

**Q1:**

$$1. \mathbf{x} + \mathbf{y} = \begin{pmatrix} 2 \\ 6 \\ 1 \\ 4 \\ 1+5i \end{pmatrix}$$

$$2. \langle \mathbf{x}, \mathbf{y} \rangle = \mathbf{x}^\dagger * \mathbf{y} = (1 \ 2 \ 3 \ 4 \ -5i) * \begin{pmatrix} 1 \\ 4 \\ -2 \\ 0 \\ 1 \end{pmatrix} \\ = 1 \times 1 + 2 \times 4 - 3 \times 2 + 0 - 1 \times 5i = 3 - 5i$$

$$3. \mathbf{y}^\dagger \mathbf{x} = (1 \ 4 \ -2 \ 0 \ 1) * \begin{pmatrix} 1 \\ 2 \\ 3 \\ 4 \\ 5i \end{pmatrix} = 1 \times 1 + 4 \times 2 - 2 \times 3 + 0 + 5i = 3 + 5i$$

$$4. \mathbf{x} \circ \mathbf{y} = \begin{pmatrix} 1 \times 1 \\ 2 \times 4 \\ 3 \times -2 \\ 4 \times 0 \\ 5i \times 1 \end{pmatrix} = \begin{pmatrix} 1 \\ 8 \\ -6 \\ 0 \\ 5i \end{pmatrix}$$

$$5. \mathbf{xy}^\dagger = \begin{pmatrix} 1 \\ 2 \\ 3 \\ 4 \\ 5i \end{pmatrix} * (1 \ 4 \ -2 \ 0 \ 1) = \begin{pmatrix} 1 * (1 \ 4 \ -2 \ 0 \ 1) \\ 2 * (1 \ 4 \ -2 \ 0 \ 1) \\ 3 * (1 \ 4 \ -2 \ 0 \ 1) \\ 4 * (1 \ 4 \ -2 \ 0 \ 1) \\ 5i * (1 \ 4 \ -2 \ 0 \ 1) \end{pmatrix} = \begin{pmatrix} 1 & 4 & -2 & 0 & 1 \\ 2 & 8 & -4 & 0 & 2 \\ 3 & 12 & -6 & 0 & 3 \\ 4 & 16 & -8 & 0 & 4 \\ 5i & 20i & -10i & 0 & 5i \end{pmatrix}$$

6. Because  $\mathbf{xy}^\dagger$  is the linear combination of  $\mathbf{y}^\dagger$  and  $\mathbf{y}^\dagger$  is a real matrix, so  $\mathbf{y}^\dagger = \mathbf{y}^T$   
and  $Rank(xy^\dagger) = Rank(y^\dagger) = Rank(y^T) = Rank(y) = 1$

**Q2:**

1.  $\|\mathbf{x}\|_2 = \left( \sum_{i,j=1}^n |x_{i,j}| \right)^{\frac{1}{2}} = \sqrt{1^2 + 2^2 + 3^2 + 4^2 + |5i^2|} = \sqrt{55} = 7.4162$

2.  $\|\mathbf{xy}^\dagger\| = \sqrt{1210} = 34.7851$

In MATLAB, use  $\text{norm}(x, 2)$

### Q3

1. For  $\mathbf{A}$ ,

$$\begin{aligned} \det(\mathbf{A}) &= 1 * \begin{vmatrix} 1 & 2 & 4 \\ 2 & 1 & 3 \\ 4 & 3 & 1 \end{vmatrix} - (-1) * \begin{vmatrix} -1 & 2 & 4 \\ 0 & 1 & 3 \\ i & 3 & 1 \end{vmatrix} + 0 * \begin{vmatrix} -1 & 1 & 4 \\ 0 & 2 & 3 \\ i & 4 & 1 \end{vmatrix} - (-i) * \begin{vmatrix} -1 & 1 & 2 \\ 0 & 2 & 1 \\ i & 4 & 3 \end{vmatrix} \\ &= 20 + 8 + 2i + 3 - 2i = 31 > 0 \end{aligned}$$

