

EECS 205003 Session 23

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Ch6 Eigenvalues and Eigenvectors

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6.1 Introduction to Eigenvalues

Eigenvalues & Eigenvectors

Eigenvalues: special numbers associated with a matrix

Eigenvectors: special vectors

Q : How special ?

Almost all vectors change direction when multiplied by A but

Eigenvectors \mathbf{x} are in the same direction as $A\mathbf{x}$

Def For an eigenvector of A (non-zero)

$A\mathbf{x} = \lambda\mathbf{x}$, λ : eigenvalue

(λ tells whether the special vector \mathbf{x} is stretched or shrunk or reversed or left unchanged)

6.1 Introduction to Eigenvalues

Eigenvalue 0

If eigenvalue $\lambda = 0$, then \exists nonzero \mathbf{x} such that $A\mathbf{x} = 0\mathbf{x} = \mathbf{0} \Rightarrow \mathbf{x}$ is in nullspace of A

\Rightarrow vectors of eigenvalue 0 makes up $N(A)$

If A is singular, then $\lambda = 0$ is an eigenvalue of A

(otherwise consider null space: $A\mathbf{x} = \mathbf{0} = 0\mathbf{x} \Rightarrow \mathbf{x} = \mathbf{0} \Rightarrow N(A) = \{\mathbf{0}\} \Rightarrow$ contradiction !)

Projection matrix P

Suppose P : projection onto a plane

For any vector on the plane, we have

$P\mathbf{x}_1 = \mathbf{x}_1 \Rightarrow \mathbf{x}_1$ is an eigenvector with eigenvalue 1

A vector \mathbf{x}_2 perpendicular to the plane $P\mathbf{x}_2 = \mathbf{0} \Rightarrow \mathbf{x}_2$ is an eigenvector with eigenvalue 0

(nonzero vector $\mathbf{x}_2 \in N(A) \Rightarrow A$ singular)

The eigenvectors of P spans the entire space (Not true for any matrix)

6.1 Introduction to Eigenvalues

$$\text{Ex: } P = \begin{bmatrix} 0.5 & 0.5 \\ 0.5 & 0.5 \end{bmatrix}$$

$$\lambda = 1 \Rightarrow P\mathbf{x} = \mathbf{x} \Rightarrow \mathbf{x} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$\lambda = 0 \Rightarrow P\mathbf{x} = 0 \Rightarrow \mathbf{x} = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

Note: Since $P = P^T$, eigenvectors are perpendicular (will prove this later)

6.1 Introduction to Eigenvalues

Ex: The reflection matrix $R = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$ has eigenvalues 1 & -1

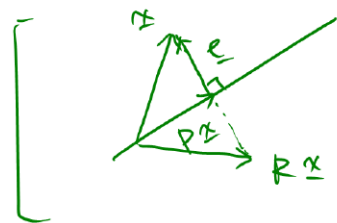
Recall: Eigenvectors for P : $\begin{bmatrix} 1 \\ 1 \end{bmatrix}$, $\begin{bmatrix} 1 \\ -1 \end{bmatrix}$

$$R\mathbf{x} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = 1 \cdot \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \lambda = 1$$

$$R\mathbf{x} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ -1 \end{bmatrix} = -1 \cdot \begin{bmatrix} 1 \\ -1 \end{bmatrix}, \lambda = -1$$

\Rightarrow same eigenvectors as P

Why? $R = 2P - I$



$$\left[\begin{aligned} \mathbf{e} &= (\mathbf{I} - P)\mathbf{x} \\ R\mathbf{x} &= \mathbf{x} - 2\mathbf{e} \\ &= \mathbf{x} - 2(\mathbf{I} - P)\mathbf{x} \\ &= (2P - \mathbf{I})\mathbf{x} \end{aligned} \right]$$

