

§1.4 Algebraic properties of matrices; Inverses

Theorem 1.4.1 properties of Matrix operations

Assuming that the sizes of the matrices are such that the indicated operations can be performed, the following rules are valid

- (a)  $A + B = B + A$       Commutativity of matrix addition
- (b)  $A + (B + C) = (A + B) + C$       Associative law of matrix addition
- (c)  $A(BC) = (AB)C$       Associative law of matrix multiplication
- (d)  $A(B + C) = AB + AC$       distributive law (left)
- (e)  $(B + C)A = BA + CA$       "      " (right)
- (f)  $A(B - C) = AB - AC$       Same as (d)
- (g)  $(B - C)A = BA - CA$       "      " (e)
- (h)  $a(B + C) = aB + aC$  }  
 (i)  $a(B - C) = aB - aC$  }
- (j)  $(a + b)C = aC + bC$  }  
 (k)  $(a - b)C = aC - bC$  }
- (l)  $a(bc) = (ab)c$
- (m)  $a(BC) = (aB)C = B(aC)$

Remarks (i) Matrix addition, matrix-matrix multiplication and scalar-matrix multiplication were defined between pairs of objects. The above laws allow combining two or more operations

$$(A + B)C \stackrel{(e)}{=} AC + BC$$

Combining two or more of the same operation

$$ABC \stackrel{(e)}{=} A(BC) \stackrel{(c)}{=} (AB)C$$

(ii) Matrix subtraction  $A - B$  is defined as  $A - B = A + (-1)B$  i.e. in terms of matrix addition and scalar multiplication by  $(-1)$ .

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$AB$  is  $m \times p$

$BC$  is  $n \times s$

proof of (c)  $A$   $m \times n$ ,  $B$   $n \times p$ ,  $C$   $p \times s$

$(AB)C$  is defined as  $AB$  first then mult. with  $C$

$A(BC)$  " " " "  $BC$  first then mult. with  $A$

$$((AB)C)_{ij} = \sum_{k=1}^p (AB)_{ik} C_{kj} = \sum_{k=1}^p \left( \sum_{\ell=1}^n A_{\ell i} B_{\ell k} \right) C_{kj}$$

$$(AB)_{ik}$$

Compute two sums:

$$= \sum_{\ell=1}^n \sum_{k=1}^p A_{\ell i} B_{\ell k} C_{kj} = \sum_{\ell=1}^n A_{\ell i} \underbrace{\sum_{k=1}^p B_{\ell k} C_{kj}}_{(BC)_{\ell j}}$$

$$= \sum_{\ell=1}^n A_{\ell i} (BC)_{\ell j} = (A(BC))_{ij}$$

Defn. the Identity matrix. For any  $n \geq 1$ , the  $n \times n$  identity matrix  $I_n$  (`eye(n)` in Matlab) or simply  $I$  is

$$I = \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & \ddots & \\ & & & 1 \end{bmatrix}, \quad I_{ij} = \begin{cases} 1 & i=j \\ 0 & i \neq j \end{cases}$$

Lemma  $AI = IA = A$  for any matrix  $A$  provided dimensions of  $A$  are compatible.

ex.  $A$  is 4 by 3 then  $I_4 A = A$   
 $A I_3 = A$ .

Defn. the zero matrix. For  $m$  and  $n$  given

The  $m$  by  $n$  zero matrix, denoted  $O$  is given by

$$O_{ij} = 0 \quad i=1, \dots, m, \quad j=1, \dots, n$$

$$O = \begin{bmatrix} 0 & \dots & 0 \\ \vdots & & \\ 0 & \dots & 0 \end{bmatrix}.$$



Theorem If  $c$  is a scalar and if the sizes of the matrices are such that the operations are valid, then

- (a)  $A + 0 = 0 + A = A$
- (b)  $A - 0 = A$
- (c)  $A - A = 0$
- (d)  $0 \cdot A = 0$
- (e) if  $cA = 0$ , then  $c = 0$  or  $A = 0$  or both  $c$  and  $A$  are zero