Because  $\det(\mathbf{A}) > 0$ ,  $\mathbf{A}$  is the full-rank matrix.  $\text{Rank}(\mathbf{A}) = 4$

For any matrix  $\mathbf{x}$ ,  $\text{Rank}(\mathbf{x}) + \text{nullity}(\mathbf{x}) = n$ , so the dimension of  $\mathbf{x}$ 's null space is:

$$\text{nullity}(\mathbf{A}) = 4 - \text{Rank}(\mathbf{A}) = 0$$

2. For  $\mathbf{B}$ ,

$$\mathbf{B} = \begin{pmatrix} 1 & 4 & 3 \\ 2 & 3 & 1 \\ 3 & 2 & -1 \end{pmatrix} \xrightarrow{rref} \begin{pmatrix} 1 & 4 & 3 \\ 0 & -5 & -5 \\ 0 & 0 & 0 \end{pmatrix}$$

$$\text{Rank}(\mathbf{B}) = 2$$

$$\text{nullity}(\mathbf{B}) = 3 - 2 = 1$$

3. For  $\mathbf{C}$ ,

$$\mathbf{C} = \begin{pmatrix} 2 & 3 \\ -3 & 0.5 \end{pmatrix} \xrightarrow{rref} \begin{pmatrix} 2 & 3 \\ 0 & 5 \end{pmatrix}$$

$$\text{Rank}(\mathbf{C}) = 2$$

$$\text{nullity}(\mathbf{C}) = 2 - 2 = 0$$

$$\det(\mathbf{C}) = \begin{vmatrix} 2 & 3 \\ -3 & 0.5 \end{vmatrix} = 1 - (-9) = 10$$

$$\mathbf{C}^{-1} = \frac{1}{\det(\mathbf{C})} \begin{pmatrix} 0.5 & -3 \\ 3 & 2 \end{pmatrix} = \begin{pmatrix} 0.05 & -0.3 \\ 0.3 & 0.2 \end{pmatrix}$$

$$\mathbf{C} + \mathbf{C}^T = \begin{pmatrix} 2 & 3 \\ -3 & 0.5 \end{pmatrix} + \begin{pmatrix} 2 & -3 \\ 3 & 0.5 \end{pmatrix} = \begin{pmatrix} 4 & 0 \\ 0 & 1 \end{pmatrix}$$

eigenvalues of  $\mathbf{C} + \mathbf{C}^T$  is  $\lambda_1 = 4, \lambda_2 = 1$

$$\mathbf{C} - \mathbf{C}^T = \begin{pmatrix} 2 & 3 \\ -3 & 0.5 \end{pmatrix} - \begin{pmatrix} 2 & -3 \\ 3 & 0.5 \end{pmatrix} = \begin{pmatrix} 0 & 6 \\ -6 & 0 \end{pmatrix}$$

eigenvalues of  $\mathbf{C} - \mathbf{C}^T$  is  $\lambda_1 = 6i, \lambda_2 = -6i$

**Q4:**

1. First, calculate the inverse of  $\mathbf{A}\mathbf{A}^T$ :

$$\mathbf{A}\mathbf{A}^T = \begin{pmatrix} 1 & 0 & 3 \\ 2 & -1 & -2 \end{pmatrix} * \begin{pmatrix} 1 & 2 \\ 0 & -1 \\ 3 & -2 \end{pmatrix} = \begin{pmatrix} 10 & -4 \\ -4 & 9 \end{pmatrix} \implies (\mathbf{A}\mathbf{A}^T)^{-1} = inv(\mathbf{A}\mathbf{A}) =$$

$$\frac{1}{74} \begin{pmatrix} 9 & 4 \\ 4 & 10 \end{pmatrix} = \begin{pmatrix} 0.1216 & 0.0541 \\ 0.0541 & 0.1351 \end{pmatrix}$$

