

EECS 16A

Module 1 Lecture 10

Logistics

- Midterm reminder 10/5
- Review : Thursday night.
- Most important: HW, discussions, notes, labs.
- Read Piazza posts!

We started the semester thinking about how to do imaging / tomography well.

Systems of equations, inversion, vector spaces

Today: How do eigenvalues help?

Properties of e-values and e-vectors:

Thm: A be an $n \times n$ matrix.

$\lambda_1, \lambda_2, \dots, \lambda_n$ distinct eigenvalues.

$\lambda_i \neq \lambda_j$ for all i, j

$\lambda_1 \neq \lambda_2, \lambda_1 \neq \lambda_3 \dots \lambda_1 \neq \lambda_n$

$\lambda_2 \neq \lambda_3 \dots \lambda_2 \neq \lambda_n$

\vdots

$\vec{v}_1, \vec{v}_2 \dots \vec{v}_n$ be the corresponding eigenvectors

$$A \cdot \vec{v}_i = \lambda_i \vec{v}_i$$

Then: $\vec{v}_1, \vec{v}_2 \dots \vec{v}_n$ form a basis for \mathbb{R}^n .

$$\mathbb{R}^2, \quad n=2$$

Thm: A : 2×2 matrix.

λ_1, λ_2 , $\lambda_1 \neq \lambda_2$ eigenvalues.

\vec{u}_1, \vec{u}_2 are eigenvectors.

\vec{u}_1, \vec{u}_2 form a basis for \mathbb{R}^2 .

Proof:

Known: $A \in \mathbb{R}^{2 \times 2}$

$$A \cdot \vec{u}_1 = \lambda_1 \cdot \vec{u}_1$$

$$A \cdot \vec{u}_2 = \lambda_2 \cdot \vec{u}_2$$

$$\vec{u}_1, \vec{u}_2 \neq \vec{0}$$

$$\lambda_1 \neq \lambda_2$$

To show: \vec{u}_1, \vec{u}_2 form a basis.

Def of basis:

- ① \vec{u}_1, \vec{u}_2 are linearly independent.
- ② \vec{u}_1, \vec{u}_2 span all of \mathbb{R}^2 .

Use a proof by contradiction.

If possible, let \vec{u}_1 and \vec{u}_2 be linearly dep.

$$\alpha_1 \vec{u}_1 + \alpha_2 \vec{u}_2 = \vec{0}$$

Not all α_i are equal to 0.

rearrange \rightarrow $\vec{u}_1 = -\frac{\alpha_2}{\alpha_1} \cdot \vec{u}_2$ (*)

Say $\alpha_1 \neq 0$

(*) Multiply by A. ●

$$A \cdot \vec{u}_1 = A \cdot \left(\frac{-\alpha_2}{\alpha_1} \right) \cdot \vec{u}_2$$

$$= \left(\frac{-\alpha_2}{\alpha_1} \right) \cdot A \cdot \vec{u}_2$$

$$A \vec{u}_1 = \left(\frac{-\alpha_2}{\alpha_1} \right) \cdot \lambda_2 \cdot \vec{u}_2 \quad (**)$$

$$A \cdot \vec{u}_1 = \lambda_1 \vec{u}_1$$

$$= \lambda_1 \left(\frac{-\alpha_2}{\alpha_1} \right) \cdot \vec{u}_2 \quad \left[\text{Using } (*) \right]$$

$$A \vec{u}_1 = \lambda_1 \left(\frac{-\alpha_2}{\alpha_1} \right) \cdot \vec{u}_2 \quad (***)$$

\Rightarrow (**) and (***) \Rightarrow

$$\left(\frac{-\alpha_2}{\alpha_1} \right) \lambda_2 \cdot \vec{u}_2 = \lambda_1 \left(\frac{-\alpha_2}{\alpha_1} \right) \cdot \vec{u}_2$$

$\Rightarrow \lambda_2 = \lambda_1 \quad !!! \quad \text{Contradiction!}$

Therefore: \vec{u}_1 and \vec{u}_2 must be
linearly independent!

Now: \vec{u}_1, \vec{u}_2 .

Question: Can I have a set of more
than 2 vectors in a 2 dimensional space
that are linearly independent?

$\implies \vec{u}_1$ and \vec{u}_2 must span \mathbb{R}^2 .

\implies They must form a basis!

