

An important and less trivial consequence of the rank formula is the following.

1.28 Theorem (Rank of the transpose). *Let $\mathbf{A} \in M_{m,n}$. Then we have*

- (i) *the maximum number of linearly independent columns and the maximum number of linearly independent rows are equal, i.e.,*

$$\text{Rank } \mathbf{A} = \text{Rank } \mathbf{A}^T,$$

- (ii) *let $p := \text{Rank } \mathbf{A}$. Then there exists a nonsingular $p \times p$ square submatrix of \mathbf{A} .*

Proof. (i) Let $\mathbf{A} = [a_j^i]$, let $\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n$ be the columns of \mathbf{A} and let $p := \text{Rank } \mathbf{A}$. We assume without loss of generality that the first p columns of \mathbf{A} are linearly independent and we define \mathbf{B} as the $m \times p$ submatrix formed by these columns, $\mathbf{B} := [\mathbf{a}_1 \mid \mathbf{a}_2 \mid \dots \mid \mathbf{a}_p]$. Since the remaining columns of \mathbf{A} depend linearly on the columns of \mathbf{B} , we have

$$a_j^k = \sum_{i=1}^p r_j^i a_i^k \quad \forall k = 1, \dots, m, \quad \forall j = p+1, \dots, n$$

for some $\mathbf{R} = [r_j^i] \in M_{p, n-p}(\mathbb{K})$. In terms of matrices,

$$[\mathbf{a}_{p+1} \mid \mathbf{a}_{p+2} \mid \dots \mid \mathbf{a}_n] = [\mathbf{a}_1 \mid \dots \mid \mathbf{a}_p] \mathbf{R} = \mathbf{B} \mathbf{R},$$

hence

$$\mathbf{A} = [\mathbf{B} \mid \mathbf{B} \mathbf{R}] = \mathbf{B} [\text{Id}_p \mid \mathbf{R}].$$

Taking the transposes, we have $\mathbf{A}^T \in M_{n,m}(\mathbb{K})$, $\mathbf{B}^T \in M_{p,m}(\mathbb{K})$ and

$$\mathbf{A}^T = \begin{pmatrix} \boxed{\text{Id}_p} \\ \boxed{\mathbf{R}^T} \end{pmatrix} \mathbf{B}^T. \tag{1.6}$$

Since $[\text{Id}_p \mid \mathbf{R}]^T$ is trivially injective, we infer that $\ker \mathbf{A}^T = \ker \mathbf{B}^T$, hence by the rank formula

$$\text{Rank } \mathbf{A}^T = m - \dim \ker \mathbf{A}^T = m - \dim \ker \mathbf{B}^T = \text{Rank } \mathbf{B}^T,$$

and we conclude that

$$\text{Rank } \mathbf{A}^T = \text{Rank } \mathbf{B}^T \leq \min(m, p) = p = \text{Rank } \mathbf{A}.$$

Finally, by applying the above to the matrix \mathbf{A}^T , we get the opposite inequality $\text{Rank } \mathbf{A} = \text{Rank } (\mathbf{A}^T)^T \leq \text{Rank } \mathbf{A}^T$, hence the conclusion.

- (ii) With the previous notation, we have $\text{Rank } \mathbf{B}^T = \text{Rank } \mathbf{B} = p$. Thus \mathbf{B} has a set of p independent rows. The submatrix \mathbf{S} of \mathbf{B} made by these rows is a square $p \times p$ matrix with $\text{Rank } \mathbf{S} = \text{Rank } \mathbf{S}^T = p$, hence nonsingular. \square

1.29 ¶. Let $\mathbf{A} \in M_{m,n}(\mathbb{K})$, let $A(\mathbf{x}) := \mathbf{A}\mathbf{x}$ and let $(\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n)$ be a basis of \mathbb{K}^n . Show the following:

- (i) A is injective if and only if the vectors $A(\mathbf{v}_1), A(\mathbf{v}_2), \dots, A(\mathbf{v}_n)$ of \mathbb{K}^m are linearly independent,
- (ii) A is surjective if and only if $\{A(\mathbf{v}_1), A(\mathbf{v}_2), \dots, A(\mathbf{v}_n)\}$ spans \mathbb{K}^m ,
- (iii) A is bijective iff $\{A(\mathbf{v}_1), A(\mathbf{v}_2), \dots, A(\mathbf{v}_n)\}$ is a basis of \mathbb{K}^m .