$$\text{Then } \mathbf{A}^T(\mathbf{A}\mathbf{A}^T)^{-1} = \begin{pmatrix} 1 & 2 \\ 0 & -1 \\ 3 & -2 \end{pmatrix} * \begin{pmatrix} 9 & 4 \\ 4 & 10 \end{pmatrix} = \begin{pmatrix} 0.2297 & 0.3243 \\ -0.0541 & -0.1351 \\ 0.2568 & -0.1081 \end{pmatrix}$$

$$2. \mathbf{A}\mathbf{A}^+ = \begin{pmatrix} 1 & 0 & 3 \\ 2 & -1 & -2 \end{pmatrix} * \begin{pmatrix} 0.2297 & 0.3243 \\ -0.0541 & -0.1351 \\ 0.2568 & -0.1081 \end{pmatrix} = I_2. \text{ So its dimension is 2.}$$

$$3. \mathbf{A}^+\mathbf{A} = \begin{pmatrix} 0.2297 & 0.3243 \\ -0.0541 & -0.1351 \\ 0.2568 & -0.1081 \end{pmatrix} * \begin{pmatrix} 1 & 0 & 3 \\ 2 & -1 & -2 \end{pmatrix} = \begin{pmatrix} 0.8784 & -0.3243 & 0.0405 \\ -0.3243 & 0.1351 & 0.1081 \\ 0.0405 & 0.1081 & 0.9865 \end{pmatrix}$$

$$\stackrel{rref}{=} \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix}. \text{ Its dimension is } 3 \times 2.$$

**Q5:**

$\mathbf{ABC}(\mathbf{ABC})^T = \mathbf{I} \implies \mathbf{ABC} = (.)_{M \times Q}$  is semi-orthogonal if  $M < Q$  or orthogonal if  $M = Q$ . So when  $\mathbf{A}, \mathbf{B}, \mathbf{C}$  are semi-orthogonal so that  $\mathbf{AA}^T = \mathbf{I}_M$ ;  $\mathbf{BB}^T = \mathbf{I}_N$ ;  $\mathbf{CC}^T = \mathbf{I}_P$  if  $M < N, N < P, P < Q$  or orthogonal so that  $\mathbf{AA}^T = \mathbf{BB}^T = \mathbf{CC}^T = \mathbf{I}_M$  if  $M = N = P = Q$ , then:

$$\mathbf{ABC}(\mathbf{ABC})^T = \mathbf{AB}(\mathbf{CC}^T)\mathbf{B}^T\mathbf{A}^T = \mathbf{A}(\mathbf{BI}_P\mathbf{B}^T)\mathbf{A}^T = \mathbf{AI}_N\mathbf{A}^T = \mathbf{I}_M$$

So its  $\dim(.) = M$

**Q6:**

$$\mathbf{A}\mathbf{A}^\dagger = \frac{1}{\sqrt{2}} \begin{pmatrix} 1+i & 1-i & 1+i \\ 1-i & 1+i & 1-i \\ 1+i & 1-i & 1+i \end{pmatrix} * \frac{1}{\sqrt{2}} \begin{pmatrix} 1-i & 1+i & 1-i \\ 1+i & 1-i & 1+i \\ 1-i & 1+i & 1-i \end{pmatrix} = \begin{pmatrix} 3 & 0 & 3 \\ 0 & 2 & 0 \\ 3 & 0 & 3 \end{pmatrix}$$

$$\mathbf{B}\mathbf{B}^\mathbf{T} = \mathbf{B} * \mathbf{B} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$$

$$\mathbf{C}\mathbf{C}^\mathbf{T} = \frac{1}{3} \begin{pmatrix} 1 & 2 & 2 \\ -2 & -1 & 2 \\ 2 & -2 & 1 \end{pmatrix} * \frac{1}{3} \begin{pmatrix} 1 & -2 & 2 \\ 2 & -1 & -2 \\ 2 & 2 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

Because  $\mathbf{B}$  is a symmetric matrix,  $\mathbf{B} = \mathbf{B}^\mathbf{T} \implies \mathbf{B}\mathbf{B}^\mathbf{T} = \mathbf{B}\mathbf{B} = \mathbf{B}^2$

$\mathbf{A}$  is not a Hermitian matrix, not conjugate symmetric, therefore  $\mathbf{A}\mathbf{A}^\dagger \neq \mathbf{A}^2$ .

