

## Midterm Review Sheet I – Linear Algebra

[Sources: *Linear Algebra and its Applications*, 4th Edition, Gilbert Strang, Brooks/Cole; and *Linear Algebra, Geodesy, and GPS*, Gilbert Strang and Kai Borre, Wellesley-Cambridge Press]

**Exam information:**

- **Date and Location:** The midterm exam will take place on Thursday, October 28, from 10 to 11.50AM at LO 1322
- **Topics:** Linear Algebra (Chapter 2) and Probability (Chapter 3)
- **Notes:** You can bring one page of notes, one side of an 8.5 in.  $\times$  11 in. sheet.

Additional Notes in Linear Algebra**Some Examples and Definitions:**– **Examples of Vector Spaces:**

- $\mathbb{R}^n$  Spaces:  $\mathbb{R}^2$  – *all vectors* of two real components (including the zero vector), *i.e.*, the plane;  $\mathbb{R}^3$  – all vectors with three real components;  $\dots$ ;  $\mathbb{R}^n$  – all vectors of  $n$  real components.
- Function Spaces:  $\mathbf{F}$  – the space of *all real functions*  $f(x)$ ;  $C^2$  – the space of all real continuous functions with continuous first and second derivatives.
- Matrix Spaces:  $\mathbf{M}$  – The space of *all real 2 by 2 matrices*.
- Other Vector Spaces:  $\mathbf{Z}$  – The vector space consisting only of a *zero vector*.

**Definition.** A *subspace* or a vector space is a set of vectors (including the zeros vector) of a vector space that satisfies two conditions: If  $\mathbf{u}$  and  $\mathbf{v}$  are vectors in the subspace, and  $c$  is any scalar, then

- $\mathbf{u} + \mathbf{w}$  is also in the subspace
- $c\mathbf{u}$  is also in the subspace

**Examples:** (1)  $\mathbb{R}^2$  is a subspace of  $\mathbb{R}^3$ ,  $\mathbb{R}^2$  consists of all vectors in  $\mathbb{R}^3$  whose third component is zero. (2) Any line on the plane through  $(0,0)$  is a subspace of  $\mathbb{R}^2$ . (3) All upper triangular matrices in  $\mathbf{M}$  form a subspace of  $\mathbf{M}$ .

**Definition.** The *column space* of a matrix  $A$  consists of all linear combinations of the columns of  $A$ . These combinations are the vectors  $A\mathbf{x}$ .

**Remarks:** (1) If  $A \in \mathbb{R}^{m \times n}$  its columns have  $m$  entries, so the columns belong to  $\mathbb{R}^m$  and the column space of  $A$  is a subspace of  $\mathbb{R}^m$ . (2) The system  $A\mathbf{x} = \mathbf{b}$  has (at least) a solution if  $\mathbf{b}$

is in the column space of  $A$ . The solution  $\mathbf{x}$  is the vector whose entries are the coefficients of the linear combination.

**Definition.** The column space of  $A$  is called the *range of  $A$*  and it is denoted by  $R(A)$ .

**Example:** The column space of the matrix

$$A = \begin{bmatrix} 1 & 0 \\ 4 & 3 \\ 2 & 3 \end{bmatrix}$$

consist of all linear combinations

$$x_1 \begin{bmatrix} 1 \\ 4 \\ 2 \end{bmatrix} + x_2 \begin{bmatrix} 0 \\ 3 \\ 3 \end{bmatrix}, \quad x_1, x_2 \in \mathbb{R}$$

These linear combinations define a plane through the origin in  $\mathbb{R}^3$ , a subspace of  $\mathbb{R}^3$ .

**Definition.** The *nullspace* (or *kernel*) of a matrix  $A \in \mathbb{R}^{m \times n}$  consists of *all solutions* to  $A\mathbf{x} = \mathbf{0}$  and it is denoted by  $N(A)$ . These vectors  $\mathbf{x}$  are in  $\mathbb{R}^n$ .

**Remark:** The system  $A\mathbf{x} = \mathbf{b}$  has a solution if  $\mathbf{b}$  is in the range of  $A$ , and the solution is unique if the only element in the nullspace of  $A$  is  $\mathbf{0}$ .

**Definition.** The row space of a matrix  $A$  is the vector space spanned by its rows. It coincides with the range of  $A^T$ .

**Definition.** The *rank*,  $r$ , of a matrix  $A$  is the number of linearly independent rows/columns of  $A$ ;  $r \leq \min\{m, n\}$ .

**Definition.** A basis for a vector space is a collection of linearly independent vectors that span the space. The dimension of a space is the number of vectors in any basis of the space.