Proof (a)  $(A + 0)_{ij} = A_{ij} + 0_{ij} = A_{ij} + 0 = A_{ij}$

(e) 
$$cA = \begin{bmatrix} ca_{11} & ca_{12} & \dots & ca_{1n} \\ ca_{21} & ca_{22} & \dots & ca_{2n} \\ \vdots & \vdots & & \vdots \\ ca_{m1} & ca_{m2} & \dots & ca_{mn} \end{bmatrix} \stackrel{\text{Suppose}}{=} \begin{bmatrix} 0 & 0 & \dots & 0 \\ 0 & 0 & \dots & 0 \\ \vdots & \vdots & & \vdots \\ 0 & 0 & \dots & 0 \end{bmatrix}$$

If  $c = 0$ , then done with proof. If  $c \neq 0$ , then  $a_{ij} = 0$   
 $i = 1, \dots, m, j = 1, \dots, n.$  ✓

Defn A matrix  $A$  is nonzero, written  $A \neq 0$  if  $A$  has at least one nonzero entry. In other words if  $A$  is not the zero matrix!!

Remark It is possible for the product of two nonzero matrices to be zero.

Ex.  $A = \begin{bmatrix} 0 & 1 \\ 0 & 2 \end{bmatrix}, B = \begin{bmatrix} 1 & 3 \\ 0 & 0 \end{bmatrix}, AB = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$   
 $\neq 0 \qquad \qquad \qquad \neq 0$

In view of this, the cancellation law valid for real numbers, is not valid in general for matrices.

For real numbers  $a, b, c$  if  $ac = bc$  and  $c \neq 0$   
 then  $a = b$ .

For matrices  $A, B, C, AC = BC, C \neq 0$  matrix  
 does not necessarily imply that  $A = B$

Ex. Suppose  $AB=0$  with  $A \neq 0, B \neq 0$

let  $C$  be any matrix, same size as  $A$

$$(A+C)B = \overbrace{AB}^0 + CB = CB$$

we have  $(A+C)B = CB$  but  $A+C \neq C$  (why?)

Theorem 1.4.3 If  $R$  is the rref of a square  $n \times n$  matrix then either  $R$  has a row of zeros or it is the identity matrix

proof. There are two mutually exclusive possibilities:

Either  $R$  has a row of zeros or it does not.

we have to show that in the 2nd case, i.e. no zero rows  $R = I$ . Indeed, if there are no zero rows, every row has a leading 1. It is clear that the leading ones are along the diagonal. Also, since each column containing a leading one has zeros with the exception of the leading one (property (4)),  $R = I$

$$\underline{n=2} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \quad \underline{n=3} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}, \quad \underline{n=4} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

Defn. let  $A$  be a square matrix. Suppose there exists a square matrix  $B$  of the same size such that

$$AB = BA = I,$$

then we say that  $A$  is invertible or nonsingular and call  $B$  the inverse of  $A$ .

Remark clearly the definition is symmetric, in the sense that  $A$  is invertible iff  $B$  is



Ex.

$$A = \begin{bmatrix} 2 & -5 \\ -1 & 3 \end{bmatrix}, \quad B = \begin{bmatrix} 3 & 5 \\ 1 & 2 \end{bmatrix}$$

$$AB = \begin{bmatrix} (2)(3) + (-5)(1) = 1 & 2(5) + (-5)(2) = 0 \\ (-1)(3) + (3)(1) = 0 & (-1)(5) + 3(2) = 1 \end{bmatrix} = I_2$$

$$BA = \begin{bmatrix} 3(2) + 5(-1) = 1 & 3(-5) + 5(3) = 0 \\ 1(2) + 2(-1) = 0 & 1(-5) + 2(3) = 1 \end{bmatrix} = I_2 \quad \checkmark$$

Remark Turns out <sup>Verifying</sup>  $AB = I$  or  $BA = I$  is enough.

