

# MA 265 Final Exam

- Thursday, 8/2
- 3:30pm-5:30pm
- FRNY G140
  
- 20 multiple choice problems
- Covers all lessons
  - no special emphasis on material since exam 2

## 7.1 Diagonalization of Symmetric Matrices (continued)

If  $A$  is symmetric ( $A^T = A$ ) then  $A = PDP^{-1} = PDP^T$

where  $P$  is an orthogonal matrix whose columns are orthonormal eigenvectors of  $A$  and  $D$  is a diagonal matrix with the eigenvalues on the main diagonal.

example

$$A = \begin{bmatrix} 2 & -1 & 1 \\ -1 & 2 & -1 \\ 1 & -1 & 2 \end{bmatrix}$$

$$\lambda = 1, 4$$

↙ repeated  
1, 1, 4

$$\underline{\lambda = 1} \quad (A - \lambda I)\vec{v} = \vec{0}$$

$$\begin{bmatrix} 1 & -1 & 1 & 0 \\ -1 & 1 & -1 & 0 \\ 1 & -1 & 1 & 0 \end{bmatrix} \sim \begin{bmatrix} 1 & -1 & 1 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

$$x_2 \text{ free} \quad x_3 \text{ free} \quad x_1 = x_2 - x_3$$

$$\vec{x} = x_2 \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} + x_3 \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix} \quad \vec{v}_1 = \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} \quad \vec{v}_2 = \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$$

$$\lambda = 4$$

$$\begin{bmatrix} -2 & -1 & 1 & 0 \\ -1 & -2 & -1 & 0 \\ 1 & -1 & -2 & 0 \end{bmatrix} \sim \dots \quad \vec{v}_3 = \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}$$

normalize them

$$\vec{u}_1 = \frac{\vec{v}_1}{\|\vec{v}_1\|} = \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \\ 0 \end{bmatrix}$$

$$\vec{u}_2 = \frac{\vec{v}_2}{\|\vec{v}_2\|} = \begin{bmatrix} -1/\sqrt{2} \\ 0 \\ 1/\sqrt{2} \end{bmatrix}$$

$$\vec{u}_3 = \frac{\vec{v}_3}{\|\vec{v}_3\|} = \begin{bmatrix} 1/\sqrt{3} \\ -1/\sqrt{3} \\ 1/\sqrt{3} \end{bmatrix}$$

make  $\vec{u}_2$  orthogonal to the rest

$$\begin{aligned} \vec{w}_2 &= \vec{u}_2 - \langle \vec{u}_2, \vec{u}_1 \rangle \vec{u}_1 - \langle \vec{u}_2, \vec{u}_3 \rangle \vec{u}_3 \\ &= \begin{bmatrix} -1/\sqrt{2} \\ 0 \\ 1/\sqrt{2} \end{bmatrix} - \frac{1}{2} \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \\ 0 \end{bmatrix} = \begin{bmatrix} -3/2\sqrt{2} \\ -1/2\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix} \end{aligned}$$

normalize it :  $\vec{w}_2 = \dots = \begin{bmatrix} 1/\sqrt{6} \\ 2/\sqrt{6} \\ 1/\sqrt{6} \end{bmatrix}$

$$P = \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{3} & 1/\sqrt{6} \\ 1/\sqrt{2} & -1/\sqrt{3} & 2/\sqrt{6} \\ 0 & 1/\sqrt{3} & 1/\sqrt{6} \end{bmatrix} \quad D = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 4 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Some properties of symmetric matrices

If  $A$  is symmetric, then  $(A\vec{x}) \cdot \vec{y} = \vec{x} \cdot (A\vec{y})$

why? If  $A$  is symmetric, then  $A = A^T$

$$\begin{aligned} (A\vec{x}) \cdot \vec{y} &= (A\vec{x})^T \vec{y} = \vec{x}^T A^T \vec{y} \\ &= \vec{x}^T (A\vec{y}) \\ &= \vec{x} \cdot (A\vec{y}) \end{aligned}$$

If  $A$  is symmetric,  $A^2$  is symmetric

why?  $A = A^T$

$$(A^2)^T = (AA)^T = A^T A^T = (A^T)^2 = A^2$$

what about  $A^3$ ?

