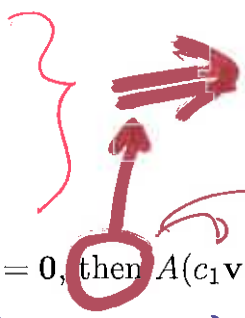


= span { }

$v_1, v_2$  are solns to  $A\vec{x} = \vec{0}$   
 homo



Any linear combi of solns is also a soln to  $A\vec{x} = \vec{0}$   
 homo

Suppose  $Av_1 = 0$  and  $Av_2 = 0$ , then  $A(c_1v_1 + c_2v_2) = 0$

$$A(c_1\vec{v}_1 + c_2\vec{v}_2) = A(c_1\vec{v}_1) + A(c_2\vec{v}_2) = c_1A\vec{v}_1 + c_2A\vec{v}_2 = \vec{0} + \vec{0} = \vec{0}$$

NOTE: Nullspace of  $A = \text{span} \left\{ \begin{bmatrix} -3 \\ 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -4 \\ 0 \\ 0 \\ 1 \end{bmatrix} \right\}$

2.8 Subspaces of  $R^n$ . = Vector space

Long definition emphasizing important points:

Defn: Let  $W$  be a nonempty subset of  $R^n$ . Then  $W$  is a subspace of  $R^n$  if and only if the following three conditions are satisfied:

- i.) 0 is in  $W$ ,
- ii.) if  $v_1, v_2$  in  $W$ , then  $v_1 + v_2$  in  $W$ ,
- iii.) if  $v$  in  $W$ , then  $cv$  in  $W$  for any scalar  $c$ .

closed under linear combinations

Short definition: Let  $W$  be a nonempty subset of  $R^n$ . Then  $W$  is a subspace of  $R^n$  if  $v_1, v_2$  in  $W$  implies  $c_1v_1 + c_2v_2$  in  $W$ ,

= span

Note that if  $S$  is a subspace, then

- if  $v_1, v_2, \dots, v_n$  in  $S$ , then  $c_1v_1 + c_2v_2 + \dots + c_nv_n$  is in  $S$ .
- $0v = 0$  is in  $S$ .

Defn: Let  $S$  be a subspace of  $R^k$ . A set  $T$  is a basis for  $S$  if

- i.)  $T$  is linearly independent and
- ii.)  $T$  spans  $S$ .

but not overly large don't want extra vectors

large enough to span for span

A subspace is a vector space

Subspace of  $\mathbb{R}^n =$  Vector space

is a set closed under linear combinations

I.e.  $S$  is a <sup>finite dimensional</sup> subspace

$$\Leftrightarrow S = \text{span} \{v_1, \dots, v_m\}$$

0-dim subspace

$$\{\vec{0}\} = \text{span} \{\vec{0}\}$$

not l.i.



Note  $\vec{0}$  is always in  $S$

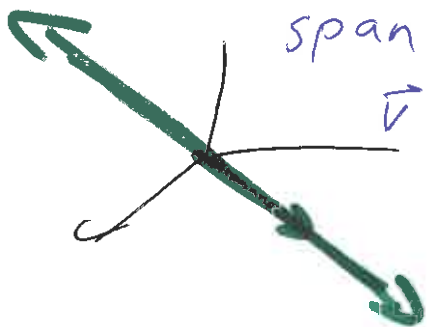
$$\vec{0} = 0\vec{v}_1 + 0\vec{v}_2 + \dots + 0\vec{v}_m$$

in  $S$

1-dim subspace

$$\text{span} \{v\}$$

$$v \neq 0$$



2-dim subspace

$$\text{span} \{v, w\}$$

l.i.



2d plane

etc

hyperplane  
 $\text{span} \{v_1, \dots, v_m\}$

Ex: Nullspace

Solution set to  $A\vec{x} = \vec{0}$

$$\vec{x} = x_i \vec{v} + x_j \vec{w} \text{ etc}$$

where  $x_i, x_j$  are free variables

$$\text{Nullspace} = \text{Span} \{ \vec{v}, \vec{w} \}$$

Dim of Nullspace = Nullity  
= # of free variables

---

Ex: Col space of  $A = [\vec{a}_1 \dots \vec{a}_n]$

Span  $\{ \vec{a}_1, \dots, \vec{a}_n \}$

Simplify answer  
take only pivot columns of

RANK =

Dim of Col space = # of pivots of  $A$

Basis for  $S$  is smallest set of vectors that span  $S$   
 The simplified answer

## 2.9: Basis and Dimension

Defn: Let  $S$  be a subspace of  $R^k$ . A set  $T$  is a **basis** for  $S$  if

i.)  $T$  is linearly independent and

ii.)  $T$  spans  $S$ .

