

# Econ 508-A

## Matrix Algebra and Linear transformations

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## NOTATION

$$A_{M \times N} = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1N} \\ a_{21} & a_{22} & \dots & a_{2N} \\ \vdots & \vdots & \vdots & \vdots \\ a_{M1} & a_{M2} & \dots & a_{MN} \end{bmatrix}$$

$$= [\mathbf{a}_{ij}]$$

$$= [\mathbf{a}^1 \dots \mathbf{a}^N], \text{ where } \mathbf{a}^i \in \mathbb{R}^M, i = 1, \dots, N,$$

# BASIC OPERATIONS

1. Scalar multiplication:

$$\alpha A = [\alpha a_{ij}]$$

2. Addition:

$$A + B = [a_{ij} + b_{ij}]$$

3. Multiplication:

$$A_{M \times N} B_{N \times Q} = D_{M \times Q} = [d_{ij}]$$

$$\text{where } d_{ij} = \sum_{k=1}^N a_{ik} b_{kj}$$

# LAW OF MATRIX ALGEBRA

Theorem:

1.  $A + B = B + A$
2.  $(A + B) + C = A + (B + C)$
3.  $(AB)C = A(BC)$
4.  $A(B + C) = AB + AC$
5.  $(A + B)C = AC + BC$

# TRANSPOSE

Definition: Transpose of  $A_{M \times N} = [a_{ij}]$  :

$$A'_{N \times M} = [a'_{ij}] = [a_{ji}]$$

Theorem:

1.  $(A + B)' = A' + B'$
2.  $(AB)' = B' A'$
3.  $(\alpha A)' = \alpha A', \alpha \in \mathbb{R}$

# SPECIAL CLASSES OF MATRICES

1. Square matrix:

$$A_{N \times N} = [a_{ij}];$$

$a_{ii}$  : diagonal elements

2. Diagonal matrix:

$$A_{N \times N} = [a_{ij}]; a_{ij} = 0 \text{ for } i \neq j$$

3. Identity matrix:

$$I_N = [a_{ij}]; a_{ii} = 1, a_{ij} = 0 \text{ for } i \neq j;$$

$$I_M A_{M \times N} = A_{M \times N} = A_{M \times N} I_N$$

4. Symmetric matrix:

$$A_{N \times N} = [a_{ij}]; a_{ij} = a_{ji} \text{ for every } i, j. A' = A$$

# TRACE

Definition: given  $A_{N \times N}$ ,

$$\text{Tr}(A) = \sum_{i=1}^N a_{ii}$$

Theorem:

Given  $A_{N \times N}, B_{N \times N}$ ,

1.  $\text{tr}(A + B) = \text{tr}(A) + \text{tr}(B)$
2.  $\text{tr}(AB) = \text{tr}(BA)$

# DETERMINANTS

Given  $A_{N \times N}$ ,  $|A|$  denotes the determinant of  $A$

Definition:  $(i, j)$ th **minor** of  $A$  :

$A_{ij}$  :  $(N - 1) \times (N - 1)$  matrix obtained by deleting row  $i$  and col  $j$

Definition:  $(i, j)$ th **cofactor** of  $A$ :

$$c_{ij}(A) = (-1)^{(i+j)} |A_{ij}|$$

Definition:

$$|A| = \sum_{k=1}^N a_{ik} c_{ik}(A)$$

(expansion along  $i$ th row ( $k = j$ ) or  $j$ th col ( $k = i$ ))



# PROPERTIES OF DETERMINANTS

Theorem:

1.  $|A| = |A^T|$
2.  $|AB| = |A| |B|$
3.  $|I_N| = 1$
4.  $|kA| = k^N |A|$
5.  $|A| = 0$  if  $A$  has a row or col of 0's
6. If we multiply a row (col) of  $A$  by  $k$  to get  $\hat{A}$ ,  
$$|\hat{A}| = k |A|$$
7. If we add a multiple of a row (col) to another row (col) to get  $\hat{A}$ , 
$$|\hat{A}| = |A|$$

# INVERSE

Definition: given  $A_{N \times N}$ , its inverse, denoted  $A^{-1}$  is the  $N \times N$  matrix such that

$$A^{-1}A = AA^{-1} = I_N$$

Theorem:

1.  $A^{-1}$  exists  $\Leftrightarrow |A| \neq 0$
2. If  $A^{-1}$  exists, it is unique

We say that  $A$  is **nonsingular** if  $A^{-1}$  exists.

# PROPERTIES OF INVERSES

Theorem:

$$1. \quad (A^{-1})^{-1} = A$$

$$2. \quad (AB)^{-1} = B^{-1}A^{-1}$$

$$3. \quad |A^{-1}| = \frac{1}{|A|}$$

$$4. \quad (A')^{-1} = (A^{-1})'$$

# COMPUTATION OF THE INVERSE

Definition: the Adjoint of an  $N \times N$  matrix  $A$  is given by

$$\text{Adj}A = [C(A)]^T,$$

$$\text{where } C(A) = [c_{ij}(A)]$$

The inverse of  $A$ ,  $A^{-1}$  is given by:

$$A^{-1} = \frac{1}{|A|} \text{Adj}A$$

# LINEAR DEPENDENCE AND INDEPENDENCE

Definition: nonzero vectors  $\{x^1, \dots, x^K\} \in \mathbb{R}^N$  are **linearly dependent** if  $\exists$  scalars  $\alpha_1, \dots, \alpha_K$  not all 0 such that:

$$0_N = \sum_{i=1}^K \alpha_i x^i$$

where  $0_N$  is the  $N$ -dimensional zero vector.

