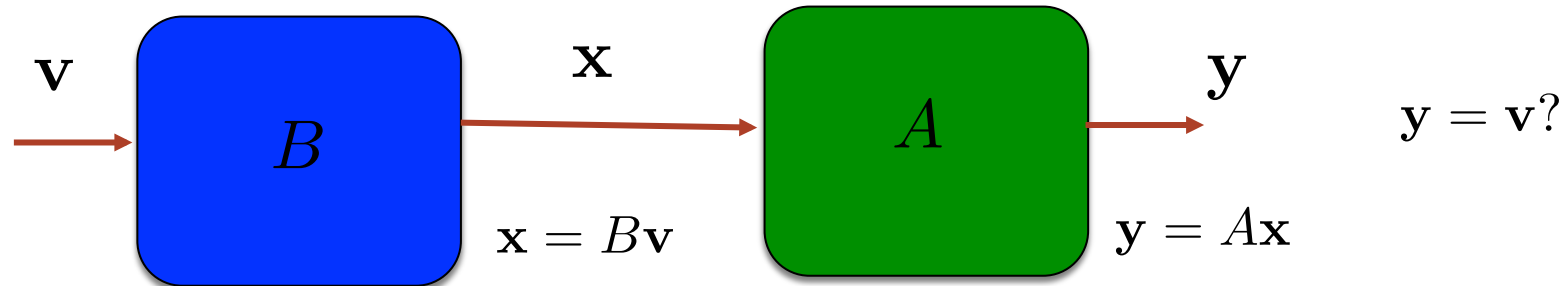


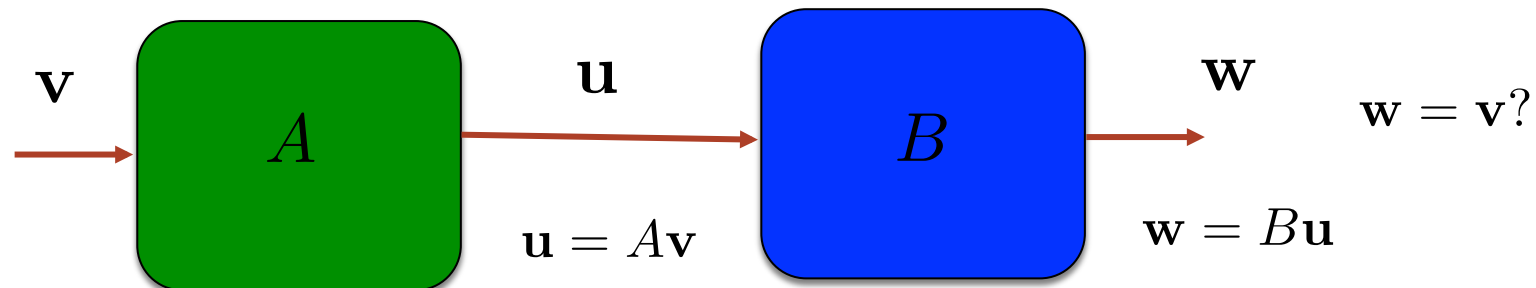
## 2.3 Invertibility and Elementary Matrices

From matrix **multiplication** to matrix **inverse**.

Let  $\mathbf{v}, \mathbf{x}, \mathbf{y} \in \mathcal{R}^n$ . Suppose  $A$  and  $B$  are  $n \times n$  matrices.



Given an  $n \times n$  matrix  $B$ , does there exist an  $n \times n$  matrix  $A$  such that  $\mathbf{y} = \mathbf{v}$  for any  $\mathbf{v} \in \mathcal{R}^n$ ?



Furthermore, can this matrix also satisfy that  $\mathbf{w} = \mathbf{v}$  for any  $\mathbf{v} \in \mathcal{R}^n$ ?

## Definitions

An  $n \times n$  matrix  $A$  is called **invertible** if there exists an  $n \times n$  matrix  $B$  such that  $AB = BA = I_n$ . In this case,  $B$  is called an **inverse** of  $A$ .

