

# Advanced Linear Algebra: Mid-Term

April 22, 2022

## Solution 1

1. Proof:

$$\begin{aligned}
 i^2 &= \begin{bmatrix} 1 & -1 & & \\ & & 1 & -1 \\ & & & 1 \end{bmatrix}^2 = \begin{bmatrix} 1 & -1 & & \\ & & 1 & -1 \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & -1 & & \\ & & 1 & -1 \\ & & & 1 \end{bmatrix} = \begin{bmatrix} -1 & & & \\ & -1 & & \\ & & -1 & \\ & & & -1 \end{bmatrix} \\
 j^2 &= \begin{bmatrix} & -1 & & \\ 1 & & 1 & \\ & & & 1 \\ & -1 & & \end{bmatrix}^2 = \begin{bmatrix} & -1 & & \\ 1 & & 1 & \\ & & & 1 \\ & -1 & & \end{bmatrix} \begin{bmatrix} & -1 & & \\ 1 & & 1 & \\ & & & 1 \\ & -1 & & \end{bmatrix} = \begin{bmatrix} -1 & & & \\ & -1 & & \\ & & -1 & \\ & & & -1 \end{bmatrix} \\
 k^2 &= \begin{bmatrix} & & -1 & \\ & 1 & & \\ 1 & & & \\ & & & 1 \end{bmatrix}^2 = \begin{bmatrix} & & -1 & \\ & 1 & & \\ 1 & & & \\ & & & 1 \end{bmatrix} \begin{bmatrix} & & -1 & \\ & 1 & & \\ 1 & & & \\ & & & 1 \end{bmatrix} = \begin{bmatrix} -1 & & & \\ & -1 & & \\ & & -1 & \\ & & & -1 \end{bmatrix} \\
 ijk &= \begin{bmatrix} & -1 & & \\ 1 & & & \\ & & 1 & -1 \\ & & & 1 \end{bmatrix} \begin{bmatrix} & & -1 & \\ 1 & & 1 & \\ & & & 1 \\ & -1 & & \end{bmatrix} \begin{bmatrix} & & & -1 \\ & 1 & & \\ & & 1 & \\ 1 & & & \end{bmatrix} \\
 &= \begin{bmatrix} & -1 & & \\ 1 & & & \\ & & 1 & -1 \\ & & & 1 \end{bmatrix} \begin{bmatrix} & -1 & & \\ 1 & & & \\ & & 1 & -1 \\ & & & 1 \end{bmatrix} = \begin{bmatrix} -1 & & & \\ & -1 & & \\ & & -1 & \\ & & & -1 \end{bmatrix}
 \end{aligned}$$

2. (Actually, it is known as Hamilton product)

$$\begin{aligned}
 L_q &= \begin{bmatrix} r & -x & -y & -z \\ x & r & -z & y \\ y & z & r & -x \\ z & -y & x & r \end{bmatrix}, R_q = \begin{bmatrix} r & -x & -y & -z \\ x & r & z & -y \\ y & -z & r & x \\ z & y & -x & r \end{bmatrix} \\
 L_q R_q &= \begin{bmatrix} r & -x & -y & -z \\ x & r & -z & y \\ y & z & r & -x \\ z & -y & x & r \end{bmatrix} \begin{bmatrix} r & -x & -y & -z \\ x & r & z & -y \\ y & -z & r & x \\ z & y & -x & r \end{bmatrix} \\
 &= \begin{bmatrix} r^2 - x^2 - y^2 - z^2 & -2rx & -2ry & -2rz \\ 2rx & r^2 - x^2 + y^2 + z^2 & -2xy & -2xz \\ 2ry & -2xy & r^2 + x^2 - y^2 + z^2 & -2yz \\ 2rz & -2xz & -2yz & r^2 + x^2 + y^2 - z^2 \end{bmatrix} \\
 R_q L_q &= \begin{bmatrix} r & -x & -y & -z \\ x & r & z & -y \\ y & -z & r & x \\ z & y & -x & r \end{bmatrix} \begin{bmatrix} r & -x & -y & -z \\ x & r & -z & y \\ y & z & r & -x \\ z & -y & x & r \end{bmatrix}
 \end{aligned}$$

$$= \begin{bmatrix} r^2 - x^2 - y^2 - z^2 & -2rx & -2ry & -2rz \\ 2rx & r^2 - x^2 + y^2 + z^2 & -2xy & -2xz \\ 2ry & -2xy & r^2 + x^2 - y^2 + z^2 & -2yz \\ 2rz & -2xz & -2yz & r^2 + x^2 + y^2 - z^2 \end{bmatrix}$$

So,  $L_q R_q = R_q L_q$ .