To show I can reach

any $\vec{x} \in \mathbb{R}^2$, using \vec{u}_1, \vec{u}_2

$$\left[\vec{u}_1 \quad \vec{u}_2 \right] \vec{x} \leftarrow V = \left[\vec{u}_1 \quad \vec{u}_2 \right]$$

Q.E.D.

V is an invertible matrix \Rightarrow

$\left[V \mid \vec{x} \right]$ has a unique solution.

Matrix transformations

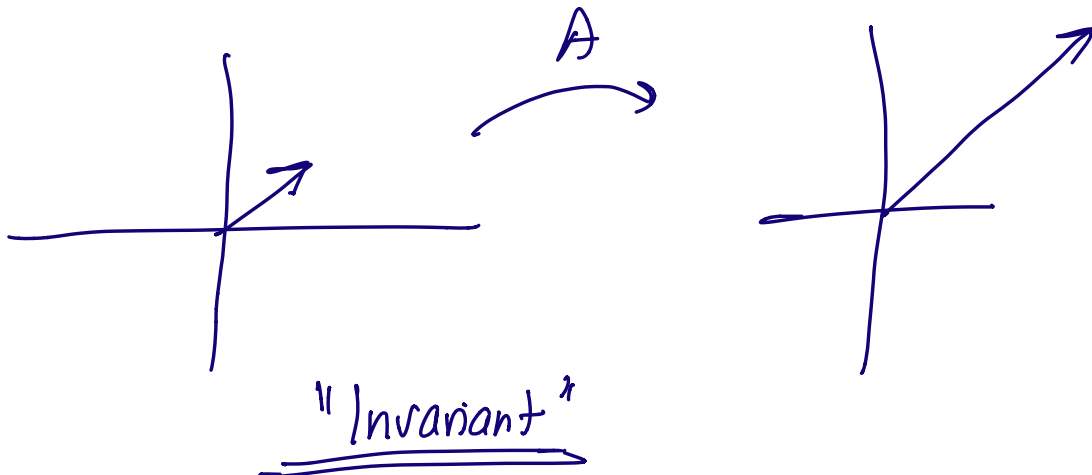
$$\vec{x}[t+1] = A \cdot \vec{x}[t] \quad \text{"Dynamical system"}$$

Steady state: $\vec{x} = A \cdot \vec{x}$

\vec{x} is an eigenvector with eigenvalue 1.

In general: \vec{x} eigenvector with eigenvalue λ

$$A\vec{x} = \lambda\vec{x}$$



A : 2×2 matrix 2 eigenvectors.

\vec{x} that is not an eigenvector of A .

$$A \cdot \vec{x}$$

$\vec{u}_1, \vec{u}_2, \dots, \vec{u}_n$ are the eigenvectors of A
 $\lambda_1, \lambda_2, \dots, \lambda_n$ are e-values (distinct).

We can write \vec{x} as a linear combination of
 $\vec{u}_1, \vec{u}_2, \dots, \vec{u}_n$!

$$\vec{x} = \alpha_1 \vec{u}_1 + \alpha_2 \vec{u}_2 + \dots + \alpha_n \vec{u}_n$$

$$\begin{aligned} A \cdot \vec{x} &= A (\alpha_1 \vec{u}_1 + \alpha_2 \vec{u}_2 + \dots + \alpha_n \vec{u}_n) \\ &= \alpha_1 A \vec{u}_1 + \alpha_2 A \vec{u}_2 + \dots + \alpha_n A \vec{u}_n \\ &= \alpha_1 \lambda_1 \vec{u}_1 + \alpha_2 \lambda_2 \vec{u}_2 + \dots + \alpha_n \lambda_n \vec{u}_n \end{aligned}$$

$$\begin{aligned} A^2 \cdot \vec{x} &= A \cdot A \vec{x} \\ &= A (\end{aligned}$$

$$= A \alpha_1 \lambda_1 \vec{u}_1 + \dots + A \alpha_n \lambda_n \vec{u}_n$$

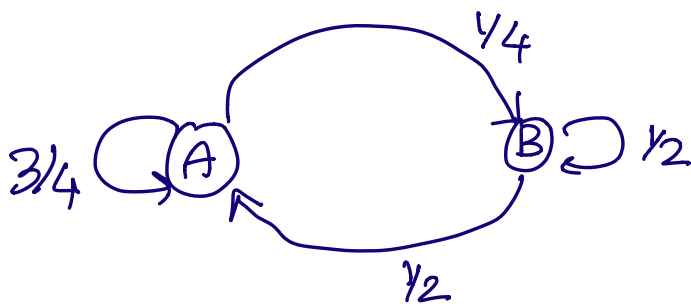
$$= \alpha_1 \lambda_1^2 \vec{u}_1 + \dots + \alpha_n \lambda_n^2 \vec{u}_n$$

