

MATH 4063-5023
Homework Set 7

1. Find the characteristic polynomials and minimal polynomials of the following matrices.

(a) $\begin{pmatrix} 2 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{pmatrix}$

- The characteristic polynomial of this matrix is

$$p_{\mathbf{A}}(\lambda) = \det \begin{pmatrix} 2-\lambda & 0 & 0 \\ 0 & 2-\lambda & 0 \\ 0 & 0 & 2-\lambda \end{pmatrix} = (2-\lambda)^3$$

The minimal polynomial will always be a factor of the characteristic polynomial; in fact, it will be the factor $m_{\mathbf{A}}(x)$ of smallest degree such $m_{\mathbf{A}}(\mathbf{A}) = \mathbf{0}$. The possible factors of $p_{\mathbf{A}}(\lambda)$ are $(2-\lambda)$, $(2-\lambda)^2$, $(2-\lambda)^3$. We have

$$(2 \cdot \mathbf{I} - \lambda)|_{\lambda=\mathbf{A}} = \begin{pmatrix} 2 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{pmatrix} - \begin{pmatrix} 2 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} = \mathbf{0}$$

so in fact the factor of lowest degree is the minimal polynomial. Thus

$$m_{\mathbf{A}}(x) = 2 - x$$

(b) $\begin{pmatrix} 2 & 1 & 1 \\ 0 & 2 & 1 \\ 0 & 0 & 2 \end{pmatrix}$

- The characteristic polynomial is again $p_{\mathbf{A}}(x) = (2-x)^3$, and we have the same initial possibilities for the minimal polynomial. We evaluate each of these possibilities at $x = \mathbf{A}$

$$(2 \cdot 1 - x)|_{x=\mathbf{A}} = \begin{pmatrix} 0 & 1 & 1 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix} \neq \mathbf{0}$$

$$(2 \cdot 1 - x)^2|_{x=\mathbf{A}} = \begin{pmatrix} 0 & 1 & 1 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} 0 & 1 & 1 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} \neq \mathbf{0}$$

$$(2 \cdot 1 - x)^3|_{x=\mathbf{A}} = \begin{pmatrix} 0 & 1 & 1 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} 0 & 1 & 1 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} 0 & 1 & 1 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} = \mathbf{0}$$

So the minimal polynomial is $(2-x)^3$

2. Find the eigenvalues of the following matrices, and then for each eigenvalue, find a basis for the corresponding eigenspace, and state the algebraic and geometric multiplicity of eigenvalue.

(a) $\begin{pmatrix} 1 & 2 \\ 0 & 3 \end{pmatrix}$

- First we find the eigenvalues. This is done by finding the roots of the characteristic polynomial:

$$0 = \det \begin{pmatrix} 1-\lambda & 2 \\ 0 & 3-\lambda \end{pmatrix} = (1-\lambda)(3-\lambda) \Rightarrow \lambda = 1, 3$$

To find the eigenvectors corresponding to an eigenvalue $\lambda = r$, we solve the homogeneous linear system $(\mathbf{A} - r\mathbf{I})\mathbf{x} = \mathbf{0}$. Any basis for this solution space will provide us with a set of linearly independent eigenvectors spanning the corresponding eigenspace.

$\lambda = 1$:

$$(\mathbf{A} - (1)\mathbf{I})\mathbf{x} = \mathbf{0} \iff \begin{pmatrix} 1-1 & 2 \\ 0 & 3-1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

or

$$\begin{pmatrix} 0 & 2 \\ 0 & 2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

The coefficient matrix for this homogeneous linear system row reduces to the following matrix in Reduced Row Echelon Form

$$\begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \Rightarrow \text{solution is } x_2 = 0$$

The component x_1 of the solution vector is left as a free parameter. Thus,

$$\mathbf{x} = \begin{pmatrix} x_1 \\ 0 \end{pmatrix} = x_1 \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

Thus, $[1, 0]$ will be a basis for the solution space, hence a basis for the $\lambda = 1$ eigenspace.

$\lambda = 3$:

$$(\mathbf{A} - (3)\mathbf{I})\mathbf{x} = \mathbf{0} \iff \begin{pmatrix} 1-3 & 2 \\ 0 & 3-3 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} -2 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

It's pretty clear that the solution of this homogeneous linear system will have $x_1 = 0$, while x_2 will be a free parameter. Thus,

$$\mathbf{x} = \begin{pmatrix} 0 \\ x_2 \end{pmatrix} = x_2 \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

and so $[0, 1]$ will be a basis for the solution space, and so also a basis for the $\lambda = 3$ eigenspace.