3.

$$L_q R_{\bar{q}} = \begin{bmatrix} r^2 + x^2 + y^2 + z^2 & 0 & 0 & 0 \\ 0 & r^2 + x^2 - y^2 - z^2 & 2xy - 2rz & 2xz + 2ry \\ 0 & 2xy + 2rz & r^2 - x^2 + y^2 - z^2 & 2yz - 2xr \\ 0 & 2xz - 2yr & 2yz - 2rx & r^2 - x^2 - y^2 + z^2 \end{bmatrix}$$

Say  $q = r + xi + yj + zk$ , its matrix representation is

$$\begin{bmatrix} r & -x & -y & -z \\ x & r & -z & y \\ y & z & r & -x \\ z & -y & x & r \end{bmatrix}.$$

Then its conjugate  $\bar{q}$  has matrix representation of

$$\begin{bmatrix} r & x & y & z \\ -x & r & z & -y \\ -y & -z & r & x \\ -z & y & -x & r \end{bmatrix}$$

which is exactly the transpose of its conjugate form.

By letting  $Q = \begin{bmatrix} r^2 + x^2 - y^2 - z^2 & 2xy - 2rz & 2xz + 2ry \\ 2xy + 2rz & r^2 - x^2 + y^2 - z^2 & 2yz - 2xr \\ 2xz - 2yr & 2yz - 2rx & r^2 - x^2 - y^2 + z^2 \end{bmatrix}$  (here it is not necessary orthogonal, but its orthogonality will be shown below),

$$L_q R_{\bar{q}} = \begin{bmatrix} 1 & 0 \\ 0 & Q \end{bmatrix}, (L_q R_{\bar{q}})^T (L_q R_{\bar{q}}) = \begin{bmatrix} 1 & 0 \\ 0 & Q^T \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & Q \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & Q^T Q \end{bmatrix}$$

So  $Q$  is orthogonal iff  $L_q R_{\bar{q}}$  is.

Due to the fact that  $q$  is unit quaternion, the matrix representation of  $q\bar{q}$  has to be the identity matrix, therefore

$$(L_q R_{\bar{q}})^T (L_q R_{\bar{q}}) = R_{\bar{q}}^T L_q^T L_q R_{\bar{q}} = R_q L_{\bar{q}} L_q R_{\bar{q}} = R_q R_{\bar{q}} = I$$

Hence  $Q$  is orthogonal. ■

## Solution 2

1. Let  $X = \begin{bmatrix} X_{11} & X_{12} \\ X_{21} & X_{22} \end{bmatrix}$ , then

$$\begin{bmatrix} R & \\ & N \end{bmatrix} \begin{bmatrix} X_{11} & X_{12} \\ X_{21} & X_{22} \end{bmatrix} = \begin{bmatrix} X_{11} & X_{12} \\ X_{21} & X_{22} \end{bmatrix} \begin{bmatrix} R' & \\ & N' \end{bmatrix} \\ \Leftrightarrow \begin{bmatrix} RX_{11} & RX_{12} \\ NX_{21} & NX_{22} \end{bmatrix} = \begin{bmatrix} X_{11}R' & X_{12}N' \\ X_{21}R' & X_{22}N' \end{bmatrix}$$

So  $RX_{12} = X_{12}N'$  and  $NX_{21} = X_{21}R'$ .

By the uniqueness of the solutions of Sylvester's equations,  $X_{12} = 0$ ,  $X_{21} = 0$ .

$X_{11}R'X_{11}^{-1} = R \Leftrightarrow X_{11}(R')^{-1}X_{11}^{-1} = R^{-1}$ .

$$\Rightarrow X \begin{bmatrix} (R')^{-1} & \\ & 0 \end{bmatrix} X^{-1} = \begin{bmatrix} X_{11}(R')^{-1}X_{11}^{-1} & \\ & 0 \end{bmatrix} = \begin{bmatrix} R^{-1} & \\ & 0 \end{bmatrix}$$

2.

$$AA^{(D)} = X \begin{bmatrix} A_R & 0 \\ 0 & A_N \end{bmatrix} X^{-1} X \begin{bmatrix} A_R^{-1} & 0 \\ 0 & 0 \end{bmatrix} X^{-1} = X \begin{bmatrix} I & 0 \\ 0 & 0 \end{bmatrix} X^{-1}$$

$$A^{(D)}A = X \begin{bmatrix} A_R^{-1} & 0 \\ 0 & 0 \end{bmatrix} X^{-1} X \begin{bmatrix} A_R & 0 \\ 0 & A_N \end{bmatrix} X^{-1} = X \begin{bmatrix} I & 0 \\ 0 & 0 \end{bmatrix} X^{-1}$$