Theorem 1.4.4 Uniqueness of  $\mathbb{R}$  inverse. Suppose  $B$  and  $C$  satisfy  $AB = BA = I$  and  $AC = CA = I$ . Then  $B = C$ .  
proof.

$$\begin{aligned} AB = I &\Rightarrow C(AB) = CI = C \\ &\quad \underbrace{(CA)}_I B = C \\ &\quad B = I \cdot B = C \quad \checkmark \end{aligned}$$

Theorem <sup>1.4.6</sup> The product of any finite number of square matrices of  $\mathbb{R}$  same size is invertible if every matrix in  $\mathbb{R}$  product is invertible. Furthermore,  $\mathbb{R}$  inverse of  $\mathbb{R}$  product is  $\mathbb{R}$  product of inverses in reverse order.  
proof.

Consider  $\mathbb{R}$  product  $B = A_1 A_2 \dots A_m$ . Assume  $A_i$ 's are all invertible, i.e.  $A_i A_i^{-1} = I$ . Then

$$\begin{aligned} &(A_1 A_2 \dots A_m)(A_m^{-1} \dots A_2^{-1} A_1^{-1}) \\ &= A_1 A_2 \dots \underbrace{A_m A_m^{-1}}_I \dots A_2^{-1} A_1^{-1} \quad \text{associativity} \\ &= A_1 A_2 \dots \underbrace{A_{m-1} A_{m-1}^{-1}}_I \dots A_2^{-1} A_1^{-1} \\ &\vdots \\ &= A_1 A_1^{-1} = I \quad \checkmark \end{aligned}$$

Remark If one or more  $A_i$ 's are singular, then so is  $\mathbb{R}$  product. ▣

Theorem 1.4.7 If  $A$  is invertible, then

- (a)  $A^{-1}$  is invertible and  $(A^{-1})^{-1} = A$
- (b) For any integer  $n \geq 1$ ,  $A^n$  is invertible and  $(A^n)^{-1} = (A^{-1})^n$ .
- (c) If  $k \neq 0$  scalar, then  $kA$  is invertible and  $(kA)^{-1} = \frac{1}{k} A^{-1}$ .

proof (a)  $A^{-1}A = I$  says also  $(A^{-1})^{-1} = A$ !

(b) follows from Thm 1.4.6, take  $A_i = A, i = 1, \dots, n$

$$(c) \quad (kA)\left(\frac{1}{k}A^{-1}\right) = kA \frac{1}{k}A^{-1} = \underbrace{k \cdot \frac{1}{k}}_1 \underbrace{AA^{-1}}_I = I \checkmark.$$

Theorem 1.4.8 If  $A$  is invertible, then so is  $A^T$  and  $(A^T)^{-1} = (A^{-1})^T$ .

proof.

$$A^T (A^{-1})^T = \left( (A^{-1})A \right)^T \quad \text{Thm 1.4.8 (e)}$$
$$= I^T = I.$$

### §1.5 Elementary matrices.

Recall the elementary row operations:

Type 1 Multiply row  $i$  of  $A$  by a constant  $c \neq 0$

Type 2 Interchange/permute rows  $i$  and  $j$  of  $A$ .

Type 3 Add  $\mu$  times row  $i$  of  $A$  to row  $j$  of  $A$  [ $i \neq j$ ]

To each elementary row operation we can associate a corresponding  $n \times n$  matrix called an elementary matrix.