$$A^t \vec{x} = \alpha_1 \lambda_1^t \vec{u}_1 + \alpha_2 \lambda_2^t \vec{u}_2 + \dots + \alpha_n \lambda_n^t \vec{u}_n$$

$$\lambda_i = 1$$

$$\lambda_i < 1$$

$$\lambda_i > 1$$



$$Q = \begin{bmatrix} 3/4 & 1/2 \\ 1/4 & 1/2 \end{bmatrix}$$

$$Q - \lambda I = \begin{bmatrix} 3/4 - \lambda & 1/2 \\ 1/4 & 1/2 - \lambda \end{bmatrix}$$

Eigenvals, E-vectors of Q.

$$\det(Q - \lambda I) = \left(\frac{3}{4} - \lambda\right)\left(\frac{1}{2} - \lambda\right) - \frac{1}{4} \cdot \frac{1}{2}$$

$$= \frac{3}{8} + \lambda^2 - \frac{5}{4} \lambda - \frac{1}{8}$$

$$= \frac{1}{4} - \frac{5}{4} \lambda + \lambda^2 \leftarrow$$

$$= (\lambda - \frac{1}{4})(\lambda - 1)$$

→ HW. → Practice

$$\left. \begin{array}{l} \lambda_1 = 1, \vec{u}_1 = \begin{bmatrix} 2 \\ 1 \end{bmatrix} \\ \lambda_2 = \frac{1}{4}, \vec{u}_2 = \begin{bmatrix} -1 \\ 1 \end{bmatrix} \end{array} \right\} \begin{array}{l} \text{Compute} \\ \text{Nul}(Q - \lambda I) \end{array}$$

$\begin{bmatrix} 2 \\ 1 \end{bmatrix}, \begin{bmatrix} -1 \\ 1 \end{bmatrix}$ → Basis for \mathbb{R}^2

Let \vec{x} be any vector in \mathbb{R}^2

$$\vec{x} = \alpha_1 \begin{bmatrix} 2 \\ 1 \end{bmatrix} + \alpha_2 \begin{bmatrix} -1 \\ 1 \end{bmatrix}$$

$$\lambda_1 = 1, \vec{u}_1$$

$$\vec{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$\lambda_2 = \frac{1}{4}, \vec{u}_2$$

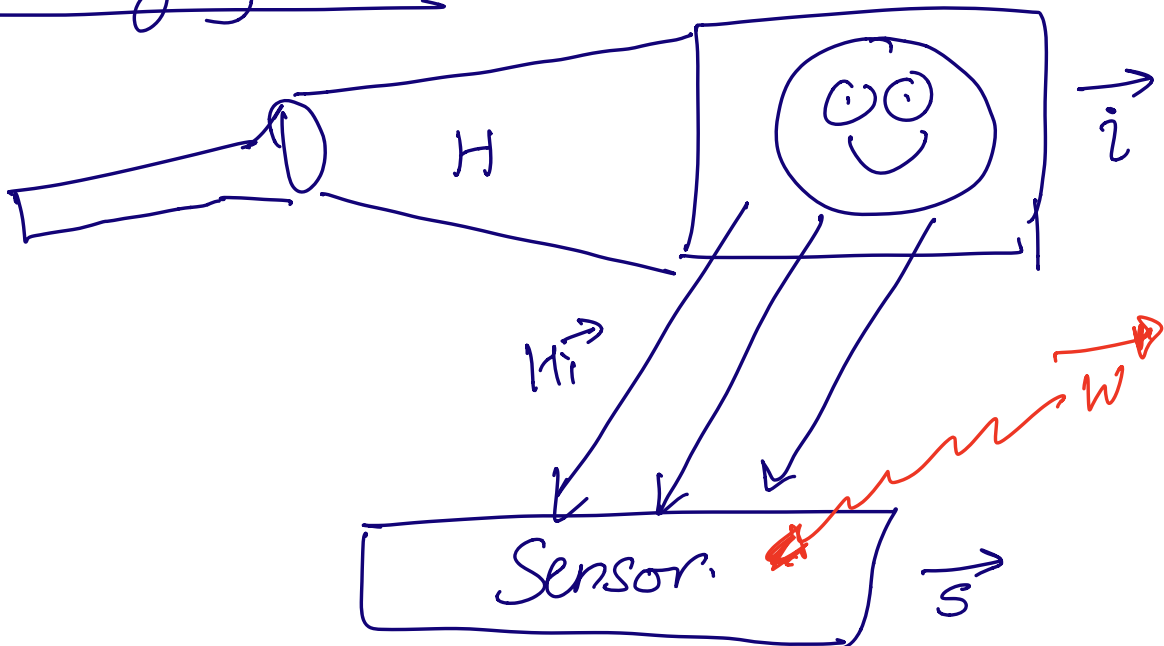
$$\vec{x} = \alpha_1 \vec{u}_1 + \alpha_2 \vec{u}_2$$

