

# Preliminary Linear Algebra

# Notation

$\forall$  for all

$\exists$  there exists

$\ni$  such that

$\therefore$  therefore

$\because$  because

$\square$  end of proof (QED)

# Notation

$A \implies B$      $A$  implies  $B$

$A \iff B$      $A$  if and only if (iff)  $B$

$a \in B$      $a$  is an element of the set  $B$

$A \subset B$      $A$  is a proper subset of  $B$

$A \subseteq B$      $A$  is a subset of  $B$

$\mathbb{R}^n$     Euclidean  $n$ -space

# Matrix Notation

•  $\mathbf{A} = [a_{ij}] = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix}$  is a matrix with  $m$  rows and  $n$  columns.

- The entry in the  $i^{th}$  row and  $j^{th}$  column of  $\mathbf{A}$  is  $a_{ij}$ .

# Square Matrices and Vectors

- Matrix  $A$  is said to be square if  $m = n$ .  
 $m \times n$
- A matrix with one column is called a vector.
- A matrix with one row is called a row vector.

# Notation

In these notes, bold uppercase letters are used to denote matrices, and bold lowercase letters are used to denote vectors.

# Examples

- $\mathbf{0}$  or  $\mathbf{0}_n$  is a vector of zeros.
- $\mathbf{1}$  or  $\mathbf{1}_n$  is a vector of ones.
- $\mathbf{I}$  or  $\mathbf{I}_n$  or  $\mathbf{I}_{n \times n}$  is the identity matrix. For example,

$$\mathbf{I}_3 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} = \text{diag}(1, 1, 1).$$

# Special Types of Square Matrices

- A square matrix  $A$  is upper triangular if  $a_{ij} = 0, \forall i > j$ .
- A square matrix  $A$  is lower triangular if  $a_{ij} = 0, \forall i < j$ .
- A square matrix  $A$  is diagonal if  $a_{ij} = 0, \forall i \neq j$ .
- Write one example for each of these types of matrices.



# Examples

- Upper triangular 
$$\begin{bmatrix} 1 & 0 & 4 & 6 \\ 0 & 2 & -3 & 5 \\ 0 & 0 & 8 & 2 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

- Lower triangular 
$$\begin{bmatrix} 1 & 0 & 0 \\ 2 & 4 & 0 \\ 3 & 5 & 6 \end{bmatrix}$$

- Diagonal 
$$\begin{bmatrix} 4 & 0 & 0 \\ 0 & 9 & 0 \\ 0 & 0 & -13 \end{bmatrix}.$$

# Matrix Transpose

If  $\mathbf{A} = [a_{ij}]$ , the transpose of  $\mathbf{A}$ , denoted  $\mathbf{A}'$ , is the matrix  $\mathbf{B} = [b_{ij}]$ ,  
where  $b_{ij} = a_{ji}$ ,  $\forall i = 1, \dots, m; \quad j = 1, \dots, n$ .

That is,  $\mathbf{B} = \mathbf{A}'$  is the matrix whose columns are the rows of  $\mathbf{A}$  and whose rows are the columns of  $\mathbf{A}$ .

# A Symmetric Matrix

A square matrix  $A$  is symmetric if  $A = A'$ .

$n \times n$

# Examples

- Find the transpose of

$$\begin{bmatrix} 4 & -2 \\ 3 & 7 \end{bmatrix}.$$

- Provide an example of a symmetric matrix.

# Examples

- $\begin{bmatrix} 4 & -2 \\ 3 & 7 \end{bmatrix}' = \begin{bmatrix} 4 & 3 \\ -2 & 7 \end{bmatrix}.$

- The matrix

$$\begin{bmatrix} 4 & 2 & -1 \\ 2 & 0 & 3 \\ -1 & 3 & 5 \end{bmatrix}$$

is symmetric.

# Matrix Multiplication

- Suppose

$$\mathbf{A}_{m \times n} = [a_{ij}] = \begin{bmatrix} \mathbf{a}'_{(1)} \\ \vdots \\ \mathbf{a}'_{(m)} \end{bmatrix} = [\mathbf{a}_1, \dots, \mathbf{a}_n]$$

and

$$\mathbf{B}_{n \times k} = [b_{ij}] = \begin{bmatrix} \mathbf{b}'_{(1)} \\ \vdots \\ \mathbf{b}'_{(n)} \end{bmatrix} = [\mathbf{b}_1, \dots, \mathbf{b}_k].$$

- Then

$$\begin{aligned} \mathbf{A}_{m \times n} \mathbf{B}_{n \times k} &= \mathbf{C}_{m \times k} = [c_{ij}] = \sum_{l=1}^n a_{il} b_{lj} = [c_{ij} = \mathbf{a}'_{(i)} \mathbf{b}_j] \\ &= [\mathbf{A} \mathbf{b}_1, \dots, \mathbf{A} \mathbf{b}_k] = \begin{bmatrix} \mathbf{a}'_{(1)} \mathbf{B} \\ \vdots \\ \mathbf{a}'_{(m)} \mathbf{B} \end{bmatrix} = \sum_{l=1}^n \mathbf{a}_{(l)} \mathbf{b}'_{(l)}. \end{aligned}$$

# Matrix Multiplication

- Suppose

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

and

$$\mathbf{B} = \begin{bmatrix} 5 & 6 \\ 7 & 8 \end{bmatrix}.$$

- Work out  $\mathbf{AB}$  using  $\mathbf{AB} = \sum_{l=1}^n \mathbf{a}_l \mathbf{b}'_{(l)}$ .