6.1 Introduction to Eigenvalues

If \mathbf{x} is an eigenvector of P

then $P\mathbf{x} = \lambda\mathbf{x} \Rightarrow 2P\mathbf{x} = 2\lambda\mathbf{x}$

$$\text{--)} \quad I\mathbf{x} = \mathbf{x}$$

$$(2P - I)\mathbf{x} = (2\lambda - 1)\mathbf{x}$$

$$\Rightarrow \quad R\mathbf{x} = (2\lambda - 1)\mathbf{x}$$

So same eigenvector for R but eigenvalue: $\lambda \rightarrow 2\lambda - 1$

$$\begin{bmatrix} 1 \\ 1 \end{bmatrix} : 2(1) - 1 = 1$$

$$\begin{bmatrix} 1 \\ -1 \end{bmatrix} : 2(0) - 1 = -1$$

6.1 Introduction to Eigenvalues

The equation for eigenvalues

An $n \times n$ matrix will have n eigenvalues $\lambda_1, \lambda_2, \dots, \lambda_n$

Solve $A\mathbf{x} = \lambda\mathbf{x}$ to obtain eigenvalues & eigenvectors

$$\Rightarrow (A - \lambda I)\mathbf{x} = \mathbf{0}$$

In order for \mathbf{x} to be an eigenvector, $A - \lambda I$ must be singular

$$\Rightarrow \det(A - \lambda I) = 0 \text{ (characteristic polynomial)}$$

(involves only λ , not \mathbf{x})

6.1 Introduction to Eigenvalues

To obtain eigenvectors

For each eigenvalue λ , solve $(A - \lambda I)\mathbf{x} = \mathbf{0}$ or $A\mathbf{x} = \lambda\mathbf{x}$
(in nullspace of $A - \lambda I$)

Ex: $A = \begin{bmatrix} 1 & 2 \\ 2 & 4 \end{bmatrix}$ (singular)

When A is singular, $\lambda = 0$ is one of eigenvalues

Since $A\mathbf{x} = \mathbf{0}\mathbf{x} = \mathbf{0}$ has solutions, vectors in $N(A)$ are eigenvectors

By eigenvalue equation,

$$\det(A - \lambda I) = \det \begin{bmatrix} 1 - \lambda & 2 \\ 2 & 4 - \lambda \end{bmatrix} = (1 - \lambda)(4 - \lambda) - 4$$

$$= \lambda^2 - 5\lambda = \lambda(\lambda - 5) = 0$$

$$\Rightarrow \lambda = 0 \text{ (as expected) or } \lambda = 5$$

6.1 Introduction to Eigenvalues

Now, find eigenvectors

$$(A - 0I)\mathbf{x} = \begin{bmatrix} 1 & 2 \\ 2 & 4 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} -2 \\ 1 \end{bmatrix} \text{ for } \lambda_1 = 0$$

$$(A - 5I)\mathbf{x} = \begin{bmatrix} -4 & 2 \\ 2 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \end{bmatrix} \text{ for } \lambda_2 = 5$$

(Matrix $A - 0I$ & $A - 5I$ are singular since $\lambda = 0, \lambda = 5$ are eigenvalues
 $(-2, 1), (1, 2)$ are in the nullspaces)

6.1 Introduction to Eigenvalues

Note: $\begin{bmatrix} 1 & 2 \\ 2 & 4 \end{bmatrix}$ has same eigenvector as $B = \begin{bmatrix} 0 & 2 \\ 2 & 3 \end{bmatrix}$

$$A\mathbf{x} = (B + I)\mathbf{x} = \lambda\mathbf{x} + \mathbf{x} = (\lambda + 1)\mathbf{x}$$

\Rightarrow eigenvalues of A are one plus eigenvalues of B
but eigenvectors stay the same

Bad news:

Elimination does not preserve λ 's

$$A = \begin{bmatrix} 1 & 2 \\ 2 & 4 \end{bmatrix} \text{ has } \lambda = 0, \lambda = 5$$

$$U = \begin{bmatrix} 1 & 2 \\ 0 & 0 \end{bmatrix} \text{ has } \lambda = 0, \lambda = 1$$

6.1 Introduction to Eigenvalues

Fact Eigenvalues of U sit on its diagonal (pivots)

Recall: $\det U = u_{11} \cdots u_{nn}$

so $\det(U - \lambda I) = (u_{11} - \lambda) \cdots (u_{nn} - \lambda) = 0$

$\Rightarrow \lambda = u_{11}, \lambda = u_{22}, \dots, \lambda = u_{nn}$

Eigenvalues are changed during row operations !

