## ExAM 3 KEY 2270-4

**Definition**: An abstract vector space V is a data set of packages called **vectors** together with operations of addition (+) and scalar multiplication  $(\cdot)$  satisfying the following eight (8) rules:

Closure: If  $\vec{x}$  and  $\vec{y}$  are in V, then  $\vec{x} + \vec{y}$  is defined and in V.

- $(1) \vec{\mathbf{x}} + \vec{\mathbf{y}} = \vec{\mathbf{y}} + \vec{\mathbf{x}}$
- (2)  $\vec{x} + (\vec{y} + \vec{z}) = (\vec{x} + \vec{y}) + \vec{z}$
- (3) There is a zero vector  $\vec{0}$  in V with  $\vec{x} + \vec{0} = \vec{x}$ .
- (4) There is a vector  $-\vec{\mathbf{x}}$  in V with  $\vec{\mathbf{x}} + (-\vec{\mathbf{x}}) = \vec{\mathbf{0}}$ .

Closure: If c = constant and  $\vec{\mathbf{x}}$  is in V, then  $c \cdot \vec{\mathbf{x}}$  is defined and in V.

- (5)  $a(\vec{\mathbf{x}} + \vec{\mathbf{y}}) = a\vec{\mathbf{x}} + a\vec{\mathbf{y}}$
- (6)  $(a+b) \cdot \vec{\mathbf{x}} = a \cdot \vec{\mathbf{x}} + b \cdot \vec{\mathbf{x}}$
- (7)  $(ab) \cdot \vec{\mathbf{x}} = a \cdot (b \cdot \vec{\mathbf{x}})$
- $(8) 1 \cdot \vec{\mathbf{x}} = \vec{\mathbf{x}}$

**Definition**. If vectors  $\vec{b}_1, \vec{b}_2, \ldots, \vec{b}_n$  are a basis for subspace X of an abstract vector space V, and  $\vec{\mathbf{x}} = c_1\vec{\mathbf{b}}_1 + c_2\vec{\mathbf{b}}_2 + \cdots + c_n\vec{\mathbf{b}}_n$  is a given linear combination of these vectors, then the uniquely determined constants  $c_1, c_2, \ldots, c_n$  are called the *coordinates of*  $\vec{\mathbf{x}}$  relative to the basis  $\vec{\mathbf{b}}_1, \vec{\mathbf{b}}_2, \ldots, \vec{\mathbf{b}}_n$  and the *coordinate map* is the isomorphism

$$\vec{\mathbf{x}} = c_1 \vec{\mathbf{b}}_1 + c_2 \vec{\mathbf{b}}_2 + \dots + c_n \vec{\mathbf{b}}_n \to \begin{pmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{pmatrix}.$$

**Definition**: A subset S of a vector space V is a subspace of V provided

- (1) The zero vector is in S
- (2) If vectors  $\vec{\mathbf{x}}$  and  $\vec{\mathbf{y}}$  are in S, then  $\vec{\mathbf{x}} + \vec{\mathbf{y}}$  is in S.
- (3) If vector  $\vec{\mathbf{x}}$  is in S and c is any scalar, then  $c\vec{\mathbf{x}}$  is in S.

**Definition**: Vectors  $\vec{\mathbf{v}}_1, \dots, \vec{\mathbf{v}}_p$  in an abstract vector space V are said to be **independent** in V provided solving the equation  $c_1\vec{\mathbf{v}}_1 + \dots + c_p\vec{\mathbf{v}}_p = \vec{\mathbf{0}}$  for scalars  $c_1, \dots, c_p$  has only the zero solution  $c_1 = \dots = c_p = 0$ .

Some problems have two solutions.

**Problem 1.** (100 points) Let V be the vector space of all functions on  $(-\infty, \infty)$ . Define  $W = \operatorname{span}\{x, e^x\}$ . Assume known that  $x, e^x$  are independent functions. Define subspace  $S = \operatorname{span}\{\vec{\mathbf{v}}_1, \vec{\mathbf{v}}_2\}$  where

$$\vec{\mathbf{v}}_1: \ y = x + e^x, \quad \vec{\mathbf{v}}_2: \ y = x - e^x$$

(a) [20%] Explain why S is contained in W, that is, provide details for why linear combinations of vectors  $\vec{\mathbf{v}}_1, \vec{\mathbf{v}}_2$  are in W.