← NOT TOO LARGE (SIMPLIFIED)  
 remove extra vectors

Examples

a.)  $\left\{ \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}, \begin{bmatrix} 3 \\ 4 \\ 0 \end{bmatrix} \right\}$  is a basis for  $\text{span}\left\{ \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}, \begin{bmatrix} 3 \\ 4 \\ 0 \end{bmatrix} \right\}$  ✓

span & li

GOLDOLOCKS APPROVED

b.)  $\left\{ \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}, \begin{bmatrix} 3 \\ 4 \\ 0 \end{bmatrix}, \begin{bmatrix} 4 \\ 6 \\ 0 \end{bmatrix} \right\}$  is NOT a basis for  $\text{span}\left\{ \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}, \begin{bmatrix} 3 \\ 4 \\ 0 \end{bmatrix} \right\}$

TOO LARGE not l.i.c.

c.)  $\left\{ \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix} \right\}$  is NOT a basis for  $\text{span}\left\{ \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}, \begin{bmatrix} 3 \\ 4 \\ 0 \end{bmatrix} \right\}$

TOO SMALL does not span

Defn: A vector space is called **finite-dimensional** if it has a basis consisting of a finite number of vectors. Otherwise,  $V$  is **infinite dimensional**.

$\{1, t, t^2, t^3, \dots\}$

Thm: All basis for a finite-dimensional vector space have the same number of elements.

Defn:  $\dim(V)$  = the **dimension** of a finite-dimensional vector space  $V$  = the number of vectors in any basis for  $S$ . If  $\dim(V) = n$ , then  $V$  is said to be  $n$ -dimensional.

Let  $S = \text{span} \left\{ \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}, \begin{bmatrix} 3 \\ 4 \\ 0 \end{bmatrix} \right\}$

$$S \subset \mathbb{R}^3$$

Basis for  $S$

Not basis for  $S$

$$\left\{ \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}, \begin{bmatrix} 3 \\ 4 \\ 0 \end{bmatrix} \right\}$$

$$\left\{ \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix} \right\} \text{ Does not span}$$

$$\left\{ \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}, \begin{bmatrix} 6 \\ 8 \\ 0 \end{bmatrix} \right\}$$

$$\left\{ \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}, \begin{bmatrix} 3 \\ 4 \\ 0 \end{bmatrix}, \begin{bmatrix} 5 \\ 6 \\ 0 \end{bmatrix} \right\}$$

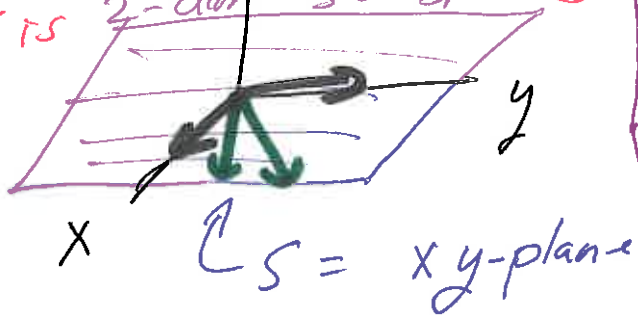
Not l.i.

$$\left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \right\}$$

$$\left\{ \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\}$$

These vectors are in  $S$   
 Span 2-dim space in  $S$   
 $S$  is 2-dim so  $\text{span} = S$

Is lin indep  
 But  $\text{span} \left\{ \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\} \neq S$



Not  $S$   
 It is a basis for  $\text{span} \left\{ \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\}$

rank  $A$  = Rank of a matrix  $A$  = dimension of Col  $A$   
 = number of pivot columns of  $A$ .

nullity of  $A$  = dimension of Nul  $A$  = number of free variables.

