EECS 16A Designing Information Devices and Systems I Fall 2020 Discussion 5B

1. Steady and Unsteady States

(a) You're given the matrix M:

$$\mathbf{M} = \begin{bmatrix} \frac{1}{2} & \frac{1}{2} & -\frac{1}{2} \\ 0 & 1 & -2 \\ 0 & 0 & 2 \end{bmatrix}$$

Which generates the next state of a physical system from its previous state: $\vec{x}[k+1] = \mathbf{M}\vec{x}[k]$. (\vec{x} could describe either people or water.) Find the eigenspaces associated with the following eigenvalues:

- i. span(\vec{v}_1), associated with $\lambda_1 = 1$
- ii. span(\vec{v}_2), associated with $\lambda_2 = 2$
- iii. span(\vec{v}_3), associated with $\lambda_3 = \frac{1}{2}$

Answer:

i.
$$\lambda = 1$$
:

$$\begin{bmatrix} \mathbf{M} - \mathbf{I} & \vec{0} \end{bmatrix} = \begin{bmatrix} -\frac{1}{2} & \frac{1}{2} & -\frac{1}{2} & 0 \\ 0 & 0 & -2 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \stackrel{G.E.}{\to} \begin{bmatrix} 1 & -1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$
$$\vec{v}_1 = \alpha \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}, \alpha \in \mathbb{R}$$

ii. $\lambda = 2$

$$\begin{bmatrix} \mathbf{M} - 2\mathbf{I} & \vec{0} \end{bmatrix} = \begin{bmatrix} \frac{-3}{2} & \frac{1}{2} & -\frac{1}{2} & 0 \\ 0 & -1 & -2 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \stackrel{G.E.}{\to} \begin{bmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 2 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$
$$\vec{v}_2 = \beta \begin{bmatrix} -1 \\ -2 \\ 1 \end{bmatrix}, \beta \in \mathbb{R}$$

iii. $\lambda = \frac{1}{2}$

$$\begin{bmatrix} \mathbf{M} - \frac{1}{2}\mathbf{I} & \vec{0} \end{bmatrix} = \begin{bmatrix} 0 & \frac{1}{2} & -\frac{1}{2} & 0 \\ 0 & \frac{1}{2} & -2 & 0 \\ 0 & 0 & \frac{3}{2} & 0 \end{bmatrix} \stackrel{G.E.}{\to} \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$
$$\vec{v}_{3} = \gamma \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \gamma \in \mathbb{R}$$

(b) Define $\vec{x} = \alpha \vec{v}_1 + \beta \vec{v}_2 + \gamma \vec{v}_3$, a linear combination of the eigenvectors. For each of the cases in the table, determine if

 $\lim_{n\to\infty}\mathbf{M}^n\vec{x}$

converges. If it does, what does it converge to?

α	β	γ	Converges?	$\lim_{n\to\infty}\mathbf{M}^n\vec{x}$
0	0	$\neq 0$		
0	$\neq 0$	0		
0	$\neq 0$	$\neq 0$		
$\neq 0$	0	0		
$\neq 0$	0	$\neq 0$		
$\neq 0$	$\neq 0$	0		
$\neq 0$	$\neq 0$	$\neq 0$		

Answer:

 $\mathbf{M}^{n}\vec{x} = \mathbf{M}^{n}(\alpha\vec{v}_{1} + \beta\vec{v}_{2} + \gamma\vec{v}_{3})$ = $\alpha\mathbf{M}^{n}\vec{v}_{1} + \beta\mathbf{M}^{n}\vec{v}_{2} + \gamma\mathbf{M}^{n}\vec{v}_{3}$ = $1^{n}\alpha\vec{v}_{1} + 2^{n}\beta\vec{v}_{2} + \left(\frac{1}{2}\right)^{n}\gamma\vec{v}_{3}$