The Choose  $\vec{v} \in S$  where  $\vec{v} = c_1 \vec{J}_1 + c_2 \vec{J}_2$  for  $c_1, c_2 \in \mathbb{R}$ . We find  $\vec{v} = c_1 (x + e^x) + c_2 (x - e^x)$   $= c_1 x + c_1 e^x + c_2 x - c_2 e^x$   $= (c_1 + c_2) \times + (c_1 - c_2) e^x$ Since  $c_1 + c_2, c_1 - c_2 \in \mathbb{R}$ ,  $\vec{v} \in \mathbb{N}$ . Therefore  $S \in \mathbb{N}$ .  $\square$ 

(b) [40%] Prove that W = S. Therefore  $\dim(S) = \dim(W) = 2$ , which proves independence of vectors  $\vec{\mathbf{v}}_1: y = x + e^x$ ,  $\vec{\mathbf{v}}_2: y = x - e^x$ .

If choose  $\vec{n} \in M$  where  $\vec{n} \cdot y = qx + c_z e^x$  for  $q, c_z \in \mathbb{R}$ . We show there exists  $d_1, d_2 \in \mathbb{R}$  such that  $\begin{cases} c_1 = d_1 + d_2 \\ c_2 = d_1 - d_3 \end{cases}$ 

has a unique solution. We find

|A| = | 1 -1 = -1-1 = -2 = 0

thus A' exists and the system (1) has

a unique solution by the Invertible Matrix

Theorem. Hence  $\vec{N}: y = (d_1 + d_2) \times + (d_1 - d_2) e^{x}$   $= d_1(x + e^{x}) + d_2(x - e^{x})$   $= d_1 \vec{V}_1 + d_2 \vec{V}_2$ 

and BES. Therefore SEW and W.C.S. meaning S=W. It follows that dim 5 = dim W=2, meaning is and vz are independent.

**Problem 1.** (100 points) Let V be the vector space of all functions on  $(-\infty, \infty)$ . Define  $W = \operatorname{span}\{x, e^x\}$ . Assume known that  $x, e^x$  are independent functions. Define subspace  $S = \operatorname{span}\{\vec{\mathbf{v}}_1, \vec{\mathbf{v}}_2\}$  where

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 $\bigwedge$  (a) [20%] Explain why S is contained in W, that is, provide details for why linear combinations of vectors  $\vec{\mathbf{v}}_1, \vec{\mathbf{v}}_2$  are in W.

(b) [40%] Prove that W = S. Therefore  $\dim(S) = \dim(W) = 2$ , which proves independence of vectors  $\vec{\mathbf{v}}_1 : y = x + e^x$ ,  $\vec{\mathbf{v}}_2 : y = x - e^x$ .

Let 
$$\overline{s} \in S$$
. From part (a),  $\overline{s} \in W$ .

Let  $\overline{s} \in W$ .  $\Rightarrow \overline{s} = d_1 \times + d_2 e^{\times} = c_1(x + e^{\times}) + c_2(x - e^{\times}) = (c_1 + c_2)x + (c_1 - c_2)e^{\times}$ 
 $\Rightarrow \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \end{bmatrix} = \begin{bmatrix} d_1 \\ d_2 \end{bmatrix}$ 

Since  $\begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} = -2 \neq 0$ , there are unique solutions for every tombination of  $d_1 \in d_2 = x = x = 0$ .

Since  $\overline{s} \in W \in S \in S = x = x = x = x = x = x$ .

(c) [40%] Define vector  $\vec{\mathbf{v}}$  in S by equation  $y=2x+3e^x$ . Show how to compute  $d_1, d_2$  in the equation  $\vec{\mathbf{v}}=d_1\vec{\mathbf{v}}_1+d_2\vec{\mathbf{v}}_2$ , using coordinate map methods. The definitions are  $\vec{\mathbf{v}}_1: y=x+e^x$ ,  $\vec{\mathbf{v}}_2: y=x-e^x$ .

