

MA2001

LIVE LECTURE 10

Q&A: log in to Pollev.com/vtpoll

Q&A: log in to PolLEv.com/vtpoll

Topics for week 10

5.3 Best Approximation

5.4 Orthogonal Matrices

6.1 Eigenvalues and Eigenvectors

Least Squares Solution

- \mathbf{u} is the "best approximated" solution to an inconsistent $\mathbf{Ax} = \mathbf{b}$
- \mathbf{u} is the least squares solution to $\mathbf{Ax} = \mathbf{b}$
- \mathbf{u} is a solution of $\mathbf{A}^T \mathbf{Ax} = \mathbf{A}^T \mathbf{b}$ always consistent
- \mathbf{u} is a solution of $\mathbf{Ax} = \mathbf{p}$, the projection of \mathbf{b} onto column space of \mathbf{A}
always consistent

We can always find least squares solution to any $\mathbf{Ax} = \mathbf{b}$

Least squares solution to $\mathbf{Ax} = \mathbf{b}$ may be unique or infinite

Find projection using least squares solution

Find **projection** of \mathbf{w} onto a subspace $V = \text{span}\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$

i. Form matrix \mathbf{A} using $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$ as columns.

Column space of $\mathbf{A} = V$

ii. Find the least squares solution of $\mathbf{Ax} = \mathbf{w}$.

iii. Find the solutions of $\mathbf{A}^T \mathbf{Ax} = \mathbf{A}^T \mathbf{w}$.

iv. Take any solution \mathbf{u} in (iii).

v. \mathbf{Au} gives the projection of \mathbf{w} onto V .

Orthogonal Matrices

\mathbf{A} is $n \times n$ matrix

- \mathbf{A} is an **orthogonal** matrix
- $\mathbf{A}^{-1} = \mathbf{A}^T$
- $\mathbf{A}\mathbf{A}^T = \mathbf{I}$ \mathbf{A}^T is also **orthogonal** matrix
- The rows of \mathbf{A} form an orthonormal basis for \mathbf{R}^n .
- The columns of \mathbf{A} form an orthonormal basis for \mathbf{R}^n .

$$\mathbf{A} = \begin{pmatrix} \mathbf{a}_1 \\ \vdots \\ \mathbf{a}_n \end{pmatrix} \quad \mathbf{A}^T = (\mathbf{a}_1^T \quad \dots \quad \mathbf{a}_n^T) \quad \mathbf{A}\mathbf{A}^T = \begin{pmatrix} \mathbf{a}_1\mathbf{a}_1^T & \mathbf{a}_1\mathbf{a}_2^T & \mathbf{a}_1\mathbf{a}_3^T \\ \mathbf{a}_2\mathbf{a}_1^T & \mathbf{a}_2\mathbf{a}_2^T & \mathbf{a}_2\mathbf{a}_3^T \\ \mathbf{a}_3\mathbf{a}_1^T & \mathbf{a}_3\mathbf{a}_2^T & \mathbf{a}_3\mathbf{a}_3^T \end{pmatrix} = \begin{pmatrix} \mathbf{a}_1 \cdot \mathbf{a}_1 & \mathbf{a}_1 \cdot \mathbf{a}_2 & \mathbf{a}_1 \cdot \mathbf{a}_3 \\ \mathbf{a}_2 \cdot \mathbf{a}_1 & \mathbf{a}_2 \cdot \mathbf{a}_2 & \mathbf{a}_2 \cdot \mathbf{a}_3 \\ \mathbf{a}_3 \cdot \mathbf{a}_1 & \mathbf{a}_3 \cdot \mathbf{a}_2 & \mathbf{a}_3 \cdot \mathbf{a}_3 \end{pmatrix}$$

True or False

$\mathbf{A} = (\mathbf{c}_1 \ \mathbf{c}_2 \ \dots \ \mathbf{c}_k)$ is an $n \times k$ matrix such that the columns $\{\mathbf{c}_1, \mathbf{c}_2, \dots, \mathbf{c}_k\}$ of \mathbf{A} form an orthonormal set.

