

EECS 16A
2/20

- Eigenvalues/vectors
- Imaging lab
- Determinants

Module 1
①
Wrap-up!

$$A = \begin{bmatrix} 1 & 2 \\ 3 & 1 \end{bmatrix} \quad \det(A - \lambda I) = (\lambda - 5)(\lambda + 1)$$
$$\lambda_1 = 5 \quad \lambda_2 = -1 .$$

$$E_5 = \text{span} \left\{ \underbrace{\begin{bmatrix} 1 \\ 1 \end{bmatrix}}_{\vec{v}_1}, \underbrace{\begin{bmatrix} 1 \\ -1 \end{bmatrix}}_{\vec{v}_2} \right\}$$
$$E_{-1} = \left\{ \begin{bmatrix} 1 \\ -1 \end{bmatrix} \right\} .$$

Note: Any $\vec{x} \in \mathbb{R}^2$ can be written as

$$\underbrace{A \cdot A \cdot A \cdots A}_{K} \quad \vec{x} = \alpha_1 \vec{v}_1 + \alpha_2 \vec{v}_2 .$$

$$A \vec{x} = \underbrace{A}_{K} (\alpha_1 \vec{v}_1 + \alpha_2 \vec{v}_2) = \alpha_1 5 \vec{v}_1 + \alpha_2 (-1) \vec{v}_2 .$$

(2)

$$\text{Consider } \vec{x}(k) = A \vec{x}(k-1)$$

$$\vec{x}(0) = \begin{bmatrix} 1 \\ 2 \end{bmatrix} = 2\vec{v}_1 + 0 \cdot \vec{v}_2$$

$$\vec{x}(100) = 2^{100} \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

"unstable"
system behavior

$$\vec{x}(10) = \begin{pmatrix} 3 \\ -3 \end{pmatrix} = 3\vec{v}_2$$

$$\vec{x}(100) = (-1)^{100} \begin{pmatrix} 3 \\ -3 \end{pmatrix}$$

(3)

When we design systems, we generally want systems to have "stable" behavior

$$\vec{x}(n) = A \vec{x}(n-1)$$

↑
design

i.e. all eigenvalues have magnitude < 1 .

(or ≤ 1 , which may give rise to some nontrivial steady state).

(9)

Thm: Let $A \in \mathbb{R}^{n \times n}$ have distinct eigenvalues $\lambda_1, \dots, \lambda_n$. Let $\vec{v}_1, \dots, \vec{v}_n$ be nonzero eigenvectors corresp. to $\lambda_1, \dots, \lambda_n$.

Then $\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n\}$ are linearly indep.

Pf (for 2×2 case):

Know: $A\vec{v}_1 = \lambda_1\vec{v}_1 \quad | \quad \vec{v}_1, \vec{v}_2 \neq 0$

WTS:

\vec{v}_2 not scalar multiple
of \vec{v}_1

$A\vec{v}_2 = \lambda_2\vec{v}_2 \quad | \quad \lambda_1 \neq \lambda_2$

Let's suppose that $\vec{v}_1 = \alpha\vec{v}_2$.

$$\lambda_1\vec{v}_1 = A\vec{v}_1 = A\alpha\vec{v}_2 = \alpha A\vec{v}_2 = \alpha\lambda_2\vec{v}_2 = \lambda_2\vec{v}_1$$

$$\Rightarrow \lambda_1 = \lambda_2 \quad \xrightarrow{\text{"contradiction"}}$$

(5)

$$\vec{x}(k+1) = A \vec{x}(k), \quad A \in \mathbb{R}^{n \times n}$$

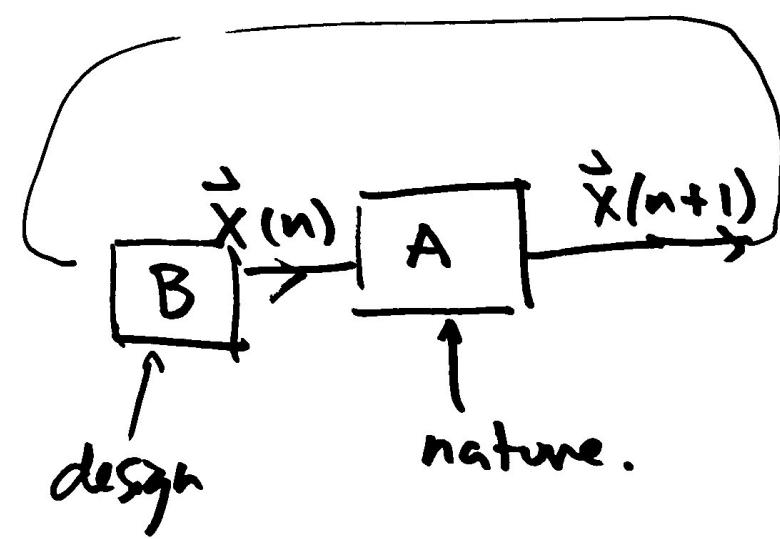
with distinct eigenvalues
 $\lambda_1, \dots, \lambda_n$

