

→ Now:  $a_m b_m = 0$  ( $\because$  at least one of them will be zero)

&  $\min(a_1 b_m, b_1 a_m) = 0$ . ( $\because$  at least one of  $(a_m, b_m)$  is zero)

→ So we can trivially say:  $a_m b_m \leq c_m \leq \min(a_1 b_m, b_1 a_m)$

→ Now, we know that  $c_1 \leq a_1 b_1$  will hold regardless of the case that A & B are invertible or not.

→ Now, for proving  $c_1 \geq \max(a_1 b_m, b_1 a_m)$ , we consider the further two cases

① Both A & B are not invertible  $\Rightarrow a_m = 0$  &  $b_m = 0 \Rightarrow \max(a_1 b_m, b_1 a_m) = 0$   
 $\Rightarrow c_1 \geq \max(a_1 b_m, b_1 a_m)$  trivially

② One of A & B is not invertible  $\Rightarrow$  But B is. So  $C = AB \Rightarrow C B^{-1} = A$

$$\begin{aligned} & c_1 \geq a_1 b_m \\ & \downarrow \\ & c_1 \geq \max(a_1 b_m, b_1 a_m) \quad (\text{as } a_m = 0) \end{aligned}$$

$$\boxed{\quad} + \boxed{\quad} + \boxed{\quad} + \boxed{\quad} = \boxed{\quad}$$

→ So, here also we can say :  $\boxed{a_1 b_1 \geq c_1 \geq \max(a_1 b_m, b_1 a_m)}$

⇒ So for any two matrix A, B & a matrix C  
s.t.  $C = AB$ , we proved the asked inequalities.

$$\boxed{\quad} + \boxed{\quad} + \boxed{\quad} + \boxed{\quad} = \boxed{\quad}$$

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A. Q. 2:

$\Rightarrow$  Let us define  $T' = \{v \in C^m \mid u^* v = 0 \text{ for all } u \in S\}$ , we need to prove that  $T = T'$ .

$\Rightarrow$  Now, from the definition of Orthogonal Subspace; if  $S$  &  $T$  are orthogonal then  $\forall u \in S \text{ and } \forall v \in T$ ,  $u^* v = 0$ . Now suppose some  $t \in T$ . Then we can trivially see that  $t \in T'$  too. Then  $\boxed{T \subseteq T'}$

$\Rightarrow$  Now, let some  $t \in T'$ . Since  $t \in C^m$  & we have that  $S + T = C^m$ ,  $\exists u \in S$  &  $\exists v \in T$  s.t.  $t = u + v$ . Now, from the definition of  $T'$  we have  $p^* t = 0$  ~~with  $\forall p \in S$~~ .

$\Rightarrow$  Thus:  $p^* u + p^* v = 0 \Rightarrow \boxed{p^* u = 0 \quad \forall p \in S}$

$$\boxed{T \subseteq T'} \Leftarrow \boxed{t = u + v} \Leftarrow \boxed{u = 0}$$

$\hookrightarrow$  Thus, we can see that  $\boxed{T' \subseteq T}$

$\Rightarrow$  Thus, we get  $\boxed{T = T'}$ . Note that till now, we only used the condition of  $S \perp T$  &  $S + T = C^m$ .

• If we add an additional constraint of  $S \cap T = \{0\}$  then the set  $T$  still remains the same as it already satisfies this constraint.

$$( \forall v \in C^m \text{ s.t. } v \in T \Rightarrow v^* v = 0 \Rightarrow \|v\|_2 = 0 \Rightarrow v = 0 )$$

&  $v \in S$

[ ] + [ ] + [ ] + [ ] = [ ]

$$(\text{using } \|x\|_2^2 = x^T x)$$

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A. 3: (a)  $\Rightarrow \text{using } \|x_i\|_2 = \|y_i\|_2 \Rightarrow x_i^T x_i = y_i^T y_i \quad \forall i$

①

$$\Rightarrow \text{using } \|x_i - x_j\|_2 = \|y_i - y_j\|_2 \Rightarrow \|x_i\|_2^2 + \|y_j\|_2^2 - 2 x_i^T x_j$$