Can we conclude that

(I) $\mathbf{A}^T \mathbf{A} = \text{identity matrix}$ and (II) $\mathbf{A} \mathbf{A}^T = \text{identity matrix}$?

$$\mathbf{A}^T \mathbf{A} = \begin{pmatrix} \mathbf{c}_1^T \\ \vdots \\ \mathbf{c}_k^T \end{pmatrix} (\mathbf{c}_1 \ \dots \ \mathbf{c}_k) = \begin{pmatrix} \mathbf{c}_1^T \mathbf{c}_1 & \dots & \mathbf{c}_1^T \mathbf{c}_k \\ \vdots & & \vdots \\ \mathbf{c}_k^T \mathbf{c}_1 & \dots & \mathbf{c}_k^T \mathbf{c}_k \end{pmatrix} \quad \text{True}$$

$$\mathbf{A} \mathbf{A}^T = \begin{pmatrix} \mathbf{r}_1 \\ \vdots \\ \mathbf{r}_n \end{pmatrix} (\mathbf{r}_1^T \ \dots \ \mathbf{r}_n^T) = \begin{pmatrix} \mathbf{r}_1 \mathbf{r}_1^T & \dots & \mathbf{r}_1 \mathbf{r}_n^T \\ \vdots & & \vdots \\ \mathbf{r}_n \mathbf{r}_1^T & \dots & \mathbf{r}_n \mathbf{r}_n^T \end{pmatrix} \quad \text{False}$$

Exercise 5 Q32 (Tutorial)

A be an orthogonal matrix:

Orthogonal matrix
preserves
norm, distance, angles

a) $||\mathbf{u}|| = ||\mathbf{A}\mathbf{u}||$;

b) $d(\mathbf{u}, \mathbf{v}) = d(\mathbf{A}\mathbf{u}, \mathbf{A}\mathbf{v})$;

c) the angle between \mathbf{u} and \mathbf{v} is equal to the angle between $\mathbf{A}\mathbf{u}$ and $\mathbf{A}\mathbf{v}$.

- Express each part using dot product
- Regard the dot product as matrix multiplication
- Use the fact that **A** is orthogonal

\mathbf{u} and \mathbf{v} regarded as column matrices

$$\mathbf{u} \cdot \mathbf{v} = \mathbf{u}^T \mathbf{v}$$

- Basis
- Orthogonal basis
- Orthonormal basis

Exercise 5 Q33 (Tutorial)

\mathbf{A} be an orthogonal matrix and let $S = \{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n\}$ be a basis for \mathbf{R}^n .

(a) Show that $T = \{\mathbf{A}\mathbf{u}_1, \mathbf{A}\mathbf{u}_2, \dots, \mathbf{A}\mathbf{u}_n\}$ is a basis for \mathbf{R}^n .

- Refer Ex 3.30(b)

(b) If S is orthogonal, show that T is orthogonal.

- Show $\mathbf{A}\mathbf{u}_i \cdot \mathbf{A}\mathbf{u}_j = 0$ for any $i \neq j$ using matrix multiplication

(c) If S is orthonormal, is T orthonormal?

- Refer to Q32

Transition matrix between orthonormal bases

Let $S = \{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n\}$ and $T = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$ be two bases for \mathbf{R}^n

Transition matrix from S to T: $\mathbf{P} = ([\mathbf{u}_1]_T \ [\mathbf{u}_2]_T \cdots [\mathbf{u}_n]_T)$

S is an orthonormal basis for \mathbf{R}^n
T is the standard basis for \mathbf{R}^n

S is the standard basis for \mathbf{R}^n
T is an orthonormal basis for \mathbf{R}^n

S is an orthonormal basis for \mathbf{R}^n
T is an orthonormal basis for \mathbf{R}^n

$$\mathbf{P} = (\mathbf{u}_1 \ \mathbf{u}_2 \ \cdots \ \mathbf{u}_n)$$

$$\mathbf{P} = \begin{pmatrix} \mathbf{v}_1 \\ \mathbf{v}_2 \\ \vdots \\ \mathbf{v}_n \end{pmatrix}$$

$$\mathbf{P} = \begin{pmatrix} \mathbf{u}_1 \cdot \mathbf{v}_1 & \mathbf{u}_2 \cdot \mathbf{v}_1 & \cdots & \mathbf{u}_k \cdot \mathbf{v}_1 \\ \mathbf{u}_1 \cdot \mathbf{v}_2 & \mathbf{u}_2 \cdot \mathbf{v}_2 & \cdots & \mathbf{u}_k \cdot \mathbf{v}_2 \\ \vdots & \vdots & & \vdots \\ \mathbf{u}_1 \cdot \mathbf{v}_k & \mathbf{u}_2 \cdot \mathbf{v}_k & \cdots & \mathbf{u}_k \cdot \mathbf{v}_k \end{pmatrix}$$

