

Linear independence

Consider the vectors $\begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \end{bmatrix}$ in \mathbb{R}^2 .

The span of these vectors is \mathbb{R}^2 , but notice 2 things:

1.) $\text{Span}\left\{\begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix}\right\} = \mathbb{R}^2$ as well.

2.) The vector $\begin{bmatrix} a \\ b \end{bmatrix}$ can be written in many different ways as a linear combination of these 3 vectors:

$$\begin{aligned} \begin{bmatrix} a \\ b \end{bmatrix} &= a \begin{bmatrix} 1 \\ 0 \end{bmatrix} + b \begin{bmatrix} 0 \\ 1 \end{bmatrix} = (a-b) \begin{bmatrix} 1 \\ 0 \end{bmatrix} + b \begin{bmatrix} 1 \\ 1 \end{bmatrix} \\ &= 1 \begin{bmatrix} 1 \\ 0 \end{bmatrix} + (1-a+b) \begin{bmatrix} 0 \\ 1 \end{bmatrix} + (a-1) \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \text{ etc.} \end{aligned}$$

Choosing a linearly independent set of vectors ensures that neither of these things can occur:

Def: A set of vectors $\{\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k\}$ is linearly independent if it satisfies the following:

$$\text{If } t_1 \vec{x}_1 + t_2 \vec{x}_2 + \dots + t_k \vec{x}_k = \vec{0}, \text{ then } t_1 = t_2 = \dots = t_k = 0.$$

Otherwise it is linearly dependent.

Theorem: If $\{\vec{x}_1, \dots, \vec{x}_k\}$ is linearly independent, then every vector in $\text{span}\{\vec{x}_1, \dots, \vec{x}_k\}$ has a unique representation

as a linear combination of the \vec{x}_i .

Why? If $\vec{v} = a_1 \vec{x}_1 + \dots + a_k \vec{x}_k$ and
 $\vec{v} = b_1 \vec{x}_1 + \dots + b_k \vec{x}_k$, then

$$a_1 \vec{x}_1 + \dots + a_k \vec{x}_k = b_1 \vec{x}_1 + \dots + b_k \vec{x}_k$$

$$\Rightarrow (a_1 - b_1) \vec{x}_1 + \dots + (a_k - b_k) \vec{x}_k = \vec{0}.$$

Thus, since they are linearly independent,

$a_i - b_i = 0 \Rightarrow a_i = b_i$ for each i , so there is only one way to write \vec{v} as a linear combination of $\vec{x}_1, \dots, \vec{x}_k$.

Ex: Is $\left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix} \right\}$ linearly independent?

Assume $a \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + b \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} + c \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix} = \vec{0}$.

We want to see if a, b, c must be 0. This becomes a system of equations:

$$\left[\begin{array}{ccc|c} 1 & 1 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 1 & 2 & 0 \end{array} \right] \rightarrow \left[\begin{array}{ccc|c} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{array} \right]$$

so $a = b = c = 0$ is the only solution. So they are linearly independent.

Ex: Are $\begin{bmatrix} 1 \\ 1 \end{bmatrix}$, $\begin{bmatrix} 1 \\ -1 \end{bmatrix}$, $\begin{bmatrix} 0 \\ 1 \end{bmatrix}$ linearly indep?

If $a \begin{bmatrix} 1 \\ 1 \end{bmatrix} + b \begin{bmatrix} 1 \\ -1 \end{bmatrix} + c \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \vec{0}$, then we have

$$\left[\begin{array}{ccc|c} 1 & 1 & 0 & 0 \\ 1 & -1 & 1 & 0 \end{array} \right]. \quad \text{Since this is homogeneous in}$$

3 variables, and there are just 2 equations, there must be infinitely many (and thus nontrivial) solutions. Thus, they are not linearly independent.