Definition:  $\{x^1, \dots, x^K\} \in \mathbb{R}^N$  are **linearly independent** if

$$0_N = \sum_{i=1}^K \alpha_i x^i \Rightarrow \alpha_k = 0 \quad \forall k = 1, \dots, K$$

# RANK

Definition: the column (row) rank of  $A_{M \times N}$  is the max number of linearly independent cols (rows).

$$\begin{aligned}\text{col rank}(A_{M \times N}) &\leq N \\ \text{row rank}(A_{M \times N}) &\leq M\end{aligned}$$

Theorem:  $\text{col rank}(A) = \text{row rank}(A) = \text{rank}(A)$

$$\text{Lemma: } \text{rank}(A) \leq \min \{N, M\}$$

$$\text{Corollary: } \text{rank}(A') = \text{rank}(A)$$

Definition:  $A_{M \times N}$ ,  $M \leq N$  is of **full rank** iff

$$\text{rank}(A) = M$$

# RANK OF SQUARE MATRICES

Theorem:  $|A| \neq 0 \Leftrightarrow \text{rank}(A_{N \times N}) = N \Leftrightarrow A^{-1}$  exists

# SIMULTANEOUS LINEAR EQUATIONS

A linear equation:

$$a_1x_1 + a_2x_2 + \dots + a_Nx_N = b$$

A system of  $M$  simultaneous linear equations:

$$a_{11}x_1 + a_{12}x_2 + \dots + a_{1N}x_N = b_1$$

$$a_{21}x_1 + a_{22}x_2 + \dots + a_{2N}x_N = b_2$$

$$\vdots$$

$$a_{M1}x_1 + a_{M2}x_2 + \dots + a_{MN}x_N = b_M$$

Can be represented in matrix terms:

$$A_{M \times N} x_{N \times 1} = b_{M \times 1}$$



# INTERPRETATION OF $Ax$

A linear combination of the columns of  $A$ :

$$\begin{aligned} Ax &= a_{11}x_1 + a_{12}x_2 + \dots + a_{1N}x_N \\ &\quad a_{21}x_1 + a_{22}x_2 + \dots + a_{2N}x_N \\ &\quad \vdots \\ &\quad a_{M1}x_1 + a_{M2}x_2 + \dots + a_{MN}x_N \end{aligned}$$

$$= \begin{bmatrix} a_{11} \\ a_{21} \\ \vdots \\ a_{M1} \end{bmatrix} x_1 + \dots + \begin{bmatrix} a_{1N} \\ a_{2N} \\ \vdots \\ a_{MN} \end{bmatrix} x_N$$

$$= a^1 x_1 + \dots + a^N x_N$$

## EXAMPLE

Consider the system of equations of the form  $Ax = b$  given by

$$\begin{bmatrix} 1 & 0 & 1 \\ 5 & 4 & 9 \\ 2 & 4 & 6 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}$$

For which values of  $b \in \mathbb{R}^3$  is the system solvable?

Re-write the system: 
$$\begin{bmatrix} 1 \\ 5 \\ 2 \end{bmatrix} x_1 + \begin{bmatrix} 0 \\ 4 \\ 4 \end{bmatrix} x_2 + \begin{bmatrix} 1 \\ 9 \\ 6 \end{bmatrix} x_3 = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}$$

The system is solvable iff  $b$  can be written as a linear combination of the columns of  $A$ .

## EXAMPLE (CONTINUED)

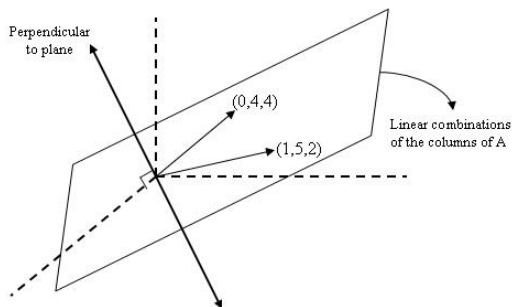
What if  $b = 0$ ?

- ▶ The trivial solution is always  $x = 0$ .
- ▶ Can there be other solutions?

Notice that

$$\begin{bmatrix} 1 & 0 & 1 \\ 5 & 4 & 9 \\ 2 & 4 & 6 \end{bmatrix} \begin{bmatrix} \alpha \\ \alpha \\ -\alpha \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

# GEOMETRIC INTUITION



$Ax = b$  can be solved iff  $b$  lies in the plane that is spanned by the column vectors of  $A$ .

The plane is called the **column space** of  $A$ .

The line perpendicular to the plane is called the **null space** of  $A$ .

# COLUMN SPACE AND NULL SPACE (HANDWAVY)

The **column space** of a matrix  $A_{M \times N}$  is the set of vectors that can be written as a linear combination of the columns of  $A$ .

The dimension of the column space is the rank of  $A$ .

The **null space or kernel** of  $A$  is the set of vectors  $x$  such that  $Ax = 0$ .

The dimension of the null space of  $A$  is equal to  $N - \text{Rank}(A)$ .

# SUMMARY

Given  $A$ , a  $N \times N$  matrix:

Nonsingular	Singular
$A$ is invertible	$A$ is not invertible
Cols are independent	Cols are dependent
Rows are independent	Rows are dependent
$Ax = 0$ has one solution $x = 0$	$Ax = 0$ has infinitely many solutions
$Ax = b$ has one solution $x = A^{-1}b$	$Ax = b$ has no solution or infinitely many
$A$ has rank $N$	$A$ has rank $R < N$
Col space of $A$ is $\mathbb{R}^N$	Col space of $A$ has dimension $R < N$
Null space of $A$ has dimension 0	Null space of $A$ has dimension $N - R$

## References:

Section 1.3 Sundaram

Chapters 8 and 9 S&B

Turkington Ch 1, 2

Strang Ch 2