\mathbf{P} is an
orthogonal
matrix

True or False

$$S = \{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n\} \text{ and } T = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$$

two orthonormal bases for \mathbf{R}^n

$$\mathbf{A} = (\mathbf{u}_1 \ \mathbf{u}_2 \ \dots \ \mathbf{u}_n) \text{ and } \mathbf{B} = (\mathbf{v}_1 \ \mathbf{v}_2 \ \dots \ \mathbf{v}_n) \text{ orthogonal matrices}$$

\mathbf{P} the transition matrix from S to T orthogonal matrices

True or false: $\mathbf{PA} = \mathbf{B}$

Recall: $\mathbf{P}[\mathbf{w}]_S = [\mathbf{w}]_T$

Correct relation: $\mathbf{A} = \mathbf{BP}$

HW3 Q5(iv): $\mathbf{BC} = \mathbf{A}$

$$\begin{array}{c} (\mathbf{B} \mid \mathbf{A}) \rightarrow (\mathbf{I} \mid \mathbf{P}) \\ \text{Pre-multiply by } \mathbf{B}^{-1} \downarrow \nearrow \\ (\mathbf{B}^{-1}\mathbf{B} \mid \mathbf{B}^{-1}\mathbf{A}) \end{array}$$

\mathbf{A}, \mathbf{B} are not square matrices

Q&A: log in to PolleEv.com/vtpoll

Orthonormal matrix ?

What is the difference between
orthogonal matrix and orthonormal matrix ?



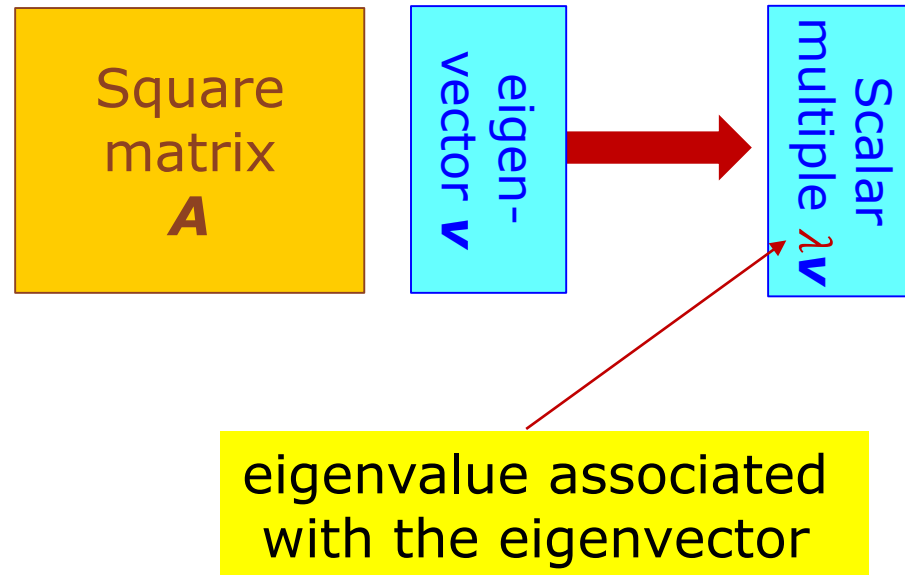
The rows/columns of **A**
form an orthonormal set



No such thing as
orthonormal matrix !