$$\Rightarrow A^{(D)}A = AA^{(D)}.$$

$$A^{(D)}AA^{(D)} = X \begin{bmatrix} I & 0 \\ 0 & 0 \end{bmatrix} X^{-1} X \begin{bmatrix} A_R^{-1} & 0 \\ 0 & 0 \end{bmatrix} X^{-1} = X \begin{bmatrix} A_R^{-1} & 0 \\ 0 & 0 \end{bmatrix} X^{-1} = A^{(D)}$$

$$\begin{aligned} A^{(D)}A^{k+1} - A^k &= X \begin{bmatrix} A_R^{-1} & 0 \\ 0 & 0 \end{bmatrix} X^{-1} X \begin{bmatrix} A_R^{k+1} & 0 \\ 0 & A_N^{k+1} \end{bmatrix} X^{-1} - X \begin{bmatrix} A_R^k & 0 \\ 0 & A_N^k \end{bmatrix} X^{-1} \\ &= X \begin{bmatrix} A_R^k & 0 \\ 0 & 0 \end{bmatrix} X^{-1} - X \begin{bmatrix} A_R^k & 0 \\ 0 & 0 \end{bmatrix} X^{-1} = 0 \end{aligned}$$

3. Claim:

$$(\mathbf{ab}^*)^{(D)} = \begin{cases} \frac{\mathbf{ab}^*}{(\mathbf{b}^*\mathbf{a})^2}, & \mathbf{b}^*\mathbf{a} \neq 0 \\ 0, & \mathbf{b}^*\mathbf{a} = 0 \end{cases}$$

Proof: Assume that  $\lambda \neq 0$ .

$$\det(\lambda I_n - \mathbf{ab}^*) = \lambda^n \det(1 - \frac{1}{\lambda} \mathbf{b}^*\mathbf{a}) = \lambda^{n-1}(\lambda - \mathbf{b}^*\mathbf{a})$$

So  $\lambda = \mathbf{b}^*\mathbf{a}$  is the only non-zero eigenvalue, when  $\mathbf{b}^*\mathbf{a} \neq 0$ .

When  $\mathbf{b}^*\mathbf{a} = 0$ , the characteristic polynomial has no non-trivial zeros.

Hence, when  $\mathbf{b}^*\mathbf{a} = 0$ ,

$$(\mathbf{ab}^*)^{(D)} = X[0]X^{-1} = 0$$

When  $\mathbf{b}^*\mathbf{a} \neq 0$ ,  $\lambda = \mathbf{b}^*\mathbf{a}$  is the only non-zero eigenvalue, so

$$\mathbf{ab}^* = X \begin{bmatrix} \mathbf{b}^*\mathbf{a} & 0 \\ 0 & 0 \end{bmatrix} X^{-1}$$

This implies

$$(\mathbf{ab}^*)^{(D)} = X \begin{bmatrix} \frac{1}{\mathbf{b}^*\mathbf{a}} & 0 \\ 0 & 0 \end{bmatrix} X^{-1} = \frac{1}{(\mathbf{b}^*\mathbf{a})^2} X \begin{bmatrix} \mathbf{b}^*\mathbf{a} & 0 \\ 0 & 0 \end{bmatrix} X^{-1} = \frac{\mathbf{ab}^*}{(\mathbf{b}^*\mathbf{a})^2}$$

4.  $r$  in the context below is the size of the block matrix  $A_R$ .

To begin, prove a lemma first.

**Lemma:**

$S(k) : A_R^{-k}$  is a polynomial of  $A_R$  (The following denote  $A_R^{-k} = p_k(A_R)$ ),  $\forall k \in \mathbf{Z}^+$ .

$S(1) :$

Consider the characteristic polynomial of  $A_R$ :

$$p_{A_R}(\lambda) = \sum_{i=0}^r a_i \lambda^i,$$

where  $a_0 = (-1)^r \det(A_R) \neq 0$ . Also,

$$\begin{aligned} p_{A_R}(A_R) &= \sum_{i=0}^r a_i A_R^i = a_0 I + \sum_{i=1}^r a_i A_R^i = 0 \\ \Rightarrow \left( \sum_{i=0}^{r-1} \left( -\frac{a_{i+1}}{a_0} \right) A_R^i \right) A_R &= I \Rightarrow A_R^{-1} = \sum_{i=0}^{r-1} \left( -\frac{a_{i+1}}{a_0} \right) A_R^i = p_1(A_R) \end{aligned}$$

So,  $A_R^{-1}$  is a polynomial of  $A_R$ , with  $\deg p_1 \leq r-1$ .