$\mathbf{C}$  is not a symmetric matrix, therefore  $\mathbf{C}\mathbf{C}^\mathbf{T} \neq \mathbf{C}^2$ .

**Q7:**

$$\frac{\partial f(\mathbf{x})}{\partial \mathbf{x}} = e^{-\frac{1}{2}\mathbf{x}^T \mathbf{K}^{-1} \mathbf{x}} * \left(-\frac{1}{2}\mathbf{x}^T \mathbf{K}^{-1} \mathbf{x}\right)' = e^{-\frac{1}{2}\mathbf{x}^T \mathbf{K}^{-1} \mathbf{x}} * -\frac{1}{2}(2\mathbf{K}^{-1} \mathbf{x}) = e^{-\frac{1}{2}\mathbf{x}^T \mathbf{K}^{-1} \mathbf{x}} \mathbf{K}^{-1} \mathbf{x}$$

As for its dimension,  $\mathbf{K}^{-1} \mathbf{x} = (.)_{N \times 1}$  and  $\mathbf{x}^T \mathbf{K}^{-1} \mathbf{x} = (.)_{1 \times 1}$   
 $\implies \dim\left(\frac{\partial f(\mathbf{x})}{\partial \mathbf{x}}\right) = \dim(\mathbf{K}^{-1} \mathbf{x}) = N \times 1$

**Q8:**

For  $\mathbf{A}$ ,  $\det(\mathbf{A} - \lambda I) = 0 \implies \text{eig}(\mathbf{A})$  in MATLAB: Then its eigenvalues are:  
 $\lambda_1 = 11.0632, \lambda_2 = 2.922, \lambda_3 = -0.9791 + 0.3559i, \lambda_4 = -0.9791 - 0.3559i$   
 $\mathbf{A}$ 's corresponding diagonal matrix  $\mathbf{D}$  is:

$$\mathbf{D} = \begin{pmatrix} 11.0632 & & & \\ & 2.922 & & \\ & & -0.9791 + 0.3559i & \\ & & & -0.9791 - 0.3559i \end{pmatrix}$$

The corresponding nonsingular matrix  $\mathbf{\Lambda}$  (via  $[\mathbf{\Lambda}, \mathbf{D}] = \text{eig}(\mathbf{A})$ ):

$$\mathbf{\Lambda} = \begin{pmatrix} -0.5140 & 0.5688 & 0.5378 + 0.0107i & 0.5378 - 0.0107i \\ -0.3665 & -0.7124 & 0.4333 - 0.1353i & 0.4333 + 0.1353i \\ -0.5291 & 0.2034 & -0.6356 & -0.6356 \\ -0.5671 & -0.3573 & -0.2862 + 0.1366i & -0.2862 - 0.1366i \end{pmatrix}$$

so that  $\mathbf{\Lambda D \Lambda}^{-1} = \mathbf{A}$

For  $\mathbf{B}$ , same operation as  $\mathbf{A}$ , then its eigenvalues are:

$$\lambda_1 = -2.1487, \lambda_2 = 0.0770, \lambda_3 = 4.3554, \lambda_4 = 9.7164$$

$\mathbf{B}$ 's diagonal matrix  $\mathbf{D}$  is:

$$\mathbf{D} = \begin{pmatrix} -2.1487 & & & \\ & 0.0770 & & \\ & & 4.3554 & \\ & & & 9.7164 \end{pmatrix}$$

$$\text{and corresponding nonsingular matrix } \mathbf{\Lambda} = \begin{pmatrix} 0.2041 & 0.6932 & -0.3318 & 0.6064 \\ 0.6640 & -0.2284 & 0.6081 & 0.3704 \\ -0.1256 & -0.6754 & -0.4595 & 0.5630 \\ -0.7083 & 0.1054 & 0.5559 & 0.4221 \end{pmatrix}$$

Before calculating the eigenvalues, its easy to predict  $\mathbf{B}$  has real eigenvalues, since its a **real symmetric matrix**.