Properties:

1. For  $n = 1$ , the definition reduces to the multiplicative inverse ( $ab = ba = 1$ ).
2. If  $B$  is an inverse of  $A$ , then  $A$  is an inverse of  $B$ , i.e.,  $A$  and  $B$  are inverses to each other.

Example:

$$\begin{array}{l} A = \begin{bmatrix} 1 & 2 \\ 3 & 5 \end{bmatrix} \\ B = \begin{bmatrix} -5 & 2 \\ 3 & -1 \end{bmatrix} \end{array} \Rightarrow \begin{array}{l} AB = \begin{bmatrix} 1 & 2 \\ 3 & 5 \end{bmatrix} \begin{bmatrix} -5 & 2 \\ 3 & -1 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = I_2 \\ BA = \begin{bmatrix} -5 & 2 \\ 3 & -1 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 3 & 5 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = I_2 \end{array}$$

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Question: Does each  $n \times n$  matrix have an inverse?

Some example matrices that have no inverse:

1.  $O \in \mathcal{M}_{n \times n}$ .

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

since  $\begin{bmatrix} 1 & 2 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} a & b \\ c & d \end{bmatrix} = \begin{bmatrix} a + 2c & b + 2d \\ 0 & 0 \end{bmatrix} \neq \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$

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Property:

If  $A$  is an invertible matrix, then  $A$  has exactly one inverse (denoted  $A^{-1}$ ).

*Proof*

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## Theorem 2.2

Let  $A$  and  $B$  be  $n \times n$  matrices.

- (a) If  $A$  is invertible, then  $A^{-1}$  is invertible and  $(A^{-1})^{-1} = A$ .
- (b) If  $A$  and  $B$  are invertible, then  $AB$  is invertible and  $(AB)^{-1} = B^{-1}A^{-1}$ .
- (c) If  $A$  is invertible, then  $A^T$  is invertible and  $(A^T)^{-1} = (A^{-1})^T$ .

***Proof***

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

## Corollary

Let  $A_1, A_2, \dots, A_k$  be  $n \times n$  invertible matrix. Then the product  $A_1 A_2 \cdots A_k$  is also invertible, and

$$(A_1 A_2 \cdots A_k)^{-1} = (A_k)^{-1} (A_{k-1})^{-1} \cdots (A_1)^{-1}$$

***Proof***

Symbolically, the inverse may be used to solve matrix equations:

$$\begin{aligned} A\mathbf{x} &= \mathbf{b} \\ A^{-1}(A\mathbf{x}) &= A^{-1}\mathbf{b} \\ (A^{-1}A)\mathbf{x} &= A^{-1}\mathbf{b} \\ I_n\mathbf{x} &= A^{-1}\mathbf{b} \\ \mathbf{x} &= A^{-1}\mathbf{b} \end{aligned} \quad \begin{array}{l} x_1 + 2x_2 = 4 \\ 3x_1 + 5x_2 = 7 \\ \underbrace{\hspace{10em}} \\ A\mathbf{x} = \mathbf{b} \end{array}$$
$$\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \mathbf{x} = A^{-1}\mathbf{b} = \begin{bmatrix} -5 & 2 \\ 3 & -1 \end{bmatrix} \begin{bmatrix} 4 \\ 7 \end{bmatrix} = \begin{bmatrix} -6 \\ 5 \end{bmatrix}$$

However, this method is computationally inefficient.

## Elementary Matrices

- Every elementary row operation can be performed by matrix multiplication.