Basis theorem: Let  $H$  be a  $p$ -dimensional subspace of  $\mathbb{R}^n$ .

i.) If  $H = \text{span}\{w_1, \dots, w_p\}$ , then  $\{w_1, \dots, w_p\}$  is a basis for  $H$ .

ii.) If  $v_1, \dots, v_p$  are linearly independent vectors in  $H$ ,  
 then  $\{v_1, \dots, v_p\}$  is a basis for  $H$ .

**Rank( $A$ ) + nullity( $A$ ) = Number of columns of  $A$ .**  
*# of pivot + # free variables = # of columns of  $A$*

Ex. 1) Suppose  $A$  is a  $5 \times 7$  matrix.

If Rank( $A$ ) = 4, then nullity( $A$ ) =  $7 - 4 = 3$

$Ax = \mathbf{0}$  has  $\infty$  solutions.

$Ax = \mathbf{b}$  has none or  $\infty$  solutions.

$\uparrow$  3 free variables

If Rank( $A$ ) = 5, then nullity( $A$ ) =  $7 - 5 = 2$

$Ax = \mathbf{0}$  has  $\infty$  solutions.

$Ax = \mathbf{b}$  has  $\infty$  solutions.

since  $\downarrow$  homos free variables

If Rank( $A$ ) = 5, the column space of  $A = \mathbb{R}^5$

*pivot in each row of coef matrix + free variable*



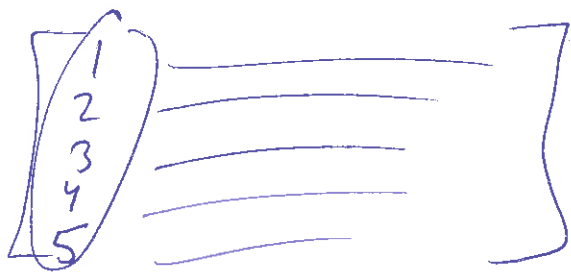
*no row of all 0's in coef matrix*

$$\begin{array}{c}
 7 \\
 \boxed{\begin{array}{c} \dots A \\ 0000000 \end{array}}
 \end{array}$$

4 pivots  $\Rightarrow 7 - 4$  free variables

~~RANK~~ RANK = 4  $\Rightarrow$  Nullity = 3

Col  $A$  is a 4-dimensional subspace of  $\mathbb{R}^5$



Null  $A$  is a 3-dimensional subspace of  $\mathbb{R}^7$

$$\begin{array}{c}
 \boxed{5} \times \boxed{7} \\
 \left[ \begin{array}{c} \dots \\ \dots \end{array} \right] \begin{array}{c} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \\ x_6 \\ x_7 \end{array} = \begin{array}{c} b_1 \\ b_2 \\ b_3 \\ b_4 \\ b_5 \end{array} \\
 \boxed{7} \times \boxed{1} \qquad \boxed{5} \times \boxed{1}
 \end{array}$$

EX:  $A$   $5 \times 7$   $\begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$

$$\text{Rank } A = 5$$

Col  $A$  is a 5-dim subspace  
of  $\mathbb{R}^5$

$$\Rightarrow \text{Col } A = \mathbb{R}^5$$



Thm 8': If  $A$  is a **SQUARE**  $n \times n$  matrix, then the following are equivalent.

- a.)  $A$  is invertible.
- b.) The row-reduced echelon form of  $A$  is  $I_n$ , the identity matrix.
- c.) An echelon form of  $A$  has  $n$  leading entries [I.e., every column of an echelon form of  $A$  is a leading entry column – no free variables]. (A square  $\Rightarrow A$  has leading entry in every column if and only if  $A$  has leading entry in every row).
- d.) The column vectors of  $A$  are linearly independent.
- e.)  $Ax = 0$  has only the trivial solution.
- f.)  $Ax = b$  has at most one sol'n for any  $b$ .
- g.)  $Ax = b$  has a unique sol'n for any  $b$ .
- h.)  $Ax = b$  is consistent for every  $n \times 1$  matrix  $b$ .
- i.)  $Ax = b$  has at least one sol'n for any  $b$ .
- j.) The column vectors of  $A$  span  $R^n$ .  
[every vector in  $R^n$  can be written as a linear combination of the columns of  $A$ ].
- k.) There is a square matrix  $C$  such that  $CA = I$ .
- l.) There is a square matrix  $D$  such that  $AD = I$ .
- m.)  $A^T$  is invertible.
- n.)  ~~$A$  is expressible as a product of elementary matrices.~~

