

Math 211

Lecture #20
Bases of a Subspace

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Subspaces of \mathbf{R}^n

Definition: A nonempty subset V of \mathbf{R}^n that has the properties

1. if \mathbf{x} and \mathbf{y} are vectors in V , $\mathbf{x} + \mathbf{y}$ is in V ,
2. if a is a scalar, and \mathbf{x} is in V , then $a\mathbf{x}$ is in V ,

is called a *subspace* of \mathbf{R}^n .

- The nullspace of a matrix is a subspace.
- We are looking for a good way to describe a subspace.

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The Span of a Set of Vectors

In every example we have seen the subspace has been the set of all linear combinations of a few vectors.

Definition: The *span* of a set of vectors is the set of all linear combinations of those vectors. The span of the vectors $\mathbf{v}_1, \mathbf{v}_2, \dots$, and \mathbf{v}_k is denoted by

$$\text{span}(\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k).$$

Proposition: If $\mathbf{v}_1, \mathbf{v}_2, \dots$, and \mathbf{v}_k are all vectors in \mathbf{R}^n , then $V = \text{span}(\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k)$ is a subspace of \mathbf{R}^n .

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[null\(A\)](#)

[null\(B\)](#)

Linear Dependence in 2- & 3-D

We need a condition that will keep unneeded vectors out of a spanning list. We will work toward a general definition.

- Two vectors are *linearly dependent* if one is a scalar multiple of the other.
- Three vectors \mathbf{v}_1 , \mathbf{v}_2 , and \mathbf{v}_3 are *linearly dependent* if one is a linear combination of the other two.
- ♦ Example: $\mathbf{v}_1 = (1, 0, 0)^T$, $\mathbf{v}_2 = (0, 1, 0)^T$, and $\mathbf{v}_3 = (1, 2, 0)^T$

$$\mathbf{v}_3 = \mathbf{v}_1 + 2\mathbf{v}_2.$$

- ♦ Notice that $\mathbf{v}_1 + 2\mathbf{v}_2 - \mathbf{v}_3 = \mathbf{0}$.

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Linear Dependence

- Three vectors are linearly dependent if there is a non-trivial linear combination of them which equals the zero vector.
- ♦ Non-trivial means that at least one of the coefficients is not 0.
- A set of vectors is linearly dependent if there is a non-trivial linear combination of them which equals the zero vector.

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Linear Independence

Definition: The vectors \mathbf{v}_1 , \mathbf{v}_2 , \dots , and \mathbf{v}_k are *linearly independent* if the only linear combination of them which is equal to the zero vector is the one with all of the coefficients equal to 0.

- In symbols,

$$\begin{aligned} c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + \dots + c_k\mathbf{v}_k &= \mathbf{0} \\ \Rightarrow c_1 = c_2 = \dots = c_k &= 0. \end{aligned}$$

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[Three vectors](#)

[More vectors](#)

Linear Independence?

How do we decide if a set of vectors is linearly independent? Are the vectors

$$\mathbf{v}_1 = \begin{pmatrix} 1 \\ -2 \\ 0 \\ 2 \end{pmatrix}, \mathbf{v}_2 = \begin{pmatrix} -1 \\ -3 \\ 2 \\ 0 \end{pmatrix}, \mathbf{v}_3 = \begin{pmatrix} 5 \\ 0 \\ -4 \\ 6 \end{pmatrix}$$

linearly independent?

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We look at linear combinations of the vectors

$$c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + c_3\mathbf{v}_3 = \mathbf{0}$$

$$\Leftrightarrow [\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3]\mathbf{c} = \mathbf{0} \quad \text{where } \mathbf{c} = (c_1, c_2, c_3)^T$$

$$\Leftrightarrow \mathbf{c} \in \text{null}([\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3]).$$

- $\mathbf{c} = (-3, 2, 1)^T \in \text{null}([\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3])$,
 $\Rightarrow -3\mathbf{v}_1 + 2\mathbf{v}_2 + \mathbf{v}_3 = \mathbf{0}$.
- $\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3$ are linearly dependent.