Q&A: log in to PolLEv.com/vtpoll

Eigenvalue and Eigenvector Visualization



May not give the complete set of eigenvalues

Finding eigenvalue (given eigenvector)

$$\mathbf{A} = \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 1 & 1 \end{pmatrix} \quad \text{Are these eigenvectors of } \mathbf{A}: \begin{pmatrix} 1 \\ -1 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} ?$$

$$\begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 1 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ -1 \\ 0 \end{pmatrix} = \begin{pmatrix} 1 \\ -1 \\ 0 \end{pmatrix}$$

$\begin{pmatrix} 1 \\ -1 \\ 0 \end{pmatrix}$ is an eigenvector

with eigenvalue 1

$$\begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 1 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} = \begin{pmatrix} 2 \\ 0 \\ 2 \end{pmatrix}$$

$\begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}$ is an eigenvector

with eigenvalue 2

$$\begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 1 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 2 \\ 1 \\ 2 \end{pmatrix}$$

$\begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}$ is not an eigenvector

$$\begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 1 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

$\begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}$ is an eigenvector

with eigenvalue 0

$n \times n$ matrix has n eigenvalues, counting multiplicities

Finding eigenvalues (without eigenvector)

$$\mathbf{A} = \begin{pmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{pmatrix}$$

So the eigenvalues of \mathbf{A} are 1 (repeated) and -1 (repeated).

Characteristic polynomial:

$$\det(\lambda \mathbf{I} - \mathbf{A}) = \begin{vmatrix} \lambda & 0 & -1 & 0 \\ 0 & \lambda & 0 & -1 \\ -1 & 0 & \lambda & 0 \\ 0 & -1 & 0 & \lambda \end{vmatrix}$$

$$= \lambda \times \begin{vmatrix} \lambda & 0 & -1 \\ 0 & \lambda & 0 \\ -1 & 0 & \lambda \end{vmatrix} - 1 \times \begin{vmatrix} 0 & \lambda & -1 \\ -1 & 0 & 0 \\ 0 & -1 & \lambda \end{vmatrix}$$

$$= \lambda(\lambda^3 - \lambda) - (1)(\lambda^2 - 1)$$

$$= \lambda^2(\lambda^2 - 1) - (1)(\lambda^2 - 1)$$

$$= (\lambda^2 - 1)(\lambda^2 - 1)$$

$$= (\lambda - 1)(\lambda + 1)(\lambda - 1)(\lambda + 1) = (\lambda - 1)^2(\lambda + 1)^2$$

Multiplicities of the eigenvalues

Different perspectives of eigenvalues

- λ is an eigenvalue of \mathbf{A}
- $\mathbf{A}\mathbf{v} = \lambda\mathbf{v}$ for some nonzero column vector \mathbf{v}
- $(\lambda\mathbf{I} - \mathbf{A})\mathbf{v} = \mathbf{0}$ for some nonzero column vector \mathbf{v}
- $\det(\lambda\mathbf{I} - \mathbf{A}) = 0$

Solve this equation to find the eigenvalues of \mathbf{A}

\mathbf{v} is a non-trivial solution of the homogeneous system
 $(\lambda\mathbf{I} - \mathbf{A})\mathbf{x} = \mathbf{0}$

True or False

Let λ be an eigenvalue of an invertible matrix A .
Then

1. 2λ is an eigenvalue of $2A$.
2. λ^{-1} is an eigenvalue of A^{-1} .