$\forall i, j \text{ s.t. } i \neq j$

$$\begin{aligned} (\text{using } \|x - y\|_2^2 &= (x^T - y^T)(x - y)) \\ &= \|x\|_2^2 + \|y\|_2^2 - 2 x^T y \end{aligned} = \left( \|y_i\|_2^2 + \|y_j\|_2^2 - 2 y_i^T y_j \right)$$

$$x_i^T x_i = y_i^T y_i$$

②

$$\Rightarrow \text{Thus, we get } x_i^T x_i = y_i^T y_i \text{ for all } i, j$$

$\Rightarrow$  Now, looking at  $X^T X$ , we can see that

$$(X^T X)_{ij} = x_i^T x_j$$

(Similarly for  $Y^T Y$ )

$$\Rightarrow \text{using this, we get } X^T X = Y^T Y$$

$$X^T X = Y^T Y$$

(Since they are real matrices)

⇒ Now, if we consider the <sup>(Reduced)</sup> QR factorization of  $X \& Y$ :

$$X = \tilde{Q}_x \tilde{R}_x \quad \& \quad Y = \tilde{Q}_y \tilde{R}_y \Rightarrow \boxed{\tilde{R}_x^T \tilde{Q}_x^T \tilde{Q}_x \tilde{R}_x = \tilde{R}_y^T \tilde{Q}_y^T \tilde{Q}_y \tilde{R}_y}$$

$$\boxed{\tilde{R}_x^T \tilde{R}_x = \tilde{R}_y^T \tilde{R}_y}$$

$$(\because \tilde{Q}_x^T \tilde{Q}_x = \tilde{Q}_y^T \tilde{Q}_y = I)$$

⇒ Now Since  $\tilde{R}_x$  is a  $n \times n$  Upper Triangular matrix:

$$\tilde{R}_x^T \tilde{R}_x = \begin{bmatrix} r_{11} & & & \\ r_{12} r_{22} & \ddots & & \\ \vdots & & \ddots & \\ 0_{nn} & & & r_{nn} \end{bmatrix} \begin{bmatrix} r_{11} & r_{12} & \cdots & r_{1n} \\ r_{21} & r_{22} & \cdots & r_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ 0_{n1} & 0_{n2} & \cdots & r_{nn} \end{bmatrix} = \begin{bmatrix} r_{11}^2 & r_{11} \cdot r_{12} & \cdots & r_{11} \cdot r_{1n} \\ r_{11} \cdot r_{12} & r_{22}^2 & \cdots & r_{22} \cdot r_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ r_{11} \cdot r_{1n} & r_{22} \cdot r_{2n} & \cdots & r_{nn}^2 \end{bmatrix}$$

Zero Entries

$$\begin{bmatrix} r_{22}^2 + r_{12}^2 & r_{22} r_{1n} + r_{12} r_{2n} \\ r_{12} r_{1n} + r_{22} r_{2n} & r_{nn}^2 \end{bmatrix}$$

$(n-1 \times n-1)$   
Segment

Similarly for the  $(n-2 \times n-2)$   
Sub matrix & so on...

$\Rightarrow$  Now, thus, if we try to get  $\boxed{\hat{R}_x^T \hat{R}_x = \hat{R}_y^T \hat{R}_y}$ , we can

see that  $\boxed{(\hat{x}_x)_{11}^2 = (\hat{x}_y)_{11}^2} \Rightarrow \boxed{(\hat{x}_x)_{11} = (\hat{x}_y)_{11}}$

Comparing the  
1<sup>st</sup> Diagonal entry

Then we compare the first row / column

& get  $\boxed{(\hat{x}_x)_{11} (\hat{x}_x)_{12} = (\hat{x}_y)_{11} (\hat{x}_y)_{12}}$

Then we compare the second diagonal element and get  $\boxed{(\hat{x}_x)_{22} = (\hat{x}_y)_{22}}$

$\boxed{(\hat{x}_x)_{1j} = (\hat{x}_y)_{1j} \quad \forall j \geq 1}$

Then, we see the second row / column & on comparing & using the row rules for  $\hat{x}_x$  &  $\hat{x}_y$ ,

we get:  $\boxed{(\hat{x}_x)_{2j} = (\hat{x}_y)_{2j} \quad \forall j \geq 2}$

And so on

$\Rightarrow$  Thus by comparing like this; we get:  $\boxed{\hat{R}_x = \hat{R}_y}$  Hence Proved.

