

Chapter 1

Systems of Linear Equations

1.1 HW1 Complex Stuff

Exercise 1.1.1.

1. For $n = 2$, we have

$$A_2 = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix},$$

and for each even $n > 2$, we have

$$A_n = \begin{bmatrix} A_2 & & & \\ & A_2 & & \\ & & \ddots & \\ & & & A_2 \end{bmatrix},$$

such that $A_n^2 = -I$.

2. If $A^2 = -I$, then A^2 has the only eigenvalue -1 (with a multiplicity of n). On the other hand, if n is odd and A is real, then the n -degree real-coefficient eigenpolynomial of A does have at least one real root. Therefore A has at least one real eigenvalue, whose square is -1 , which is impossible.

Exercise 1.1.2.

1. For any $k = a + bi \in \mathbb{C}$ where $a, b \in \mathbb{R}$ and $\mathbf{v} \in \mathbb{R}^n$,

$$\begin{aligned} B(k\mathbf{v}) &= B((a + bi)\mathbf{v}) = B(a\mathbf{v} + bi\mathbf{v}) = B(a\mathbf{v} + bA\mathbf{v}) = aB\mathbf{v} + bBA\mathbf{v}, \\ kB\mathbf{v} &= (a + bi)(B\mathbf{v}) = a(B\mathbf{v}) + bi(B\mathbf{v}) = aB\mathbf{v} + bAB\mathbf{v}. \end{aligned}$$

Therefore, $B(k\mathbf{v}) = kB\mathbf{v}$ if and only if $AB = BA$.

2. No. For example, let

$$A = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}, \quad X = \begin{bmatrix} 1 & 2 \\ -1 & -1 \end{bmatrix}.$$

Then $A^2 = X^2 = -I$, but

$$\begin{aligned} AX &= \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ -1 & -1 \end{bmatrix} = \begin{bmatrix} -1 & -1 \\ -1 & -2 \end{bmatrix} \\ \neq XA &= \begin{bmatrix} 1 & 2 \\ -1 & -1 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix} = \begin{bmatrix} -2 & 1 \\ 1 & -1 \end{bmatrix}. \end{aligned}$$

3. Since $C^2 = I$, C have n complex eigenvalues whose square are 1. Hence the eigenvalues must be 1 or -1 , and because they are all real, C is diagonalizable. Suppose the eigenspaces for 1 and -1 are F and G respectively. For each $\mathbf{v} \in F$, since $CA = -AC$, we have

$$C(A\mathbf{v}) = (CA)\mathbf{v} = (-AC)\mathbf{v} = -A(C\mathbf{v}) = -A\mathbf{v},$$

i.e. $A\mathbf{v}$ is in G . Note that A is invertible ($A^{-1} = -A$), so A gives a bijection between F and G , and therefore they have the same dimension.

4. An example is

$$A = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix},$$

and

$$C = \begin{bmatrix} -1 & 0 \\ 0 & 1 \end{bmatrix} \text{ or } \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}.$$

Exercise 1.1.3.

1. It is real linear but not complex linear, as for each element c in any vector, only if k is real do we have $\overline{kc} = k\overline{c}$.
2. \mathbb{C} -linear implies \mathbb{R} -linear, because if $L(k\mathbf{v}) = kL\mathbf{v}$ for all $k \in \mathbb{C}$ and $\mathbf{v} \in V$, then this is also true for any $k \in \mathbb{R}$, since a real number is as well a complex number.
3. An \mathbb{R} -basis for \mathbb{C}^2 is

$$\left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} i \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ i \end{bmatrix} \right\}.$$

It is 4-real-dimension. And a \mathbb{C} -basis for \mathbb{C}^2 is

$$\left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right\}.$$

It is 2-complex-dimension.

