

1 Dimensions of Vector Spaces

We already have a pretty good concept of dimensions already, but we'll go through some formal definitions here.

The dimension of a vector space is based on the linearly independent set that makes up its basis. For example, in \mathbb{R}^3 , there are 3 basis vectors, so there are 3 dimensions.

Assume $\{\vec{x}, \vec{y}, \vec{z}\}$ is a basis for V . What can we say about $\{\vec{x}, \vec{y}, \vec{z}, \vec{w}\}$? We know that if we add a vector to our basis, it must be linearly dependent; \vec{w} is a linear combination of the others.

Theorem 1 *If a vector space V has a basis $B = \{\vec{b}_1, \dots, \vec{b}_n\}$, then any set in V containing more than n vectors must be linearly dependent.*

Theorem 2 *If a vector space V has a basis of n vectors, then V is **finite-dimensional**, and the dimension, $\dim V$, is the number of vectors in a basis V .*

The dimension of a zero vector space $\{\vec{0}\}$ is defined to be 0.

If V is not spanned by a finite set, V is infinite-dimensional.

Some Common Vector Space Dimensions

$$\dim \mathbb{R}^n = n$$

$$\dim \mathbb{P}_n = n + 1$$

Example

Let the x - y plane in \mathbb{R}^3 be defined as $P = \text{Span}\{\vec{x}, \vec{y}\}$. The basis of P is the set \vec{x}, \vec{y} .

$$\dim P = 2$$

Example

Let W be the set of all vectors of the form

$$\begin{bmatrix} a - b \\ b - c \\ c - a \\ b \end{bmatrix}$$

The set of vectors that span W can be written using parametric form.

What is the dimension of W ?

We should put this in parametric vector form, then set up a matrix and do Gaussian elimination.

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

We can tell from here that the maximum dimension possible is 3 because there are 3 columns. But it could be lower. Let's reduce

$$\text{RREF}(A) = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix}$$

$$\dim W = 3$$

Subspace of a Finite-Dimensional Space

Theorem 3 *Let H be a subspace of a finite-dimensional vector space V . Any linearly independent set in H can be expanded into a basis for H .*

H is also finite-dimensional and $\dim H \leq \dim V$.

Theorem 4 (The Basis Theorem) *Let V be a p -dimensional vector space, $p \geq 1$.*

Any linearly independent set of exactly p elements in V is automatically a basis for V .

2 Column and Null Space

The **rank** of an $m \times n$ matrix A is the dimension of the column space.

The **nullity** of A is the dimension of the null space.

Theorem 5 (The Rank Theorem)

$$\text{rank } A + \text{nullity } A = \# \text{ of columns in } A$$

Example

$$A = \begin{bmatrix} 1 & 3 & -4 & 2 & -1 & 6 \\ 0 & 0 & 1 & -3 & 7 & 0 \\ 0 & 0 & 0 & 1 & 4 & -3 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

Determine the dimensions of $\text{Nul } A$, $\text{Col } A$.

To find $\text{rank } A$, we count the vectors in the column space, which are all the columns with pivot positions. $\text{rank } A = 3$.

For the nullity, there are a few ways we can find it. We know that $\text{rank } A + \text{nullity } A = \# \text{ of columns in } A$, so $6 - 3 = 3$. We can also count the free variables to find the null space (I think), which is also 3.

Example

If a 6×3 matrix A has rank 3, find $\text{nullity } A$ and $\text{rank } A^T$.

We know that there are 3 columns, so by the rank theorem, $\text{nullity } A = 0$.

For $\text{rank } A^T$, imagine transposing A . We would have 3 rows and 6 columns. However, we know that there are only 3 basic variables, so $\text{rank } A^T = 3$.