- For examples: Let  $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$

- 1. Multiplying row 2 of  $A$  by the scalar  $k$ :

$$\begin{bmatrix} & \end{bmatrix} \begin{bmatrix} a & b \\ c & d \end{bmatrix} = \begin{bmatrix} a & b \\ kc & kd \end{bmatrix}$$

- 2. Interchanging rows 1 and 2 of  $A$ :

$$\begin{bmatrix} & \end{bmatrix} \begin{bmatrix} a & b \\ c & d \end{bmatrix} = \begin{bmatrix} c & d \\ a & b \end{bmatrix}$$

- 3. Adding  $k$  times row 1 to row 2:

$$\begin{bmatrix} & \end{bmatrix} \begin{bmatrix} a & b \\ c & d \end{bmatrix} = \begin{bmatrix} a & b \\ ka + c & kb + d \end{bmatrix}$$



## Definition

An  $n \times n$  matrix  $E$  is called an **elementary matrix** if  $E$  can be obtained from  $I_n$  by a **single elementary row operation**.

$$\text{Example: } E_1 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix}, E_2 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & -4 & 0 \\ 0 & 0 & 1 \end{bmatrix}, E_3 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 2 & 0 & 1 \end{bmatrix}.$$

## Proposition

Let  $A$  be an  $m \times n$  **matrix**, and let  $E$  be an  $m \times m$  **elementary matrix** obtained by performing an **elementary row operation** on  $I_m$ . Then the product  $EA$  can be obtained from  $A$  by performing the identical elementary row operation on  $A$ .

$$\text{Example: } A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \\ 5 & 6 \end{bmatrix} \Rightarrow E_3 A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 2 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 3 & 4 \\ 5 & 6 \end{bmatrix} = \begin{bmatrix} 1 & 2 \\ 3 & 4 \\ 7 & 10 \end{bmatrix}.$$

Question: Is an elementary matrix always invertible?

Every elementary matrix is **invertible**. Furthermore, the inverse of an elementary matrix is also an elementary matrix.

$$\text{Example: } E_1 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix}, E_2 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & -4 & 0 \\ 0 & 0 & 1 \end{bmatrix}, E_3 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 2 & 0 & 1 \end{bmatrix}.$$

Then,

$$E_1^{-1} = \begin{bmatrix} \phantom{0} & \phantom{0} & \phantom{0} \\ \phantom{0} & \phantom{0} & \phantom{0} \\ \phantom{0} & \phantom{0} & \phantom{0} \end{bmatrix},$$

$$E_2^{-1} = \begin{bmatrix} \phantom{0} & \phantom{0} & \phantom{0} \\ \phantom{0} & \phantom{0} & \phantom{0} \\ \phantom{0} & \phantom{0} & \phantom{0} \end{bmatrix},$$

$$E_3^{-1} = \begin{bmatrix} \phantom{0} & \phantom{0} & \phantom{0} \\ \phantom{0} & \phantom{0} & \phantom{0} \\ \phantom{0} & \phantom{0} & \phantom{0} \end{bmatrix}.$$

In general, if  $E \in \mathcal{R}^{n \times n}$  is an elementary matrix that corresponds to some elementary row operation, then  $E^{-1}$  is the elementary matrix that corresponds to the **reverse** elementary row operation.

Think: How do we transform a matrix  $A$  to its reduced row echelon form by multiplying it with elementary matrices?

## Theorem 2.3

Let  $A$  be an  $m \times n$  matrix with reduced row echelon form  $R$ . Then there exists an invertible  $m \times m$  matrix  $P$  such that  $PA = R$ .

*Proof*

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## Corollary

The matrix equation  $A\mathbf{x} = \mathbf{b}$  has the same solution set as  $R\mathbf{x} = \mathbf{c}$ , where  $\begin{bmatrix} R & \mathbf{c} \end{bmatrix}$  is the reduced echelon form of  $\begin{bmatrix} A & \mathbf{b} \end{bmatrix}$ .

*Proof*

## Theorem E.1 (Column Correspondence Theorem)

Let  $A = [\mathbf{a}_1 \ \mathbf{a}_2 \ \cdots \ \mathbf{a}_n]$  be an  $m \times n$  matrix whose reduced row echelon form is  $R = [\mathbf{r}_1 \ \mathbf{r}_2 \ \cdots \ \mathbf{r}_n]$ . Then

(a) If  $\mathbf{a}_j$  is a linear combination of other columns of  $A$ , then  $\mathbf{r}_j$  is a linear combination of the corresponding columns of  $R$  with the same coefficients.