A A-

**Expected in (c)**: Calculations of  $d_1, d_2$  are to be done using column vectors from  $\mathbb{R}^2$  and  $2 \times 2$  matrices, not functions from V. Zero credit for not using column vectors and coordinate maps.

$$\overline{V} = d_1 \overline{V}_1 + d_2 \overline{V}_2 = d_1 (X + e^X) + d_2 (X - e^X) = d_1 \begin{bmatrix} 1 \\ 1 \end{bmatrix} + d_2 \begin{bmatrix} -1 \\ -1 \end{bmatrix}$$

$$\overline{V} = 2X + 3e^X = \begin{cases} c_1 = 2 \\ c_2 = 3 \end{cases} \quad \text{Coordinate map used, but not explicitly.}$$

$$\begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} d_1 \\ d_2 \end{bmatrix} = \begin{bmatrix} 2 \\ 3 \end{bmatrix} = \begin{cases} 1 & 1 \\ 2 \end{bmatrix} \sim \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \sim \begin{bmatrix} 1 & 1/2 \\ 0 & 1/2 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 0 & 5/2 \\ 0 & 1 & 1/2 \end{bmatrix} \sim \begin{cases} d_1 = \frac{5}{2} \\ d_2 = \frac{1}{2} \end{cases}$$

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**Definition**: A subset S of a vector space V is a subspace of V provided

- (1) The zero vector is in S
- (2) If vectors  $\vec{\mathbf{x}}$  and  $\vec{\mathbf{y}}$  are in S, then  $\vec{\mathbf{x}} + \vec{\mathbf{y}}$  is in S.
- (3) If vector  $\vec{\mathbf{x}}$  is in S and c is any scalar, then  $c\vec{\mathbf{x}}$  is in S.

## Problem 2. (100 points)

(a) [60%] Let V be an abstract vector space. Let  $\vec{\mathbf{v}}_1, \vec{\mathbf{v}}_2$  be two vectors in V. Define S to be the set of all linear combinations of  $\vec{\mathbf{v}}_1 + \vec{\mathbf{v}}_2$ ,  $\vec{\mathbf{v}}_1 - \vec{\mathbf{v}}_2$ . Prove that S is a subspace of V, using only the definition of subspace.

Expected: A proof uses the symbols  $\vec{\mathbf{v}}_1, \vec{\mathbf{v}}_2$  and the 8 rules of a vector space, plus theorems like  $0\vec{\mathbf{v}} = \vec{\mathbf{0}}$ . Symbols  $\vec{\mathbf{v}}_1, \vec{\mathbf{v}}_2$  are not assumed to be column vectors.

For 
$$0 \in S$$
.  
 $S = 5pan \{ V_1 + V_2, V_1 - V_2 \} = c_1(V_1 + V_2) + c_2(V_1 - V_2)$   
If  $q = c_2 = 0$ ,  $\Rightarrow S = 0$ 

Let 
$$\bar{\chi} \in S \in \bar{V} \in S = \sum_{i=1}^{N} \bar{\chi} = c_1(\bar{V}_1 + \bar{V}_2) + (2(\bar{V}_1 - \bar{V}_2))$$
  
 $\bar{V} = d_1(\bar{V}_1 + \bar{V}_2) + d_2(\bar{V}_1 - \bar{V}_2)$ 

$$\begin{array}{lll} X+Y &= C_1(\bar{V}_1+\bar{V}_2)+C_2(\bar{V}_1-\bar{V}_2)+d_1(\bar{V}_1+\bar{V}_2)+d_2(\bar{V}_1-\bar{V}_2)\\ \text{ing(i) from rules of a vector space:} \\ \bar{X}+\bar{Y} &= C_1(\bar{V}_1+\bar{V}_2)+d_1(\bar{V}_1+\bar{V}_2)+G_2(\bar{V}_1-\bar{V}_2)+O_2(\bar{V}_1-\bar{V}_2) \end{array}$$

sing (6) from rules of vector space:

Let 
$$\alpha \in \mathbb{R} = \lambda \times \mathbb{R} = \alpha \left[ c_1 (V_1 + V_2) + c_2 (V_1 - V_2) \right]$$
  
Let  $\alpha \in \mathbb{R} = \lambda \times \mathbb{R} = \alpha \left[ c_1 (V_1 + V_2) + c_2 (V_1 - V_2) \right]$ 

ising Rule (5):

By definition of a subspace, S is a subspace of V.

**Definition**: A subset S of a vector space V is a subspace of V provided

- (1) The zero vector is in S
- (2) If vectors  $\vec{x}$  and  $\vec{y}$  are in S, then  $\vec{x} + \vec{y}$  is in S.
- (3) If vector  $\vec{\mathbf{x}}$  is in S and c is any scalar, then  $c\vec{\mathbf{x}}$  is in S.



(b) [40%] Let V be the vector space of all  $2 \times 2$  matrices. Invent an example of a non-void subset S of V that satisfies (1) and (2) but fails the third item (3).

Let 
$$S = \begin{bmatrix} x & 0 \\ 4 & 6 \end{bmatrix}$$
 S.E.  $X \ge 0 \leqslant 4 \ge 0$ .