# Matrix Multiplication

$$\begin{aligned}\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \begin{bmatrix} 5 & 6 \\ 7 & 8 \end{bmatrix} &= \begin{bmatrix} 1 \\ 3 \end{bmatrix} [5 \ 6] + \begin{bmatrix} 2 \\ 4 \end{bmatrix} [7 \ 8] \\ &= \begin{bmatrix} 5 & 6 \\ 15 & 18 \end{bmatrix} + \begin{bmatrix} 14 & 16 \\ 28 & 32 \end{bmatrix} \\ &= \begin{bmatrix} 19 & 22 \\ 43 & 50 \end{bmatrix} .\end{aligned}$$



# Transpose of a Matrix Product



$$(AB)' = B'A'$$

- The transpose of a product is the product of the transposes in reverse order.

# Scalar Multiplication of a Matrix

If  $c \in \mathbb{R}$ , then  $c$  times the matrix  $A$  is the matrix whose  $i, j^{th}$  element is  $c$  times the  $i, j^{th}$  element of  $A$ ; i.e.,

$$c\underset{m \times n}{A} = c\underset{m \times n}{[a_{ij}]} = \underset{m \times n}{[ca_{ij}]} = \begin{bmatrix} ca_{11} & ca_{12} & \cdots & ca_{1n} \\ ca_{21} & ca_{22} & \cdots & ca_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ ca_{m1} & ca_{m2} & \cdots & ca_{mn} \end{bmatrix}.$$

# Linear Combination

If  $\mathbf{x}_1, \dots, \mathbf{x}_n \in \mathbb{R}^m$  and  $c_1, \dots, c_n \in \mathbb{R}$ , then

$$\sum_{i=1}^n c_i \mathbf{x}_i = c_1 \mathbf{x}_1 + \dots + c_n \mathbf{x}_n$$

is a linear combination (LC) of  $\mathbf{x}_1, \dots, \mathbf{x}_n$ .

The coefficients of the LC are  $c_1, \dots, c_n$ .

# Linear Independence and Linear Dependence

- A set of vectors  $\mathbf{x}_1, \dots, \mathbf{x}_n$  is linearly independent (LI) if

$$\sum_{i=1}^n c_i \mathbf{x}_i = \mathbf{0} \iff c_1 = \dots = c_n = 0.$$

- A set of vectors  $\mathbf{x}_1, \dots, \mathbf{x}_n$  is linearly dependent (LD) if

$$\exists c_1, \dots, c_n \text{ not all } 0 \ni \sum_{i=1}^n c_i \mathbf{x}_i = \mathbf{0}.$$

- Prove or disprove: any set of vectors that contains the vector  $\mathbf{0}$  is LD.
- Prove or disprove: The following set of vectors is LI.

$$\begin{bmatrix} 1 \\ -5 \\ 3 \end{bmatrix}, \begin{bmatrix} 7 \\ 4 \\ 1 \end{bmatrix}, \begin{bmatrix} 9 \\ -6 \\ 7 \end{bmatrix}.$$

- Suppose  $\mathbf{x}_1 = \mathbf{0}$  and  $\mathbf{x}_2, \dots, \mathbf{x}_n$  are any other vectors of the same dimension as  $\mathbf{x}_1$ .
- If we take  $c_1 = 1, c_2 = \dots = c_n = 0$ , then  $\sum_{i=1}^n c_i \mathbf{x}_i = \mathbf{0}$  and  $c_1, c_2, \dots, c_n$  are not all zero.
- Thus, any set of vectors containing  $\mathbf{0}$  is LD.

- If we take  $c_1 = 2, c_2 = 1, c_3 = -1$  then

$$c_1 \begin{bmatrix} 1 \\ -5 \\ 3 \end{bmatrix} + c_2 \begin{bmatrix} 7 \\ 4 \\ 1 \end{bmatrix} + c_3 \begin{bmatrix} 9 \\ -6 \\ 7 \end{bmatrix} = \mathbf{0}.$$

Thus the set of vectors is LD.

- One way to arrive at such solution is to search for a solution to the system of the equations:

$$c_1 + 7c_2 + 9c_3 = 0$$

$$-5c_1 + 4c_2 - 6c_3 = 0$$

$$3c_1 + c_2 + 7c_3 = 0.$$

## Fact V1:

The nonzero vectors  $x_1, \dots, x_n$  are LD  $\iff x_j$  is a LC of  $x_1, \dots, x_{j-1}$  for some  $j \in \{2, \dots, n\}$ .



