

Definitions

- Pivot of A A column in $\text{rref}(A)$ which contains a leading one has a corresponding column in A , called a pivot column of A .
- Basis of V It is an independent set v_1, \dots, v_k from data set V whose linear combinations generate all data items in V . Generally, a basis is discovered by taking partial derivatives on symbols representing arbitrary constants.

Main Results

Theorem 1 (Dimension)

If a vector space V has a basis v_1, \dots, v_p and also a basis u_1, \dots, u_q , then $p = q$. The **dimension** of V is this unique number p .

Theorem 2 (The Pivot Theorem)

- The pivot columns of a matrix A are linearly independent.
- A non-pivot column of A is a linear combination of the pivot columns of A .

Definitions

$\text{rank}(A)$ The number of leading ones in $\text{rref}(A)$

$\text{nullity}(A)$ The number of columns of A minus $\text{rank}(A)$

Main Results

Theorem 3 (Rank-Nullity Equation)

$\text{rank}(A) + \text{nullity}(A) = \text{column dimension of } A$

Theorem 4 (Row Rank Equals Column Rank)

The number of independent rows of a matrix A equals the number of independent columns of A . Equivalently, $\text{rank}(A) = \text{rank}(A^T)$.

Theorem 5 (Pivot Method)

Let A be the augmented matrix of $\mathbf{v}_1, \dots, \mathbf{v}_k$. Let the leading ones in $\text{rref}(A)$ occur in columns i_1, \dots, i_p . Then a largest independent subset of the k vectors $\mathbf{v}_1, \dots, \mathbf{v}_k$ is the set

$$\mathbf{v}_{i_1}, \mathbf{v}_{i_2}, \dots, \mathbf{v}_{i_p}.$$

Definitions

$\text{kernel}(\mathbf{A}) = \text{nullspace}(\mathbf{A}) = \{\mathbf{x} : \mathbf{A}\mathbf{x} = \mathbf{0}\}.$

$\text{Image}(\mathbf{A}) = \text{colspace}(\mathbf{A}) = \{\mathbf{y} : \mathbf{y} = \mathbf{A}\mathbf{x} \text{ for some } \mathbf{x}\}.$

$\text{rowspace}(\mathbf{A}) = \text{colspace}(\mathbf{A}^T) = \{\mathbf{w} : \mathbf{w} = \mathbf{A}^T\mathbf{y} \text{ for some } \mathbf{y}\}.$

$\dim(V)$ is the number of elements in a basis for V .

How to Compute Null, Row, Column Space

Null Space. Compute $\text{rref}(\mathbf{A})$. Write out the general solution \mathbf{x} to $\mathbf{A}\mathbf{x} = \mathbf{0}$, where the free variables are assigned parameter names t_1, \dots, t_k . Report the basis for $\text{nullspace}(\mathbf{A})$ as the list $\partial_{t_1}\mathbf{x}, \dots, \partial_{t_k}\mathbf{x}$.

Column Space. Compute $\text{rref}(\mathbf{A})$. Identify the pivot columns i_1, \dots, i_k . Report the basis for $\text{colspace}(\mathbf{A})$ as the list of columns i_1, \dots, i_k of \mathbf{A} .

Row Space. Compute $\text{rref}(\mathbf{A}^T)$. Identify the lead variable columns i_1, \dots, i_k . Report the basis for $\text{rowspace}(\mathbf{A})$ as the list of rows i_1, \dots, i_k of \mathbf{A} .

Alternatively, compute $\text{rref}(\mathbf{A})$, then $\text{rowspace}(\mathbf{A})$ has a *different* basis consisting of the list of nonzero rows of $\text{rref}(\mathbf{A})$.

Theorem 6 (Dimension Identities)

(a) $\dim(\text{nullspace}(A)) = \dim(\text{kernel}(A)) = \text{nullity}(A)$

(b) $\dim(\text{colspace}(A)) = \dim(\text{Image}(A)) = \text{rank}(A)$

(c) $\dim(\text{rowspace}(A)) = \text{rank}(A)$

(d) $\dim(\text{kernel}(A)) + \dim(\text{Image}(A)) = \text{column dimension of } A$

(e) $\dim(\text{kernel}(A)) + \dim(\text{kernel}(A^T)) = \text{column dimension of } A$

Theorem 7 (Equivalence Test for Bases)

Define augmented matrices

$$B = \text{aug}(\mathbf{v}_1, \dots, \mathbf{v}_k), \quad C = \text{aug}(\mathbf{u}_1, \dots, \mathbf{u}_\ell), \quad W = \text{aug}(B, C).$$

Then relation

$$k = \ell = \text{rank}(B) = \text{rank}(C) = \text{rank}(W)$$

implies

1. $\mathbf{v}_1, \dots, \mathbf{v}_k$ is an independent set.
2. $\mathbf{u}_1, \dots, \mathbf{u}_\ell$ is an independent set.
3. $\text{span}\{\mathbf{v}_1, \dots, \mathbf{v}_k\} = \text{span}\{\mathbf{u}_1, \dots, \mathbf{u}_\ell\}$

In particular, $\text{colspace}(B) = \text{colspace}(C)$ and each set of vectors is an equivalent basis for this vector space.

Proof: Because $\text{rank}(B) = k$, then the first k columns of W are independent. If some column of C is independent of the columns of B , then W would have $k + 1$ independent columns, which violates $k = \text{rank}(W)$. Therefore, the columns of C are linear combinations of the columns of B . Then vector space $\text{colspace}(C)$ is a subspace of vector space $\text{colspace}(B)$. Because both vector spaces have dimension k , then $\text{colspace}(B) = \text{colspace}(C)$. The proof is complete.

Equivalent Bases: Computer Illustration

The following maple code applies the theorem to verify that two bases are equivalent:

1. The basis is determined from the `colspace` command in maple.
2. The basis is determined from the pivot columns of A .

In maple, the report of the column space basis is identical to the nonzero rows of $\text{rref}(A^T)$.

```
with(linalg):
A:=matrix([[1,0,3],[3,0,1],[4,0,0]]);
colspace(A);          # Solve Ax=0, basis v1,v2 below
v1:=vector([2,0,-1]);v2:=vector([0,2,3]);
rref(A);              # Find the pivot cols=1,3
u1:=col(A,1); u2:=col(A,3); # pivot col basis
B:=augment(v1,v2); C:=augment(u1,u2);
W:=augment(B,C);
rank(B),rank(C),rank(W); # Test requires all equal 2
```

A False Test for Equivalent Bases

The relation

$$\text{rref}(B) = \text{rref}(C)$$

holds for a substantial number of matrices B and C . However, it does not imply that each column of C is a linear combination of the columns of B .

For example, define

$$B = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 1 \end{pmatrix}, \quad C = \begin{pmatrix} 1 & 1 \\ 0 & 1 \\ 1 & 0 \end{pmatrix}.$$

Then

$$\text{rref}(B) = \text{rref}(C) = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix},$$

but $\text{col}(C, 2)$ is not a linear combination of the columns of B . This means $\text{colspace}(B) \neq \text{colspace}(C)$.

Geometrically, the column spaces are planes in \mathbf{R}^3 which intersect only along the line L through the two points $(0, 0, 0)$ and $(1, 0, 1)$.