- (2) => If a = S = a => a + a = S since x + x 20 & y + y 2 20
  - (1) Additionally, OES since x can equal 0 & y can equal 0,
  - (3) However, If c=-1,  $-\overline{x}$  is not in S, since  $x \ge 0 \le y \ge 0$ , so it fails (3).

**Problem 3.** (100 points) Let A be a  $4 \times 3$  matrix. Assume the determinant of  $A^T A$  is zero. Prove that the nullspace of A contains a nonzero vector.

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A So ATA is not invertible. By the invertible Matrix theorem, an invertible matrix must have a non-zero determinant.

So, ATA has dependent cols.

ATA x = 0 has nontrivial soin

 $x^T A^T A \vec{x} = x^T \vec{0} \Rightarrow (A\vec{x})^T A \vec{x} = \vec{0} \Rightarrow ||A\vec{x}||^2 = \vec{0} \Rightarrow A\vec{x} = \vec{0}$ 

Since  $A\vec{x} = \vec{0}$  Cols blc
has solwhere  $\vec{x} \neq \vec{0}$ , then  $\vec{x} \neq 0$ 

the Mullspace has a nonzero vector

Problem 4. (100 points)

(a) [40%] Define  $\vec{\mathbf{y}} = \begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix}$ ,  $\vec{\mathbf{u}} = \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}$ . Find the orthogonal projection vector  $\vec{\mathbf{v}}$  (the shadow projection vector) of  $\vec{\mathbf{y}}$  onto the direction of  $\vec{\mathbf{u}}$ .

(b) [60%] Let  $\vec{\mathbf{v}}_1 = \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}$ ,  $\vec{\mathbf{v}}_2 = \begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix}$ ,  $\vec{\mathbf{x}} = \begin{pmatrix} 1 \\ 1 \\ 2 \end{pmatrix}$ . Check that  $\vec{\mathbf{v}}_1$ ,  $\vec{\mathbf{v}}_2$  are orthogonal and then compute  $\vec{\mathbf{v}}$  = the vector projection of  $\vec{\mathbf{x}}$  onto the subspace  $S = \operatorname{span}\{\vec{\mathbf{v}}_1, \vec{\mathbf{v}}_2\}$ .

Reminder:  $\vec{\mathbf{v}}$  is the sum of two shadow projections.

$$\frac{\vec{V}_{1} \cdot \vec{V}_{2}}{\vec{V}_{1} \cdot \vec{V}_{3}} = \frac{\vec{V}_{1} \cdot \vec{X}}{\vec{V}_{1} \cdot \vec{V}_{1}} = \frac{1}{1+1+0} = \frac{1}{1+1+0} = \frac{1}{1+1+0} = \frac{1}{1+1+0} = \frac{1}{1+1+0} = \frac{1}{1+1+1} = \frac{1}{1+1+$$

**Problem 5.** (100 points) Let A be an  $m \times n$  matrix and  $\vec{b}$  an  $m \times 1$  vector. Let W be the column space of A. Linear equations  $A^T A \vec{z} = A^T \vec{b}$  are the **normal equations** for the problem  $A\vec{\mathbf{x}} = \vec{\mathbf{b}}$ .

(a) [30%] Let 
$$A = \begin{pmatrix} 1 & 1 \\ 1 & 1 \\ 1 & 0 \end{pmatrix}$$
,  $\vec{\mathbf{b}} = \begin{pmatrix} 1 \\ -1 \\ -1 \end{pmatrix}$ .

Display toolkit steps that verify  $A\vec{\mathbf{x}} = \vec{\mathbf{b}}$  has no solution.

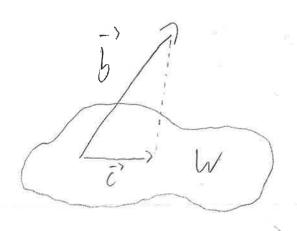
Display toolkit steps that verify 
$$Ax = b$$
 has no solution.

$$A_{\lambda}^{-1} = b - 1$$

$$A_{\lambda}^{-1$$

(b) [30%] Let  $\vec{c} = A\vec{z}$  where  $\vec{z}$  is the unique theoretical solution of the normal equations. Explain with a figure:  $\vec{c}$  is the nearest point to  $\vec{b}$  in the column space W.