**Definition.** The *nullity* of a matrix  $A$  is the dimension of its nullspace

**Fundamental Theorem of Linear Algebra**  $A \in \mathbb{R}^{m \times n}$

1.  $R(A)$  – column space of  $A$ ; dimension  $r$ .
2.  $N(A)$  – nullspace of  $A$ ; dimension  $n - r$ .
3.  $R(A^T)$  – row space of  $A^T$ ; dimension  $r$ .
4.  $N(A^T)$  – nullspace of  $A^T$ ; dimension  $m - r$ .

**Singular Value Decomposition.** (*SVD*) of a matrix  $A \in \mathbb{R}^{m \times n}$ ,  $A = UDV^T$ , where:

- $U \in \mathbb{R}^{m \times m}$  – orthogonal matrix whose columns are (1) the eigenvectors of  $AA^T$ , (2) its first  $r$  columns form a basis for  $R(A)$ , and (3) its last  $m - r$  columns form a basis for  $N(A^T)$ .
- $D \in \mathbb{R}^{m \times n}$  – diagonal matrix with the  $r$  non-singular values of  $A$  which are the square root of the nonzero eigenvalues of both  $AA^T$  and  $A^T A$ .
- $V \in \mathbb{R}^{n \times n}$  – orthogonal matrix whose columns are (1) the eigenvectors of  $A^T A$ , (2) its first  $r$  columns a basis for  $R(A^T)$ , and (3) its last  $n - r$  columns form a basis for  $N(A)$ .

- **Note:**  $AV = UD$ . That is, when  $A$  multiplies column  $v_j$  of  $V$ , it produces  $d_j$  times a column of  $U$ .

### Problems in Linear Algebra

**Problem 1.** Determine whether the following statements are **True** or **False**. Justify your answer.

- If  $A \in \mathbb{R}^{m \times n}$ , then  $A$  and  $A^T$  have the same nullspaces.
- If the columns of a matrix are linearly dependent so are the rows.
- The column space of a 2 by 2 matrix is the same as its row space.
- The columns of a matrix are a basis for its column space.
- Suppose the columns of a 4 by 4 matrix are a basis for  $\mathbb{R}^4$ , then the equation  $A\mathbf{x} = \mathbf{0}$  has only the solution  $\mathbf{x} = \mathbf{0}$

**Problem 2.** Describe the column spaces and nullspaces of the matrices

$$A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \quad B = \begin{bmatrix} 1 & 2 \\ 2 & 4 \end{bmatrix}, \quad \text{and} \quad C = \begin{bmatrix} 1 & 2 & 3 \\ 0 & 0 & 4 \end{bmatrix}.$$

**Problem 3.** For which numbers  $c$  and  $d$  do these matrices have rank  $r = 2$ ?

$$A = \begin{bmatrix} 1 & 2 & 5 & 0 & 5 \\ 0 & 0 & c & 2 & 2 \\ 0 & 0 & 0 & d & 2 \end{bmatrix} \quad \text{and} \quad B = \begin{bmatrix} c & d \\ d & c \end{bmatrix}$$

**Problem 4.** Consider the system  $A\mathbf{x} = \mathbf{b}$  with

$$A = \begin{bmatrix} 2 & \alpha \\ 4 & 8 \end{bmatrix} \quad \text{and} \quad \mathbf{b} = \begin{bmatrix} 16 \\ \beta \end{bmatrix}.$$

For what values of  $\alpha$  and  $\beta$  does the system have:

- No solution?
- Infinitely many solutions?
- Exactly one solution?

**Problem 5.** The matrix  $B$  has eigenvalues 0, 1, and 2. Find:

- The rank of  $B$ .
- The determinant of  $B^T B$ .
- The eigenvalues of  $B^T B$ .
- The eigenvalues of  $(B + I)^{-1}$

**Problem 6.** Suppose  $A = \mathbf{u}\mathbf{v}^T$  is a column times a row (a *rank-1* matrix)

- (a) Show that  $\mathbf{u}$  is an eigenvector of  $A$  by multiplying  $A$  times  $\mathbf{u}$ . What is the corresponding eigenvalue?
- (b) What are the other eigenvalues of  $A$ ? Why?
- (c) Compute the trace of  $A$  by (i) adding its diagonal entries, and (ii) adding its eigenvalues

**Problem 7.** Singular Value Decomposition:  $A = UDV^T$ . Suppose  $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n$ , and  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$ , are two orthonormal bases for  $\mathbb{R}^n$ . Construct the matrix that transforms each  $\mathbf{v}_j$  into  $\mathbf{u}_j$  to give  $A\mathbf{v}_1 = \mathbf{u}_1, \dots, A\mathbf{v}_n = \mathbf{u}_n$ .

**Problem 8.** Find  $UDV^T$  if  $A$  has orthogonal columns  $\mathbf{w}_1, \dots, \mathbf{w}_n$  of lengths  $d_1, \dots, d_n$ .