α	β	γ	Converges?	$\lim_{n\to\infty}\mathbf{M}^n\vec{x}$
0	0	$\neq 0$	Yes	Ō
0	$\neq 0$	0	No	-
0	$\neq 0$	eq 0	No	-
$\neq 0$	0	0	Yes	$\alpha \vec{v}_1$
$\neq 0$	0	$\neq 0$	Yes	$\alpha \vec{v}_1$
$\neq 0$	$\neq 0$	0	No	-
$\neq 0$	$\neq 0$	$\neq 0$	No	-

2. Eigenvalues and Special Matrices – Visualization

As seen earlier, an eigenvector \vec{v} belonging to a square matrix **A** is a nonzero vector that satisfies

$$\mathbf{A}\vec{v} = \lambda\vec{v}$$

where λ is a scalar known as the **eigenvalue** corresponding to eigenvector \vec{v} . Rather than mechanically compute the eigenvalues and eigenvectors, answer each part here by reasoning about the matrix at hand.

 (a) Does the identity matrix in ℝⁿ have any eigenvalues λ ∈ ℝ? What are the corresponding eigenvectors?
Answer: Multiplying the identity matrix with any vector in ℝⁿ produces the same vector, that is, Ix = x = 1 ⋅ x. Therefore, λ = 1. Since x can be any vector in ℝⁿ, the corresponding eigenvectors are all vectors in ℝⁿ.

(b) Does a diagonal matrix
$$\begin{bmatrix} d_1 & 0 & 0 & \cdots & 0 \\ 0 & d_2 & 0 & \cdots & 0 \\ 0 & 0 & d_3 & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & d_n \end{bmatrix}$$
 in \mathbb{R}^n have any eigenvalues $\lambda \in \mathbb{R}$? What are the corresponding eigenvectors?

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Answer: Since the matrix is diagonal, multiplying the diagonal matrix with any standard basis vector \vec{e}_i produces $d_i\vec{e}_i$, that is, $\mathbf{D}\vec{e}_i = d_i\vec{e}_i$. Therefore, the eigenvalues are the diagonal entries d_i of \mathbf{D} , and the corresponding eigenvector associated with $\lambda = d_i$ is the standard basis vector \vec{e}_i .

(c) Conceptually, does a rotation matrix in \mathbb{R}^2 by angle θ have any eigenvalues $\lambda \in \mathbb{R}$? For which angles is this the case?

Answer: In a conceptual sense, there are three cases:

- **Rotation by** 0°: (more accurately, any integer multiple of 360°), which yields a rotation matrix $\mathbf{R} = \mathbf{I}$: This will have one eigenvalue of +1 because it doesn't affect any vector ($\mathbf{R}\vec{x} = \vec{x}$). The eigenspace associated with it is \mathbb{R}^2 .
- **Rotation by** 180°: (more accurately, any angle of $180^\circ + n \cdot 360^\circ$ for integer *n*), which yields a rotation matrix $\mathbf{R} = -\mathbf{I}$: This will have one eigenvalue of -1 because it "flips" any vector ($\mathbf{R}\vec{x} = -\vec{x}$). The eigenspace associated with it is \mathbb{R}^2 .
- Any other rotation: there aren't any real eigenvalues. The reason is, if there were any real eigenvalue $\lambda \in \mathbb{R}$ for a non-trivial rotation matrix, it means that we can get $\mathbf{R}\vec{x} = \lambda\vec{x}$ for some $\vec{x} \neq \vec{0}$, which means that by rotating a vector, we scaled it. This is a contradiction (again, unless $\mathbf{R} = \mathbf{I}$). Refer to Figure 1 for a visualization.



Figure 1: Rotation will never scale any non-zero vector (by a real number) unless it is rotation by an integer multiple of 360° (identity matrix) or the rotation angle is $\theta = 180^\circ + n \cdot 360^\circ$ for any integer n (–I).