This is done by applying the elementary row operation to the  $n \times n$  identity matrix

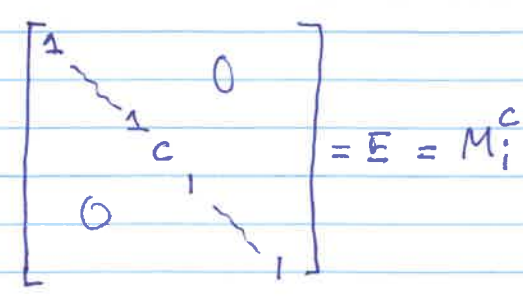
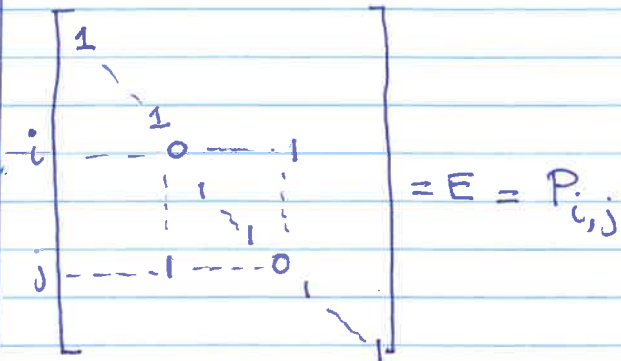
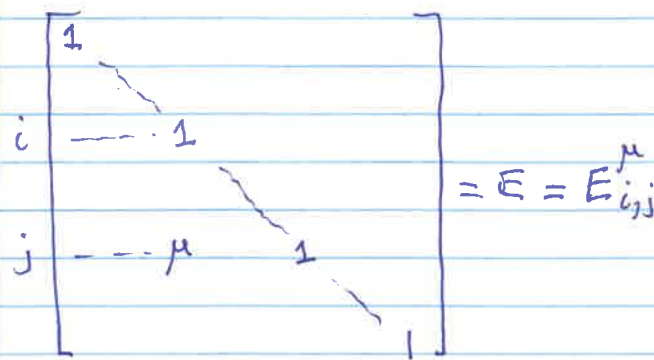
Ex. Add 4 times row 2 to row 4.  $\leftarrow$  elem. row operation

$$A = \begin{bmatrix} 1 & 2 & 5 \\ -3 & 9 & 7 \\ 8 & 0 & 2 \\ 1 & 9 & -3 \end{bmatrix}$$

We take  $n=4$ . In general  $n$  is the number of rows of the matrix  $A$  to which the elementary row operation is to be applied.

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \xrightarrow[4 \times r_2 + r_4]{\text{elem. row op}} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 4 & 0 & 1 \end{bmatrix} = E.$$

More generally

<u>Elementary row operation</u>	<u>Corresponding <math>n \times n</math> elementary matrix</u>
<u>Type 1</u> Multiply row $i$ by $c \neq 0$	 $= E = M_i^c$
<u>Type 2</u> Interchange rows $i$ and $j$	 $= E = P_{i,j}$
<u>Type 3</u> Add $\mu$ times row $i$ to row $j$ , $j \neq i$	 $= E = E_{i,j}^\mu$

Note The notation  $M_i^c$ ,  $P_{i,j}$ ,  $E_{i,j}^\mu$  is my own and is introduced for convenience.

As you see, elementary matrices are very easy to construct. They are also very useful in proving many important facts. Moreover, they can be used to show why/some important algorithms work. e.g. finding the inverse of a matrix.



In Matlab, an easy way to generate elementary matrices can be done as follows

Type 1  $E = \text{eye}(m)$  creates  $m \times m$  identity matrix  
 $E(i,i) = c$  puts  $c$  in the  $(i,i)$  location

Type 2  $E = \text{eye}(m)$   
 $E(i,i) = E(i,j) = 0$   
 $E(i,j) = E(j,i) = 1$

(There may be an easier way!)

Type 3  $E = \text{eye}(m)$   
 $E(i,j) = \mu$

It turns out that there is a mathematically powerful way to represent elementary matrices using the identity matrix together with so-called rank-one matrices.

