

Linear Independence

Linear Algebra

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This work was initially funded by an Institutional OER Grant from the Colorado Department of Higher Education (CDHE). For similar OER materials in other courses funded by this project in the Department of Mathematical and Statistical Sciences at the University of Colorado Denver, visit <https://github.com/CU-Denver-MathStats-OER>

Redundancy when taking Span

We've seen that the **Span** of a set $S = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$ of vectors does not change if we add a vector \mathbf{w} that is a linear combination of $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$.

So there is redundancy in S for the **Span** if a vector $\mathbf{v}_j \in S$ can be written as a linear combination of the other vectors in S .

How can we tell if a vector in S can be written as a linear combination of the other vectors?

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For **each** vector $\mathbf{v}_j \in S$, we can solve

$$x_1\mathbf{v}_1 + x_2\mathbf{v}_2 + \cdots + x_{j-1}\mathbf{v}_{j-1} + \overbrace{\hspace{1cm}}^{\mathbf{v}_j \text{ missing}} + x_{j+1}\mathbf{v}_{j+1} + \cdots + x_n\mathbf{v}_n = \mathbf{v}_j.$$

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This might require solving **n systems** of linear equations!

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In particular, suppose that $c_j \neq 0$. Then we solve for \mathbf{v}_j :

$$c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + \cdots + c_j\mathbf{v}_j + \cdots + c_n\mathbf{v}_n = \mathbf{0}$$

$$c_j\mathbf{v}_j = -c_1\mathbf{v}_1 - c_2\mathbf{v}_2 + \cdots - c_{j-1}\mathbf{v}_{j-1} - c_{j+1}\mathbf{v}_{j+1} - \cdots - c_n\mathbf{v}_n$$

$$\mathbf{v}_j = -\frac{c_1}{c_j}\mathbf{v}_1 - \frac{c_2}{c_j}\mathbf{v}_2 - \cdots - \frac{c_{j-1}}{c_j}\mathbf{v}_{j-1} - \frac{c_{j+1}}{c_j}\mathbf{v}_{j+1} - \cdots - \frac{c_n}{c_j}\mathbf{v}_n.$$

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So \mathbf{v}_j can be written as a **linear combination** of the other vectors $\mathbf{v}_1, \dots, \mathbf{v}_{j-1}, \mathbf{v}_{j+1}, \dots, \mathbf{v}_n$.

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So \mathbf{v}_j can be written as a **linear combination** of the other vectors $\mathbf{v}_1, \dots, \mathbf{v}_{j-1}, \mathbf{v}_{j+1}, \dots, \mathbf{v}_n$. This is **if and only if**: if \mathbf{v}_j can be written as a linear combination of the other vectors, then there exists a nontrivial solution to the homogeneous system.

Revisiting Vector Equations

If A is an $m \times n$ matrix with columns vectors $\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n$ (each \mathbf{a}_i is in \mathbb{R}^m), the homogeneous matrix equation $A\mathbf{x} = \mathbf{0}$ has the same solution set as the vector equation

$$x_1\mathbf{a}_1 + x_2\mathbf{a}_2 + \dots + x_n\mathbf{a}_n = \mathbf{0},$$

which has corresponding augmented matrix

$$[\mathbf{a}_1 \quad \mathbf{a}_2 \quad \dots \quad \mathbf{a}_n \quad \mathbf{0}].$$

- ▶ Homogeneous linear systems always have a trivial solution $x_1 = x_2 = \dots = x_n = 0$.
- ▶ If a non-trivial solution exists, then:
 - ▶ At least one $x_j \neq 0$.
 - ▶ At least one \mathbf{a}_j can be written as a linear combination of the other column vectors.
- ▶ If the trivial solution is the only solution, then it is not possible to write any column vector as a linear combination of the other column vectors.

Example

Determine whether the equation

$$x_1 \begin{bmatrix} 2 \\ 1 \\ 0 \end{bmatrix} + x_2 \begin{bmatrix} -1 \\ 6 \\ 3 \end{bmatrix} + x_3 \begin{bmatrix} 6 \\ -10 \\ -6 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

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Since x_3 is a **free variable**, nontrivial solutions exist!

Interpreting the Solution

Determine whether the equation

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has a non-trivial solution.

Solution set is

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -2x_3 \\ 2x_3 \\ x_3 \end{bmatrix} = x_3 \begin{bmatrix} -2 \\ 2 \\ 1 \end{bmatrix}$$

So we have

$$-2 \begin{bmatrix} 2 \\ 1 \\ 0 \end{bmatrix} + 2 \begin{bmatrix} -1 \\ 6 \\ 3 \end{bmatrix} + 1 \begin{bmatrix} 6 \\ -10 \\ -6 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

which gives

$$\begin{bmatrix} 6 \\ -10 \\ -6 \end{bmatrix} = 2 \begin{bmatrix} 2 \\ 1 \\ 0 \end{bmatrix} + -2 \begin{bmatrix} -1 \\ 6 \\ 3 \end{bmatrix}$$

Linear Independence

Let $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$ denote a set of n vectors each in \mathbb{R}^m . We say the set of vectors is **linearly independent** if the vector equation $x_1\mathbf{v}_1 + x_2\mathbf{v}_2 + \dots + x_n\mathbf{v}_n = \mathbf{0}$ has only a trivial solution.

Otherwise, a set of vectors $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$ is said to be **linearly dependent** if there exists weights c_1, c_2, \dots, c_n **not all zero** such that

$$c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + \dots + c_n\mathbf{v}_n = \mathbf{0}.$$

Example

The set $\left\{ \begin{bmatrix} 2 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ 6 \\ 3 \end{bmatrix}, \begin{bmatrix} 6 \\ -10 \\ -6 \end{bmatrix} \right\}$ is **linearly dependent**.