(b)  $\Rightarrow$  By taking off  $Q_x = Y_1, Q_x_2 = Y_2, \dots, Q_x_n = Y_n$ ; we can see that  $\boxed{Q_x = Y}$

$\Rightarrow$  Now since  $\boxed{\hat{R}_x = \hat{R}_y} \Rightarrow \boxed{R_x = R_y}$  (the last row are just extended by 0)

Full QR

$\Rightarrow$  Now, using  $\boxed{X = Q_x R_x}$  &  $\boxed{Y = Q_y R_y}$ ; we can see that

$$Q(Q_x R_x) = Q_y R_y \Rightarrow$$

$$QQ_x = Q_y$$

Use  $R_x = R_y$  2  
the fact that  
they are Upper Triangular



$$Q = Q_y Q_x^T$$

$\because Q_x$  is a square matrix  
with orthogonal columns

$$Q \text{ is orthogonal} \Leftrightarrow Q^T Q = I$$

$$(Q_x Q_y^T Q_y Q_x^T = Q_x Q_x^T = I)$$

Ans 4: a)  $\Rightarrow$  we have:  $FA = A - \frac{2VV^*}{V^*V} A$

$$FA = A + V \left( \frac{-2V^*A}{V^*V} \right) = A + Vw^*$$

i) Compute  $F$  first & then multiply with  $A$

$$F = I - \frac{2VV^*}{V^*V}$$

Needs  $m^2$  multiplications  
Subtraction  
Needs  $m^2$  multiplications &  $(m-1)$  additions  
Needs  $2m^2$  multiplications (with  $\alpha = VV^*$ )

Note: 1, 2, 3, 4. are the order of computation of the terms.

Thus we get  $VV^*$  first, then  $V^*V$ , then  $\frac{2(VV^*)}{(V^*V)}$ , & then  $I - \beta$ .

$$\boxed{\quad} + \boxed{\quad} + \boxed{\quad} + \boxed{\quad} = \boxed{\quad}$$

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→ Thus, we need :  $\boxed{\sim 4m^2 \text{ flops}}$  to compute F

→ Then to get FA, we will need m multiplications &  $(m-1)$  additions to compute each element of IFA

$$\xrightarrow{\quad} \boxed{\sim (2m)(mn) \text{ flops}}$$

→ Thus the asymptotic operation count here :  $\boxed{\sim 2m^2 n \text{ flops}}$

ii) Computing  $W^*$  first, then  $VW^*$ , & then matrix addition

$$* W^* = \begin{matrix} 3. \\ - 2 \frac{V^* A}{V^* V} \end{matrix} \xrightarrow{\quad} (2m-1)(n) \text{ flops}.$$

$\bullet$  2n flops

$\xrightarrow{\quad}$  Needs  $\boxed{\sim 2mn \text{ flops}}$  to be computed

→  $VW^*$  → Needs  $\boxed{\sim mn \text{ flops}}$  to be computed

→  $A + VW^*$  → Needs  $\boxed{\sim mn \text{ flops}}$  to be computed

→ Thus the asymptotic operation count here :  $\boxed{\sim 4mn \text{ flops}}$

⇒ Thus, the second way is much more efficient than the first.

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$$\boxed{\quad} + \boxed{\quad} + \boxed{\quad} + \boxed{\quad} = \boxed{\quad}$$

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(b)  $\Rightarrow$  Given  $\frac{\|\tilde{V} - V\|_2}{\|V\|_2} = O(\epsilon_m)$ , we have  $\tilde{V} = V(1 + \epsilon_1)$

$\epsilon_1$  is  $O(\epsilon_m)$

$\Rightarrow$  Now, if we consider the Error Part in the different operators involved here; Then we get:

$$\hat{F} = I - 2 \left( \underbrace{\frac{VV^*}{V^*V} \left( (1+\epsilon_1)^2 (1+\epsilon_3) \right)}_{\text{Can be reduced to}} (1+\epsilon_5) \right) (1+\epsilon_5)$$

All  $\epsilon_i$  is  $O(\epsilon_m)$

$$\hat{F} = I (1 + \epsilon_5) - 2 \left( \frac{VV^*}{V^*V} \right) (1 + \epsilon_6 + O(\epsilon_m))$$

