

南京航空航天大学

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2012 ~ 2013 学年 《 Matrix Theory 》 期中考试试题

考试日期： 2012 年 11 月 试卷类型： 试卷代号：

学院						学号						姓名					
题号	1	2	3	4	5	6	7	8	9	10	总分						
得分																	

参考答案

Part I (必做题, 70 分)

#1. For the given matrix

$$A = \begin{pmatrix} 1 & 3 & -2 & 1 \\ 2 & 1 & 3 & 2 \\ 3 & 4 & 5 & 6 \end{pmatrix}$$

- (1) find the reduced row echelon form (简化行阶梯形) of A ;
- (2) find a basis for the row space of A ;
- (3) find a basis for the column space of A ;
- (4) find a basis for the nullspace of A .

Solution:

$$(1) \quad A = \begin{pmatrix} 1 & 3 & -2 & 1 \\ 2 & 1 & 3 & 2 \\ 3 & 4 & 5 & 6 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 3 & -2 & 1 \\ 0 & -5 & 7 & 0 \\ 0 & 0 & 4 & 3 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 3 & 0 & 5/2 \\ 0 & 1 & 0 & 21/20 \\ 0 & 0 & 1 & 3/4 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 0 & -13/20 \\ 0 & 1 & 0 & 21/20 \\ 0 & 0 & 1 & 3/4 \end{pmatrix} = U$$

(2) $(1, 0, 0, -13/20)$, $(0, 1, 0, 21/20)$, $(0, 0, 1, 3/4)$ form a basis for the row space of A , since row elementary operations do not change the row space of A .

(3) The 1st, 2nd, and 3rd column vectors of U are linearly independent. The 4th column vector is a linear combination of the first three column vectors. Hence, $(1, 2, 3)^T$, $(3, 1, 4)^T$, $(-2, 3, 5)^T$ form a basis for the column space of A .

(4) Solving the system $Ax = \mathbf{0}$, we obtain a basis $(13, -21, -15, 20)^T$ for the nullspace of A .

#2. Let $\mathbf{e}_1, \mathbf{e}_2$ be a basis for \mathbf{R}^2 and $\mathbf{b}_1, \mathbf{b}_2, \mathbf{b}_3$ be a basis for \mathbf{R}^3 ,

where $(\mathbf{e}_1, \mathbf{e}_2) = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$, $(\mathbf{b}_1, \mathbf{b}_2, \mathbf{b}_3) = \begin{pmatrix} 1 & 1 & 0 \\ 1 & 0 & 1 \\ 0 & 1 & 1 \end{pmatrix}$.

Let $L: \mathbf{R}^2 \rightarrow \mathbf{R}^3$ be the linear transformation defined by

$$L(\mathbf{x}) = x_1 \mathbf{b}_1 + x_2 \mathbf{b}_2 + (x_1 + x_2) \mathbf{b}_3.$$

Find the matrix A representing L with respect to the bases $[\mathbf{e}_1, \mathbf{e}_2]$ and $[\mathbf{b}_1, \mathbf{b}_2, \mathbf{b}_3]$.

Solution:

$$L(\mathbf{e}_1) = \mathbf{b}_1 + \mathbf{b}_3$$

$$L(\mathbf{e}_2) = \mathbf{b}_2 + \mathbf{b}_3$$

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

#3. Let \mathbf{S} be the two dimensional subspace of \mathbf{R}^3 spanned by

$$\mathbf{u}_1 = (1, 0, 2)^T \text{ and } \mathbf{u}_2 = (0, 1, -2)^T$$

Find a basis for \mathbf{S}^\perp and determine the projection matrix P that projects vectors in \mathbf{R}^3 onto \mathbf{S}^\perp .

