

## 7.6 The pivot theorem

# Basis problem

- Now address the problem of extracting a basis in  $S$  for the  $\text{Span}(S)$ .
- The row operations changes the column spaces.
- If  $A$  and  $B$  are row equivalent, then  $Ax=0$ ,  $Bx=0$  have the same set of solutions.
- $Ax=0 \leftrightarrow x_1a_1+x_2a_2+\dots+x_na_n=0$ .
- $Bx=0 \leftrightarrow x_1b_1+x_2b_2+\dots+x_nb_n=0$ .

**Theorem 7.6.1** *Let  $A$  and  $B$  be row equivalent matrices.*

- (a) If some subset of column vectors from  $A$  is linearly independent, then the corresponding column vectors from  $B$  are linearly independent, and conversely.*
- (b) If some subset of column vectors from  $B$  is linearly dependent, then the corresponding column vectors from  $A$  are linearly dependent, and conversely. Moreover, the column vectors in the two matrices have the same dependency relationships.*

- Proof: If necessary form  $A'$  from the set of column vectors of  $A$ .
- Thus our strategy is to ref  $A$  and choose the pivot columns as basis and transfer back to  $A$ .
- Example 1.

# Pivot theorem

**Definition 7.6.2** The column vectors of a matrix  $A$  that lie in the column positions where the leading 1's occur in the row echelon forms of  $A$  are called the ***pivot columns*** of  $A$ .

**Theorem 7.6.3 (The Pivot Theorem)** *The pivot columns of a nonzero matrix  $A$  form a basis for the column space of  $A$ .*

- Proof: We see that leading 1s are at every position in the pivot column vectors.

# Pivot algorithm

**Algorithm 1** If  $W$  is the subspace of  $R^n$  spanned by  $S = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_s\}$ , then the following procedure extracts a basis for  $W$  from  $S$  and expresses the vectors of  $S$  that are not in the basis as linear combinations of the basis vectors.

**Step 1.** Form the matrix  $A$  that has  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_s$  as successive column vectors.

**Step 2.** Reduce  $A$  to a row echelon form  $U$ , and identify the columns with the leading 1's to determine the pivot columns of  $A$ .

**Step 3.** Extract the pivot columns of  $A$  to obtain a basis for  $W$ . If appropriate, rewrite these basis vectors in comma-delimited form.

**Step 4.** If it is desired to express the vectors of  $S$  that are not in the basis as linear combinations of the basis vectors, then continue reducing  $U$  to obtain the reduced row echelon form  $R$  of  $A$ .

**Step 5.** By inspection, express each column vector of  $R$  that does not contain a leading 1 as a linear combination of preceding column vectors that contain leading 1's. Replace the column vectors in these linear combinations by the corresponding column vectors of  $A$  to obtain equations that express the column vectors of  $A$  that are not in the basis as linear combinations of basis vectors.

## Example 2

- Given  $W = \text{span}(S)$ .  $S$  finite.
- (a) Extract basis in  $S$ .
- (b) Express other vectors in  $S$

# Basis for the fundamental spaces

- $A$   $m \times n$   $\rightarrow$   $U$  upper echelon  $\rightarrow$   $R$  ref.
- 1.  $\text{row}(A)$ : basis nonzero rows of  $U$  or  $R$ .
- 2.  $\text{col}(A)$ : pivot columns of  $A$ .
- 3.  $\text{null}(A)$ : canonical solutions from  $Rx=0$ .
- 4.  $\text{null}(A^T)$ : Solve  $A^T x=0$ .
- $A$   $m \times n$  rank  $k$ .  $\dim \text{null}(A^T) = m-k$ . Why? If  $k=m$ ,  $\dim=0$ .
- Another method using row operations only.

**Algorithm 2** If  $A$  is an  $m \times n$  matrix with rank  $k$ , and if  $k < m$ , then the following procedure produces a basis for  $\text{null}(A^T)$  by elementary row operations on  $A$ .

**Step 1.** Adjoin the  $m \times m$  identity matrix  $I_m$  to the right side of  $A$  to create the partitioned matrix  $[A \mid I_m]$ .

**Step 2.** Apply elementary row operations to  $[A \mid I_m]$  until  $A$  is reduced to a row echelon form  $U$ , and let the resulting partitioned matrix be  $[U \mid E]$ .

**Step 3.** Repartition  $[U \mid E]$  by adding a horizontal rule to split off the zero rows of  $U$ . This yields a matrix of the form

$$\left[ \begin{array}{c|c} V & E_1 \\ \hline 0 & E_2 \end{array} \right] \begin{array}{l} k \\ m - k \end{array}$$

$n \quad m$

where the margin entries indicate sizes.

**Step 4.** The row vectors of  $E_2$  form a basis for  $\text{null}(A^T)$ .

- Example 3:



# Column-row factorization

**Theorem 7.6.4 (Column-Row Factorization)** *If  $A$  is a nonzero  $m \times n$  matrix of rank  $k$ , then  $A$  can be factored as*

$$A = CR \quad (1)$$

*where  $C$  is the  $m \times k$  matrix whose column vectors are the pivot columns of  $A$  and  $R$  is the  $k \times n$  matrix whose row vectors are the nonzero rows in the reduced row echelon form of  $A$ .*

- Proof:  $EA=R_0$ .  $E$   $m \times m$  matrix a product of elementary matrices.
  - $R_0$  ref of  $A$ .  $m \times n$ -matrix
  - Let  $R$  be the  $k \times n$ -matrix of nonzero rows of  $R_0$ .
  - Then let  $E^{-1}=[C|D]$   $C$   $m \times k$ .  $D$   $m \times (m-k)$

$$R_0 = \begin{bmatrix} R \\ O \end{bmatrix}$$

- Proof continued:

- $A = E^{-1}R = [C \mid D] \begin{bmatrix} R \\ O \end{bmatrix} = CR + DO = CR$

- C consists of pivot columns of A.
  - Multiplying by  $E^{-1}$  to  $R_0$  returns to A.
  - Restrict to pivot columns of R  $\rightarrow$  pivot columns of A.
  - Pivot columns of R form I of  $k \times k$  size.
  - CR restricted  $CI = C$ . Thus C is the pivot columns of A.
- Example 4.

# Column-row expansion

- We can write the above as the sum of vector products...

**Theorem 7.6.5 (Column-Row Expansion)** *If  $A$  is a nonzero matrix of rank  $k$ , then  $A$  can be expressed as*

$$A = \mathbf{c}_1\mathbf{r}_1 + \mathbf{c}_2\mathbf{r}_2 + \cdots + \mathbf{c}_k\mathbf{r}_k \quad (4)$$

*where  $\mathbf{c}_1, \mathbf{c}_2, \dots, \mathbf{c}_k$  are the successive pivot columns of  $A$  and  $\mathbf{r}_1, \mathbf{r}_2, \dots, \mathbf{r}_k$  are the successive nonzero row vectors in the reduced row echelon form of  $A$ .*

## Column-row rule (Theorem 3.8.1)

$$A^{m \times s} = [c_1, c_2, \dots, c_s], B^{s \times n} = \begin{bmatrix} r_1 \\ r_2 \\ \vdots \\ r_s \end{bmatrix}$$

$$(AB)_{ij} = \sum_{k=1}^s A_{ik} B_{kj} = \sum_{k=1}^s c_{ik} r_{kj}$$

$$AB = \sum_{k=1}^s c_k r_k = c_1 r_1 + c_2 r_2 + \cdots + c_s r_s$$