Assume  $S(k)$  holds,

$$A_R^{-k} = p_k(A_R) = \sum_{i=0}^{r-1} a_{ik} A_R^i = a_{0k} I + \sum_{i=1}^{r-1} a_{ik} A_R^i$$

Then for  $S(k+1)$ ,

$$A_R^{-(k+1)} = A_R^{-1} p_k(A_R) = a_{0k} A_R^{-1} + \sum_{i=1}^{r-1} a_{ik} A_R^{i-1} = a_{0k} \sum_{i=0}^{r-1} \left( -\frac{a_{i+1}}{a_0} \right) A_R^i + \sum_{i=1}^{r-1} a_{ik} A_R^{i-1} = p_{k+1}(A_R)$$

is again a polynomial, so  $S(k) \Rightarrow S(k+1)$  and  $S(n)$  holds,  $\forall n \in \mathbf{Z}^+$ .

Let  $p(A) = A^{n-r} p_{n-r+1}(A)$ ,

$$\begin{aligned} p(A) &= A^{n-r} p_{n-r+1}(A) = X \begin{bmatrix} A_R^{n-r} & 0 \\ 0 & A_N^{n-r} \end{bmatrix} X^{-1} X \begin{bmatrix} p_{n-r+1}(A_R) & 0 \\ 0 & p_{n-r+1}(A_N) \end{bmatrix} X^{-1} \\ &= X \begin{bmatrix} A_R^{n-r} & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} A_R^{-(n-r+1)} & 0 \\ 0 & p_{n-r+1}(A_N) \end{bmatrix} X^{-1} = X \begin{bmatrix} A_R^{-1} & 0 \\ 0 & 0 \end{bmatrix} X^{-1} = A^{(D)} \end{aligned}$$

5.  $AB = BA$ , and  $A^{(D)}$  is a polynomial of  $A$ ,  $e^{-A^{(D)}Bt}$  is a polynomial of  $A^{(D)}B$ , so all these four matrices commute. Therefore,

$$\begin{aligned} Av' + Bv &= A(e^{-A^{(D)}Bt} AA^{(D)}v_0)' + B(e^{-A^{(D)}Bt} AA^{(D)}v_0) \\ &= -AA^{(D)}B e^{-A^{(D)}Bt} AA^{(D)}v_0 + B e^{-A^{(D)}Bt} AA^{(D)}v_0 \\ &= -B e^{-A^{(D)}Bt} AA^{(D)} AA^{(D)}v_0 + B e^{-A^{(D)}Bt} AA^{(D)}v_0 \\ &= -B e^{-A^{(D)}Bt} AA^{(D)}v_0 + B e^{-A^{(D)}Bt} AA^{(D)}v_0 = 0 \end{aligned}$$

### Solution 3

1.

$$(I_m - AB)^{-1} = I_m + AB + ABAB + ABABAB + \dots$$

2.

$$(I_m - AB)^{-1} = \sum_{k=0}^{+\infty} (AB)^k = I_m + \sum_{k=1}^{+\infty} (AB)^k = I_m + A \left( \sum_{k=1}^{+\infty} (BA)^{k-1} \right) B = I_m + A(I_n - BA)^{-1} B$$

3. Let  $p(x) = \sum_{k=0}^n a_k x^k$ , then

$$Ap(BA) = A \left( \sum_{k=0}^n a_k (BA)^k \right) = \sum_{k=0}^n a_k A(BA)^k = \sum_{k=0}^n a_k (AB)^k A = p(AB)A$$

4. Assume that  $g$  is well-defined (i.e. analytic) on the spectrum of  $X$  and  $Y$ .

$$\text{W.L.O.G., let } A = \begin{bmatrix} X & \\ & Y \end{bmatrix}$$

Let  $s$  be the number of distinct eigenvalues of  $A$ ,  $\lambda_i$ ,  $i = 1, 2, \dots, s$  be the eigenvalues of  $A$ .

By Hermite interpolation, it is possible to find a polynomial  $p$  such that  $\deg p = \text{size } t$  of the largest Jordan block such that

$$p^{(t')}(\lambda_i) = g^{(t')}(\lambda_i), \forall t' : 0 \leq t' < t, \forall i : 1 \leq i \leq s$$

Then immediately we have,

$$\begin{aligned} g(J_i) &= \begin{bmatrix} g(\lambda_i) & g'(\lambda_i)/1! & g''(\lambda_i)/2! & \dots & g^{(n-1)}(\lambda_i)/(n-1)! \\ & g(\lambda_i) & g'(\lambda_i)/1! & \dots & g^{(n-2)}(\lambda_i)/(n-2)! \\ & & \ddots & \ddots & \vdots \\ & & & g(\lambda_i) & g'(\lambda_i) \\ & & & & g(\lambda_i) \end{bmatrix} \\ &= \begin{bmatrix} p(\lambda_i) & p'(\lambda_i)/1! & p''(\lambda_i)/2! & \dots & p^{(n-1)}(\lambda_i)/(n-1)! \\ & p(\lambda_i) & p'(\lambda_i)/1! & \dots & p^{(n-2)}(\lambda_i)/(n-2)! \\ & & \ddots & \ddots & \vdots \\ & & & p(\lambda_i) & p'(\lambda_i) \\ & & & & p(\lambda_i) \end{bmatrix} = p(J_i) \end{aligned}$$