(b) $\begin{pmatrix} 2 & 0 & 0 \\ 2 & 1 & 0 \\ 2 & 2 & 2 \end{pmatrix},$

- In this (and subsequent) parts, I'll just summarize the calculations following the steps used in part (a).

$$\det \begin{pmatrix} 2-\lambda & 0 & 0 \\ 2 & 1-\lambda & 0 \\ 2 & 2 & 2-\lambda \end{pmatrix} = (1-\lambda)(2-\lambda)^2 \Rightarrow \lambda = 1, 2$$

$\lambda = 1$:

$$(\mathbf{A} - (1)\mathbf{I}) = \begin{pmatrix} 1 & 0 & 0 \\ 2 & 0 & 0 \\ 2 & 2 & 1 \end{pmatrix} \xrightarrow{\text{row reduction}} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & \frac{1}{2} \\ 0 & 0 & 0 \end{pmatrix} \xrightarrow{\text{solutions}} \begin{cases} x_1 = 0 \\ x_2 = -\frac{1}{2}x_3 \end{cases}$$

$$\xrightarrow{\text{solution vector}} \mathbf{x} = x_3 \begin{pmatrix} 0 \\ -\frac{1}{2} \\ 1 \end{pmatrix} \xrightarrow{\text{basis for } \lambda = 1 \text{ eigenspace}} \begin{pmatrix} 0 \\ -\frac{1}{2} \\ 1 \end{pmatrix}$$

$\lambda = 2$:

$$(\mathbf{A} - (2)\mathbf{I}) = \begin{pmatrix} 0 & 0 & 0 \\ 2 & -1 & 0 \\ 2 & 2 & 0 \end{pmatrix} \xrightarrow{\text{row reduction}} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix} \xrightarrow{\text{solutions}} \begin{cases} x_1 = 0 \\ x_2 = 0 \end{cases}$$

$$\xrightarrow{\text{solution vector}} \mathbf{x} = x_3 \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \xrightarrow{\text{basis for } \lambda = 2 \text{ eigenspace}} \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$$

$$(c) \begin{pmatrix} -3 & 0 & 4 \\ 0 & -1 & 0 \\ -2 & 7 & 3 \end{pmatrix}$$

•

$$\det(\mathbf{A} - \lambda\mathbf{I}) = \det \begin{pmatrix} -3 - \lambda & 0 & 4 \\ 0 & -1 - \lambda & 0 \\ -2 & 7 & 3 - \lambda \end{pmatrix} = -\lambda^3 - \lambda^2 + \lambda + 1 = -(\lambda - 1)(\lambda + 1)^2 \Rightarrow \lambda = 1, -1$$

$$\lambda = 1 :$$

$$(\mathbf{A} - (1)\mathbf{I}) = \begin{pmatrix} -4 & 0 & 4 \\ 0 & -2 & 0 \\ -2 & 7 & 2 \end{pmatrix} \xrightarrow{\text{row reduction}} \begin{pmatrix} 1 & 0 & -1 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix} \xrightarrow{\text{solutions}} \begin{cases} x_1 = x_3 \\ x_2 = 0 \end{cases}$$

$$\xrightarrow{\text{solution vector}} \mathbf{x} = x_3 \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} \xrightarrow{\text{basis for } \lambda = 1 \text{ eigenspace}} \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}$$

$$\lambda = -1 :$$

$$(\mathbf{A} - (-1)\mathbf{I}) = \begin{pmatrix} -2 & 0 & 4 \\ 0 & 0 & 0 \\ -2 & 7 & 4 \end{pmatrix} \xrightarrow{\text{row reduction}} \begin{pmatrix} 1 & 0 & -2 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix} \xrightarrow{\text{solutions}} \begin{cases} x_1 = 2x_3 \\ x_2 = 0 \end{cases}$$

$$\xrightarrow{\text{solution vector}} \mathbf{x} = x_3 \begin{pmatrix} 2 \\ 0 \\ 1 \end{pmatrix} \xrightarrow{\text{basis for } \lambda = -1 \text{ eigenspace}} \begin{pmatrix} 2 \\ 0 \\ 1 \end{pmatrix}$$

$$(d) \begin{pmatrix} 1 & 2 \\ -1 & 3 \end{pmatrix} \text{ (you'll have to work over } \mathbb{C} \text{ for this one.)}$$