Example

Determine whether the set of vectors $S = \left\{ \begin{bmatrix} 1 \\ 2 \end{bmatrix}, \begin{bmatrix} 3 \\ 4 \end{bmatrix} \right\}$ is linearly independent.

$$x_1 \begin{bmatrix} 1 \\ 2 \end{bmatrix} + x_2 \begin{bmatrix} 3 \\ 4 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 3 & 0 \\ 2 & 4 & 0 \end{bmatrix} \xrightarrow{\text{RREF}} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$

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$$\begin{bmatrix} 1 & 3 & 0 \\ 2 & 4 & 0 \end{bmatrix} \xrightarrow{\text{RREF}} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$

There is a unique solution, the trivial solution $\mathbf{x} = \mathbf{0}$, so the vectors are **linearly independent**.

The columns of matrix A are linearly independent if and only if the matrix equation $A\mathbf{x} = \mathbf{0}$ has **only the trivial solution**.

The Invertible Matrix Theorem redux 2

Let A be a square $n \times n$ matrix. Then all of the following statements are **equivalent**.

- (a) A is an invertible matrix.
- (b) A is row equivalent to the $n \times n$ identity matrix I_n .
- (c) A has n pivots.
- (d) The equation $A\mathbf{x} = \mathbf{0}$ has only the trivial solution.
- (e) The columns of A form a linearly independent set.
- (f) The linear transformation $\mathbf{x} \mapsto A\mathbf{x}$ mapping \mathbb{R}^n into \mathbb{R}^n is one-to-one.
- (g) The equation $A\mathbf{x} = \mathbf{b}$ has at least one solution for each \mathbf{b} in \mathbb{R}^n .
- (h) The columns of A span \mathbb{R}^n .
- (i) The linear transformation $\mathbf{x} \mapsto A\mathbf{x}$ maps \mathbb{R}^n onto \mathbb{R}^n .
- (j) There is an $n \times n$ matrix C such that $CA = I_n$.
- (k) There is an $n \times n$ matrix D such that $AD = I_n$.
- (l) A^T is an invertible matrix.
- (m) $\det A \neq 0$.

What if S contains $\mathbf{0}$?

Theorem

If a set $S = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$ of vectors in \mathbb{R}^m contains the **zero vector** $\mathbf{0}$, then the set is linearly dependent.

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Proof.

Without any loss of generality, suppose $\mathbf{v}_1 = \mathbf{0}$. Then we see the equation

$$x_1\mathbf{v}_1 + x_2\mathbf{v}_2 + \dots + x_n\mathbf{v}_n = \mathbf{0}$$

has a non-trivial solution. Namely, we set $x_1 = 1$ (or any non-zero value) and set $x_i = 0$ for each $i > 1$. This gives

$$1\mathbf{0} + 0\mathbf{v}_2 + \dots + 0\mathbf{v}_n = \mathbf{0} + \mathbf{0} + \dots + \mathbf{0} = \mathbf{0}.$$



Small Sets Vectors

1. If possible, give an example of a set of 1 vector in \mathbb{R}^3 so the set is linearly independent.
2. If possible, give an example of a set of 2 vectors in \mathbb{R}^3 so the set is linearly independent.
3. If possible, give an example of a set of 3 vectors in \mathbb{R}^3 so the set is linearly independent.
4. If possible, give an example of a set of 4 vectors in \mathbb{R}^3 so the set is linearly independent.

Number of Vectors in Set S and the Dimension of the Vectors in S

Theorem

Any set $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$ of vectors in \mathbb{R}^m is **linearly dependent** if $n > m$. That is, if a set contains more vectors than there are entries in each vector, then the set is linearly dependent.

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The columns of matrix A are linearly independent if and only if the matrix equation $A\mathbf{x} = \mathbf{0}$ has **only the trivial solution**.

Proof.

Let $A = [\mathbf{v}_1 \ \mathbf{v}_2 \ \dots \ \mathbf{v}_n]$ denote the $m \times n$ matrix. Then the equation $A\mathbf{x} = \mathbf{0}$ has m equations and n variables.

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Let $A = [\mathbf{v}_1 \ \mathbf{v}_2 \ \dots \ \mathbf{v}_n]$ denote the $m \times n$ matrix. Then the equation $A\mathbf{x} = \mathbf{0}$ has m equations and n variables. Since $n > m$, we have more variables than we have equations, so there must be at least one free variable. Thus $A\mathbf{x} = \mathbf{0}$ has a non-trivial solution, and the set $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$ is linearly dependent. □

Characterization of Linearly Dependent Sets

Theorem

An indexed set $S = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$ of two or more vectors is **linearly dependent** if and only if at least one of the vectors in S is a **linear combination of the others**. In fact, if S is linearly dependent and $\mathbf{v}_1 \neq \mathbf{0}$, then some \mathbf{v}_j (with $j > 1$) is a linear combination of the preceding vectors $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_{j-1}$.

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Extra statement: If S is linearly dependent, then by definition there exists weights c_1, c_2, \dots, c_n not all zero such that $c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + \dots + c_n\mathbf{v}_n = \mathbf{0}$. Suppose j is the largest index where $c_j \neq 0$.

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$$\mathbf{v}_j = -\frac{c_1}{c_j}\mathbf{v}_1 - \frac{c_2}{c_j}\mathbf{v}_2 - \dots - \frac{c_{j-1}}{c_j}\mathbf{v}_{j-1} - \frac{0}{c_j}\mathbf{v}_{j+1} - \dots - \frac{0}{c_j}\mathbf{v}_n.$$

Thus we see \mathbf{v}_j is a linear combination of the vectors $\mathbf{v}_1, \dots, \mathbf{v}_{j-1}$. □