Due to the fact that  $g$ ,  $X$  and  $Y$  are fixed,  $A$  can have its Jordan decomposition:

$$A = BJB^{-1}$$

such that

$$\begin{aligned} \begin{bmatrix} g(X) & & \\ & g(Y) & \\ & & \ddots \end{bmatrix} &= g(A) = Bg(J)B^{-1} = B \begin{bmatrix} g(J_1) & & \\ & g(J_2) & \\ & & \ddots \end{bmatrix} B^{-1} = \begin{bmatrix} p(J_1) & & \\ & p(J_2) & \\ & & \ddots \end{bmatrix} B^{-1} \\ &= p(A) = \begin{bmatrix} p(X) & & \\ & p(Y) & \\ & & \ddots \end{bmatrix} \end{aligned}$$

5.  $f$  is defined on  $AB$  and  $BA$ , so  $\exists p$  s.t.

$$p(AB) = f(AB), p(BA) = f(BA)$$

Therefore,

$$Af(BA) = Ap(BA) = p(AB)A = f(AB)A$$

6.

$$I_m + Af(BA)B = I_m + f(AB)AB = I_m + \left(\sum_{k=0}^{+\infty} (AB)^k\right)(AB) = \sum_{k=0}^{+\infty} (AB)^k = f(AB)$$

#### Solution 4

$$1. \text{ Let } X = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & & \vdots \\ a_{(n-1)1} & a_{(n-1)2} & \cdots & a_{(n-1)n} \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{bmatrix}.$$

$$\text{Then } XN = \begin{bmatrix} 0 & a_{11} & \cdots & a_{1(n-1)} \\ 0 & a_{21} & \cdots & a_{2(n-1)} \\ \vdots & \vdots & & \vdots \\ 0 & a_{n1} & \cdots & a_{n(n-1)} \end{bmatrix}, \quad NX = \begin{bmatrix} a_{21} & a_{22} & \cdots & a_{2n} \\ a_{31} & a_{32} & \cdots & a_{3n} \\ \vdots & \vdots & & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \\ 0 & 0 & \cdots & 0 \end{bmatrix}.$$

By observation,

$$a_{i1} = 0, (i = 2, 3, \dots, n), a_{1j} = 0, (j = 1, 2, \dots, n-1), a_{ij} = a_{(i-1)(j-1)}$$

Hence,  $X$  is an upper triangular matrix, with equal elements on every upper diagonals, which is a linear combination of powers of nilpotent matrices.

2.

$$\begin{aligned} Y = e^N &= \sum_{k=0}^{+\infty} \frac{1}{k!} N^k = \sum_{k=0}^{n-1} \frac{1}{k!} N^k, \quad Y - I = \sum_{k=1}^{n-1} \frac{1}{k!} N^k \\ (Y - I)^k &= N^k \sum_{i_1=1}^{n-1} \cdots \sum_{i_k=1}^{n-1} \frac{1}{i_1! \cdots i_k!} N^{i_1 + \cdots + i_k - k} \end{aligned}$$

Suppose by contradiction that  $Y, Y - I, (Y - I)^2, \dots, (Y - I)^{n-1}$  are not independent matrices, then W.L.O.G. we may let

$$Y = \sum_{k=1}^{n-1} a_k (Y - I)^k$$

That immediately results in

$$I + \sum_{k=1}^{n-1} \frac{1}{k!} N^k = Y = \sum_{k=1}^{n-1} a_k N^k \sum_{i_1=1}^{n-1} \cdots \sum_{i_k=1}^{n-1} \frac{1}{i_1! \cdots i_k!} N^{i_1 + \cdots + i_k - k}$$

then,  $I$  cannot be represented by linear combinations of the other matrices, which is a contradiction (so  $Y$  is independent to other matrices).

Similarly, if we let

$$(Y - I)^j = \sum_{k=j+1}^{n-1} a_k (Y - I)^k$$

then,

$$(N^j) \sum_{i_1=1}^{n-1} \dots \sum_{i_j=1}^{n-1} \frac{1}{i_1! \dots i_j!} N^{i_1 + \dots + i_j - j} = (Y - I)^j = \sum_{k=j+1}^{n-1} a_k N^k \sum_{i_1=1}^{n-1} \dots \sum_{i_k=1}^{n-1} \frac{1}{i_1! \dots i_k!} N^{i_1 + \dots + i_k - k}$$

With regards to  $(Y - I)^j$ , the elements of the  $N^j$  component cannot be expressed by the linear combinations of the matrices of  $(Y - I)^{j+1}, (Y - I)^{j+2}, \dots$  which is another contradiction.

