

# Eigenvalue Problems

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December 20, 2016

## Theorem

For  $B \in M_2(\mathbf{R})$ , the following conditions are equivalent:

- **(i)**  $B$  is regular.
- **(ii)**  $B\vec{v} = \vec{0} \Rightarrow \vec{v} = \vec{0}$
- **(iii)**  $|B| \neq 0$

## Theorem

For  $B \in M_2(\mathbf{R})$ , the following conditions are equivalent:

- **NOT(i)**  $B$  is not regular (singular).
- **NOT(ii)** There exists  $\vec{v} \neq \vec{0}$  satisfying  $B\vec{v} = \vec{0}$ .
- **NOT(iii)**  $|B| = 0$

## Example (1)

Let  $A = \begin{pmatrix} 1 & 2 \\ 4 & 3 \end{pmatrix}$ . We consider the system of equations:

$$A \begin{pmatrix} x \\ y \end{pmatrix} = \lambda \begin{pmatrix} x \\ y \end{pmatrix}$$

First remark that the system is equivalent to

$$(\#) \quad (\lambda I_2 - A) \begin{pmatrix} x \\ y \end{pmatrix} = \vec{0}, \quad \text{where} \quad \lambda I_2 - A = \begin{pmatrix} \lambda - 1 & -2 \\ -4 & \lambda - 3 \end{pmatrix}$$

We also remark that

$$|\lambda I_2 - A| = \begin{vmatrix} \lambda - 1 & -2 \\ -4 & \lambda - 3 \end{vmatrix} = (\lambda + 1)(\lambda - 5)$$

## Example (2)

In case  $\lambda \neq -1, 5$   $(\#) \Leftrightarrow \begin{pmatrix} x \\ y \end{pmatrix} = \vec{0}$

In case  $\lambda = -1$

$$(\#) \Leftrightarrow \begin{pmatrix} -2 & -2 \\ -4 & -4 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \vec{0} \Leftrightarrow x + y = 0$$

We put  $y = t$  to get the solution

$$\begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} -t \\ t \end{pmatrix} = t \begin{pmatrix} -1 \\ 1 \end{pmatrix}$$

## Example (3)

In case  $\lambda = 5$

$$(\#) \Leftrightarrow \begin{pmatrix} 4 & -2 \\ -4 & 2 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \vec{0} \Leftrightarrow 2x - y = 0$$

We put  $y = t$  to get the solution

$$\begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} t \\ 2t \end{pmatrix} = t \begin{pmatrix} 2 \\ 1 \end{pmatrix}$$

# Diagonalization (1)

We define two vectors

$$\vec{p}_1 = \begin{pmatrix} -1 \\ 1 \end{pmatrix}, \quad \vec{p}_2 = \begin{pmatrix} 1 \\ 2 \end{pmatrix}$$

and a  $2 \times 2$  matrix

$$P = (\vec{p}_1 \ \vec{p}_2) = \begin{pmatrix} -1 & 1 \\ 1 & 2 \end{pmatrix}$$

The matrix  $P$  is regular since

$$|P| = \begin{vmatrix} -1 & 1 \\ 1 & 2 \end{vmatrix} = -3 \neq 0$$

## Diagonalization (2)

$$\begin{aligned}AP &= A(\vec{p}_1 \ \vec{p}_2) = (A\vec{p}_1 \ A\vec{p}_2) \\&= (-\vec{p}_1 \ 5\vec{p}_2) \\&= (\vec{p}_1 \ \vec{p}_2) \begin{pmatrix} -1 & 0 \\ 0 & 5 \end{pmatrix} = P \begin{pmatrix} -1 & 0 \\ 0 & 5 \end{pmatrix}\end{aligned}$$

We multiply  $P^{-1}$  from the left to get

$$P^{-1}AP = \begin{pmatrix} -1 & 0 \\ 0 & 5 \end{pmatrix}$$

This process is called a diagonalization of  $A$ . To see how it dows mean, we introduce the coordinate transform given by

$$\begin{pmatrix} x \\ y \end{pmatrix} = \xi \vec{p}_1 + \eta \vec{p}_2 = (\vec{p}_1 \ \vec{p}_2) \begin{pmatrix} \xi \\ \eta \end{pmatrix} = P \begin{pmatrix} \xi \\ \eta \end{pmatrix}$$