## Proof of Fact V1:

( $\implies$ ) Suppose there exist  $c_1, \dots, c_n$  such that  $\sum_{i=1}^n c_i \mathbf{x}_i = \mathbf{0}$ . Let

$$j = \max\{i : c_i \neq 0\}.$$

Since  $\mathbf{x}_1, \dots, \mathbf{x}_n$  are nonzero,  $j > 1$ . Then

$$\begin{aligned} \sum_{i=1}^j c_i \mathbf{x}_i = \mathbf{0} &\implies \sum_{i=1}^{j-1} c_i \mathbf{x}_i = -c_j \mathbf{x}_j. \\ &\implies \sum_{i=1}^{j-1} \frac{-c_i}{c_j} \mathbf{x}_i = \mathbf{x}_j. \end{aligned}$$

( $\impliedby$ ) Suppose  $\mathbf{x}_j = \sum_{i=1}^{j-1} c_i \mathbf{x}_i$ , then  $\sum_{i=1}^n d_i \mathbf{x}_i = \mathbf{0}$ , where

$$d_i = \begin{cases} c_i & \text{if } i < j \\ -1 & \text{if } i = j \\ 0 & \text{if } i > j. \end{cases}$$



# Orthogonality

- The two vectors  $\mathbf{x}, \mathbf{y}$  are orthogonal to each other if their inner product is zero, i.e.,

$$\mathbf{x}'\mathbf{y} = \mathbf{y}'\mathbf{x} = \sum_{i=1}^n x_i y_i = 0.$$

- The length of a vector, also known as its Euclidean norm, is

$$\|\mathbf{x}\| := \sqrt{\mathbf{x}'\mathbf{x}} = \sqrt{\sum_{i=1}^n x_i^2}.$$

- The vectors  $\mathbf{x}_1, \dots, \mathbf{x}_n$  are mutually orthogonal if

$$\mathbf{x}_i' \mathbf{x}_j = 0, \quad \forall i \neq j.$$

- The vectors  $\mathbf{x}_1, \dots, \mathbf{x}_n$  are mutually orthonormal if

$$\mathbf{x}_i' \mathbf{x}_j = 0 \quad \forall i \neq j, \text{ and } \|\mathbf{x}_i\| = 1 \quad \forall i = 1, \dots, n.$$

- Write down a set of mutually orthogonal but not mutually orthonormal vectors.
- Write down a set of mutually orthonormal vectors.

- $\begin{bmatrix} 2 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix}$  are mutually orthogonal but not mutually orthonormal.
- $\begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix}$  are mutually orthonormal.
- $\begin{bmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{bmatrix}, \begin{bmatrix} \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} \end{bmatrix}$  are mutually orthonormal.

# Orthogonal Matrix

- A square matrix with mutually orthonormal columns is called an orthogonal matrix.

- Show that if  $Q$  is orthogonal, then  $Q'Q = I$ .
- Show that if  $Q$  is orthogonal and  $x$  is any vector of appropriate dimension, then  $\|Qx\| = \|x\|$ .

- $Q'Q = [q_i'q_j]$ , where  $Q = [q_1, \dots, q_n]$ .
- By orthogonality of  $Q$ ,  $q_1, \dots, q_n$  are mutually orthonormal.
- Thus,

$$q_i'q_j = 0 \quad \forall i \neq j$$

and

$$\|q_i\| = 1 \quad \forall i = 1, \dots, n.$$

$$\therefore Q'Q = I.$$



$$\begin{aligned}
\|Qx\| &= \sqrt{(Qx)'Qx} \\
&= \sqrt{x'Q'Qx} \\
&= \sqrt{x'Ix} \\
&= \sqrt{x'x} \\
&= \|x\|.
\end{aligned}$$

An orthogonal matrix  $Q$  is sometimes called a rotation matrix because if a vector  $x$  is premultiplied by  $Q$ , the result  $(Qx)$  is the vector  $x$  rotated to a new position in  $\mathbb{R}^n$ .

# Vector Space

A vector space  $\mathcal{S}$  is a set of vectors that is closed under addition (i.e., if  $\mathbf{x}_1 \in \mathcal{S}, \mathbf{x}_2 \in \mathcal{S}$ , then  $\mathbf{x}_1 + \mathbf{x}_2 \in \mathcal{S}$ ) and closed under scalar multiplication (i.e., if  $c \in \mathbb{R}, \mathbf{x} \in \mathcal{S}$ , then  $c\mathbf{x} \in \mathcal{S}$ ).