$$(A^3)^T = (AA^2)^T = (A^2)^T A^T = A^2 A = A^3$$

If  $A$  is orthogonally diagonalizable and invertible,  
then  $A^{-1}$  is also orthogonally diagonalizable

$$A = P D P^{-1} = P D P^T$$

$$A^{-1} = (P D P^{-1})^{-1} = (P^{-1})^{-1} D^{-1} P^{-1}$$

$$= P D^{-1} P^{-1}$$

also diagonal  
e.g.  $D = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{bmatrix}$

$$\begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 2 & 0 & 0 & 0 \\ 0 & 0 & 3 & 0 & 0 \end{bmatrix}$$

$$\sim \begin{bmatrix} 1 & 0 & 0 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 & 1/2 & 0 \\ 0 & 0 & 1 & 0 & 0 & 1/3 \end{bmatrix}$$

$D^{-1}$

## Spectral Decomposition

the set of eigenvalues of  $A$  is called the spectrum of  $A$ .

If  $A$  is orthogonally diagonalizable, then

$$A = P D P^T$$

$$= \begin{bmatrix} \vec{u}_1 & \vec{u}_2 & \dots & \vec{u}_n \end{bmatrix} \begin{bmatrix} \lambda_1 & & & 0 \\ & \ddots & & \\ 0 & & & \lambda_n \end{bmatrix} \begin{bmatrix} \vec{u}_1^T \\ \vdots \\ \vec{u}_n^T \end{bmatrix}$$

$$= \begin{bmatrix} \lambda_1 \vec{u}_1 & \lambda_2 \vec{u}_2 & \dots & \lambda_n \vec{u}_n \end{bmatrix} \begin{bmatrix} \vec{u}_1^T \\ \vdots \\ \vec{u}_n^T \end{bmatrix}$$

$$= \lambda_1 \vec{u}_1 \vec{u}_1^T + \lambda_2 \vec{u}_2 \vec{u}_2^T + \dots + \lambda_n \vec{u}_n \vec{u}_n^T$$

if  $A$  is  $n \times n$   
 then  $\vec{u}_i$  is  $n \times 1$   
 and  $\vec{u}_i \vec{u}_i^T$  is  $n \times n$

$$A\vec{x} = \lambda_1 \underbrace{\vec{u}_1 \vec{u}_1^T}_{\text{projection of } \vec{x} \text{ onto subspace spanned by } \vec{u}_1} \vec{x} + \lambda_2 \vec{u}_2 \vec{u}_2^T \vec{x} + \dots + \lambda_n \vec{u}_n \vec{u}_n^T \vec{x}$$

projection  
of  $\vec{x}$  onto  
subspace  
spanned by  $\vec{u}_1$

example

$$A = \begin{bmatrix} 7 & 4 & -4 \\ 4 & 5 & 0 \\ -4 & 0 & 9 \end{bmatrix}$$

$$\lambda_1 = 13, \quad \vec{u}_1 = \begin{bmatrix} -2/3 \\ -1/3 \\ 2/3 \end{bmatrix}$$

$$\lambda_2 = 7, \quad \vec{u}_2 = \begin{bmatrix} 1/3 \\ 2/3 \\ 2/3 \end{bmatrix}$$

$$\lambda_3 = 1, \quad \vec{u}_3 = \begin{bmatrix} -2/3 \\ 2/3 \\ -1/3 \end{bmatrix}$$

$$P = \begin{bmatrix} \vec{u}_1 & \vec{u}_2 & \vec{u}_3 \end{bmatrix}$$

$$D = \begin{bmatrix} 13 & & 0 \\ 0 & 7 & \\ 0 & & 1 \end{bmatrix}$$

$$\begin{aligned}
 A &= PDP^T = \lambda_1 \vec{u}_1 \vec{u}_1^T + \lambda_2 \vec{u}_2 \vec{u}_2^T + \lambda_3 \vec{u}_3 \vec{u}_3^T \\
 &= 13 \begin{bmatrix} 4/9 & 2/9 & -4/9 \\ 2/9 & 4/9 & -2/9 \\ -4/9 & -2/9 & 4/9 \end{bmatrix} + 7 \begin{bmatrix} 1/9 & 2/9 & 2/9 \\ 2/9 & 4/9 & 4/9 \\ 2/9 & 4/9 & 4/9 \end{bmatrix} + 1 \begin{bmatrix} 4/9 & -4/9 & 2/9 \\ -4/9 & 4/9 & -2/9 \\ 2/9 & -2/9 & 4/9 \end{bmatrix}
 \end{aligned}$$