# Diagonalization (3)

In this situation the map

$$\begin{pmatrix} x' \\ y' \end{pmatrix} = A \begin{pmatrix} x \\ y \end{pmatrix}$$

can be clarified by using the coordinate transform as follows.

$$\begin{aligned} \begin{pmatrix} \xi' \\ \eta' \end{pmatrix} &= P^{-1} \begin{pmatrix} x' \\ y' \end{pmatrix} \\ &= P^{-1} A \begin{pmatrix} x \\ y \end{pmatrix} \\ &= P^{-1} A P \begin{pmatrix} \xi \\ \eta \end{pmatrix} = \begin{pmatrix} -1 & 0 \\ 0 & 5 \end{pmatrix} \begin{pmatrix} \xi \\ \eta \end{pmatrix} \end{aligned}$$

# Definitions

Let  $A$  be a  $2 \times 2$  matrix:

$$A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$$

Then  $\alpha \in \mathbf{R}$  is called an eigenvalue of  $A$  if there exists  $\begin{pmatrix} x \\ y \end{pmatrix} \neq \vec{0}$  satisfying

$$(\#) \quad A \begin{pmatrix} x \\ y \end{pmatrix} = \alpha \begin{pmatrix} x \\ y \end{pmatrix}$$

In this situation  $\begin{pmatrix} x \\ y \end{pmatrix} \neq \vec{0}$  satisfying  $(\#)$  is called an eigen vector of  $A$  for the eigenvalue  $\alpha$ .

Remark that

$$(\#) \quad A \begin{pmatrix} x \\ y \end{pmatrix} = \alpha \begin{pmatrix} x \\ y \end{pmatrix} \Leftrightarrow (\#)' \quad (\alpha I_2 - A) \begin{pmatrix} x \\ y \end{pmatrix} = \vec{0}$$

and that

$$B\vec{v} = \vec{0} \quad \text{for some } \vec{v} \neq \vec{0} \quad \Leftrightarrow \quad |B| = 0.$$

Accordingly it follows that

$$\alpha \text{ is an eigen value of } A \quad \Leftrightarrow \quad |\alpha I_2 - A| = 0$$

# Eigenpolynomial

We define the eigenpolynomial of  $A$  by

$$\begin{aligned}\Phi_A(\lambda) &:= |\lambda I_2 - A| = \begin{vmatrix} \lambda - a & -b \\ -c & \lambda - d \end{vmatrix} \\ &= \lambda^2 - (a + d)\lambda + ad - bc\end{aligned}$$

Remark that

$$\alpha \text{ is an eigenvalue of } A \quad \Leftrightarrow \quad \Phi_A(\alpha) = 0$$

# Theorem

## Theorem

We assume that  $\Phi_A(\lambda)$  is factorized by

$$\Phi_A(\lambda) = (\lambda - \alpha)(\lambda - \beta)$$

with  $\alpha, \beta \in \mathbf{R}$  satisfying  $\alpha \neq \beta$ . Moreover we assume that the two vectors  $\vec{p}_1, \vec{p}_2$  satisfy

$$A\vec{p}_1 = \alpha\vec{p}_1, \quad A\vec{p}_2 = \beta\vec{p}_2$$

with  $\vec{p}_1 \neq \vec{0}, \vec{p}_2 \neq \vec{0}$ . Then

$$P = (\vec{p}_1 \ \vec{p}_2)$$

is regular.

# Proof

Remark that

$$P \text{ is regular} \Leftrightarrow \left( c_1 \vec{p}_1 + c_2 \vec{p}_2 = \vec{0} \Rightarrow c_1 = c_2 = 0 \right).$$

We assume that

$$c_1 \vec{p}_1 + c_2 \vec{p}_2 = \vec{0}$$

and multiply the both hand sides by  $(\beta I_2 - A)$  to get

$$c_1(\beta - \alpha)\vec{p}_1 = \vec{0}$$

It follows from  $\beta - \alpha \neq 0$  and  $\vec{p}_1 \neq \vec{0}$  that

$$c_1 = 0$$

In this situation we get

$$c_2 \vec{p}_2 = \vec{0}$$

It follows from  $\vec{p}_2 \neq \vec{0}$  that

$$c_2 = 0.$$

## Example (1)

Let  $A = \begin{pmatrix} 2 & 2 \\ 2 & 5 \end{pmatrix}$ . Then

$$\Phi_A(\lambda) = \begin{vmatrix} \lambda - 2 & -2 \\ -2 & \lambda - 5 \end{vmatrix} = (\lambda - 1)(\lambda - 6)$$

Thus the eigenvalues of  $A$  are  $\lambda = 1, 6$ . We find the eigenvectors of  $A$  as follows.