(Along with  $1 + \epsilon_5$ ) Ignored

with  $\epsilon_5, \epsilon_6$  in  $O(\epsilon_m)$

$$\|\hat{F} - F\|_2 = \left\| I \epsilon_5 - 2 \epsilon_6 \left( \frac{VV^*}{V^*V} \right) \right\|_2 \leq \epsilon_5 \|I\|_2 + 2 \epsilon_6 \left\| \frac{VV^*}{V^*V} \right\|_2$$

$$\|\hat{F} - F\|_2 \leq \epsilon_5 + 2 \epsilon_6 \leq k \epsilon_m$$

(as  $\epsilon_5, \epsilon_6$  in  $O(\epsilon_m)$ )

Thus,  $\|\hat{F} - F\|_2 = O(\epsilon_m)$  L.P.

[ ] + [ ] + [ ] + [ ] = [ ]

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$\Rightarrow$  If we take Error gr by  $f\ell$  operator as  $\epsilon_7$  in  $O(\epsilon_m)$ , we get

$$\Rightarrow f\ell(\hat{F}A) = \left( A(1+\epsilon_5) - 2 \left( \frac{vv^*A}{v^*v} \right) (1+\epsilon_6) \right) (1+\epsilon_7)$$

$$= A(1+\epsilon_8) - 2 \left( \frac{vv^*A}{v^*v} \right) (1+\epsilon_9) \quad \text{with } \epsilon_8, \epsilon_9 \text{ in } O(\epsilon_m)$$

$$\Rightarrow \text{Now, } F^2 = \left( I - \frac{2vv^*}{v^*v} \right) \left( I - \frac{2vv^*}{v^*v} \right) = I - \frac{4vv^*}{v^*v} + 4 \frac{vv^*}{v^*v} \cdot \frac{vv^*}{v^*v} = I$$

$$\& \|F\|_2 \leq \|I\|_2 + 2 \left\| \frac{vv^*}{v^*v} \right\|_2 = 3$$

$$\& FA = A - \frac{2vv^*}{v^*v} A$$

(~~error~~)

$$\Rightarrow \text{by the above } \epsilon_i \text{'s } \Rightarrow f\ell(\hat{F}A) = \left( A - \frac{2vv^*A}{v^*v} \right) + \left( A\epsilon_8 - \frac{2\epsilon_9 vv^*A}{v^*v} \right)$$

$$f\ell(\hat{F}A) = FA + IA \left( \epsilon_8 - 2\epsilon_9 \left( \frac{vv^*}{v^*v} \right) \right)$$

$$f\ell(\hat{F}A) = F \left( A + FA \left( \epsilon_8 - 2\epsilon_9 \left( \frac{vv^*}{v^*v} \right) \right) \right) \quad \begin{matrix} \text{(why)} \\ I = F.F \end{matrix}$$

↓

$$\|\delta A\|_2 \leq \epsilon_8 \|F\|_2 \|A\|_2 + 2\epsilon_9 \|F\|_2 \|A\|_2 \left\| \frac{vv^*}{v^*v} \right\|_2$$

$$\delta A = FA \left( \epsilon_8 - 2\epsilon_9 \left( \frac{vv^*}{v^*v} \right) \right)$$

in  $O(\epsilon_m)$

$$\frac{\|\delta A\|_2}{\|A\|_2} \leq \left( 3\epsilon_8 + 6\epsilon_9 \right) \Rightarrow$$

$$\frac{\|\delta A\|_2}{\|A\|_2} = O(\epsilon_m)$$

H.P.

A<sub>5</sub> : (a)  $\Rightarrow$  we desire to minimize the residual norm  
 $\|r\|_2 = \|b - Ax\|_2$ . We claim that this is possible  
iff  $r \perp \text{range}(A) \Rightarrow A^* r = 0$

\* Then, from the definition of Orthogonal Projectors,  
the above statement would imply:

$y = Ax = (P)b$  would be the unique  
point minimizing  $\|b - y\|_2$   
Orthogonal Projector onto Range(A)

\* The Proof of this claim can be provided by taking  
some  $z \neq y$  to be another point in Range(A).