**Proof:**  $\mathbf{B}$  is a real symmetric matrix, so  $\mathbf{B} = \mathbf{B}^T = \bar{\mathbf{B}}$ . Assume  $\lambda$  is a complex eigenvalue and  $\mathbf{x}$  is the complex eigenvector of  $\mathbf{B}$ , and  $\bar{\lambda}$  and  $\bar{\mathbf{x}}$  are the corresponding conjugate complex and conjugate vector, then we have

$$\mathbf{B}\bar{\mathbf{x}} = \bar{\mathbf{B}}\bar{\mathbf{x}} = (\bar{\mathbf{B}}\mathbf{x}) = (\bar{\lambda}\mathbf{x}) = \bar{\lambda}\bar{\mathbf{x}}$$

$$\bar{\mathbf{x}}^T \mathbf{B} \mathbf{x} = \bar{\mathbf{x}}^T (\mathbf{B} \mathbf{x}) = \bar{\mathbf{x}}^T \lambda \mathbf{x} = \lambda \bar{\mathbf{x}}^T \mathbf{x} \quad (1)$$

$$\bar{\mathbf{x}}^T \mathbf{B} \mathbf{x} = (\mathbf{x}^T \mathbf{B}^T) \mathbf{x} = (\mathbf{B} \bar{\mathbf{x}})^T \mathbf{x} = (\bar{\lambda} \bar{\mathbf{x}})^T \mathbf{x} = \bar{\lambda} \bar{\mathbf{x}}^T \mathbf{x} \quad (2)$$

Let (1)-(2), we get  $(\lambda - \bar{\lambda}) \bar{\mathbf{x}}^T \mathbf{x} = 0$ . Since  $\mathbf{x}$  is not 0,  $\lambda = \bar{\lambda}$ . Therefore,  $\lambda$  is real.

As for  $\mathbf{A}$ , the characteristic polynomials format of its eigenvalues is  $\det(\mathbf{A} - \lambda \mathbf{E}) = x^4 - 12x^3 + 6x^2 + 48x + 35 = 0$ . According to **Descartes' rule of signs**, we could get the number of positive and negative real roots in a polynomial. Let  $f(x) = \underbrace{x^4 - 12x^3 + 6x^2 + 48x + 35}$  and  $f(-x) = \underbrace{x^4 + 12x^3 + 6x^2 - 48x + 35}$ , there are both 2 sign changes, indicating 2 or 0 positive roots and 2 or 0 negative roots.



Therefore,  $numberofcomplexroots = n - p - q \geq 0$ , where  $n$  is the degree of polynomials (here is 4), and  $p$  and  $q$  are the number of positive and negataive roots respectively. It's high likely that  $\mathbf{A}$  has complex roots.

**Q9:**

$$\mathbf{A} = \begin{pmatrix} 1 & 1 & 2 & 2 \\ 1 & 3 & 3 & 2 \\ 3 & 1 & 3 & 2 \\ 2 & 6 & 6 & 4 \end{pmatrix} \stackrel{rref}{=} \begin{pmatrix} 1 & 0 & 0 & -1 \\ 0 & 1 & 0 & -1 \\ 0 & 0 & 1 & 2 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$

After adding the noise  $\mathbf{N} = \begin{pmatrix} -0.0065 & -0.0085 & -0.0020 & -0.0151 \\ 0.0118 & -0.0057 & 0.0059 & 0.0088 \\ -0.0076 & -0.0056 & -0.0085 & -0.0024 \\ -0.0111 & 0.0018 & 0.0080 & 0.0017 \end{pmatrix}$ , the bottom

row of  $\mathbf{A}$  would be no longer 0. Since the noise row is less likely to be a linear combination of previous 3 rows, making the noised matrix  $\mathbf{A} + \mathbf{N}$  full rank, which means the inverse matrix exists.