In case  $\lambda = 1$

$$A \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} x \\ y \end{pmatrix} \Leftrightarrow \begin{pmatrix} -1 & -2 \\ -2 & -4 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \vec{0} \Leftrightarrow x + 2y = 0$$

Thus

$$\begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} -2y \\ y \end{pmatrix} = y \begin{pmatrix} -2 \\ 1 \end{pmatrix} \quad (y \neq 0)$$

are the eigenvectors of  $A$  for the eigenvalue  $\lambda = 1$ .

## Example (2)

In case  $\lambda = 6$

$$A \begin{pmatrix} x \\ y \end{pmatrix} = 6 \begin{pmatrix} x \\ y \end{pmatrix} \Leftrightarrow \begin{pmatrix} 4 & -2 \\ -2 & 1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \vec{0} \Leftrightarrow 2x - y = 0.$$

Thus

$$\begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} x \\ 2x \end{pmatrix} = x \begin{pmatrix} 1 \\ 2 \end{pmatrix} \quad (x \neq 0)$$

are the eigenvectors of  $A$  for the eigenvalue  $\lambda = 6$ .

## Example (3)

We choose two vectors  $\vec{q}_1 = \begin{pmatrix} -2 \\ 1 \end{pmatrix}$ ,  $\vec{q}_2 = \begin{pmatrix} 1 \\ 2 \end{pmatrix}$  and also a matrix

$$Q = (\vec{q}_1 \ \vec{q}_2) = \begin{pmatrix} -2 & 1 \\ 1 & 2 \end{pmatrix}$$

The  $Q$  is regular by the Theorem proven above and

$$\begin{aligned} AQ &= (A\vec{q}_1 \ A\vec{q}_2) = (\vec{q}_1 \ 6\vec{q}_2) \\ &= (\vec{q}_1 \ \vec{q}_2) \begin{pmatrix} 1 & 0 \\ 0 & 6 \end{pmatrix} = Q \begin{pmatrix} 1 & 0 \\ 0 & 6 \end{pmatrix} \end{aligned}$$

Thus we can diagonalize  $A$  by

$$Q^{-1}AQ = \begin{pmatrix} 1 & 0 \\ 0 & 6 \end{pmatrix}$$

# Symmetric matrices

Remark that

$$\vec{q}_1 \perp \vec{q}_2$$

This is not a coincidence. Actually we have the following theorem.

## Theorem

Let  $A$  be a  $2 \times 2$  matrix and assume that  $A$  is symmetric *i.e.*  ${}^tA = A$ . Moreover we assume that

$$A\vec{p} = \alpha\vec{p}, \quad A\vec{q} = \beta\vec{q}$$

with  $\alpha, \beta \in \mathbf{R}$  satisfying  $\alpha \neq \beta$ . Then

$$\vec{p} \perp \vec{q}$$

$$(A\vec{p}, \vec{q}) = (\vec{p}, {}^tA\vec{q}) = (\vec{p}, A\vec{q})$$

Moreover

$$(A\vec{p}, \vec{q}) = (\alpha\vec{p}, \vec{q}) = \alpha(\vec{p}, \vec{q})$$

$$(\vec{p}, A\vec{q}) = (\vec{p}, \beta\vec{q}) = \beta(\vec{p}, \vec{q})$$

These two identities lead us to

$$(\alpha - \beta)(\vec{p}, \vec{q}) = 0$$

# Why do we need symmetric matrices?

We are given a symmetric matrix  $A = \begin{pmatrix} a & c \\ c & b \end{pmatrix}$ . We introduce the quadratic form for  $A$  by

$$\left( A \begin{pmatrix} x \\ y \end{pmatrix}, \begin{pmatrix} x \\ y \end{pmatrix} \right) = ax^2 + 2cxy + by^2$$

In case  $A = \begin{pmatrix} 2 & 2 \\ 2 & 5 \end{pmatrix}$ , we have

$$\left( A \begin{pmatrix} x \\ y \end{pmatrix}, \begin{pmatrix} x \\ y \end{pmatrix} \right) = 2x^2 + 4xy + 5y^2$$

We need a more detailed analysis about the diagonalization of symmetric matrices for its application.