In other words,

$$c_1\mathbf{x}_1 + c_2\mathbf{x}_2 \in \mathcal{S} \quad \forall c_1, c_2 \in \mathbb{R}; \mathbf{x}_1, \mathbf{x}_2 \in \mathcal{S}.$$

- Is  $\{\mathbf{x} \in \mathbb{R}^n : \|\mathbf{x}\| = 1\}$  a vector space?
- Is  $\{\mathbf{x} \in \mathbb{R}^n : \mathbf{1}'\mathbf{x} = 0\}$  a vector space?
- Is  $\{\mathbf{A}\mathbf{x} : \mathbf{x} \in \mathbb{R}^m\}$  a vector space?  
 $n \times m$

- Suppose  $\mathbf{y} \in \mathbb{R}^n$ ,  $c \in \mathbb{R}$  and  $\|\mathbf{y}\| = 1$ , then

$$\begin{aligned}\|c\mathbf{y}\| &= \sqrt{(c\mathbf{y})'c\mathbf{y}} \\ &= \sqrt{c^2\mathbf{y}'\mathbf{y}} \\ &= |c|\|\mathbf{y}\| = |c|.\end{aligned}$$

- Thus  $\mathbf{y} \in \{\mathbf{x} \in \mathbb{R}^n : \|\mathbf{x}\| = 1\}$  does not imply that  $c\mathbf{y} \in \{\mathbf{x} \in \mathbb{R}^n : \|\mathbf{x}\| = 1\}$ . Therefore, this set is not a vector space.

- Let

$$\mathcal{S} = \{\mathbf{x} \in \mathbb{R}^n : \mathbf{1}'\mathbf{x} = 0\}.$$

- Suppose  $c_1, c_2 \in \mathbb{R}$  and  $\mathbf{x}_1, \mathbf{x}_2 \in \mathcal{S}$ , then

$$\mathbf{1}'(c_1\mathbf{x}_1 + c_2\mathbf{x}_2) = c_1\mathbf{1}'\mathbf{x}_1 + c_2\mathbf{1}'\mathbf{x}_2 = 0.$$

- Thus  $c_1\mathbf{x}_1 + c_2\mathbf{x}_2 \in \mathcal{S}$  and it follows that  $\mathcal{S}$  is a vector space.

- Let

$$\mathcal{S} = \left\{ \underset{n \times m}{A} \mathbf{x} : \mathbf{x} \in \mathbb{R}^m \right\}.$$

- Suppose  $c_1, c_2 \in \mathbb{R}$  and  $\mathbf{y}_1, \mathbf{y}_2 \in \mathcal{S}$ .

- $\mathbf{y}_1, \mathbf{y}_2 \in \mathcal{S} \implies \exists \mathbf{x}_1, \mathbf{x}_2 \in \mathbb{R}^m \ni$

$$\mathbf{y}_1 = A\mathbf{x}_1 \text{ and } \mathbf{y}_2 = A\mathbf{x}_2.$$

- Thus,

$$c_1\mathbf{y}_1 + c_2\mathbf{y}_2 = c_1A\mathbf{x}_1 + c_2A\mathbf{x}_2 = A(c_1\mathbf{x}_1 + c_2\mathbf{x}_2).$$

$\therefore c_1\mathbf{x}_1 + c_2\mathbf{x}_2 \in \mathbb{R}^m, c_1\mathbf{y}_1 + c_2\mathbf{y}_2 \in \mathcal{S}$ . It follows that  $\mathcal{S}$  is a vector space.

# Generators of a Vector Space

A vector space  $\mathcal{S}$  is said to be generated by a set of vectors  $x_1, \dots, x_n$  if

$$x \in \mathcal{S} \implies x = \sum_{i=1}^n c_i x_i \text{ for some } c_1, \dots, c_n \in \mathbb{R}.$$



# Span of a Set of Vectors

- The span of vectors  $\mathbf{x}_1, \dots, \mathbf{x}_n$  is the set of all LC of  $\mathbf{x}_1, \dots, \mathbf{x}_n$ , i.e.,

$$\text{span}\{\mathbf{x}_1, \dots, \mathbf{x}_n\} = \left\{ \sum_{i=1}^n c_i \mathbf{x}_i : c_1, \dots, c_n \in \mathbb{R} \right\}.$$

- $\text{span}\{\mathbf{x}_1, \dots, \mathbf{x}_n\}$  is the vector space generated by  $\mathbf{x}_1, \dots, \mathbf{x}_n$ .

Find a set of vectors that generates the space

$$\{\mathbf{x} \in \mathbb{R}^3 : \mathbf{1}'\mathbf{x} = 0\};$$

i.e., find a set of vectors whose span is

$$\mathcal{S} = \{\mathbf{x} \in \mathbb{R}^3 : \mathbf{1}'\mathbf{x} = 0\}.$$

- Let  $\mathbf{x}_1 = \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix}$ ,  $\mathbf{x}_2 = \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix}$ . Note that  $\mathbf{1}'\mathbf{x}_1 = 0$  and  $\mathbf{1}'\mathbf{x}_2 = 0$ .