• We proceed as before, except we allow complex numbers to creep into our solutions.

$$\det(\mathbf{A} - \lambda\mathbf{I}) = \det \begin{pmatrix} 1 - \lambda & 2 \\ -1 & 3 - \lambda \end{pmatrix} = \lambda^2 - 4\lambda + 5 = (\lambda - 2 + i)(\lambda - 2 - i) \Rightarrow \lambda = 2 + i, 2 - i$$

$$\lambda = 2 + i$$

$$(\mathbf{A} - (2 + i)\mathbf{I}) = \begin{pmatrix} -1 - i & 2 \\ -1 & 1 - i \end{pmatrix} \xrightarrow{R_2 \rightarrow (1 + i)R_2} \begin{pmatrix} -1 - i & 2 \\ -1 - i & 2 \end{pmatrix} \xrightarrow{R_2 \rightarrow R_2 - R_1} \begin{pmatrix} -1 - i & 2 \\ 0 & 0 \end{pmatrix}$$

$$\Rightarrow x_1 = \frac{2}{1 + i}x_2 = \frac{2(1 - i)}{(1 + i)(1 - i)}x_2 = (1 - i)x_2$$

$$\Rightarrow \mathbf{x} = x_2 \begin{pmatrix} 1 - i \\ 1 \end{pmatrix} \Rightarrow \text{basis for } \lambda = 2 + i \text{ eigenspace is } \left\{ \begin{bmatrix} 1 - i \\ 1 \end{bmatrix} \right\}$$

$$\lambda = 2 - i$$

$$(\mathbf{A} - (2 - i)\mathbf{I}) = \begin{pmatrix} -1 + i & 2 \\ -1 & 1 + i \end{pmatrix} \xrightarrow{R_2 \rightarrow (1 - i)R_2} \begin{pmatrix} -1 + i & 2 \\ -1 + i & 2 \end{pmatrix} \xrightarrow{R_2 \rightarrow R_2 - R_1} \begin{pmatrix} -1 + i & 2 \\ 0 & 0 \end{pmatrix}$$

$$\Rightarrow x_1 = \frac{2}{1 - i}x_2 = \frac{2(1 + i)}{(1 - i)(1 + i)}x_2 = (1 + i)x_2$$

$$\Rightarrow \mathbf{x} = x_2 \begin{pmatrix} 1 + i \\ 1 \end{pmatrix} \Rightarrow \text{basis for } \lambda = 2 - i \text{ eigenspace is } \left\{ \begin{bmatrix} 1 + i \\ 1 \end{bmatrix} \right\}$$

3. Let V be a vector space with basis $\{v_1, \dots, v_m\}$. Find a basis for $L(V, V)$.

• Let for each $i, j \in \{1, \dots, m\}$ let $T_{ij} : V \rightarrow V$ be the linear transformation defined by

$$T_{ij}(a_1v_1 + \dots + a_mv_m) = a_jv_i$$

This gives us m^2 different linear transformations. To demonstrate that this set of n^2 linear transformations actually constitutes a basis for $L(V, V)$, it suffices to show that a linear combination of them can produce the zero transformation only if all coefficients are equal to $0_{\mathbb{F}}$. Let

$$S = \sum_{i,j=1}^m b_{ij} T_{ij}$$

If S were the zero transformation it would have to vanish on each basis vector v_k . But

$$0 = S(v_k) = \sum_{i,j=1}^m b_{ij} T_{ij}(v_k) = \sum_{i=1}^m b_{ik} v_i \quad \Rightarrow \quad b_{ik} = 0 \quad , \quad k = 1, \dots, m$$

since the basis vectors v_i are linearly independent. Thus, a linear combination of the linear transformations T_{ij} can yield the zero transformation only if we take all the coefficients equal to zero. Hence, the T_{ij} are all linearly independent. Since there are m^2 of them (which is the dimension of $L(V, V)$), they comprise a basis for $L(V, V)$.

4. Suppose V is an n -dimensional vector space and $T : V \rightarrow V$ is an endomorphism of V with n linearly independent eigenvectors v_1, \dots, v_n with eigenvalues ξ_1, \dots, ξ_n . Set

$$f(x) = \prod_{i=1}^n (x - \xi_i)$$

(a) Show $f(T) = 0$.