(d) Now let us mechanically compute the eigenvalues of the rotation matrix in \mathbb{R}^2 . Does it agree with our findings above? As a refresher, the rotation matrix **R** has the following form:

$$\mathbf{R} = \begin{bmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{bmatrix}$$

Answer: Using our known determinant formula for $2x^2$ matrices det(A) = ad - bc we can compute the characteristic polynomial

$$\det(\mathbf{R} - \lambda \mathbf{I}) = \det \begin{bmatrix} \cos(\theta) - \lambda & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) - \lambda \end{bmatrix} = \cos(\theta)^2 + \sin(\theta)^2 - 2\cos(\theta)\lambda + \lambda^2 \equiv 0$$

From here we can first simplify $1 = \cos(\theta)^2 + \sin(\theta)^2$ and then use the quadratic formula to attain the two possible λ values.

$$\lambda = \cos(\theta) \pm \sqrt{\cos(\theta)^2 - 1} = \cos(\theta) \pm i\sqrt{1 - \cos(\theta)^2} = \cos(\theta) \pm i\sqrt{\sin(\theta)^2}$$

In exponential phase notation we can write the two eigenvalues more concisely: $\lambda = e^{\pm i\theta}$

(e) Does the reflection matrix **T** across the x-axis in $\mathbb{R}^{2\times 2}$ have any eigenvalues $\lambda \in \mathbb{R}$?

$$\mathbf{T} = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$$

Answer: Yes, both +1 and -1. Mechanically, we could go through the methods we have learned for attaining a characteristic polynomial from det $(T - \lambda I) = (1 - \lambda)(-1 - \lambda) - (0)(0)$ and recalling our eigenvalues are the roots of this polynomial (the values where this polynomial is zero). This works because matrix $T - \lambda I$ only has a nonempty null space when its determinant is zero!

$$\det(T - \lambda I) = \lambda^2 - 1 \equiv 0 \quad \rightarrow \quad \lambda = \pm 1$$

Conceptually, we can reason that a vector along the x-axis will be unaffected by **T** (in this case $\lambda = +1$), where as a vector along the y-axis gets perfectly flipped by **T** (in this case $\lambda = -1$)

NOTE: A 2 × 2 reflection matrix always has $\lambda = \pm 1$, REGARDLESS of the axis of reflection. Why? Reflecting any vector that is on the reflection axis will not affect it (eigenvalue +1). Reflecting any vector orthogonal (perpendicular) to the reflection axis will just "flip it/negate it" (eigenvalue -1). In other words, the set of vectors that lie along the axis of reflection is the eigenspace associated with the eigenvalue +1 and the set of vectors orthogonal to the axis of reflection is the eigenspace associated with the eigenvalue -1.

(f) If a matrix **M** has an eigenvalue $\lambda = 0$, what does this say about its null space? What does this say about the solutions of the system of linear equations $\mathbf{M}\vec{x} = \vec{b}$?

Answer: N(A) is not just $\vec{0}$ as we have some $\vec{v} \neq \vec{0}$ satisfying $A\vec{v} = \lambda\vec{v}$. Another way we can state this is that dim(N(A)) > 0.

Thus we can imagine if $\mathbf{M}\vec{x} = \vec{b}$ has a solution then $\mathbf{M}(\vec{x} + \vec{v}) = \vec{b}$ also solves the system, hence there are infinite solutions. Yet we also know that a nonzero null space means **M** has linearly dependent columns, so the vector \vec{b} could lie outside of this span in which case there is no solution.

In summary, there are either infinite or no solutions to the system of equations $\mathbf{M}\vec{x} = \vec{b}$

(g) (**Practice**) Does the matrix $\begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}$ have any eigenvalues $\lambda \in \mathbb{R}$? What are the corresponding eigenvectors?

Answer:

Note that the matrix has linearly dependent columns. Therefore, according to part (f), one eigenvalue is $\lambda = 0$. The corresponding eigenvector, which is equivalent to the basis vector for the null space, is $\begin{bmatrix} 1 \\ -1 \end{bmatrix}$. The other eigenvalue is, by inspection, $\lambda = 1$ with the corresponding eigenvector $\begin{bmatrix} 1 \\ 0 \end{bmatrix}$ because $\begin{bmatrix} 1 \\ 0 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$.