# SQUARE

$n \times n$  matrix

o.) The column vectors of  $A$  form a basis for  $R^n$ .  
 [every vector in  $R^n$  can be written uniquely as a linear combination of the columns of  $A$ ]. *lin index*

p.) Col  $A = R^n$ .

q.)  $\dim$  Col  $A = n$ .

r.) rank of  $A = n$ .

s.) Nul  $A = \{0\}$ ,

t.)  $\dim$  Nul  $A = 0$ .

u.)  $A$  has nullity 0.

*n pivots*  
*pivot in each row*  
 SQUARE  
*n pivots*  
*pivot in each column  $\Rightarrow$  no free variables*

**Rank(A) + nullity(A) = Number of columns of A.**

Ex. 2) Suppose  $A$  is a  $9 \times 4$  matrix.

If Rank(A) = 4, then nullity(A) =  $4 - 4 = 0$

$Ax = 0$  has unique solutions.

$Ax = b$  has at most one solutions.

none or exactly one

*no free variables*

If Rank(A) = 3, then nullity(A) =  $4 - 3 = 1$

$Ax = 0$  has  $\infty$  solutions.

$Ax = b$  has none or  $\infty$  solutions.

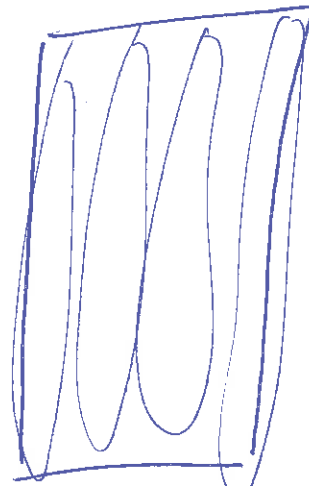
*free variables*

*rows of 0's*



A

$9 \times 4$


$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_9 \end{bmatrix}$$

$9 \times 4$     $4 \times 1$     $=$     $9 \times 1$

$$\text{Nul } A \subset \mathbb{R}^4$$

$$\text{Col } A \subset \mathbb{R}^9$$

Note: In ch. 3 all matrices are SQUARE.

---

3.1 Defn:  $\det A = \sum \pm a_{1j_1} a_{2j_2} \dots a_{nj_n}$

2 × 2 short-cut:  $\det \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} = a_{11} a_{22} - a_{12} a_{21}$

3 × 3 short-cut:  $\det \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} \begin{matrix} a_{11} & a_{12} \\ a_{21} & a_{22} \\ a_{31} & a_{32} \end{matrix}$

Note there is no short-cut for  $n \times n$  matrices when  $n > 3$ .

---

REQUIRED Definition of Determinant using cofactor expansion

Defn:  $A_{ij}$  is the matrix obtained from  $A$  by deleting the  $i$ th row and the  $j$ th column.

Defn: Let  $A = (a_{ij})$  by an  $n \times n$  square matrix. The determinant of  $A$  is