Let  $v$  be an  $m \times 1$  matrix and  $w$  a  $1 \times n$  matrix

$$v = \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_m \end{bmatrix}, \quad w = [w_1, w_2, \dots, w_n].$$

The product  $v w$  makes sense <sup>and is an  $m \times n$  matrix,</sup> It is sometimes called the exterior product of  $v$  and  $w$ . Note the

product  $w v$  makes sense only if  $m=n$  and is the usual dot product of  $w$  and  $v$ . Now by definition

$$v w = \begin{bmatrix} v_1 w_1 & v_1 w_2 & \dots & v_1 w_n \\ v_2 w_1 & v_2 w_2 & \dots & v_2 w_n \\ \vdots & \vdots & \dots & \vdots \\ v_m w_1 & v_m w_2 & \dots & v_m w_n \end{bmatrix} \quad \text{Ex}$$



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Theorem 1.5.1 Let  $E$  denote the  $m \times m$  elementary matrix corresponding to a given elementary row operation. If  $A$  is an  $m \times n$  matrix, then  $EA$  is the (same) matrix that would be obtained by applying the elementary row operation to  $A$ .

Ex.  $m=3$

Elem. row operation: Add  $-2 \times$  row 3 to row 1

Elementary  $3 \times 3$  matrix  $E = E_{3,1}^{-2} = \begin{bmatrix} 1 & 0 & -2 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$

$$\text{Let } A = \begin{bmatrix} 1 & 5 & -3 & 4 \\ 2 & 0 & -5 & 7 \\ 4 & 8 & -3 & 9 \end{bmatrix}$$

now

$$A \xrightarrow{\text{row op}} \begin{bmatrix} -7 & -11 & 3 & -14 \\ 2 & 0 & -5 & 7 \\ 4 & 8 & -3 & 9 \end{bmatrix}$$

On the other hand

$$E = I + (-2) \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \begin{bmatrix} 0 & 0 & 1 \end{bmatrix}$$

$$\Rightarrow EA = A + (-2) \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \begin{bmatrix} 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 5 & -3 & 4 \\ 2 & 0 & -5 & 7 \\ 4 & 8 & -3 & 9 \end{bmatrix}$$

$$= A + (-2) \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \begin{bmatrix} 4 & 8 & -3 & 9 \end{bmatrix} = A + \begin{bmatrix} -8 & -16 & 6 & -18 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

$$= \begin{bmatrix} -7 & -11 & 3 & -14 \\ 2 & 0 & -5 & 7 \\ 4 & 8 & -3 & 9 \end{bmatrix} \checkmark$$

proof. just as in the example above, the proof consists in a simple verification of the statement for each type.



Remark It is easy to see that every elementary row operation has an inverse which is also an elementary row operation of the same type. By inverse we mean an operation that cancels the effect of the operation.

Elem row operation

$M_i^c \leftrightarrow$  Multiply row  $i$  by  $c \neq 0$

$P_{ij} \leftrightarrow$  interchange rows  $i$  and  $j$

$E_{ij}^\mu \leftrightarrow$  Add  $\mu \times$  row  $i$  to row  $j$

inverse elementary row operation

Multiply row  $i$  by  $1/c \leftrightarrow M_i^{1/c}$

interchange rows  $i$  and  $j \leftrightarrow P_{ij}$

Add  $-\mu \times$  row  $i$  to row  $j \leftrightarrow E_{ij}^{-\mu}$

Lemma Elementary matrices are invertible. Also, the inverse of an elementary matrix, is the elementary matrix corresponding to the inverse elementary row operation. In other words

$$(M_i^c)^{-1} = M_i^{1/c}$$

$$(P_{ij})^{-1} = P_{ij}$$

$$(E_{ij}^\mu)^{-1} = E_{ij}^{-\mu}$$

proof  $M_i^c M_i^{1/c} = (I + (c-1)VW)(I + (\frac{1}{c}-1)VW)$

$$= I + (c-1 + \frac{1}{c}-1)VW + (c-1)(\frac{1}{c}-1)VWVW.$$

Note that

$$VW = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \begin{bmatrix} 0 & 1 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

$$\text{and } VWVW = V(WV)W = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \left( \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \right) \begin{bmatrix} 0 & 1 & 0 \end{bmatrix} = VW$$