# Rotation Matrices(1)

We define the rotation matrix of the angle  $\theta$  by

$$R_\theta = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix}$$

The rotation can be figured out by the identity

$$R_\theta \begin{pmatrix} \cos \alpha \\ \sin \alpha \end{pmatrix} = \begin{pmatrix} \cos(\alpha + \theta) \\ \sin(\alpha + \theta) \end{pmatrix}$$

# Rotation matrices

Moreover we find that

$$R_{\theta}^{-1} = \frac{1}{|R_{\theta}|} \begin{pmatrix} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{pmatrix} = \begin{pmatrix} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{pmatrix} = {}^t R_{\theta}$$

Accordingly

$${}^t R_{\theta} \cdot R_{\theta} = R_{\theta} \cdot {}^t R_{\theta} = I_2$$

This identity leads us to

$$(R_{\theta} \vec{v}, R_{\theta} \vec{w}) = (\vec{v}, \vec{w})$$

for any  $\vec{v}, \vec{w} \in \mathbf{R}^2$ . In fact

$$LHS = (\vec{v}, {}^t R_{\theta} R_{\theta} \vec{w}) = (\vec{v}, I_2 \vec{w}) = (\vec{v}, \vec{w})$$

# Diagonalization of symmetric matrices

Let us go back to the example  $A = \begin{pmatrix} 2 & 2 \\ 2 & 5 \end{pmatrix}$ . We can diagonalize  $A$  by a rotation matrix chosen as folloes. We define two unit vectors by

$$\vec{r}_1 = \frac{1}{\sqrt{5}} \begin{pmatrix} 2 \\ -1 \end{pmatrix}, \quad \vec{r}_2 = \frac{1}{\sqrt{5}} \begin{pmatrix} 1 \\ 2 \end{pmatrix}$$

Then  $R = (\vec{r}_1 \ \vec{r}_2)$  is a rotation matrix and we have

$$A\vec{r}_1 = \vec{r}_1, \quad A\vec{r}_2 = 6\vec{r}_2$$

Accordingly

$$AR = (A\vec{r}_1 \ A\vec{r}_2) = (\vec{r}_1 \ 6\vec{r}_2) = (\vec{r}_1 \ \vec{r}_2) \begin{pmatrix} 1 & 0 \\ 0 & 6 \end{pmatrix} = R \begin{pmatrix} 1 & 0 \\ 0 & 6 \end{pmatrix}$$

Thus we have deduced  $R^{-1}AR = \begin{pmatrix} 1 & 0 \\ 0 & 6 \end{pmatrix}$

# Diagonalization of symmetric matrices

We now consider the quadratic form of  $A$ . Remark that  $R^{-1}$  is a rotation. It follows from this fact that

$$\begin{aligned}\left(A \begin{pmatrix} x \\ y \end{pmatrix}, \begin{pmatrix} x \\ y \end{pmatrix}\right) &= \left(R^{-1}A \begin{pmatrix} x \\ y \end{pmatrix}, R^{-1} \begin{pmatrix} x \\ y \end{pmatrix}\right) \\ &= \left(R^{-1}AR \cdot R^{-1} \begin{pmatrix} x \\ y \end{pmatrix}, R^{-1} \begin{pmatrix} x \\ y \end{pmatrix}\right) \\ &= \left(\begin{pmatrix} 1 & 0 \\ 0 & 6 \end{pmatrix} \begin{pmatrix} \xi \\ \eta \end{pmatrix}, \begin{pmatrix} \xi \\ \eta \end{pmatrix}\right) = \xi^2 + 6\eta^2\end{aligned}$$

Here we used a rotational coordinate transformation defined by

$$\begin{pmatrix} \xi \\ \eta \end{pmatrix} = R^{-1} \begin{pmatrix} x \\ y \end{pmatrix}$$