\* Now, since  $(z-y)$  is orthogonal to  $(b-y)$ , we can write

$$\|b-z\|_2^2 = \|(b-y) + (y-z)\|_2^2 = \|b-y\|_2^2 + \|y-z\|_2^2$$

( $\because z-y$  &  $b-y$  are orthogonal)

Up.  $\boxed{\|b-z\|_2^2 > \|b-y\|_2^2}$

$\Rightarrow$  Thus, we are needed to obtain the orthogonal projector matrix P. As obtained in Eq (6.6) of

Trefethen & Bau, this orthogonal projector P can be

given as:  $P = \tilde{U} \tilde{U}^*$  where  $A = \tilde{U} \tilde{\Sigma} \tilde{V}^*$   
(Reduced)  
SVD

$\Rightarrow$  We can obtain  $\tilde{U}$  for the  $U$  corresponding to the full SVD of  $A$  ( $A = U\Sigma V^*$ ) by taking the first  $r$  columns of  $U$ .

$\Rightarrow$  Thus, we get the unique vector  $y = \tilde{U}\tilde{U}^*b$

b)  $\Rightarrow$  Taking the full SVD of  $A$  again, as  $A = U\Sigma V^*$ , we can make use of the fact that  $VV^* = I$

$\downarrow$

(as  $V$  is unitary)

$$\min_x \|b - Ax\|_2^2 = \min_x \|U^*(Ax - b)\|_2^2$$

~~As Matrix  
is Unitary  
so  
it has  
all  
the Value~~

$$= \min_x \|U^*(AVV^*x - b)\|_2^2$$

$$= \min_x \|U^*(U\Sigma V^*VU^*x - b)\|_2^2$$

$$= \min_x \|\Sigma V^*x - V^*b\|_2^2$$

$$\|x\|_2^2 = \|U^*x\|_2^2$$

if  $U^*$  is Unitary

$\Rightarrow$  Now taking  $V^*x$  as  $z$ , we can obtain:

$$\min_x \|b - Ax\|_2^2 = \min_z \|\Sigma z - V^*b\|_2^2$$

$$= \sum_{i=1}^r (\sigma_i z_i - u_i^* b)^2 + \sum_{i=r+1}^m (u_i^* b)^2$$

(Because Rank of  $A = r \Rightarrow \sigma_j = 0$  for  $\sigma_j \geq r+1$ )

$\Rightarrow$  So, to get the minimum value, we can see:

$$d_i z_i = u_i^* b \quad \forall i=1 \dots r$$

~~z<sub>i</sub>~~ ↓

$$z_i = \begin{cases} \frac{u_i^* b}{d_i} & 1 \leq i \leq r \quad \& i \in \mathbb{N} \\ \text{Arbitrary} & r+1 \leq i \leq n \quad \& i \notin \mathbb{N} \end{cases}$$

$$\Rightarrow \text{we have } z = v^* x \Rightarrow x = v z \quad (\because v \text{ is unitary}).$$

↳ General Solution of

The Least-Squares Problem

$\Rightarrow$  Now, if we want to minimize  $\|x\|_2$ , we know that

$$\|x\|_2 = \|v x\|_2 = \|z\|_2 \rightarrow \text{will be minimized}$$

( $\because v$  is unitary)

if  $|z_j| = 0 \quad \forall j > r+1$

Minimize  $\|x\|_2$

$$x' = v z' \quad \leftarrow z'_i = \begin{cases} \frac{u_i^* b}{d_i} & i=1, 2, \dots, r \\ 0 & i=r+1, \dots, n \end{cases}$$

+  +  +  =

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Ans 6:

(a)  $\Rightarrow$  Suppose  $[P] = \begin{bmatrix} a_0 \\ a_1 \\ a_2 \\ \vdots \\ a_n \end{bmatrix}$   $[q] = \begin{bmatrix} b_0 \\ b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix}$  ( $P, q \in P$ )

$\Rightarrow$  Now:  $(P, q) = \int_{-1}^1 P(x) q(x) dx$

$\therefore (P, q) = \int_{-1}^1 (\bar{a}_0 + \bar{a}_1 x + \dots + \bar{a}_n x^n) (b_0 + b_1 x + \dots + b_n x^n) dx$

$= \int_{-1}^1 (\bar{a}_0 b_0 + (\bar{a}_1 b_0 + \bar{a}_0 b_1)x + (\bar{a}_2 b_0 + \bar{a}_1 b_1 + \bar{a}_0 b_2)x^2 + \dots) dx$