(b) If  $\mathbf{r}_j$  is a linear combination of other columns of  $R$ , then  $\mathbf{a}_j$  is a linear combination of the corresponding columns of  $A$  with the same coefficients.

Example:

$$A = \begin{bmatrix} 1 & 2 & -1 & 2 & 1 & 2 \\ -1 & -2 & 1 & 2 & 3 & 6 \\ 2 & 4 & -3 & 2 & 0 & 3 \\ -3 & -6 & 2 & 0 & 3 & 9 \end{bmatrix} \quad R = \begin{bmatrix} 1 & 2 & 0 & 0 & -1 & -5 \\ 0 & 0 & 1 & 0 & 0 & -3 \\ 0 & 0 & 0 & 1 & 1 & 2 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

$$\begin{array}{l} \mathbf{a}_2 = 2\mathbf{a}_1 \\ \mathbf{a}_5 = -\mathbf{a}_1 + \mathbf{a}_4 \end{array} \quad \Leftrightarrow \quad \begin{array}{l} \mathbf{r}_2 = 2\mathbf{r}_1 \\ \mathbf{r}_5 = -\mathbf{r}_1 + \mathbf{r}_4 \end{array}$$

## Theorem E.1

Let  $A = [ \mathbf{a}_1 \quad \mathbf{a}_2 \quad \cdots \quad \mathbf{a}_n ]$  be an  $m \times n$  matrix whose reduced row echelon form is  $R = [ \mathbf{r}_1 \quad \mathbf{r}_2 \quad \cdots \quad \mathbf{r}_n ]$ . Then

- (a) If  $\mathbf{a}_j$  is a linear combination of other columns of  $A$ , then  $\mathbf{r}_j$  is a linear combination of the corresponding columns of  $R$  with the same coefficients.
- (b) If  $\mathbf{r}_j$  is a linear combination of other columns of  $R$ , then  $\mathbf{a}_j$  is a linear combination of the corresponding columns of  $A$  with the same coefficients.

**\*Proof** (a) Suppose  $\mathbf{a}_j = \sum_{k=1, k \neq j}^n c_k \mathbf{a}_k$ .

You show part (b).

## Properties of a Matrix in Reduced Row Echelon Form (in Appendix E)

Let  $R$  be an  $m \times n$  matrix in reduced row echelon form. Then the following statements are true.

- (a) A column of  $R$  is a pivot column if and only if it is nonzero and not a linear combination of the preceding columns of  $R$ .
- (b) The  $j$ th pivot column of  $R$  is  $\mathbf{e}_j$ , the  $j$ th standard vector of  $\mathcal{R}^m$ , and hence the pivot columns of  $R$  are linearly independent.
- (c) Every column of  $R$  is a linear combination of the pivot columns of  $R$ .

$$\begin{bmatrix} 1 & 0 & * & * & 0 & 0 & * \\ 0 & 1 & * & * & 0 & 0 & * \\ 0 & 0 & 0 & 0 & 1 & 0 & * \\ 0 & 0 & 0 & 0 & 0 & 1 & * \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

**pivot columns**

Example:

$$\begin{bmatrix} 1 & 0 & * & * & 0 & 0 & * \\ 0 & 1 & * & * & 0 & 0 & * \\ 0 & 0 & 0 & 0 & 1 & 0 & * \\ 0 & 0 & 0 & 0 & 0 & 1 & * \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

**pivot columns**

More precisely for Property (c):

Suppose that  $\mathbf{r}_j$  is not a pivot column of  $R$ , and there are  $k$  pivot columns of  $R$  preceding it.

Then  $\mathbf{r}_j$  is a linear combination of the  $k$  preceding pivot columns, and the coefficients of the linear combination are the first  $k$  entries of  $\mathbf{r}_j$ . Furthermore, the other entries of  $\mathbf{r}_j$  are zeros.