(e) [40%] Let 
$$A = \begin{pmatrix} 1 & 1 \\ 1 & 1 \\ 1 & 0 \end{pmatrix}$$
,  $\overline{b} = \begin{pmatrix} 1 \\ -1 \\ -1 \end{pmatrix}$ . Find vector  $\vec{x}$  in the normal equations.

$$V_{2} = A^{T} A \vec{z}^{2} = A^{T} \vec{b}^{2}$$

$$A_{3} = A^{T} \vec{b}^{2}$$

$$A_{4} = A^{T} \vec{b}^{2}$$

$$A_{5} = A^{T} \vec{b}^{2}$$

$$A_{7} =$$

Problem 6. (100 points) The Fundamental Theorem of Linear Algebra contains this statement: The row space and the null space of a matrix are orthogonal. This means that  $\vec{R} \cdot \vec{N} = 0$  for each vector  $\vec{R}$  in the row space and each vector  $\vec{N}$  in the null space.

The four fundamental subspaces in the Fundamental Theorem of Linear algebra are: (1) Nullspace of A, (2) Column Space of A, (3) Row space of A, 4) Nullspace of  $A^T$ .

(a) [30%] Define precisely the four fundamental subspaces. For example, the Nullspace of A is the set of all solutions  $\vec{x}$  to the matrix equation  $A\vec{x} = \vec{0}$ .

(2) The Column Space of A is the set of all I.C. of the columns of A.

(3) The Row space of A is the set of all I.c. of the rows of A (equal to the col. space of AT)

(4) The Nullspace of AT is the set of all nonzero rows of A

(b) [30%] Assume A is  $20 \times 12$  and has rank 10. Equivalently, matrix A has 10 pivots . Report the dimensions of the four fundamental subspaces.

A If 
$$m=20$$
  $n=12$   $r=10$   
dim Nullspace (A)=  $n-r=12-10=[2]$   
dim Col(A) =  $r=[10]$   
dim Pow(A) =  $r=[10]$   
dim Nullspace (A<sup>T</sup>)=[10]

if A has 10 pivot col AT has 10 pivot rows A has 10

hon zero rows

A

<sup>(1)</sup> The Mullspace of A is the set of all solutions to the homogeneous eq. + = 5.

(c) [40%] Let A be an  $m \times n$  matrix. Let  $\vec{\mathbf{C}}$  be a linear combination of the columns of A and let  $\vec{\mathbf{Y}}$  belong to the nullspace of  $A^T$ . Prove that  $\vec{\mathbf{C}} \cdot \vec{\mathbf{Y}} = 0$ , that is,  $\vec{\mathbf{C}}$  and  $\vec{\mathbf{Y}}$  are orthogonal.

The Fundamental Theorem of Algebra States that Row (A) I Mullspace (A) Similarly, ROW(AT) I Mullspace (AT) C = part of Col(A) By definition T = part of Mull (AT) of AT, Col (A) = ROW(AT) Therefore, C= part of Row (AT) and by Fund. Theorem of L.A., C. T=0 blc Row(AT) I Null(AT) and i and i are orthogonal.

(c) [40%] Let A be an  $m \times n$  matrix. Let  $\vec{\mathbf{C}}$  be a linear combination of the columns of A and let  $\vec{\mathbf{Y}}$  belong to the nullspace of  $A^T$ . Prove that  $\vec{\mathbf{C}} \cdot \vec{\mathbf{Y}} = 0$ , that is,  $\vec{\mathbf{C}}$  and  $\vec{\mathbf{Y}}$  are orthogonal.

A. 
$$\overrightarrow{C} \cdot \overrightarrow{y} = A \overrightarrow{x} \cdot \overrightarrow{y}$$
 ( $\overrightarrow{x}$  being the solution to  $\overrightarrow{A} \cdot \overrightarrow{x} + \overrightarrow{C} \cdot \overrightarrow{x}$  known to exist by the definition of linear combination)

 $\overrightarrow{A} \cdot \overrightarrow{y} = (\overrightarrow{A} \cdot \overrightarrow{x})^T \overrightarrow{y} = \overrightarrow{x}^T \overrightarrow{A}^T \overrightarrow{y}$ 
 $\overrightarrow{A} \cdot \overrightarrow{y} = \overrightarrow{O}$ , by the definition of nullspace

 $\overrightarrow{X} \cdot \overrightarrow{A} \cdot \overrightarrow{y} = \overrightarrow{X} \cdot \overrightarrow{O} = \overrightarrow{O}$ 

Thus,  $\overrightarrow{C}$  and  $\overrightarrow{y}$  are orthogonal excellent