- Because V is n -dimensional and the eigenvectors v_1, \dots, v_n are stipulated to be linearly independent, the vectors v_1, \dots, v_n will comprise a basis for V . Thus, any $v \in V$ can be uniquely written as

$$(1) \quad v = a_1 v_1 + \dots + a_n v_n \quad , \quad a_1, \dots, a_n \in \mathbb{F} \quad .$$

Now consider the operators

$$S_i = (x - \xi_i)|_{x=T} = T - \xi_i \mathbf{1}.$$

We have

$$(2) \quad S_i S_j = (T - \xi_i \mathbf{1})(T - \xi_j \mathbf{1}) = T^2 - \xi_i T - \xi_j T + \xi_i \xi_j \mathbf{1} = (T - \xi_j \mathbf{1})(T - \xi_i \mathbf{1}) = S_j S_i$$

and so these operators all commute. Also, when S_i acts on the i^{th} eigenvector we have

$$(3) \quad S_i v_i = (T - \xi_i \mathbf{1}) v_i = T(v_i) - \xi_i v_i = \xi_i v_i - \xi_i v_i = \mathbf{0}_V \quad .$$

Using (3) and (2) we then have

$$f(T) v_i = \prod_{j=1}^n (T - \xi_j \mathbf{1}) v_i = S_1 \cdots S_i \cdots S_n v_i = S_1 \cdots S_{i-1} S_{i+1} \cdots S_n S_i v_i = S_1 \cdots S_{i-1} S_{i+1} \cdots S_n \mathbf{0}_V = \mathbf{0}_V$$

So $f(T)$ vanishes on each basis vector v_i . But we can express every vector v in the form (1), we have

$$f(T) v = f(T) (a_1 v_1 + \dots + a_n v_n) = a_1 f(T) v_1 + \dots + a_n f(T) v_n = \mathbf{0}_V + \dots + \mathbf{0}_V = \mathbf{0}_V \quad .$$

□

(b) Show the minimal polynomial of T is the product of the distinct factors $(x - \xi_i)$ of f .

- Notice that if $\xi_i = \xi_j$, one doesn't need both factors S_i and S_j to produce an operator that vanishes on both v_i and v_j via the argument of equation (3) in part (a). Therefore, if we use only the S_i corresponding to distinct eigenvalues, we'll still have an operator that is a polynomial in T

and vanishes on all of V . What we need to show is that the operator of lower degree. But we also know that the minimal polynomial has to divide

$$\tilde{f}(x) = \prod_{\xi_i \neq \xi_j} (x - \xi_i)$$

since $f(T) = 0$ on V . But $\tilde{f}(x)$ is already factored into irreducible polynomials, so any polynomial that divides $\tilde{f}(x)$ must be a product of the same irreducibles, except perhaps missing some factors. However, if you remove a factor, say $(x - \xi_i)$ from $\tilde{f}(x)$, you end up with a polynomial whose corresponding operator no longer vanishes on the ξ_i -eigenspace:

$$(T - \xi_1) \cdots (T - \xi_{i-1}) (T - \xi_{i+1}) \cdots (T - \xi_n) v_i = (\xi_i - \xi_1) \cdots (\xi_i - \xi_{i-1}) (\xi_i - \xi_{i+1}) \cdots (\xi_i - \xi_n) v_i \neq 0 \quad \text{because } \xi_i \neq \xi_j$$

Thus, $\tilde{f}(x)$ is the minimal polynomial. \square

5. Show that $T \in L(V, V)$ is invertible if and only if the constant term of the minimal polynomial is not equal to zero. Come up with an algorithm for computing T^{-1} from its minimal polynomial.

