Matrix theory - Challenge

Shreeprasad Bhat AI20MTECH14011

Abstract—This document illustrates proving properties If $\lambda_1 \neq \lambda_2$, then it follows that of traingle using linear algebra

Download all latex-tikz from

https://github.com/shreeprasadbhat/matrix-theory/ blob/master//challenging problems/ challenge 03 10 2020/

1 Problem

In conic sections you have seen that

 $V = PDP^{T}$, with $P^{T}P = I$. So P is an orthogonal matrix. For what matrices V do you get this kind of decomposition where P is an orthogonal? Can you prove this result?

2 ANSWER

We get this kind of result for real symmetric matrix. Any real symmetric matrix can always be decomposed as $V = PDP^{T}$, with $P^{T}P = I$.

3 PROOF

Let V be an $n \times n$ real symmetric matrix. Its characteristic equation is

$$\lambda^2 - (a+b)\lambda + (ab - h^2) = 0 {(3.0.1)}$$

So the eigenvalues are given by

$$\frac{(a+b) \pm \sqrt{(a-b)^2 + 4h^2}}{2}$$
 (3.0.2)

These eigenvalues are equal only if

$$(a-b)^2 + 4h^2 = 0 \implies a = b, h = 0$$
 (3.0.3)

Let λ_1 and λ_2 be two eigenvalue of V. Let **u** and **v** be two corresponding eigenvectors. We have

$$\lambda_1(\mathbf{u}^T \mathbf{v}) = (\lambda_1 \mathbf{u})^T \mathbf{v}$$

$$= (V\mathbf{u})^T \mathbf{v}$$
(3.0.4)
$$= (3.0.5)$$

$$= \mathbf{u}^T(\mathbf{V}\mathbf{v}) \tag{3.0.6}$$

$$= \mathbf{u}^T (\lambda_2 \mathbf{v}) \tag{3.0.7}$$

$$= \lambda_2(\mathbf{u}^T \mathbf{v}) \tag{3.0.8}$$

$$\mathbf{u}^T \mathbf{v} = 0 \tag{3.0.9}$$

If $\lambda_1 = \lambda_2$, then from (3.0.3),

$$\mathbf{V} = \begin{pmatrix} a & 0 \\ 0 & a \end{pmatrix} \implies \mathbf{V} = aI \tag{3.0.10}$$

Thus, for real symmetric matrix eigenvectors will be orthogonal.

Let **P** be eigenvector matrix, in which very two eigenvectors are orthogonal

$$\mathbf{P} = \begin{pmatrix} | & & | \\ \mathbf{v_1} & \cdots & \mathbf{v_n} \\ | & & | \end{pmatrix}$$
 (3.0.11)

By definition of orthogonal matrix, we have,

$$\mathbf{P}\mathbf{P}^{\mathrm{T}} = \mathbf{P}^{T}\mathbf{P} = I \tag{3.0.12}$$

Consider,

$$\mathbf{VP} = \begin{pmatrix} | & & | \\ \mathbf{Vv_1} & \cdots & \mathbf{Vv_n} \\ | & & | \end{pmatrix}$$
 (3.0.13)

$$= \begin{pmatrix} | & | \\ \lambda_1 \mathbf{v_1} & \cdots & \lambda_n \mathbf{V} \mathbf{v_n} \end{pmatrix}$$
 (3.0.14)

$$= \begin{pmatrix} | & | & | \\ \mathbf{v_1} & \cdots & \mathbf{V}\mathbf{v_n} | \end{pmatrix} \begin{pmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \ddots & \vdots \\ \vdots & \ddots & \ddots & 0 \\ 0 & 0 & \cdots & \lambda_n \end{pmatrix}$$
(3.0.15)

VP = PD(3.0.16)

$$\mathbf{VPP}^T = \mathbf{PDP}^T \tag{3.0.17}$$

$$\mathbf{V} = \mathbf{P}\mathbf{D}\mathbf{P}^T \tag{3.0.18}$$

Hence proved the result.