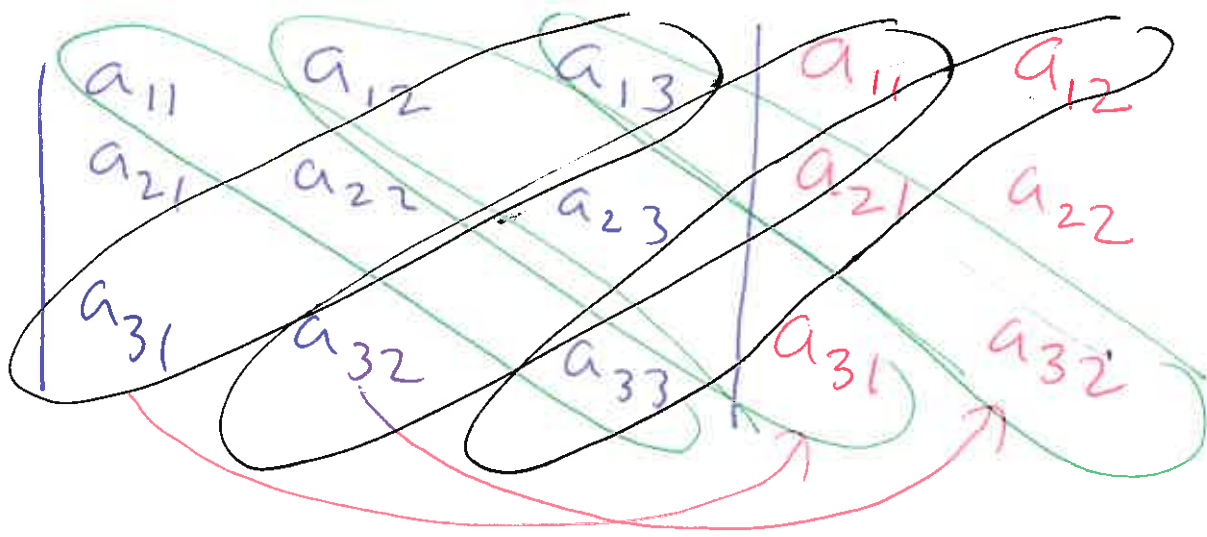
1.) If  $n = 1$ ,  $\det A = a_{11}$ .

2.) If  $n > 1$ ,  $\det A = \sum_{k=1}^n (-1)^{1+k} a_{1k} \det A_{1k}$

$$= a_{11} \det A_{11} - a_{12} \det A_{12} + \dots + (-1)^{1+n} a_{1n} \det A_{1n}$$

Note the above definition is an inductive or recursive definition.

must learn



$$a_{11} a_{22} a_{33} + a_{12} a_{23} a_{31} + a_{13} a_{21} a_{32} - a_{13} a_{22} a_{31} - a_{11} a_{23} a_{32} - a_{12} a_{21} a_{33}$$

This method only works for 3x3

Optional

$$\det \begin{bmatrix} 1 & 3 \\ 2 & 4 \end{bmatrix} = (1)(4) - (2)(3)$$

$$\begin{vmatrix} a & b \\ c & d \end{vmatrix} = ad - bc$$

Cofactor expansion

$$= ad - bc$$

$$(-1)^{1+1} ad + (-1)^{1+2} bc$$

$$\begin{vmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 0 & 8 \end{vmatrix}$$

- ① Expand along row 1  
 ② " " " 3  
 ③ " " column 2
- } shorter method

$$\begin{vmatrix} \cancel{1} & \cancel{2} & \cancel{3} \\ 4 & 5 & 6 \\ 7 & 0 & 8 \end{vmatrix}$$

$$+1 \begin{vmatrix} 5 & 6 \\ 0 & 8 \end{vmatrix} - 2 \begin{vmatrix} 4 & 6 \\ 7 & 8 \end{vmatrix} + 3 \begin{vmatrix} 4 & 5 \\ 7 & 0 \end{vmatrix}$$

$$1(5 \cdot 8 - 0 \cdot 6) - 2(4 \cdot 8 - 7 \cdot 6) + 3(4 \cdot 0 - 5 \cdot 7)$$

$$40 - 2(-10) + 3(-35)$$

$$60 - 105 = -45$$

$$\begin{vmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ \cancel{7} & 0 & 8 \end{vmatrix}$$

$$\begin{vmatrix} + & - & + \\ - & + & - \\ + & - & + \end{vmatrix}$$

$$= \cancel{7} \begin{vmatrix} 2 & 3 \\ 5 & 6 \end{vmatrix} - \cancel{0} \begin{vmatrix} 1 & 3 \\ 4 & 6 \end{vmatrix} + 8 \begin{vmatrix} 1 & 2 \\ 4 & 5 \end{vmatrix}$$

$$= 7(12 - 15) + 8(5 - 8)$$

$$= 7(-3) + 8(-3) = 15(-3) = \boxed{-45}$$

$$\begin{vmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 0 & 8 \end{vmatrix}$$