Example:

$$\left[ \begin{array}{ccc|c} 1 & 2 & -1 & 0 \\ -1 & 1 & -8 & 1 \\ 2 & -1 & 13 & -2 \\ 1 & -1 & 8 & 1 \end{array} \right] \longrightarrow \left[ \begin{array}{ccc|c} 1 & 0 & 5 & 0 \\ 0 & 1 & -3 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{array} \right]$$



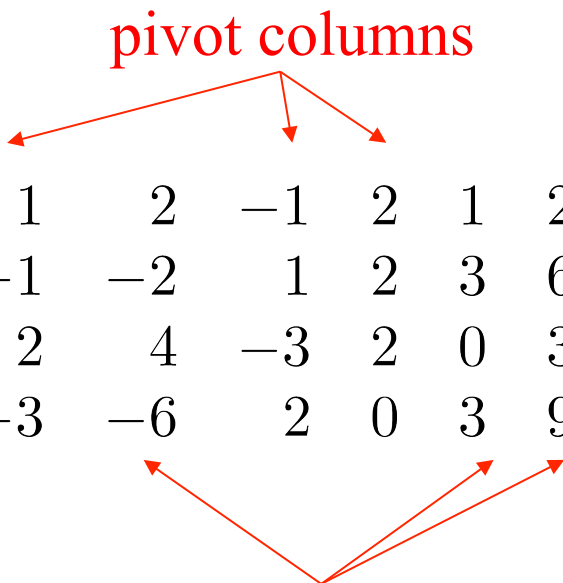
## Properties of R.R.E.F Matrices

### + Theorem E.1 (Correspondence Theorem)

→ (1) Pivot columns of  $A$  are linearly independent.

(2) Each non-pivot column is a linear comb. of pivot columns.

Example:

$$A = \begin{bmatrix} 1 & 2 & -1 & 2 & 1 & 2 \\ -1 & -2 & 1 & 2 & 3 & 6 \\ 2 & 4 & -3 & 2 & 0 & 3 \\ -3 & -6 & 2 & 0 & 3 & 9 \end{bmatrix} \quad R = \begin{bmatrix} 1 & 2 & 0 & 0 & -1 & -5 \\ 0 & 0 & 1 & 0 & 0 & -3 \\ 0 & 0 & 0 & 1 & 1 & 2 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$


non-pivot columns

$$\mathbf{a}_2 = 2\mathbf{a}_1$$

$$\mathbf{a}_5 = -\mathbf{a}_1 + \mathbf{a}_4$$

$$\mathbf{a}_6 = -5\mathbf{a}_1 - 3\mathbf{a}_3 + 2\mathbf{a}_4$$

## Theorem 2.4

The following statements are true for any matrix  $A$ .

- (a) The pivot columns of  $A$  are linearly independent.
- (b) Each nonpivot column of  $A$  is a linear combination of the previous pivot columns of  $A$ , where the coefficients of the linear combination are the entries of the corresponding column of the reduced row echelon form of  $A$ .

Example:

$$A = \begin{bmatrix} 1 & 2 & -1 & 2 & 1 & 2 \\ -1 & -2 & 1 & 2 & 3 & 6 \\ 2 & 4 & -3 & 2 & 0 & 3 \\ -3 & -6 & 2 & 0 & 3 & 9 \end{bmatrix} \quad R = \begin{bmatrix} 1 & 2 & 0 & 0 & -1 & -5 \\ 0 & 0 & 1 & 0 & 0 & -3 \\ 0 & 0 & 0 & 1 & 1 & 2 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

**pivot columns**

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### *Proof*

Suppose  $R$  is the r.r.e.f. of  $A$ .

Then there exists invertible  $m \times m$  matrix  $P$  such that  
 $R = PA$ .

Let  $\mathbf{r}_{p1} = \mathbf{e}_1, \mathbf{r}_{p2} = \mathbf{e}_2, \dots$ , be pivot columns of  $R$ .