Thus,  $\mathbf{x}_1, \mathbf{x}_2 \in \mathcal{S}$  so that  $\text{span}\{\mathbf{x}_1, \mathbf{x}_2\} \subseteq \mathcal{S}$ .

- Now suppose  $\mathbf{y} = \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} \in \mathcal{S}$ .

Then  $0 = \mathbf{1}'\mathbf{y} = y_1 + y_2 + y_3 \implies y_3 = -y_1 - y_2$  so that

$$y_1\mathbf{x}_1 + y_2\mathbf{x}_2 = \begin{bmatrix} y_1 \\ 0 \\ -y_1 \end{bmatrix} + \begin{bmatrix} 0 \\ y_2 \\ -y_2 \end{bmatrix} = \begin{bmatrix} y_1 \\ y_2 \\ -y_1 - y_2 \end{bmatrix} = \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} = \mathbf{y}.$$

$\therefore \mathcal{S} \subseteq \text{span}\{\mathbf{x}_1, \mathbf{x}_2\}$ , and  $\mathcal{S} = \text{span}\{\mathbf{x}_1, \mathbf{x}_2\}$ .

# Basis of a Vector Space

If a vector space  $\mathcal{S}$  is generated by LI vectors  $\mathbf{x}_1, \dots, \mathbf{x}_n$ , then  $\mathbf{x}_1, \dots, \mathbf{x}_n$  form a basis for  $\mathcal{S}$ .

## Fact V2:

Suppose  $\mathbf{a}_1, \dots, \mathbf{a}_n$  form a basis for a vector space  $\mathcal{S}$ . If  $\mathbf{b}_1, \dots, \mathbf{b}_k$  are LI vectors in  $\mathcal{S}$ , then  $k \leq n$ .

## Proof of Fact V2:

- Because  $\mathbf{a}_1, \dots, \mathbf{a}_n$  form a basis for  $\mathcal{S}$  and  $\mathbf{b}_1 \in \mathcal{S}$ ,  $\mathbf{b}_1 = \sum_{i=1}^n c_i \mathbf{a}_i$  for some  $c_1, \dots, c_n \in \mathbb{R}$ . Thus,  $\mathbf{a}_1, \dots, \mathbf{a}_n, \mathbf{b}_1$  are LD by Fact V1.
- Again, using V1, we have  $\mathbf{a}_j$  a LC of  $\mathbf{b}_1, \mathbf{a}_1, \dots, \mathbf{a}_{j-1}$  for some  $j \in \{1, 2, \dots, n\}$ .

- Thus,  $\mathbf{b}_1, \mathbf{a}_1, \dots, \mathbf{a}_{j-1}, \mathbf{a}_{j+1}, \dots, \mathbf{a}_n$  generate  $\mathcal{S}$ . It follows that  $\mathbf{b}_1, \mathbf{a}_1, \dots, \mathbf{a}_{j-1}, \mathbf{a}_{j+1}, \dots, \mathbf{a}_n, \mathbf{b}_2$  is a LD set of vectors by V1.
- Again by V1, one of the vectors  $\mathbf{b}_1, \mathbf{b}_2, \mathbf{a}_1, \dots, \mathbf{a}_{j-1}, \mathbf{a}_{j+1}, \dots, \mathbf{a}_n$  is a LC of the preceding vectors. It is not  $\mathbf{b}_2$   $\because \mathbf{b}_1, \dots, \mathbf{b}_k$  are LI.

- Thus  $\mathbf{b}_1, \mathbf{b}_2$  and  $n - 2$  of  $\mathbf{a}_1, \dots, \mathbf{a}_n$  generate  $\mathcal{S}$ .
- If  $k > n$ , we can continue adding  $\mathbf{b}$  vectors and deleting  $\mathbf{a}$  vectors to get  $\mathbf{b}_1, \dots, \mathbf{b}_n$  generates  $\mathcal{S}$ . However, then V1 would imply  $\mathbf{b}_1, \dots, \mathbf{b}_{n+1}$  are LD. This contradicts LI of  $\mathbf{b}_1, \dots, \mathbf{b}_k \therefore k \leq n$ . □



## Fact V3:

If  $\{a_1, \dots, a_n\}$  and  $\{b_1, \dots, b_k\}$  each provide a basis for a vector space  $S$ , then  $n = k$ .

*Proof:* By V2, we have  $k \leq n$  and  $n \leq k$ .  $\therefore k = n$ . □

# Dimension of a Vector Space

A basis for a vector space is not unique, but the number of vectors in the basis, known as dimension of the vector space, is unique.

Find  $\dim(\mathcal{S})$ , the dimension of vector space  $\mathcal{S}$ , for

$$\mathcal{S} = \{\mathbf{x} \in \mathbb{R}^3 : \mathbf{1}'\mathbf{x} = 0\}.$$

- As demonstrated previously,

$$\text{span} \left\{ \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix} \right\} = \mathcal{S}.$$

- Because  $\left\{ \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix} \right\}$  is a LI set of vectors,  $\left\{ \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix} \right\}$  forms a basis for  $\mathcal{S}$ .