Therefore, these matrices are linearly independent.

In addition, since  $N^n = 0$ , but  $N^{n-1} \neq 0$ , so dimension of the solution space of  $X$  is  $n$ , corresponding to these  $n$  matrices, therefore these matrices form a basis of the solution space of  $X$ .

3. Suppose  $X$  is a solution to  $e^X = e^N$ . Then

$$YX = e^N X = e^X X = \sum_{k=0}^{+\infty} \frac{X^k}{k!} \cdot X = X \cdot \sum_{k=0}^{+\infty} \frac{X^k}{k!} = X e^X = X e^N = XY$$

Moreover,  $[X, Y - I] = [X, (Y - I)^2] = \dots = [X, (Y - I)^{n-1}] = 0$ .

Also,  $N = a_0 Y + \sum_{i=1}^{n-1} a_i (Y - I)^i$ , so

$$XN = a_0 XY + \sum_{i=1}^{n-1} a_i X(Y - I)^i = a_0 YX + \sum_{i=1}^{n-1} a_i (Y - I)^i X = NX$$

So,  $[X, N] = 0$ , and

$$e^{-N} e^X = I = e^X e^{-N} \Rightarrow e^{X-N} = I$$

Hence,

$$X - N = 2\pi k i I \Rightarrow X = N + 2\pi k i I, k \in \mathbb{Z}$$

4. Before searching for such a pair of  $(A, B)$ , the following lemma would be proved first:

$$\forall A : A^2 = -4\pi^2 I \Rightarrow e^A = I$$

Proof:

$$\begin{aligned} e^A &= \sum_{k=0}^{+\infty} \frac{A^{2k}}{(2k)!} + \sum_{k=0}^{+\infty} \frac{A^{2k+1}}{(2k+1)!} \\ &= \sum_{k=0}^{+\infty} \frac{(-1)^k (2\pi)^{2k}}{(2k)!} I + \frac{1}{2\pi} \sum_{k=0}^{+\infty} \frac{(-1)^k (2\pi)^{2k+1}}{(2k+1)!} A = \cos 2\pi I + \frac{1}{2\pi} \sin 2\pi A = I \end{aligned}$$

Let  $X = 2\pi \cdot \begin{bmatrix} 1 & 1 \\ -2 & -1 \end{bmatrix}$ ,  $Y = 2\pi \cdot \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$ . Clearly,

$$[X, Y] = 4\pi^2 \left( \begin{bmatrix} 1 & -1 \\ -1 & 2 \end{bmatrix} - \begin{bmatrix} 2 & 1 \\ 1 & 1 \end{bmatrix} \right) = 4\pi^2 \cdot \begin{bmatrix} -1 & -2 \\ -2 & 1 \end{bmatrix} \neq 0$$

, however  $e^X = I = e^Y$

5. Suppose  $X$  exists, then

$$\sin^2(X) + \cos^2(X) = \left( \frac{e^{iX} - e^{-iX}}{2i} \right)^2 + \left( \frac{e^{iX} + e^{-iX}}{2} \right)^2 = \frac{2I - e^{2iX} - e^{-2iX}}{4} + \frac{e^{2iX} + e^{-2iX} + 2I}{4} = I$$

Therefore,

$$\cos^2(X) = I - \sin^2(X) = \begin{bmatrix} 0 & -3992 \\ 0 & 0 \end{bmatrix}$$

This implies  $\cos^2(X)$  is nilpotent, so is  $\cos(X)$ . However, it means the entry of -3992 should actually be zero, which is a contradiction. ■

## Solution 5

1. Let  $\lambda_n$  be any one of the eigenvalues of  $X_n$ , where  $v$  is the eigenvector associates with it, i.e.  $X_n v = \lambda_n v$ .

Then, this immediately yields  $X_n^{-1} v = v/\lambda$ .

That means  $X_{n+1}$  and  $X_n$  both have the eigenvector  $v$ .

Since  $X_{n+1} = \frac{1}{2}(X_n + X_n^{-1})$ , if  $\lambda_{n+1}$  satisfies  $X_{n+1} v = \lambda_{n+1} v$ , then  $\lambda_{n+1} = \frac{1}{2}(\lambda_n + 1/\lambda_n)$ .