Then,  $\mathbf{a}_{p1} = P\mathbf{e}_1, \mathbf{a}_{p2} = P\mathbf{e}_2, \dots$ , are also l.i.

(b) Suppose  $\mathbf{r}_j$  is a nonpivot column of  $R$ , then by property (c),

$$\mathbf{r}_j = c_1\mathbf{r}_{p1} + c_2\mathbf{r}_{p2} + \dots$$

$$\text{Then } \mathbf{a}_j = P^{-1}\mathbf{r}_j = P^{-1}(c_1\mathbf{r}_{p1} + \dots) = c_1\mathbf{a}_{p1} + c_2\mathbf{a}_{p2} + \dots$$

## \*Uniqueness of the reduced row echelon form of a matrix:

### **Theorem 1.4** (Introduced in Section 1.3 and partly proved in Section 1.4)

Every matrix can be transformed into **one** and **only one** in **reduced echelon form** by means of a sequence of **elementary row operations**.

**Proof** Let  $A \in \mathcal{R}^{m \times n}$  and  $R$  be a reduced row echelon form of  $A$ .

**Claim I:** Contents and positions of the pivot columns of  $R$  are **uniquely** determined by  $A$ .

**Claim II:** Contents and positions of the nonpivot columns of  $R$  are **uniquely** determined by  $A$ .

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**Claim I:** Contents and positions of the pivot columns of  $R$  are **uniquely** determined by  $A$ .

**Proof**

By Property (a) of  $R$  and the column correspondence property, we know that  $\mathbf{r}_i$  is a pivot column of  $R$

$\Leftrightarrow \mathbf{r}_i \neq \mathbf{0}$  and  $\mathbf{r}_i$  is not a l.c. of the preceding columns of  $R$  (Property a)

$\Leftrightarrow \mathbf{a}_i \neq \mathbf{0}$  and  $\mathbf{a}_i$  is not a linear combination of the preceding columns of  $A$ .

Thus **positions** of the pivot columns of  $R$  are uniquely determined by positions of  $A$ 's nonzero columns that are not linear combinations of the preceding columns.

By the Property (b) of  $R$ , the contents of the pivot columns of  $R$  are fixed ( $\mathbf{e}_1, \mathbf{e}_2, \dots$  in  $\mathcal{R}^m$ ).

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**Claim II:** Contents and positions of the nonpivot columns of  $R$  are **uniquely** determined by  $A$ .

#### *Proof*

Suppose  $\mathbf{r}_{n_1} = \mathbf{e}_1, \mathbf{r}_{n_2} = \mathbf{e}_2, \dots, \mathbf{r}_{n_k} = \mathbf{e}_k$  are the pivot columns of  $R$ , and  $\mathbf{r}_j$  is a non-pivot column of  $R$ .

$\Rightarrow \mathbf{r}_{n_1}, \mathbf{r}_{n_2}, \dots, \mathbf{r}_{n_k}$  are L.I. (property b) and  $\mathbf{r}_j = c_1\mathbf{r}_{n_1} + c_2\mathbf{r}_{n_2} + \dots + c_k\mathbf{r}_{n_k}$  for some  $c_1, c_2, \dots, c_k$ . (property c)

$\Rightarrow$  By the column correspondence property, (actually, by Thm 2.4b)

$\mathbf{a}_{n_1}, \mathbf{a}_{n_2}, \dots, \mathbf{a}_{n_k}$  are L.I. and

$$\mathbf{a}_j = c_1\mathbf{a}_{n_1} + c_2\mathbf{a}_{n_2} + \dots + c_k\mathbf{a}_{n_k}.$$

Thus  $c_1, c_2, \dots, c_k$  are (and hence  $\mathbf{r}_j$  is) uniquely determined by  $A$ .

## Homework Set for 2.3

Section 2.3: Problems 1, 3, 5, 9, 11, 15, 17, 19, 21, 23, 27, 29, 31, 33, 35, 37, 39, 41, 43, 45, 49, 51.