- Thus,

$\dim(\mathcal{S}) = 2$  (even though dimension of vectors in  $\mathcal{S}$  is 3).

Consider the set  $\{\mathbf{0}\}_{n \times 1}$ . Is this a vector space? If so, what is its dimension?

- $\{\mathbf{0}\}$  is a vector space because

$$c_1\mathbf{x}_1 + c_2\mathbf{x}_2 = \mathbf{0} \in \{\mathbf{0}\} \quad \forall c_1, c_2 \in \mathbb{R} \text{ and } \forall \mathbf{x}_1, \mathbf{x}_2 \in \{\mathbf{0}\}.$$

- The set  $\{\mathbf{0}\}$  generates the vector space  $\{\mathbf{0}\}$ . However,  $\{\mathbf{0}\}$  is not a LI set of vectors.  $\dim(\{\mathbf{0}\}) = 0$ .

## Fact V4:

Suppose  $\mathbf{a}_1, \dots, \mathbf{a}_n$  are LI vectors in a vector space  $\mathcal{S}$  with dimension  $n$ .  
Then  $\mathbf{a}_1, \dots, \mathbf{a}_n$  form a basis for  $\mathcal{S}$ .

## Proof of Fact V4:

- It suffices to show that  $\mathbf{a}_1, \dots, \mathbf{a}_n$  generate  $\mathcal{S}$ .
- Let  $\mathbf{a}$  denote an arbitrary vector in  $\mathcal{S}$ .
- By V2,  $\mathbf{a}_1, \dots, \mathbf{a}_n, \mathbf{a}$  are LD. By V1,  $\mathbf{a} = \sum_{i=1}^n c_i \mathbf{a}_i$  for some  $c_1, \dots, c_n \in \mathbb{R}$ .
- Thus

$$\mathcal{S} = \left\{ \sum_{i=1}^n c_i \mathbf{x}_i : c_1, \dots, c_n \in \mathbb{R} \right\},$$

and the result follows. □



## Fact V5:

If  $\mathbf{a}_1, \dots, \mathbf{a}_k$  are LI vectors in an  $n$ -dimensional vector space  $\mathcal{S}$ , then there exists a basis for  $\mathcal{S}$  that contains  $\mathbf{a}_1, \dots, \mathbf{a}_k$ .

## Proof of Fact V5:

- $k \leq n$  by V2.
- If  $k = n$ , the result follows from V4.
- Suppose  $k < n$ . Then, there exist  $\mathbf{a}_{k+1} \in \mathcal{S}$  such that  $\mathbf{a}_1, \dots, \mathbf{a}_{k+1}$  are LI. Because if not,  $\mathbf{a}_1, \dots, \mathbf{a}_k$  would generate  $\mathcal{S}$  (by V1), and thus be a basis of dimension  $k < n$ , which is impossible by V3. Similarly, we can continue to add vectors to  $\{\mathbf{a}_1, \dots, \mathbf{a}_{k+1}\}$  until we have  $\mathbf{a}_1, \dots, \mathbf{a}_n$  LI vectors. The result follows from V4.  $\square$

## Fact V6:

If  $\mathbf{a}_1, \dots, \mathbf{a}_k$  are LI and orthonormal vectors in  $\mathbb{R}^n$ , then there exist  $\mathbf{a}_{k+1}, \dots, \mathbf{a}_n$  such that  $\mathbf{a}_1, \dots, \mathbf{a}_n$  are LI and orthonormal.

*Proof:* The proof is left as an exercise for the reader.



# Rank of a Matrix

It can be shown that

- the (maximum) number of LI rows of a matrix  $A$  is the same as the (maximum) number of LI columns of  $A$ .  
 $m \times n$
- This number of LI rows or columns is known as the rank of  $A$  and is denoted  $rank(A)$  or  $r(A)$ .  
 $m \times n$

- If  $r(\underset{m \times n}{A}) = m$ ,  $\underset{m \times n}{A}$  is said to have full row rank.
- If  $r(\underset{m \times n}{A}) = n$ ,  $\underset{m \times n}{A}$  is said to have full column rank.

# Inverse of a Matrix

- If  $r(\underset{n \times n}{A}) = n$ , there exists a matrix  $\underset{n \times n}{B}$  such that  $\underset{n \times n}{A} \underset{n \times n}{B} = \underset{n \times n}{I}$ .
- Such a matrix  $B$  is called the inverse of  $A$  and is denoted  $A^{-1}$ .