$$\begin{aligned}
 A\vec{x} &= A \cdot \alpha_1 \vec{u}_1 + A \cdot \alpha_2 \vec{u}_2 \\
 &= \alpha_1 \cdot 1 \cdot \vec{u}_1 + \alpha_2 \cdot \frac{1}{4} \cdot \vec{u}_2
 \end{aligned}$$

$$A^t \vec{x} = \alpha_1 (1)^t \cdot \vec{u}_1 + \alpha_2 \left(\frac{1}{4}\right)^t \cdot \vec{u}_2$$

$$\lim_{t \rightarrow \infty} A^t \vec{x} = \alpha_1 \vec{u}_1 + 0$$

Imaging Lab



$$\underbrace{\vec{s}}_{\text{Sensor}} = H \cdot \underbrace{\vec{z}}_{\text{image}}$$

H : measurement
mask / matrix

~~$$\vec{z} = H^{-1} \vec{s}$$~~

H^{-1} exists

$$\vec{s} = H \vec{z} + \vec{w}$$

$$\begin{aligned} \underline{H^{-1} \vec{s}} &= H^{-1} (H \vec{z} + \vec{w}) \\ &= \vec{z} + \underbrace{H^{-1} \vec{w}}_{\text{corruption}} \end{aligned}$$

Eigenvectors of H^{-1}

$$\vec{u}_1, \vec{u}_2, \dots, \vec{u}_n, \quad \lambda_1, \lambda_2, \dots, \lambda_n$$

$$\vec{w} = \alpha_1 \vec{u}_1 + \alpha_2 \vec{u}_2 + \dots + \alpha_n \vec{u}_n$$

$$H^{-1} \vec{w} = \alpha_1 \lambda_1 \vec{u}_1 + \alpha_2 \lambda_2 \vec{u}_2 + \dots + \alpha_n \lambda_n \vec{u}_n$$



Office Hours
Addendum: Let $\vec{u}_1, \vec{u}_2 \in \mathbb{R}^2$.

\vec{u}_1, \vec{u}_2 are linearly indep.

Prove that: \vec{u}_1, \vec{u}_2 are a basis for \mathbb{R}^2 .

$$V = \begin{bmatrix} \frac{1}{u_1} & \frac{1}{u_2} \\ 1 & 1 \end{bmatrix} \quad V \in \mathbb{R}^{2 \times 2}$$

V has linearly indep columns.

V^{-1} exist.

We need to show that for any vector $\vec{x} \in \mathbb{R}^2$ we can write \vec{x} as a linear combi of \vec{u}_1 and \vec{u}_2 .

$$\begin{bmatrix} \vec{u}_1 & \vec{u}_2 \end{bmatrix} \begin{bmatrix} \alpha_1 \\ \alpha_2 \end{bmatrix} = \begin{bmatrix} \vec{x} \end{bmatrix} \quad \leftarrow \text{Show that such } \alpha_1, \alpha_2 \text{ exist.}$$

Consider: $V^{-1} \vec{x}$

→ Since V is invertible we know that this has a unique solution.

So α_1, α_2 exist.

So $\vec{x} \in \text{span}\{\vec{u}_1, \vec{u}_2\}$.

So $\{\vec{u}_1, \vec{u}_2\}$ is a basis for \mathbb{R}^2 .