Let  $\lambda_n = x + iy$ ,  $x \neq 0$

$$\lambda_{n+1} = \frac{x(x^2 + y^2 + 1)}{2(x^2 + y^2)} + i \frac{y(x^2 + y^2 - 1)}{2(x^2 + y^2)}$$

So that implies  $\Re(\lambda_{n+1}) \neq 0$ . ■

2.

$$\frac{X_{n+1} + 1}{X_{n+1} - 1} = \frac{X_n + X_n^{-1} + 2}{X_n + X_n^{-1} - 2} = \frac{(X_n + 1)^2}{(X_n - 1)^2} = \frac{(A + 1)^{2^{n+1}}}{(A - 1)^{2^{n+1}}} \Rightarrow X_n = \frac{\frac{(A+1)^{2^n}}{(A-1)^{2^n}} + 1}{\frac{(A+1)^{2^n}}{(A-1)^{2^n}} - 1} = \frac{1 + \frac{(A-1)^{2^n}}{(A+1)^{2^n}}}{1 - \frac{(A-1)^{2^n}}{(A+1)^{2^n}}}$$

When  $\Re(A) > 0$ ,  $|A + 1| > |A - 1|$ ,

$$\lim_{n \rightarrow +\infty} X_n = \lim_{n \rightarrow +\infty} \frac{1 + \frac{(A-1)^{2^n}}{(A+1)^{2^n}}}{1 - \frac{(A-1)^{2^n}}{(A+1)^{2^n}}} = \frac{1 + 0}{1 - 0} = 1$$

When  $\Re(A) < 0$ ,  $|A + 1| < |A - 1|$ ,

$$\lim_{n \rightarrow +\infty} X_n = \lim_{n \rightarrow +\infty} \frac{\frac{(A+1)^{2^n}}{(A-1)^{2^n}} + 1}{\frac{(A+1)^{2^n}}{(A-1)^{2^n}} - 1} = \frac{0 + 1}{0 - 1} = -1$$

So in general, for one by one matrix,

$$\lim_{n \rightarrow +\infty} X_n = \text{sgn}(A)$$

3.  $A$  is diagonalizable, so is  $X_n$ .

$$X_n = B \Lambda B^{-1}$$

Then,

$$\lim_{t \rightarrow +\infty} X_t = B \lim_{t \rightarrow +\infty} \text{diag}(\lambda_{1t}, \dots, \lambda_{nt}) B^{-1}$$

When  $A = B \text{diag}(\lambda_{10}, \dots, \lambda_{n0}) B^{-1}$ ,

$$B \lim_{t \rightarrow +\infty} \text{diag}(\lambda_{1t}, \dots, \lambda_{nt}) B^{-1} = B \text{diag}(\text{sgn}(\lambda_{10}), \dots, \text{sgn}(\lambda_{n0})) B^{-1} = \text{sgn}(A)$$

4. Similar to (2), there is also a formula for  $X_n$ :

$$X_n = [I - (A - I)^{2^n} (A + I)^{-2^n}]^{-1} [I + (A - I)^{2^n} (A + I)^{-2^n}]$$

Here,  $A = I + N$ .

Moreover,  $2^{n-1} \geq n$ ,  $\forall n \in \mathbf{Z}^+$ , then

$$\begin{aligned} X_{n-1} &= [I - (I + N - I)^{2^{n-1}} (I + N + I)^{-2^{n-1}}]^{-1} [I + (I + N - I)^{2^{n-1}} (I + N + I)^{-2^{n-1}}] \\ &= [I - N^n (2I + N)^{-2^{n-1}}]^{-1} [I + N^n (2I + N)^{-2^{n-1}}] = I \end{aligned}$$

## Solution 6

Let  $A = \begin{bmatrix} a_1 & b_1 \\ c_1 & d_1 \end{bmatrix}$ ,  $B = \begin{bmatrix} a_2 & b_2 \\ c_2 & d_2 \end{bmatrix}$ .

1. On one hand,

$$AB = \begin{bmatrix} a_1 & b_1 \\ c_1 & d_1 \end{bmatrix} \begin{bmatrix} a_2 & b_2 \\ c_2 & d_2 \end{bmatrix} = \begin{bmatrix} a_1a_2 + b_1c_2 & a_1b_2 + b_1d_2 \\ c_1a_2 + d_1c_2 & c_1b_2 + d_1d_2 \end{bmatrix}$$

On the other hand,

$$f_A(f_B(x)) = \frac{a_1\left(\frac{a_2x+b_2}{c_2x+d_2}\right) + b_1}{c_1\left(\frac{a_2x+b_2}{c_2x+d_2}\right) + d_1} = \frac{a_1a_2x + a_1b_2 + b_1c_2x + b_1d_2}{c_1a_2x + c_1b_2 + d_1c_2x + d_1d_2} = \frac{(a_1a_2 + b_1c_2)x + (a_1b_2 + b_1d_2)}{(c_1a_2 + d_1c_2)x + (c_1b_2 + d_1d_2)}$$