- Prove that  $r(\underset{n \times n}{A}) = n \iff \exists \underset{n \times n}{B} \ni \underset{n \times n}{A} \underset{n \times n}{B} = \underset{n \times n}{I}.$

- Prove that  $\underset{n \times n}{A} \underset{n \times n}{B} = \underset{n \times n}{I} \implies \underset{n \times n}{B} \underset{n \times n}{A} = \underset{n \times n}{I}$

- Thus  $AA^{-1} = A^{-1}A = I.$

*Proof:*

$(\implies)$ :

- The columns of  $A$  form a basis for  $\mathbb{R}^n$  by V4. Thus, there exists a LC of columns of  $A$  that equals  $e_i$  for all  $i = 1, \dots, n$ , where  $e_i$  is the  $i^{\text{th}}$  column of the identity matrix  $I$ .
- Let  $b_i$  denote the coefficients of the LC of the columns of  $A$  that yields  $e_i$ . Then, with  $B = [b_1, \dots, b_n]$ , we have  $AB = [Ab_1, \dots, Ab_n] = [e_1, \dots, e_n] = I$ .



( $\Leftarrow$ ):

- If  $\exists \underset{n \times n}{B} \underset{n \times n}{\ni} \underset{n \times n}{A} \underset{n \times n}{B} = \underset{n \times n}{I}$ , then the columns of  $A$  generate  $\mathbb{R}^n$   
 $\therefore \forall \mathbf{x} \in \mathbb{R}^n, \mathbf{ABx} = \mathbf{Ix} = \mathbf{x}$ .
- If the columns of  $A$  were LD, then a subset of the columns of  $A$  would be LI and also generate  $\mathbb{R}^n$ .

- However, such a subset would be a basis for  $\mathbb{R}^n$  and thus must have  $n$  elements.

- Thus, the columns of  $A$  must be LI. Hence,  $r_{n \times n}(A) = n$ . □



$$\underset{n \times n}{A} \underset{n \times n}{B} = \underset{n \times n}{I} \implies \text{Columns of } \underset{n \times n}{A} \text{ are LI}$$

$$\implies \text{Rows of } \underset{n \times n}{A} \text{ are LI}$$

$$\implies \text{Rows of } \underset{n \times n}{A} \text{ are a basis for } \mathbb{R}^n$$

$$\implies \exists \underset{n \times n}{C} \ni \underset{n \times n}{C} \underset{n \times n}{A} = \underset{n \times n}{I}.$$

- Thus,

$$AB = I \implies CAB = CI$$

$$\implies IB = C$$

$$\implies B = C.$$

- $\therefore AA^{-1} = A^{-1}A = I$



# Singular / Nonsingular Matrix

- If  $r(\underset{n \times n}{A}) = n$ ,  $\underset{n \times n}{A}$  is said to be nonsingular.
- If  $r(\underset{n \times n}{A}) < n$ ,  $\underset{n \times n}{A}$  is said to be singular.

## Inverse of a Nonsingular $2 \times 2$ Matrix

- $\begin{bmatrix} a & b \\ c & d \end{bmatrix}^{-1} = \frac{1}{ad-bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$  if  $ad - bc \neq 0$ .

- $\begin{bmatrix} a & b \\ c & d \end{bmatrix}$  is singular if  $ad - bc = 0$ .

# Column Space of a Matrix

The column space of a matrix  $\underset{m \times n}{A}$ , denoted by  $\mathcal{C}(A)$ , is the vector space generated by the columns of  $\underset{m \times n}{A}$ ; i.e.,

$$\mathcal{C}(A) = \{A\mathbf{x} : \mathbf{x} \in \mathbb{R}^n\}.$$

$\dim(\mathcal{C}(\mathbf{A})) = r(\mathbf{A})$  because the largest possible subset of LI columns of  $\mathbf{A}$  is a basis for  $\mathcal{C}(\mathbf{A})$ .

- Let

$$A = \begin{bmatrix} 1 & 0 & 3 \\ 2 & 1 & 4 \\ 4 & 2 & 8 \end{bmatrix}.$$

- Find  $r(A)$ .
- Give a basis for  $\mathcal{C}(A)$ .
- Characterize  $\mathcal{C}(A)$ .



$$3 \begin{bmatrix} 1 \\ 2 \\ 4 \end{bmatrix} - 2 \begin{bmatrix} 0 \\ 1 \\ 2 \end{bmatrix} = \begin{bmatrix} 3 \\ 4 \\ 8 \end{bmatrix}.$$

- Thus, the columns of  $A$  are LD and  $r(A) < 3$ .

$$c_1 \begin{bmatrix} 1 \\ 2 \\ 4 \end{bmatrix} + c_2 \begin{bmatrix} 0 \\ 1 \\ 2 \end{bmatrix} = \begin{bmatrix} c_1 \\ 2c_1 + c_2 \\ 4c_1 + 2c_2 \end{bmatrix} = \mathbf{0} \implies c_1 = c_2 = 0. \therefore r(A) = 2.$$

- A basis for  $\mathcal{C}(\mathbf{A})$  is given by  $\left\{ \begin{bmatrix} 1 \\ 2 \\ 4 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 2 \end{bmatrix} \right\}$ .