4. \mathbb{C} -linearly independent implies \mathbb{R} -linearly independent, because if a bunch of vectors $\{\mathbf{v}_i\}$ are not \mathbb{R} -linearly independent, then we have a set of real coefficient $\{a_i\}$ all of which are not zero, such that $\sum a_i \mathbf{v}_i = \mathbf{0}$. And this is also true for complex $\{a_i\}$, since a real number is as well a complex number, and therefore $\{\mathbf{v}_i\}$ are not \mathbb{C} -linearly independent.
5. \mathbb{R} -spanning implies \mathbb{C} -spanning, because for a bunch of vectors $\{\mathbf{v}_i\}$, if any $\mathbf{v} \in V$ can be represented as $\mathbf{v} = \sum a_i \mathbf{v}_i$ where $\{a_i\}$ are all real, then this is as well true for complex $\{a_i\}$, since a real number is as well a complex number.

Exercise 1.1.4.

- 1.

$$P \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}, \quad P \begin{bmatrix} 1 \\ i \\ -1 \\ -i \end{bmatrix} = \begin{bmatrix} i \\ -1 \\ -i \\ 1 \end{bmatrix}.$$

2.

$$\begin{aligned}
PF_4 &= P \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & i & -1 & -i \\ 1 & -1 & 1 & -1 \\ 1 & -i & -1 & i \end{bmatrix} \\
&= \begin{bmatrix} 1 & i & -1 & -i \\ 1 & -1 & 1 & -1 \\ 1 & -i & -1 & i \\ 1 & 1 & 1 & 1 \end{bmatrix} \\
&= \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & i & -1 & -i \\ 1 & -1 & 1 & -1 \\ 1 & -i & -1 & i \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & i & 0 & 0 \\ 0 & 0 & -1 & 0 \\ 0 & 0 & 0 & -i \end{bmatrix} \\
&= F_4 D.
\end{aligned}$$

Therefore, the four eigenvectors and eigenvalues of P are the columns of F_4 and the diagonal entries of D respectively.

3.

$$\begin{aligned}
C \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} &= \begin{bmatrix} c_0 + c_1 + c_2 + c_3 \\ c_3 + c_0 + c_1 + c_2 \\ c_2 + c_3 + c_0 + c_1 \\ c_1 + c_2 + c_3 + c_0 \end{bmatrix}, \\
C \begin{bmatrix} 1 \\ i \\ -1 \\ -i \end{bmatrix} &= \begin{bmatrix} c_0 + c_1 i + c_2(-1) + c_3(-i) \\ c_3 + c_0 i + c_1(-1) + c_2(-i) \\ c_2 + c_3 i + c_0(-1) + c_1(-i) \\ c_1 + c_2 i + c_3(-1) + c_0(-i) \end{bmatrix} = \begin{bmatrix} c_0 - c_2 + (c_1 - c_3)i \\ c_3 - c_1 + (c_0 - c_2)i \\ c_2 - c_0 + (c_3 - c_1)i \\ c_1 - c_3 + (c_2 - c_0)i \end{bmatrix}.
\end{aligned}$$

4. Obviously $C = c_0 P^0 + c_1 P^1 + c_2 P^2 + c_3 P^3 = \sum_n c_n P^n$. For the i -th column of F (also a eigenvector of P , with eigenvalue λ_i) \mathbf{F}_i , we have

$$P^n \mathbf{F}_i = \lambda_i^n \mathbf{F}_i.$$

Note that $\mathbf{F}_i = (\lambda_i^0, \lambda_i^1, \lambda_i^2, \lambda_i^3)$, therefore

$$C \mathbf{F}_i = \sum_n c_n P^n \mathbf{F}_i = \sum_n c_n \lambda_i^n \mathbf{F}_i = (\mathbf{c} \cdot \mathbf{F}_i) \mathbf{F}_i,$$

where $\mathbf{c} = (c_0, c_1, c_2, c_3)$. Therefore the eigenvectors and the eigenvalues are \mathbf{F}_i and $(\mathbf{c} \cdot \mathbf{F}_i)$ respectively.