Therefore  $f_A \circ f_B = f_{AB}$

2. If  $k \neq 0$ , then

$$f_{kA}(x) = \frac{(ka_1)x + (kb_1)}{(kc_1)x + (kd_1)} = \frac{a_1x + b_1}{c_1x + d_1} = f_A(x)$$

Conversely, when  $f_A(x) \equiv f_B(x)$

$$\frac{a_1x + b_1}{c_1x + d_1} \equiv \frac{a_2x + b_2}{c_2x + d_2} \Leftrightarrow a_1c_2x^2 + (a_1d_2 + b_1c_2)x + b_1d_2 \equiv a_2c_1x^2 + (b_2c_1 + d_1a_2)x + b_2d_1$$

As  $x$  is an arbitrary value, let  $k_1 = a_1c_2 = a_2c_1$ ,  $k_2 = b_1d_2 = b_2d_1$ , then

$$a_2 = k_1/c_1, b_2 = k_2/d_1, c_2 = k_1/a_1, d_2 = k_2/b_1$$

$$a_1d_2 + b_2c_2 = d_1a_2 + b_2c_1 \Rightarrow \frac{a_1}{b_1}k_2 + \frac{b_1}{a_1}k_1 = \frac{d_1}{c_1}k_1 + \frac{c_1}{d_1}k_2 \Rightarrow \frac{k_1}{k_2} = \frac{a_1c_1}{b_1d_1}$$

With all these,

$$\frac{a_1c_1}{b_1d_1} = \frac{a_1c_2}{b_1d_2} \Rightarrow \frac{c_1}{c_2} = \frac{d_1}{d_2}, \frac{a_1c_1}{b_1d_1} = \frac{a_1c_2}{b_2d_1} \Rightarrow \frac{b_1}{b_2} = \frac{c_1}{c_2}, \frac{a_1c_1}{b_1d_1} = \frac{a_2c_1}{b_2d_1} \Rightarrow \frac{a_1}{a_2} = \frac{b_1}{b_2}$$

Hence

$$\frac{a_1}{a_2} = \frac{b_1}{b_2} = \frac{c_1}{c_2} = \frac{d_1}{d_2} \Leftrightarrow A = kB$$

for some constant  $k$ .

3.

$$A \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} a_1 & b_1 \\ c_1 & d_1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} a_1x + b_1y \\ c_1x + d_1y \end{bmatrix}$$

is interpreted as  $\frac{a_1x+b_1y}{c_1x+d_1y}$ , while

$$f_A\left(\frac{x}{y}\right) = \frac{a_1(x/y) + b_1}{c_1(x/y) + d_1} = \frac{a_1x + b_1y}{c_1x + d_1y}$$

So they are equal.

4. Let  $x$  be the fixed point of transformation  $f_A$ .

If  $x \in \mathbf{R}$ , then

$$\frac{ax+b}{cx+d} = x \Leftrightarrow cx^2 - (a-d)x - b = 0$$

So excluding  $\infty$ ,  $f_A$  may have every points fixed, two points fixed, one point fixed or none fixed.

$$\text{Every } x \in \mathbf{R} \text{ is fixed} \Leftrightarrow cx^2 - (a-d)x - b \equiv 0 \Leftrightarrow \begin{cases} b = c = 0 \\ a = d \neq 0 \end{cases}$$

$f_A$  has two fixed points  $\Leftrightarrow c \neq 0 \wedge (a-d)^2 + 4bc > 0$

$f_A$  has one single fixed point  $\Leftrightarrow (c \neq 0 \wedge (a-d)^2 + 4bc = 0) \vee (c = 0 \wedge a \neq d)$

$f_A$  has no fixed point  $\Leftrightarrow (c \neq 0 \wedge (a-d)^2 + 4bc < 0) \vee (c = 0 \wedge a = d \wedge b \neq 0)$

If  $x = \infty$ , then

$$f_A(x) = x \Leftrightarrow \frac{c + d/x}{a + b/x} = 0 \Leftrightarrow c = 0$$



In conclusion, these are the possibilities of the fixed points of  $f_A$ :

number of fixed points	conditions
$\infty$	$\begin{cases} b = c = 0 \\ a = d \neq 0 \end{cases}$
2	$\begin{cases} (a - d)^2 + 4bc > 0, & c \neq 0 \\ a \neq d, & c = 0 \end{cases}$
1	$\begin{cases} (a - d)^2 + 4bc = 0, & c \neq 0 \\ a = d \wedge b \neq 0, & c = 0 \end{cases}$
0	$(a - d)^2 + 4bc < 0 \wedge c \neq 0$