- $\mathbf{x} \in \mathcal{C}(\mathbf{A}) \implies \mathbf{x} = c_1 \begin{bmatrix} 1 \\ 2 \\ 4 \end{bmatrix} + c_2 \begin{bmatrix} 0 \\ 1 \\ 2 \end{bmatrix}$  for some  $c_1, c_2 \in \mathbb{R}^n$ .

- Note

$$\left\{ \mathbf{x} = \begin{bmatrix} c_1 \\ 2c_1 + c_2 \\ 4c_1 + 2c_2 \end{bmatrix} : c_1, c_2 \in \mathbb{R} \right\} = \mathcal{C}(\mathbf{A})$$

is the set of vectors in  $\mathbb{R}^3$  where the first component is arbitrary and the third component is twice the second component, i.e.,

$$\{\mathbf{x} \in \mathbb{R}^3 : 2x_2 = x_3\}.$$

## Result A.1:

$$\textit{rank}(\mathbf{AB}) \leq \min\{\textit{rank}(\mathbf{A}), \textit{rank}(\mathbf{B})\}$$

## Proof of Result A.1:

- Let  $\mathbf{b}_1, \dots, \mathbf{b}_n$  denote the columns of  $\mathbf{B}$  so that  $\mathbf{B} = [\mathbf{b}_1, \dots, \mathbf{b}_n]$ .
- Then  $\mathbf{AB} = [\mathbf{Ab}_1, \dots, \mathbf{Ab}_n]$ . This implies that the columns of  $\mathbf{AB}$  are in  $\mathcal{C}(\mathbf{A})$ .

- $\dim(\mathcal{C}(\mathbf{A}))$  is  $\text{rank}(\mathbf{A})$ .
- There does not exist a set of LI vectors in  $\mathcal{C}(\mathbf{A})$  with more than  $\text{rank}(\mathbf{A})$  vectors by Fact V2.
- It follows that  $\text{rank}(\mathbf{AB}) \leq \text{rank}(\mathbf{A})$ .

- It remains to show that

$$\text{rank}(\mathbf{AB}) \leq \text{rank}(\mathbf{B}).$$

- Note that  $\text{rank}(\mathbf{AB})$  is the same as  $\text{rank}((\mathbf{AB})') = \text{rank}(\mathbf{B}'\mathbf{A}')$ .
- Our previous argument shows that

$$\text{rank}(\mathbf{B}'\mathbf{A}') \leq \text{rank}(\mathbf{B}') = \text{rank}(\mathbf{B}).$$



Provide an example where

$$\text{rank}(\mathbf{AB}) < \min\{\text{rank}(\mathbf{A}), \text{rank}(\mathbf{B})\}.$$





$$\mathbf{A} = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}, \mathbf{B} = \begin{bmatrix} 1 \\ -1 \end{bmatrix}.$$

• Then

$$\mathbf{AB} = \mathbf{0}.$$

• Therefore,

$$\text{rank}(\mathbf{AB}) = 0, \quad \text{rank}(\mathbf{A}) = 1, \quad \text{rank}(\mathbf{B}) = 1.$$

## Result A.2:

- (a) If  $A = BC$ , then  $\mathcal{C}(A) \subseteq \mathcal{C}(B)$ ,
- (b) If  $\mathcal{C}(A) \subseteq \mathcal{C}(B)$ , then there exist  $C$  such that  $A = BC$ .

## Proof of A.2(a):

- Suppose  $x \in \mathcal{C}(A)$  then  $\exists y \ni x = Ay$ .
- Now  $A = BC \implies x = BCy$ .
- Thus,  $\exists z \ni x = Bz$  (namely,  $z = Cy$ ).  $\therefore x \in \mathcal{C}(B)$ .
- We have shown  $x \in \mathcal{C}(A) \implies x \in \mathcal{C}(B)$ .  $\therefore \mathcal{C}(A) \subseteq \mathcal{C}(B)$ . □

## Proof of A.2(b):

- Let  $\mathbf{a}_1, \dots, \mathbf{a}_n$  denote the columns of  $\mathbf{A}$ .  
 $\mathcal{C}(\mathbf{A}) \subseteq \mathcal{C}(\mathbf{B}) \implies \mathbf{a}_1, \dots, \mathbf{a}_n \in \mathcal{C}(\mathbf{B})$ .
- Let  $\mathbf{c}_i$  be such that  $\mathbf{B}\mathbf{c}_i = \mathbf{a}_i \ \forall \ i = 1, \dots, n$ . Then denote  $\mathbf{C} = [\mathbf{c}_1, \dots, \mathbf{c}_n]$ .
- It follows that

$$\begin{aligned}\mathbf{BC} &= \mathbf{B}[\mathbf{c}_1, \dots, \mathbf{c}_n] \\ &= [\mathbf{B}\mathbf{c}_1, \dots, \mathbf{B}\mathbf{c}_n] \\ &= [\mathbf{a}_1, \dots, \mathbf{a}_n] = \mathbf{A}.\end{aligned}$$