Solution: $\mathbf{x} \in \mathbf{S}^\perp$ if and only if $\mathbf{x} \perp \mathbf{u}_1$ and $\mathbf{x} \perp \mathbf{u}_2$. That is

$$x_1 + 2x_3 = 0$$

$$x_2 - 2x_3 = 0$$

$\mathbf{S}^\perp = \text{span}\{(-2, 2, 1)^T\}$. An orthonormal basis for \mathbf{S}^\perp is $(-2/3, 2/3, 1/3)^T$

The projection matrix is

$$\begin{pmatrix} -2/3 \\ 2/3 \\ 1/3 \end{pmatrix} (-2/3, 2/3, 1/3) = \begin{pmatrix} 4/9 & -4/9 & -2/9 \\ -4/9 & 4/9 & 2/9 \\ -2/9 & 2/9 & 1/9 \end{pmatrix}$$

#4. For the given matrix A , find all possible values of the scalar β that make A diagonalizable or show that no such values exist.

$$A = \begin{pmatrix} 4 & 6 & -2 \\ -1 & -1 & 1 \\ 0 & 0 & \beta \end{pmatrix}$$

Solution: The eigenvalues of A are $\lambda_1 = 1$, $\lambda_2 = 2$, $\lambda_3 = \beta$.

If $\beta \neq 1$ and $\beta \neq 2$, then A is diagonalizable since A has 3 distinct eigenvalues, and hence A has 3 independent eigenvectors.

If $\beta = 1$, then 2 is a single eigenvalue and 1 is a double eigenvalue of matrix A . For the double eigenvalue,

$$A - 1 \cdot I = \begin{pmatrix} 3 & 6 & -2 \\ -1 & -2 & 1 \\ 0 & 0 & 0 \end{pmatrix}, \quad N(A - 1 \cdot I) = \text{span}\{(-2, 1, 0)^T\}, \text{ which is of dimension 1.}$$

Hence, in this case A is not diagonalizable since the geometric multiplicity of $\lambda_1 = 1$ is less than its algebraic multiplicity.

If $\beta = 2$, then 1 is a single eigenvalue and 2 is a double eigenvalue of matrix A . For this double eigenvalue,

$$A - 2 \cdot I = \begin{pmatrix} 2 & 6 & -2 \\ -1 & -3 & 1 \\ 0 & 0 & 0 \end{pmatrix} \quad N(A - 2 \cdot I) = \text{span}\{(1, 0, 1)^T, (3, -1, 0)^T\}, \text{ which is of dimension 2.}$$

Thus, in this case A is diagonalizable since for each eigenvalue of A , its algebraic multiplicity is the same as its geometric multiplicity.

In summary, A is diagonalizable if and only if $\beta \neq 1$.

Part II (选做题, 30 分)

请选择下列三题中的两题解答, 并在所选的题号上划圈, 否则按得分最低的两题计分.

#5. Let S be a subspace of \mathbf{R}^1 . Show that either $S = \{0\}$ or $S = \mathbf{R}^1$.

Proof If $S \neq \{0\}$, then there exists a nonzero element $(a) \in S$, where a is a nonzero real number. For any element $(x) \in \mathbf{R}^1$, $\frac{x}{a}$ is a real number. Thus, $\frac{x}{a}(a) = (x) \in S$ since S is closed under scalar multiplication. Thus, any element in \mathbf{R}^1 is also in S , and hence $S = \mathbf{R}^1$.

#6. If V_1, V_2, V_3 are subspaces of vector space V , show that $V_1 + V_2 + V_3$ is a direct sum if and only if

$$V_1 \cap (V_2 + V_3) = \{0\}, V_2 \cap (V_1 + V_3) = \{0\}, \text{ and } V_3 \cap (V_1 + V_2) = \{0\}.$$

(注: 不可利用书中 36 页上的 Theorem 1.7.3)

Proof $V_1 + V_2 + V_3$ is a direct sum if each vector in the sum can be *uniquely* represented as

$$\mathbf{x} = \mathbf{x}_1 + \mathbf{x}_2 + \mathbf{x}_3, \text{ where } \mathbf{x}_i \in V_i (i=1,2,3).$$

Suppose that $V_1 + V_2 + V_3$ is a direct sum. If $V_1 \cap (V_2 + V_3) \neq \{0\}$, then there is a nonzero element $\mathbf{u} \in V_1 \cap (V_2 + V_3)$.