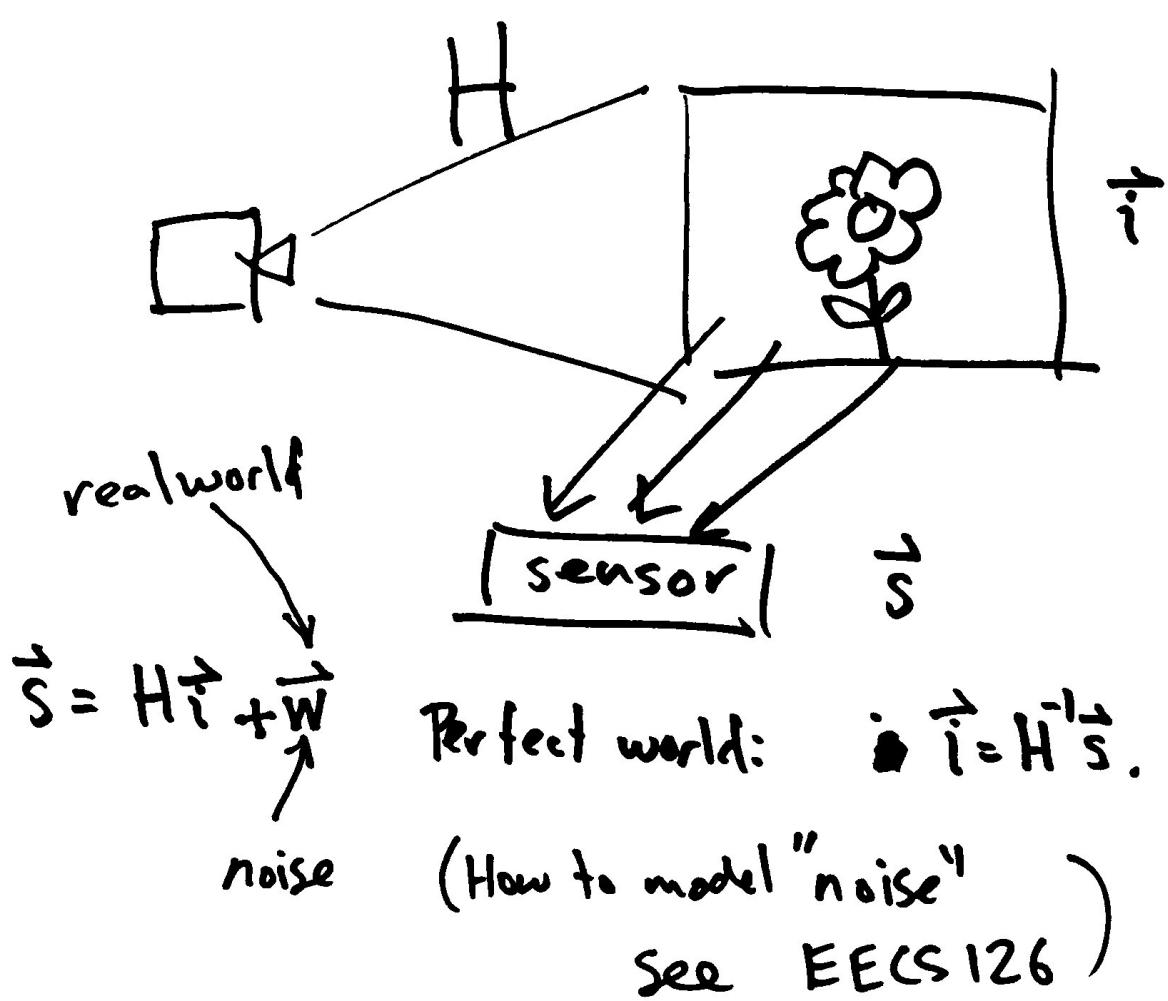
Q: $\vec{x}(100) = ?$

Step 1: write $\vec{x}(0) = \sum \alpha_i \vec{v}_i$. (why?)

Step 2: $\vec{x}(k) = A^k \vec{x}(0) = A^k \sum_{i=1}^n \alpha_i \vec{v}_i$
 $= \sum_{i=1}^n \alpha_i A^k \vec{v}_i = \sum_{i=1}^n \alpha_i (\lambda_i)^k \vec{v}_i$

(6)





(8)

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

$$\hat{\vec{i}} = H^{-1}\vec{s} = \vec{i} + H^{-1}\vec{w}.$$

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Q: What properties of H^{-1} would ensure $H^{-1}\vec{w}$ is "small"

A: Eigenvalues of H^{-1} should be "small".

Q: If H invertible, how do eigenvalues of H^{-1} relate to those of H ?

(7)

Let (λ, \vec{x}) be eval/evec for H .

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

left-mlt. by H^{-1} : $\vec{x} = \lambda H^{-1}\vec{x} \xrightarrow{\downarrow} H^{-1}\vec{x} = \frac{1}{\lambda}\vec{x}$
 $\Rightarrow (\frac{1}{\lambda}, \vec{x})$ is eval/evec for H^{-1} .
 when $\lambda \neq 0$.

Claim: If H is invertible, then all eigenvalues are nonzero. (Why). $0 \neq \det(H) = \det(H - 0 \cdot I)$

$$= p_H(0)$$

$\Rightarrow 0$ is not a root of p_H
 $\Rightarrow 0$ is not e-value of H .

Summary: If H is invertible with eigenvalues $\lambda_1, \dots, \lambda_n$, then H^{-1} has eigenvalues $\frac{1}{\lambda_1}, \dots, \frac{1}{\lambda_n}$. (10)

Also: if \vec{x} is an eigenvector of H for eigenvalue λ , then it is also an eigenvector of H^{-1} for eigenvalue $\frac{1}{\lambda}$.

Last observation of module 1.

$$P_A(\lambda) = \det(A - \lambda I) = \prod_{i=1}^n (\lambda_i - \lambda_0) \quad , \begin{matrix} \lambda_i's = \\ \text{evals} \\ \text{of } A. \end{matrix}$$

$$\det(A) = P_A(0) = \prod \lambda_i$$

$\Rightarrow \det(A) = \text{product of eigenvalues.}$

(ii)

$$\text{Ex: } \det(A^{-1}) = \prod_{i=1}^n \frac{1}{\lambda_i} = \frac{1}{\prod_{i=1}^n \lambda_i} = \frac{1}{\det(A)} .$$