- Write

$$m_T(x) = a_0 + a_1x + \cdots + a_{n-1}x^{n-1} + x^n$$

for the minimal polynomial. By definition, $m_T(x)$ is the monic polynomial of lowest degree such that

$$m_T(T)v = 0 \quad \forall v \in V \quad .$$

\Rightarrow Suppose T is invertible. Then the null space of T is $\{\mathbf{0}_V\}$ and the range of T is the entire space V . On the other hand, we have

$$(1) \quad \mathbf{0}_V = m_T(T)v \Rightarrow a_0v = -(a_1 + a_2T + \cdots + a_nT^{n-1})Tv$$

Now if the constant term a_0 equals $0_{\mathbb{F}}$, then we have

$$-(a_1 + a_2T + \cdots + a_nT^{n-1})T(v) = \mathbf{0}_V \quad \forall v \in V$$

But since the range of T is all of V , this requires

$$-(a_1 + a_2T + \cdots + a_nT^{n-1})v = 0 \quad \forall v \in V \quad .$$

But this can't be, since $-(a_1 + a_2T + \cdots + a_nT^{n-1})$ is a polynomial in T of lower degree than the minimal polynomials. So we can conclude that T being invertible requires the constant term of the minimal polynomial for T to be non-zero.

\Leftarrow Suppose $a_0 \neq 0$. We want to show that T is invertible. Equation (1) still holds, and can be rewritten as

$$v = \frac{1}{a_0} (a_1 + a_2T + \cdots + a_nT^{n-1})Tv \quad \forall v \in V$$

So

$$\frac{1}{a_0} (a_1 + a_2T + \cdots + a_nT^{n-1})T = \mathbf{I}_{L(V,V)}$$

and so

$$T^{-1} = \frac{1}{a_0} (a_1 + a_2T + \cdots + a_nT^{n-1}) \quad .$$

Hence, T is invertible.

6. Two matrices \mathbf{A} and \mathbf{B} are said to be *similar* if there exists an invertible matrix \mathbf{C} such that $\mathbf{B} = \mathbf{C}^{-1}\mathbf{A}\mathbf{C}$. Show that similar matrices have the same eigenvalues.

- The eigenvalues a matrix \mathbf{A} are the roots of the equation $\det(\mathbf{A} - x\mathbf{I}) = 0$. Suppose $\mathbf{B} = \mathbf{C}^{-1}\mathbf{A}\mathbf{C}$. Then the eigenvalues of \mathbf{B} will be the roots of

$$\begin{aligned}
 0 &= \det(\mathbf{C}^{-1}\mathbf{A}\mathbf{C} - x\mathbf{I}) \\
 &= \det(\mathbf{C}^{-1}\mathbf{A}\mathbf{C} - x\mathbf{C}^{-1}\mathbf{C}) \\
 &= \det(\mathbf{C}^{-1}(\mathbf{A} - x\mathbf{I})\mathbf{C}) \\
 &= \det(\mathbf{C}^{-1}) \det(\mathbf{A} - x\mathbf{I}) \det(\mathbf{C}) \\
 &= \frac{1}{\det(\mathbf{C})} \det(\mathbf{A} - x\mathbf{I}) \det(\mathbf{C}) \\
 &= \det(\mathbf{A} - x\mathbf{I})
 \end{aligned}$$

Since \mathbf{A} and \mathbf{B} have the same characteristic polynomial, their eigenvalues must coincide.

7. Suppose $T \in L(V, V)$ has $n = \dim V$ distinct eigenvalues. Show that there exists a basis of V consisting of eigenvectors of T . What will be the matrix of T with respect to this basis?

- The eigenvectors for distinct eigenvalues are linearly independent (see Theorem 16.11 of Lecture 16), the fact that we have as many distinct eigenvectors as the dimension of V implies that the corresponding eigenvectors will provide a basis $B = \{v_1, \dots, v_n\}$ for V . Let's write down the matrix T_{BB} of T corresponding to adapting the basis B for both the domain and codomain of the linear transformation $T : V \rightarrow V$. We have

$$T(v_i) = \lambda_i v_i \quad \Rightarrow \quad \text{the coordinate vector of } T(v_i) \text{ with respect to } B \text{ is } [0, 0, \dots, 0, \lambda_i, 0, \dots, 0]$$

where the non-zero entry occurs precisely in the i^{th} slot. Thus, the matrix \mathbf{T}_{BB} corresponding to the linear transformation T and and basis B will have the form

$$\mathbf{T}_{BB} = \begin{pmatrix} \lambda_1 & 0 & \cdots & 0 & 0 \\ 0 & \lambda_2 & & \vdots & 0 \\ \vdots & & \ddots & & \vdots \\ 0 & \cdots & 0 & \lambda_{n-1} & 0 \\ 0 & \cdots & \cdots & 0 & \lambda_n \end{pmatrix}$$

that is to say, \mathbf{T}_{BB} will be a diagonal matrix whose diagonal entries are just the eigenvalues of T . \square