Bring in the corresponding eigenvector \mathbf{v}

Start with $A\mathbf{v} = \lambda\mathbf{v}$

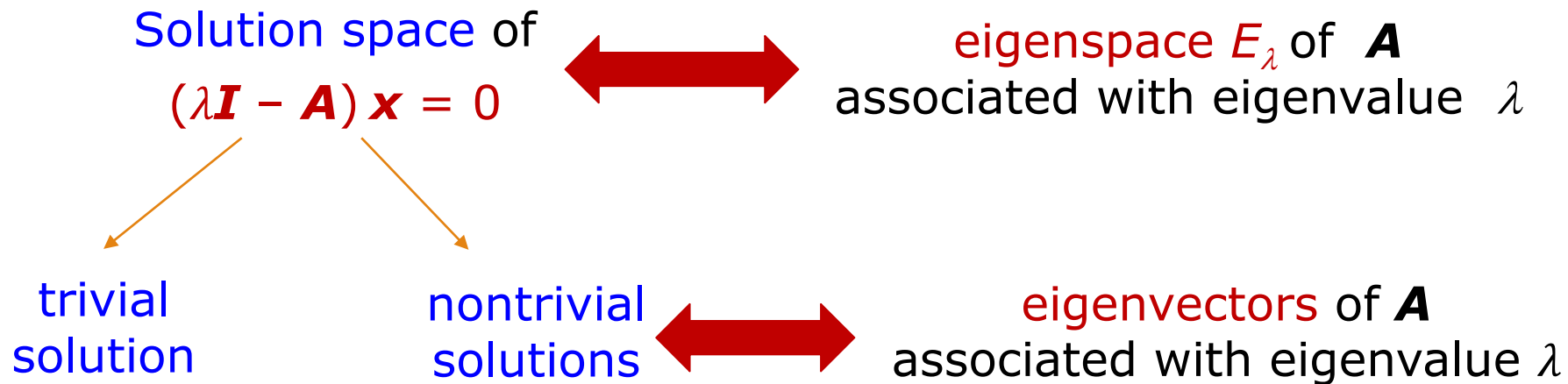
Ways To Find Eigenvalues

- If an eigenvector \mathbf{u} is given, multiply it by the matrix: $\mathbf{A}\mathbf{u} = \lambda\mathbf{u}$
- Solve characteristic equation $\det(\lambda\mathbf{I} - \mathbf{A}) = 0$
- If the matrix is triangular, take the diagonal entries
- If λ is an eigenvalue of \mathbf{A} , then
 - $c\lambda$ is an eigenvalue of $c\mathbf{A}$
 - λ is an eigenvalue of \mathbf{A}^T
 - λ^n is an eigenvalue of \mathbf{A}^n
 - λ^{-1} is an eigenvalue of \mathbf{A}^{-1} (when \mathbf{A} is invertible)

Exercise 6 Q3

Eigenspace

\mathbf{A} : $n \times n$ matrix and λ is (one of the) eigenvalue of \mathbf{A}



Zero vector is not an eigenvector

Check: Eigenspace for $\lambda = -1$: $E_{-1} = \text{span}\left\{\begin{pmatrix} -1 \\ 0 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ -1 \\ 0 \\ 1 \end{pmatrix}\right\}$

Finding eigenvector (given eigenvalue)

$$\mathbf{A} = \begin{pmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{pmatrix}$$

Solve $(\lambda \mathbf{I} - \mathbf{A}) \mathbf{x} = \mathbf{0}$ for eigenvalues $\lambda = 1$ and -1

$$\begin{pmatrix} \lambda & 0 & -1 & 0 \\ 0 & \lambda & 0 & -1 \\ -1 & 0 & \lambda & 0 \\ 0 & -1 & 0 & \lambda \end{pmatrix} \begin{pmatrix} x \\ y \\ z \\ w \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \end{pmatrix}$$