Good news: When A is $n \times n$,

(1) $\lambda_1 + \lambda_2 + \cdots + \lambda_n = a_{11} + a_{22} + \cdots + a_{nn} = \text{trace}(A)$

For 2×2 : $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$

$\det(A - \lambda I) = \lambda^2 - (a + d)\lambda + ad - bc$

$= \lambda^2 - (\text{trace} A)\lambda + \det A$

6.1 Introduction to Eigenvalues

In general, $\det(A - \lambda I) = (\lambda_1 - \lambda) \cdots (\lambda_n - \lambda)$

from LHS, check coefficient for λ^{n-1}

$$\begin{vmatrix} a_{11} - \lambda & \cdots & \cdots & a_{1n} \\ \vdots & a_{22} - \lambda & & \vdots \\ \vdots & & \ddots & \vdots \\ a_{n1} & \cdots & \cdots & a_{nn} - \lambda \end{vmatrix}$$

$$= (a_{11} - \lambda)C_{11} + a_{12}C_{12} + \cdots + a_{1n}C_{1n}$$

for C_{12} , first row & 2nd column are crossed out $\Rightarrow (a_{11} - \lambda), (a_{22} - \lambda)$
are crossed out \Rightarrow degree at most λ^{n-2}

Similarly for $C_{1j}, j \neq 1$

So λ^{n-1} comes from $(a_{11} - \lambda) \cdots (a_{nn} - \lambda)$

\Rightarrow coefficient for $\lambda^{n-1} = (-1)^n \text{trace} A$

from RHS,

$(\lambda_1 - \lambda) \cdots (\lambda_n - \lambda) \Rightarrow$ coefficient for $\lambda^{n-1} = (-1)^n (\lambda_1 + \cdots + \lambda_n)$

$\Rightarrow \lambda_1 + \cdots + \lambda_n = \text{trace} A$

6.1 Introduction to Eigenvalues

$$(2) \det(A) = \lambda_1 \lambda_2 \cdots \lambda_n$$

$$\det(A - \lambda I) = (\lambda_1 - \lambda) \cdots (\lambda_n - \lambda) = 0$$

(polynomial of degree n)

Let $\lambda = 0$, we have $\det(A) = \lambda_1 \lambda_2 \cdots \lambda_n$

A caution:

$$\text{If } A\mathbf{x} = \lambda\mathbf{x}, \quad B\mathbf{x} = \alpha\mathbf{x}$$

$$\Rightarrow (A + B)\mathbf{x} = (\lambda + \alpha)\mathbf{x}$$

So $A + B$ has eigenvalue $\lambda + \alpha$?

Not really !

Only true when A & B have the same eigenvectors

Similarly, eigenvalues of $AB \neq \lambda(A)\lambda(B)$

6.1 Introduction to Eigenvalues

Complex eigenvalues

The matrix $Q = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$ rotates every vector by 90°

trace = $0 = \lambda_1 + \lambda_2$, determinant = $1 = \lambda_1 \lambda_2$

The only real eigenvector is $\mathbf{0}$ since any other vector changes direction when multiplied by Q

$$\det(A - \lambda I) = \begin{vmatrix} -\lambda & -1 \\ 1 & -\lambda \end{vmatrix} = \lambda^2 + 1 = 0 \Rightarrow \lambda = i, -i$$

Note: If $a + bi$ is an eigenvalue $\Rightarrow a - bi$ is also eigenvalue

Note: symmetric matrices have Real eigenvalues

anti-symmetric matrices have Imaginary eigenvalues

$$(A^T = -A, \text{ like } Q)$$

6.1 Introduction to Eigenvalues

Triangular matrix & repeated eigenvalues

$$A = \begin{bmatrix} 3 & 1 \\ 0 & 3 \end{bmatrix}, \lambda_1 = 3, \lambda_2 = 3$$

To find eigenvectors,

$$(A - 3I)\mathbf{x} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \mathbf{x} = \mathbf{0}$$

$$\Rightarrow \mathbf{x}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \text{ there is NO independent eigenvector } \mathbf{x}_2$$