However, because the noise only adds trivial value to original matrix, its eigenvalues changes little. The eigenvalues of  $\mathbf{A}$  and eigenvalues of  $\mathbf{A} + \mathbf{N}$  are:

$$\mathbf{vec}_{\lambda 1} = \begin{pmatrix} 10.5625 \\ 0.2188 + 0.5752i \\ 0.2188 - 0.5752i \\ 0.0000 \end{pmatrix}, \mathbf{vec}_{\lambda 2} = \begin{pmatrix} 10.5552 \\ 0.1969 + 0.5579i \\ 0.1969 - 0.5579i \\ 0.0320 \end{pmatrix}$$

$$\Delta \mathbf{vec}_{\lambda} = \mathbf{vec}_{\lambda 1} - \mathbf{vec}_{\lambda 2} = \begin{pmatrix} 0.0073 \\ 0.0219 + 0.0173i \\ 0.0219 - 0.0173i \\ -0.0320 \end{pmatrix} \text{ and } \text{mean}(\Delta \mathbf{vec}_{\lambda}) = 0.0048$$

We can see that the eigenvalues doesn't change a lot when  $\alpha = 0.01$ . If the  $\alpha$  increase, the eigenvalues also change more and vice versa, as the following image shows.

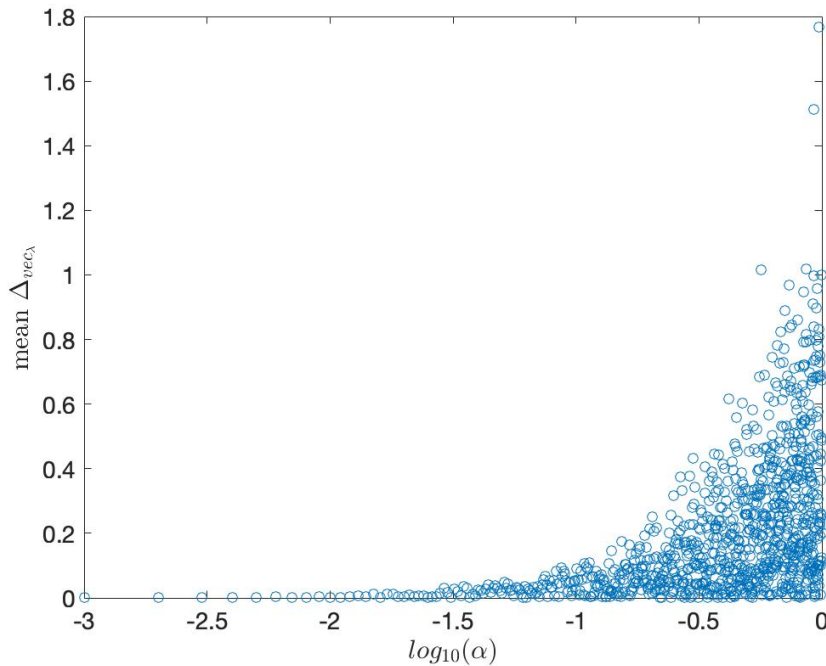


Figure 1:  $\mathbf{A}$ 's eigenvalues change under different noise amplitude  $\alpha$ . Not axis  $y$  is the  $\text{mean}(\Delta \mathbf{vec}_{\lambda})$ ; axis  $x$  is the  $\log_{10}$  value of  $\alpha$ . We tested  $\alpha$  from 0.001 to 1 with 0.001 as the stride.