Co-efficient of  $x^j = \sum_{i=0}^j (\bar{a}_i b_{j-i})$

High Power Term :  $x^{2n}$

$\Rightarrow$  Doing the above integral, we get:

$$(P, q) = 2 \left( \frac{\bar{a}_0 b_0 + (\bar{a}_2 b_0 + \bar{a}_1 b_1 + \bar{a}_0 b_2)}{3} + \frac{(\bar{a}_4 b_0 + \bar{a}_3 b_1 + \bar{a}_2 b_2 + \bar{a}_1 b_3 + \bar{a}_0 b_4)}{5} + \dots \right)$$

$$\boxed{\quad} + \boxed{\quad} + \boxed{\quad} + \boxed{\quad} = \boxed{\quad}$$

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$\Rightarrow$  In the above expression, if we consider the term  $\bar{a}_i b_j$  with  $0 \leq i, j \leq n$ , the

Co-efficient would be :  $c_{ij} = \begin{cases} 2 & \text{if } (i+j) \text{ is even} \\ i+j+1 & \\ 0 & \text{else} \end{cases}$

$$\Rightarrow \text{Thus, we have : } (P, q) = \sum_{i=0}^n \sum_{j=0}^n \bar{a}_i c_{ij} b_j$$

$$\Rightarrow \text{Now, we also know that : } (P)^T G[q] = \sum_{i=0}^n \sum_{j=0}^n \bar{a}_i g_{ij} b_j$$

$\Rightarrow$  From comparing the above two Eqs, we get

$$G_{ij} = \begin{cases} 2 & \text{if } (i+j) \text{ is even} \\ i+j+1 & \\ 0 & \text{else} \end{cases}$$

$$0 \leq i, j \leq n$$

$(n+1) \times (n+1)$  matrix

$\hookrightarrow$  Satisfies  $G_i = G_i^T$

Thus, Hermitian

$$\boxed{\quad} + \boxed{\quad} + \boxed{\quad} + \boxed{\quad} = \boxed{\quad}$$

$c_{[P]}$  will be the coefficient vector

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b)  $\Rightarrow$  The Subspace  $\langle P \rangle$  will be defined as:

$$\langle c_P \mid c \in \mathbb{C} \rangle$$

$\Rightarrow$  Projecting  $q$  orthogonally onto the Subspace  $\langle P \rangle$  w.r.t the inner  $(P, q)$  can be defined as

Having  $q = r + (q-r)$   
 s.t.  $r \in \langle P \rangle$  &  
 $(r, (q-r)) = 0$

$\Rightarrow$  Since  $r \in \langle P \rangle$ ; we can write

$$r = c_P \quad \exists c \in \mathbb{C}$$

$$\Rightarrow \text{so, } (c_P, q - c_P) = 0 \Rightarrow (c_P, q) - (c_P, c_P) = 0$$

$$\left( \text{Since } (a, b-c) = \int_{-1}^1 \bar{a}(x) (b(x) - c(x)) dx = \int_{-1}^1 \bar{a}(x) b(x) dx - \int_{-1}^1 \bar{a}(x) c(x) dx \right)$$

$$\Rightarrow \text{Then, we can write: } (c_P, q) = \int_{-1}^1 \bar{c}_P(x) q(x) dx = \bar{c} \int_{-1}^1 \bar{P}(x) q(x) dx$$

$$\text{& } (c_P, c_P) = c \cdot \bar{c} \int_{-1}^1 \bar{P}(x) P(x) dx$$

$$\Rightarrow \text{So, we have: } (c_P, q) = (c_P, c_P) \Rightarrow \bar{c} (P, q) = c \bar{c} (P, P)$$

$$c = \frac{(P, q)}{(P, P)}$$

$$\boxed{\quad} + \boxed{\quad} + \boxed{\quad} + \boxed{\quad} = \boxed{\quad}$$

$\Rightarrow$  Thus, the Projection Vector will be

$$P \cdot \left( \frac{(P, a)}{(P, P)} \right)$$

$\Rightarrow$  Now using that  $(P, a) = [P]^* G [a]$ , we get:

$$\left[ P \left( \frac{(P, a)}{(P, P)} \right) \right] = [P] \left( \frac{[P]^* G [a]}{[P]^* G \cdot [P]} \right) = \left[ \frac{[P] [P]^* G}{[P]^* G \cdot [P]} [a] \right]$$

Projector matrix  
acting on  $[a]$