

Lecture 1a: Basis and dimension

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- These notes are purposefully terse; you can generate more detailed notes, as per your writing style.
- Include a short summary of the concepts covered so far (in particular if they relate to the topic you are writing about)

RECALL: vector spaces, matrix multiplication, linear independence

Basis and dimension

A maximal linearly independent set in a vector space is called a basis. We say that a set $T \subseteq \mathcal{V}$ is maximally linearly independent if adding any vector to T makes it linearly dependent, i.e.

$T \cup \{\underline{v}\}$ is linearly dependent for any $\underline{v} \in \mathcal{V}$.

Theorem 1. If $\{\underline{v}_1, \underline{v}_2, \dots, \underline{v}_k\}$ is a basis for \mathcal{V} , then every vector $\underline{v} \in \mathcal{V}$ can be written uniquely as

$$\underline{v} = \sum_{i=1}^k \alpha_i \underline{v}_i.$$

The scalars α_i are called the coefficients or coordinates of the vector \underline{v} in the given basis.

Proof. 1. *Span:* Suppose we have a $\underline{v} \in \mathcal{V}$ that cannot be written as a linear combination of $\{\underline{v}_1, \underline{v}_2, \dots, \underline{v}_k\}$. Then $\underline{v} \cup \{\underline{v}_1, \underline{v}_2, \dots, \underline{v}_k\}$ is a linearly independent set (why?), contradicting of maximality of $\{\underline{v}_1, \underline{v}_2, \dots, \underline{v}_k\}$.

2. *Uniqueness:* Suppose there are two different ways of writing \underline{v} as a linear combination

$$\underline{v} = \sum_{i=1}^k \alpha_i \underline{v}_i, \quad \underline{v} = \sum_{i=1}^k \beta_i \underline{v}_i, \quad \text{with } (\alpha_1, \alpha_2, \dots) \neq (\beta_1, \beta_2, \dots).$$

Then $\sum_{i=1}^k (\alpha_i - \beta_i) \underline{v}_i = 0$, giving a non-trivial linear combination that is zero. This contradicts the linear independence of $\{\underline{v}_1, \underline{v}_2, \dots, \underline{v}_k\}$.

□

There could be many bases for a vector space ¹. The following result guarantees that all bases of a vector space have the same number of elements.

For example, the set $\left\{ \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \right\}$ is a basis for \mathbb{R}^3 . The set $\{1, x, x^2, x^3\}$ is a basis for the vector space consisting of all degree 3 polynomials, etc.

This result highlights the importance of a basis. There are two aspects to this theorem: First that any vector *can* be expressed as a linear combination of the vectors in the basis, and secondly, that such an expression is *unique*. This justifies the term 'basis': a set that can be used to build the vector space.

¹ As an exercise, try to construct three different bases for \mathbb{R}^3

Theorem 2. Any two bases of a vector space have the same number of elements.

Proof. Suppose we have bases $\{\underline{v}_1, \underline{v}_2, \dots, \underline{v}_k\}$ and $\{\underline{w}_1, \underline{w}_2, \dots, \underline{w}_p\}$ with $k < p$. Let $V = \begin{pmatrix} \underline{v}_1 & \underline{v}_2 & \dots & \underline{v}_k \end{pmatrix}$ and $W = \begin{pmatrix} \underline{w}_1 & \underline{w}_2 & \dots & \underline{w}_p \end{pmatrix}$ be the corresponding matrices. Then $W = VA$ for some $k \times p$ matrix A (why? ²). Thus the linear independence of columns of W and columns of V implies that

$$A\underline{x} = 0 \text{ only when } \underline{x} = 0,$$

which is a contradiction since A is a fat matrix ($k < p$) ³. \square

Because of Theorem ??, it makes sense to define the following:

Definition. THE DIMENSION of a vector space is the size of any basis in the vector space. ⁴

Matrix spaces

Recall the system interpretation of a matrix. In view of that, we define the following

1. The image space of A , written $im(A)$ ⁵ is the set of all possible outputs of A

$$im(A) = \{A\underline{x} : \underline{x} \in \mathbb{R}^n\}.$$

2. The null space of A , written $null(A)$ ⁶, is the set of all inputs that give a zero output

$$null(A) = \{\underline{x} \in \mathbb{R}^n : A\underline{x} = 0\}.$$

Characterize the column and null spaces of

$$A = \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix} \quad B = \begin{pmatrix} 1 & 2 \\ 2 & 4 \end{pmatrix}.$$

Since all bases (maximally linearly independent sets) have the same size, one could also have defined a basis as a *maximum* size linearly independent set.

² This is obtained by writing each basis vector in $\{\underline{w}_1, \underline{w}_2, \dots, \underline{w}_p\}$ in terms of the basis vectors in $\{\underline{v}_1, \underline{v}_2, \dots, \underline{v}_k\}$.

³ This result (that $A\underline{x} = 0$ always has a non zero solution when A is fat) was used in an earlier lecture as well. We can prove it using row echelon forms, which we will defer till later in the course.

⁴ What is the dimension of the set of all degree 3 polynomials f with $f(0) = 0$? You need to first check that this is a vector space.

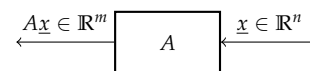


Figure 1: System interpretation of an $m \times n$ matrix A . The matrix takes an 'input' $\underline{x} \in \mathbb{R}^n$ and gives an 'output' $A\underline{x} \in \mathbb{R}^m$. The input space is \mathbb{R}^n and the output space is \mathbb{R}^m .

⁵ Note that $im(A)$ lies in the output space, i.e. $im(A) \subseteq \mathbb{R}^m$. Since $A\underline{x}$ is a linear combination of columns of A , the set $im(A)$ is also called the column space of A .

⁶ Note that $null(A)$ lies in the input space, i.e. $null(A) \subseteq \mathbb{R}^n$.

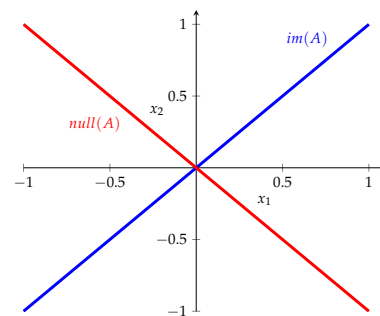


Figure 2: For the 2×2 matrix A , both $im(A)$ and $null(A)$ are in \mathbb{R}^2 , plotted above. Verify, and similarly repeat for B .