\mathbf{u} can be represented as $\mathbf{u} = \mathbf{u} + \mathbf{0} + \mathbf{0}$, where $\mathbf{u} \in V_1$, and $\mathbf{0} \in V_2, \mathbf{0} \in V_3$.

And also, \mathbf{u} can be represented as $\mathbf{u} = \mathbf{0} + \mathbf{x}_2 + \mathbf{x}_3$ since $\mathbf{u} \in V_2 + V_3$, where $\mathbf{0} \in V_1, \mathbf{x}_2 \in V_2, \mathbf{x}_3 \in V_3$. Thus, \mathbf{u} has two distinct representations. This is a contradiction. Thus, when $V_1 + V_2 + V_3$ is a direct sum, we must have $V_1 \cap (V_2 + V_3) = \{0\}$. Similarly, we can prove that $V_2 \cap (V_1 + V_3) = \{0\}$, and $V_3 \cap (V_1 + V_2) = \{0\}$.

Conversely, suppose that $V_1 \cap (V_2 + V_3) = \{0\}, V_2 \cap (V_1 + V_3) = \{0\}$, and $V_3 \cap (V_1 + V_2) = \{0\}$.

If $V_1 + V_2 + V_3$ is not a direct sum, then there exists an element $\mathbf{u} \in V_1 + V_2 + V_3$, such that

$$\mathbf{u} = \mathbf{x}_1 + \mathbf{x}_2 + \mathbf{x}_3 \text{ and } \mathbf{u} = \mathbf{y}_1 + \mathbf{y}_2 + \mathbf{y}_3, \text{ where } \mathbf{x}_i, \mathbf{y}_i \in V_i (i=1,2,3).$$

and there exist at least one integer $i \in \{1,2,3\}$ such that $\mathbf{x}_i \neq \mathbf{y}_i$. Without loss of generality, assume that $i=1$.

#7. Let σ be a linear transformation on vector space V over the complex number field, and S be a σ -invariant subspace of V . Show that there exists a nonzero vector $\mathbf{u} \in S$, such that $\sigma(\mathbf{u}) = \lambda \mathbf{u}$, where λ is a scalar. (For the definition of invariant subspace, see page 92 in the textbook.)

提示: 取 S 的一组基 $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k$, 则 $\sigma(\mathbf{u}_1), \sigma(\mathbf{u}_2), \dots, \sigma(\mathbf{u}_k)$ 都可由 $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k$ 的线性组合表示,

利用线性组合的系数矩阵.

Proof In formal multiplication, we can write

$$(\sigma(\mathbf{u}_1), \sigma(\mathbf{u}_2), \dots, \sigma(\mathbf{u}_k)) = (\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k)A$$

where A is a $k \times k$ matrix. Let λ be an eigenvalue of A , and \mathbf{x} be an eigenvector such that $A\mathbf{x} = \lambda\mathbf{x}$, where $\mathbf{x} = (x_1, x_2, \dots, x_k)^T$

Let $\mathbf{u} = x_1\mathbf{u}_1 + x_2\mathbf{u}_2 + \dots + x_k\mathbf{u}_k = (\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k)\mathbf{x}$, which is not zero.

Then $\sigma(\mathbf{u}) = x_1\sigma(\mathbf{u}_1) + x_2\sigma(\mathbf{u}_2) + \dots + x_k\sigma(\mathbf{u}_k) = (\sigma(\mathbf{u}_1), \sigma(\mathbf{u}_2), \dots, \sigma(\mathbf{u}_k))\mathbf{x}$
 $= (\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k)A\mathbf{x} = (\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k)\lambda\mathbf{x} = \lambda(\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k)\mathbf{x} = \lambda\mathbf{u}$