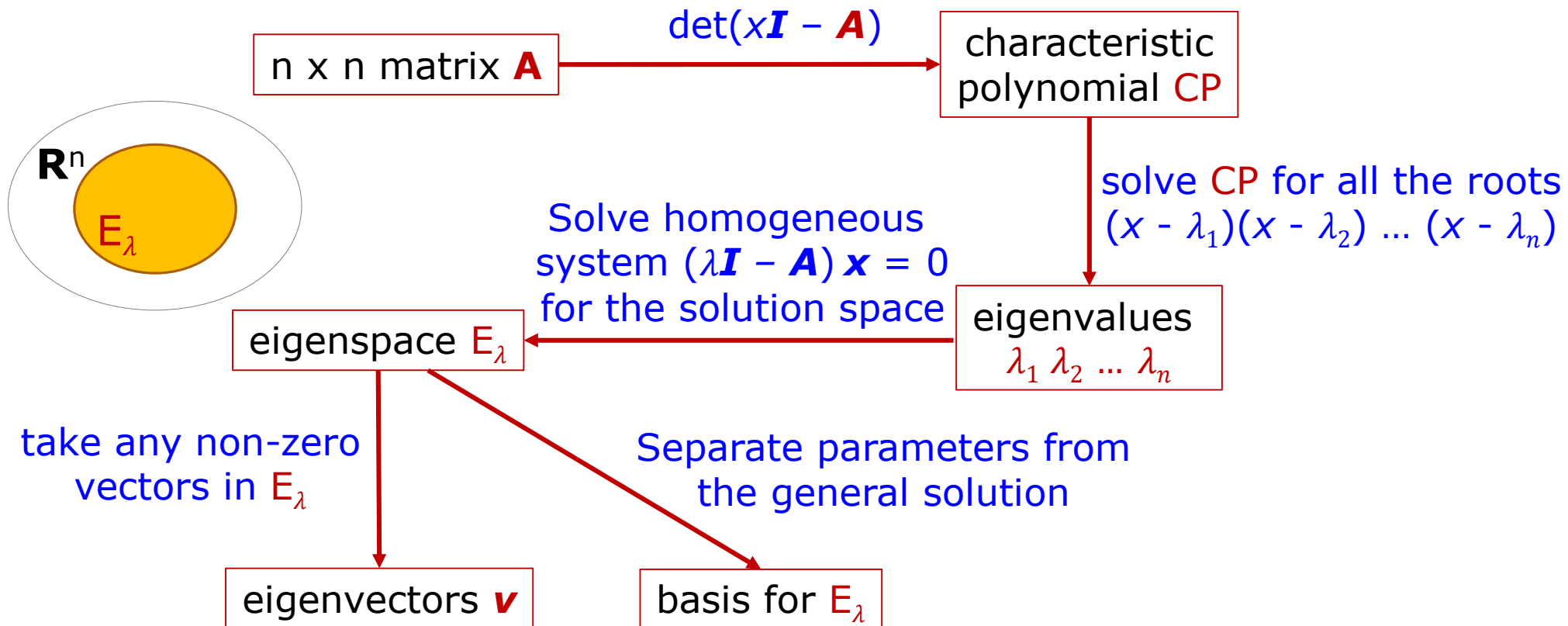
$$\lambda = 1$$

$$\left(\begin{array}{cccc|c} 1 & 0 & -1 & 0 & 0 \\ 0 & 1 & 0 & -1 & 0 \\ -1 & 0 & 1 & 0 & 0 \\ 0 & -1 & 0 & 1 & 0 \end{array} \right) \xrightarrow{G.E.} \left(\begin{array}{cccc|c} 1 & 0 & -1 & 0 & 0 \\ 0 & 1 & 0 & -1 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{array} \right) \longrightarrow w = t, z = s, y = t, x = s$$

$$\text{Gen. soln: } \begin{pmatrix} x \\ y \\ z \\ w \end{pmatrix} = \begin{pmatrix} s \\ t \\ s \\ t \end{pmatrix} = s \begin{pmatrix} 1 \\ 0 \\ 1 \\ 0 \end{pmatrix} + t \begin{pmatrix} 0 \\ 1 \\ 0 \\ 1 \end{pmatrix} \quad \text{Eigenspace for } \lambda = 1 : E_1 = \text{span}\left\{\begin{pmatrix} 1 \\ 0 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 0 \\ 1 \end{pmatrix}\right\}$$

Any non-trivial linear combination is an eigenvector associated to $\lambda = 1$

Complete process



$$\dim(U + V) = \dim(U) + \dim(V) - \dim(U \cap V)$$

Exercise 6 Q7 (modified)

A is a 3 x 3 matrix with eigenvalue λ and the associated eigenspace $E_{\lambda, \mathbf{A}}$ such that $\dim(E_{\lambda, \mathbf{A}}) = 2$.

B is another 3 x 3 matrix with eigenvalue μ and the associated eigenspace $E_{\mu, \mathbf{B}}$ such that $\dim(E_{\mu, \mathbf{B}}) = 2$.

Show that $\lambda + \mu$ is an eigenvalue of the matrix $\mathbf{A} + \mathbf{B}$.

★★★ In general, $\lambda + \mu$ is not an eigenvalue of $\mathbf{A} + \mathbf{B}$

$E_{\lambda, \mathbf{A}}$ is a subspace of \mathbf{R}^3 with dimension 2

$E_{\mu, \mathbf{B}}$ is a subspace of \mathbf{R}^3 with dimension 2

Then $E_{\lambda, \mathbf{A}} \cap E_{\mu, \mathbf{B}}$ is a subspace of \mathbf{R}^3 with dimension 1 or 2

The two eigenspaces have non-trivial intersection

Take $\mathbf{v} \in E_{\lambda, \mathbf{A}} \cap E_{\mu, \mathbf{B}}$

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

$$\mathbf{B}\mathbf{v} = \mu\mathbf{v}$$

$$\Rightarrow (\mathbf{A} + \mathbf{B})\mathbf{v} = (\lambda + \mu)\mathbf{v}$$

Exercise 6 Q26

Let $\mathbf{A} = \mathbf{A}^T$.