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[Example](#)

[Linear independence](#)

Another Example

Are the vectors

$$\mathbf{v}_1 = \begin{pmatrix} 1 \\ -2 \\ 0 \\ 2 \end{pmatrix}, \mathbf{v}_2 = \begin{pmatrix} -1 \\ -3 \\ 2 \\ 0 \end{pmatrix}, \mathbf{v}_3 = \begin{pmatrix} 5 \\ 0 \\ -4 \\ 3 \end{pmatrix}$$

linearly independent?

- $\text{null}([\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3]) = \{\mathbf{0}\}$.
- $\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3$ are linearly independent.

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[Method](#)

[Linear independence](#)

Proposition: Suppose that $\mathbf{v}_1, \mathbf{v}_2, \dots,$ and \mathbf{v}_k are vectors in \mathbf{R}^n . Set $V = [\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k]$.

1. If $\text{null}(V) = \{\mathbf{0}\}$, then $\mathbf{v}_1, \mathbf{v}_2, \dots,$ and \mathbf{v}_k are linearly independent.
2. If $\mathbf{c} = (c_1, c_2, \dots, c_k)^T$ is a nonzero vector in $\text{null}(V)$, then

$$c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + \dots + c_k\mathbf{v}_k = \mathbf{0},$$

so the vectors are linearly dependent.

Method & example

Another example

Basis of a Subspace

Definition: A set of vectors $\mathbf{v}_1, \mathbf{v}_2, \dots,$ and \mathbf{v}_k form a *basis* of a subspace V if

1. $V = \text{span}(\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k)$
2. $\mathbf{v}_1, \mathbf{v}_2, \dots,$ and \mathbf{v}_k are linearly independent.

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Span

Examples of Bases

- The vector $\mathbf{v} = (1, -1, 1)^T$ is a basis for $\text{null}(A)$.
 - ♦ $\text{null}(A)$ is the subspace of \mathbf{R}^3 with basis \mathbf{v} .
- The vectors $\mathbf{v} = (1, -1, 1, 0)^T$ and $\mathbf{w} = (0, -2, 0, 1)^T$ form a basis for $\text{null}(B)$.
 - ♦ $\text{null}(B)$ is the subspace of \mathbf{R}^4 with basis $\{\mathbf{v}, \mathbf{w}\}$.

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Basis of a Subspace

Proposition: Let V be a subspace of \mathbf{R}^n .

1. If $V \neq \{\mathbf{0}\}$, then V has a basis.
2. Every basis of V has the same number of elements.

Definition: The *dimension* of a subspace V is the number of elements in a basis of V .

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Examples

Example

Find the nullspace of

$$A = \begin{pmatrix} 3 & -3 & 1 & -1 \\ -2 & 2 & -1 & 1 \\ 1 & -1 & 0 & 0 \\ 13 & -13 & 5 & -5 \end{pmatrix}.$$

- $\text{null}(A)$ is the subspace of \mathbf{R}^4 with basis $(1, 1, 0, 0)^T$ and $(0, 0, 1, -1)^T$.
- $\text{null}(A)$ has dimension 2.

Example 1

$$A = \begin{pmatrix} 4 & 3 & -1 \\ -3 & -2 & 1 \\ 1 & 2 & 1 \end{pmatrix}$$

The nullspace of A is

$$\text{null}(A) = \{a\mathbf{v} \mid a \in \mathbf{R}\},$$

where $\mathbf{v} = (1, -1, 1)^T$.

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Example 2

$$B = \begin{pmatrix} 4 & 3 & -1 & 6 \\ -3 & -2 & 1 & -4 \\ 1 & 2 & 1 & 4 \end{pmatrix}$$

- $\text{null}(B) = \{a\mathbf{v} + b\mathbf{w} \mid a, b \in \mathbf{R}\}$, where $\mathbf{v} = (1, -1, 1, 0)^T$ and $\mathbf{w} = (0, -2, 0, 1)^T$.
- $\text{null}(B)$ consists of all linear combinations of \mathbf{v} and \mathbf{w} .

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