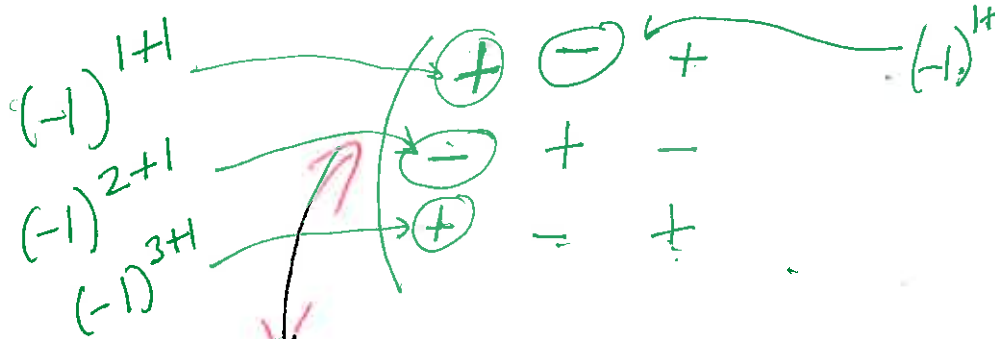
$$\begin{vmatrix} + & - & + \\ - & + & - \\ + & - & + \end{vmatrix} \quad (-1)^{1+2}$$

$$\cancel{-2} \begin{vmatrix} 4 & 6 \\ 7 & 8 \end{vmatrix} + 5 \begin{vmatrix} 1 & 3 \\ 7 & 8 \end{vmatrix} - \cancel{0} \begin{vmatrix} 1 & 3 \\ 4 & 6 \end{vmatrix}$$

$$= -2(32 - 42) + 5(8 - 21)$$

$$= -2(-10) + 5(-13) = 20 - 65 = \boxed{-45}$$





Thm: Let  $A = (a_{ij})$  by an  $n \times n$  square matrix,  $n > 1$ .  
 Then expanding along row  $i$ , *row*

$$\det A = \sum_{k=1}^n (-1)^{i+k} a_{ik} \det A_{ik}$$

*column  $n$*

Or expanding column  $j$ ,

$$\det A = \sum_{k=1}^n (-1)^{k+j} a_{kj} \det A_{kj}$$

Defn:  $\det A_{ij}$  is the  $i, j$ -minor of  $A$ .

*remove row  $i$ , column  $j$*

$(-1)^{i+j} \det A_{ij}$  is the  $i, j$ -cofactor of  $A$ .

### 3.2: Properties of Determinants

Thm: If  $A \xrightarrow{R_i \rightarrow cR_i} B$ , then  $\det B = c(\det A)$ .

Warning note:  $\det(cA) = c^n \det A$ .

Thm: If  $A \xrightarrow{R_i \leftrightarrow R_j} B$ , then  $\det B = -(\det A)$ .

Thm: If  $A \xrightarrow{R_i + cR_j \rightarrow R_i} B$ , then  $\det B = \det A$ .

*Do row ops to get more 0's*

$$1R_i + cR_j \rightarrow 1R_i$$

$$R_i + cR_j \rightarrow 1R_j \leftarrow \text{affects determ}$$

$$\left| \begin{array}{ccc|c} 1 & 2 & 3 & 7 \\ 4 & 5 & 6 & 8 \\ 7 & 0 & 8 & 8 \end{array} \right| \xrightarrow{R_3 - 7R_1} \left| \begin{array}{ccc|c} 1 & 2 & 3 & 7 \\ 4 & 5 & 6 & 8 \\ 0 & -14 & -13 & -41 \end{array} \right|$$

$7 \times 7 = 49$   
 $0 - 14 = -14$   
 $8 - 21 = -13$   
 $8 - 49 = -41$



$$\rightarrow 1R_3$$

$$1R_2 - 4R_1 \rightarrow 1R_2$$

$$\left| \begin{array}{ccc|c} 1 & 2 & 3 & 7 \\ 0 & -3 & -6 & -3 \\ 0 & -14 & -13 & -41 \end{array} \right| = +1 \left| \begin{array}{cc|c} -3 & -6 & -3 \\ -14 & -13 & -41 \end{array} \right| = 0 + 0$$

$$= 1(+39 - 84) = -45$$