$\underbrace{\begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}}_{=I}$

Hence

$$M_i^c M_i^{1/c} = I + \underbrace{\left( c-1 + \frac{1}{c} - 1 + (c-1)\left(\frac{1}{c}-1\right) \right)}_{=0} v w = I.$$

Hence by uniqueness of the inverse,  $M_i^{1/c} = (M_i^c)^{-1}$ .

Now

$$P_{ij} P_{ij} = (I - v w)(I - v w) = I - 2v w + (v w)(v w)$$

$v = \begin{bmatrix} 1 \\ \leftarrow j \\ -1 \end{bmatrix} \begin{matrix} e_i \\ \\ \end{matrix}, w = \begin{bmatrix} \downarrow i \\ 1 \\ -1 \end{bmatrix}$

Now

$$(v w)(v w) = v (w v) w \text{ by associativity.}$$

$$\text{Also, } w v = [1 \quad -1] \begin{bmatrix} 1 \\ -1 \end{bmatrix} = (1)(1) + (-1)(-1) = 2 \text{ (This is a dot prod.)}$$

Hence

$$(v w)(v w) = v (w v) w = v(2)w = 2v w.$$

Thus

$$P_{ij} P_{ij} = I - 2v w + 2v w = I. \Rightarrow (P_{ij})^{-1} = P_{ij}.$$

Finally,

$$\begin{aligned} E_{ij}^M E_{ij}^{-M} &= (I + \mu v w)(I - \mu v w) \\ &= I + \underbrace{\mu v w - \mu v w}_0 - \underbrace{(\mu v w)(\mu v w)}_{\mu^2 v w v w} \\ &= I - \mu^2 v w v w. \end{aligned}$$

Note that  $v = \begin{bmatrix} 1 \\ \leftarrow j \\ -1 \end{bmatrix} \begin{matrix} e_i \\ \\ \end{matrix}; w = \begin{bmatrix} \downarrow i \\ 1 \\ -1 \end{bmatrix}$ ,  $i \neq j$ ; hence  $w v = 0$ .

This shows that  $E_{ij}^M E_{ij}^{-M} = I$ , i.e.  $(E_{ij}^M)^{-1} = E_{ij}^{-M}$  ■

Theorem 1.53 If  $A$  is an  $n \times n$  matrix, then the following statements are equivalent.

- $A$  is invertible
- $Ax=0 \Rightarrow x=0$ , i.e.  $Ax=0$  has only the trivial solution
- $\text{rref}(A) = I$
- $A$  is the product of elementary matrices.

proof  $(a) \Rightarrow (b)$ .

A invertible means  $\exists!$  inverse of A,  $A^{-1}$  such that  $A^{-1}A = I$ .

Now multiply both sides of  $Ax=0$  by  $A^{-1}$

$$\Rightarrow \underbrace{A^{-1}A}_I x = \underbrace{A^{-1}0}_0 \Rightarrow x=0 \quad \checkmark$$

$(b) \Rightarrow (c)$

Consider the augmented matrix  $[A|0]$  of the homogeneous system  $Ax=0$ . We bring  $[A|0]$  into reduced row echelon form by a sequence of elementary row operations. We claim that there can be no zero rows in the reduced row echelon form which must be of the form  $[A|0]$ . Indeed, if the last row of  $\tilde{A}$  consisted of zeros entirely, then that would mean that there are free variables, i.e. infinitely many solutions, contradicting the assumption that the only solution is  $x=0$ . Essentially, we have shown that  $\tilde{A} = I_n$ , which was the desired result.