Suppose \mathbf{u} and \mathbf{v} are eigenvectors of \mathbf{A} associating with two different eigenvalues.

Show that \mathbf{u} and \mathbf{v} are orthogonal.

Let $\lambda \neq \mu$ be eigenvalues for \mathbf{u} and \mathbf{v}

$$\mathbf{A}\mathbf{u} = \lambda\mathbf{u} \quad \mathbf{A}\mathbf{v} = \mu\mathbf{v}$$

$$\mathbf{v} \cdot \mathbf{A}\mathbf{u} = \mathbf{A}\mathbf{u} \cdot \mathbf{v} = (\mathbf{A}\mathbf{u})^T \mathbf{v} = \mathbf{u}^T \mathbf{A}^T \mathbf{v} = \mathbf{u} \cdot \mathbf{A}^T \mathbf{v} = \mathbf{u} \cdot \mathbf{A}\mathbf{v}$$

$$\Downarrow$$

$$\mathbf{v} \cdot (\lambda\mathbf{u})$$

$$\Downarrow$$

$$\lambda(\mathbf{v} \cdot \mathbf{u})$$

$$\Downarrow$$

$$\mathbf{u} \cdot (\mu\mathbf{v})$$

$$\Downarrow$$

$$\mu(\mathbf{u} \cdot \mathbf{v})$$

$$\lambda(\mathbf{u} \cdot \mathbf{v}) = \mu(\mathbf{u} \cdot \mathbf{v}) \Leftrightarrow (\lambda - \mu)(\mathbf{u} \cdot \mathbf{v}) = 0 \Leftrightarrow \mathbf{u} \cdot \mathbf{v} = 0$$

Exercise 6 Q8 (Tutorial)

$\{ \mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n \}$ basis for \mathbf{R}^n and \mathbf{A} is an $n \times n$ matrix such that

$$\mathbf{A}\mathbf{u}_i = \mathbf{u}_{i+1} \text{ for } i = 1, 2, \dots, n-1 \text{ and } \mathbf{A}\mathbf{u}_n = \mathbf{0}$$

Show that the only eigenvalue of \mathbf{A} is 0 and find all the eigenvectors of \mathbf{A} .

Hint: Start with an eigenvalue λ and show $\lambda = 0$

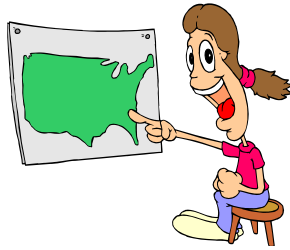
Suppose \mathbf{v} is an eigenvector of \mathbf{A} with eigenvalue λ

$$\mathbf{A}\mathbf{v} = \lambda\mathbf{v} \text{ ---} (*)$$

$$\mathbf{v} = c_1\mathbf{u}_1 + c_2\mathbf{u}_2 + \dots + c_n\mathbf{u}_n \text{ ---} (**)$$

Substitute $(**)$ into $(*)$ and use the given condition

Deduce $\lambda = 0$ by comparing the coefficients of the \mathbf{u}_i , and show $c_1 = c_2 = \dots = c_{n-1} = 0$



to be continued

Map of LA

A is an $n \times n$ matrix

A is invertible	chapter 2	A is not invertible
$\det A \neq 0$	chapter 2	$\det A = 0$
rref of A is identity matrix	chapter 1	rref of A has a zero row
AX= 0 has only the trivial solution	chapter 1	AX= 0 has non-trivial solutions
AX= B has a unique solution	chapter 1	AX= B has no solution or infinitely many solutions
rows (columns) of A are linearly independent	chapter 3	rows (columns) of A are linearly dependent
row (column) space of A = \mathbb{R}^n	chapter 4	row (column) space of A $\neq \mathbb{R}^n$
$\text{rank}(A) = n$ $\text{nullity}(A) = 0$	chapter 4	$\text{rank}(A) < n$ $\text{nullity}(A) > 0$
0 is not an eigenvalue of A	chapter 6	0 is an eigenvalue of A