$(c) \Rightarrow (d)$

(c) means  $\text{rref}(A) = I_n$  whereas

$A \xrightarrow{\text{op}^1} \xrightarrow{\text{op}^2} \dots \xrightarrow{\text{op}^s} \text{rref}(A)$  can be expressed as

$$E_s \dots E_1 A = I, \text{ where } E_i \text{ is the elementary}$$

matrix corresponding to the  $i$ -th elem. row operation.

We know that each  $E_i$  is invertible and  $E_i^{-1}$  is an elementary matrix (of the same type). We also know that the product of any finite number of invertible matrices is invertible and the inverse of the product is the product of inverses in reverse order. In other words

$$E_s \dots E_1 A = I \Rightarrow A = E_1^{-1} \dots E_s^{-1} I \\ = E_1^{-1} \dots E_s^{-1} \quad \checkmark$$



(d)  $\Rightarrow$  (a)

(d) means  $A$  is the product of elementary matrices. since such matrices are invertible, so is the product, which is  $A$ .  $\checkmark$   $\square$

Remark One efficient way to calculate the inverse of an invertible matrix  $A$  is done as follows

- (1) Apply a sequence of <sup>elem.</sup> row operations to reduce  $A$  to  $\text{rref}(A)$ . Note  $A$  invertible  $\Leftrightarrow \text{rref}(A) = I$ .
- (2) Apply the same sequence of elem. row operations to  $I$ . At the end, the resulting matrix is  $A^{-1}$ .

$$\begin{bmatrix} A \\ I \end{bmatrix} \xrightarrow{\text{op}_1} \xrightarrow{\text{op}_2} \dots \xrightarrow{\text{op}_s} \begin{bmatrix} I \\ A^{-1} \end{bmatrix}$$

(Rigorous)  
Justification

$$E_s \dots E_1 A = I$$

$$\Rightarrow A = E_1^{-1} \dots E_s^{-1} I = E_s^{-1} \dots E_1^{-1}$$

$$\Rightarrow A^{-1} = (E_s^{-1} \dots E_1^{-1})^{-1} = (E_1^{-1})^{-1} \dots (E_s^{-1})^{-1}$$

$$\Rightarrow \boxed{A^{-1} = E_s \dots E_1 = E_s \dots E_1 I}$$

Ex.

$$A = \begin{bmatrix} 1 & 2 & 3 \\ 2 & 5 & 3 \\ 1 & 0 & 8 \end{bmatrix}$$

$$\downarrow \begin{matrix} -2r_1 + r_2 \\ -r_1 + r_3 \end{matrix}$$

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

$$\downarrow 2r_2 + r_3$$

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

$$\downarrow -1 \times r_3$$

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

$$\downarrow -2r_2 + r_1$$

$$\begin{bmatrix} 1 & 0 & 9 \\ 0 & 1 & -3 \\ 0 & 0 & 1 \end{bmatrix}$$

$$\downarrow \begin{matrix} 3r_3 + r_2 \\ -9r_3 + r_1 \end{matrix}$$

$$I = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$I = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$\downarrow \begin{matrix} -2r_1 + r_2 \\ -r_1 + r_3 \end{matrix}$$

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

$$\downarrow 2r_2 + r_3$$

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

$$\downarrow -1 \times r_3$$

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

$$\downarrow -2r_2 + r_1$$

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

$$\downarrow \begin{matrix} 3r_3 + r_2 \\ -9r_3 + r_1 \end{matrix}$$

$$\begin{bmatrix} -40 & 16 & 9 \\ 13 & -5 & -3 \\ 5 & -2 & -1 \end{bmatrix} = A^{-1}$$

> may interchange order

> may interchange order

The fact that  $\text{cof}(A) = I$  shows that  $A$  is invertible.

Now, to find  $A^{-1}$  exactly the same operations that were used in the first column must be used on  $I$ . Also, the same order must be followed in the calculation of  $A^{-1}$ .

It is possible to interchange certain operations, e.g. as shown above. However it is safer not to take chances and stick to exactly the same order.