We summarize this strategy as follows:

Linear independence test:

To check if $\{\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k\}$ is linearly independent,

1.) Set up a homog. system of equations

$$t_1 \vec{x}_1 + t_2 \vec{x}_2 + \dots + t_k \vec{x}_k = \vec{0}, \text{ i.e.}$$

$$\left[\begin{array}{ccc|c} \vec{x}_1 & \vec{x}_2 & \dots & \vec{x}_k \\ \hline & & & \vec{0} \end{array} \right]$$

2.) If there is a nontrivial solution (i.e. any parameter), they are not lin. indep., otherwise they are.

Ex: The standard basis is linearly indep.:

$$[\vec{e}_1, \vec{e}_2, \dots, \vec{e}_k | \vec{0}] = [I | \vec{0}], \text{ which has only the}$$

trivial solution.

Ex: Is $\left\{ \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ -1 \\ 1 \end{bmatrix} \right\}$ linearly independent?

If

$$a \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} + b \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} + c \begin{bmatrix} 0 \\ -1 \\ 1 \end{bmatrix} = \vec{0}$$

$$\Rightarrow \left[\begin{array}{ccc|c} 1 & 1 & 0 & 0 \\ 1 & 0 & -1 & 0 \\ 0 & 1 & 1 & 0 \end{array} \right] \rightarrow \left[\begin{array}{ccc|c} 1 & 1 & 0 & 0 \\ 0 & -1 & -1 & 0 \\ 0 & 1 & 1 & 0 \end{array} \right] \rightarrow \left[\begin{array}{ccc|c} 1 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right]$$

Only 2 leading ones, so there are inf. many solutions, so it's not indep.

Ex: Can the set $\{\vec{0}, \vec{x}_1, \dots, \vec{x}_k\}$ be linearly independent?

No: $1(\vec{0}) + 0\vec{x}_1 + \dots + 0\vec{x}_k = \vec{0}$

↑
nonzero
coefficient

No set containing $\vec{0}$ is ever linearly independent.

Ex: If $\{\vec{x}, \vec{y}\}$ is linearly independent, is $\{\vec{x}+\vec{y}, \vec{x}-\vec{y}\}$?

If $a(\vec{x}+\vec{y}) + b(\vec{x}-\vec{y}) = \vec{0}$, then

$$(a+b)\vec{x} + (a-b)\vec{y} = \vec{0}, \text{ so}$$

$$a + b = 0, a - b = 0 \Rightarrow a = 0 = b, \text{ so}$$

$\{\vec{x} + \vec{y}, \vec{x} - \vec{y}\}$ is also lin. independent.

Ex: If $\{\vec{v}, \vec{w}\}$ is linearly dependent, and both \vec{v} and \vec{w} are nonzero, then

$$s\vec{v} + t\vec{w} = \vec{0} \text{ for some } s, t \neq 0.$$

$\Rightarrow s\vec{v} = -t\vec{w}$, so \vec{v} and \vec{w} must be parallel.

Conversely, if we assume \vec{v} and \vec{w} are parallel vectors then $\vec{v} = k\vec{w}$, some $k \neq 0$, so

$$\vec{v} - k\vec{w} = \vec{0}, \text{ so } \{\vec{v}, \vec{w}\} \text{ is linearly dependent.}$$

Thus, $\{\vec{v}, \vec{w}\}$ is lin. dependent if and only if \vec{v} and \vec{w} are parallel.

What about 3 vectors?

Ex: If $\{\vec{u}, \vec{v}, \vec{w}\}$ is dependent (in \mathbb{R}^3), then

$$a\vec{u} + b\vec{v} + c\vec{w} = \vec{0}, \text{ where } a, b, \text{ or } c \text{ is nonzero.}$$

Assume $a \neq 0$.

$$\text{Then } a\vec{u} = -b\vec{v} - c\vec{w} \Rightarrow \vec{u} = \frac{-b}{a}\vec{v} - \frac{c}{a}\vec{w}, \text{ so}$$

\vec{u} is in $\text{span}\{\vec{v}, \vec{w}\}$, which is a plane or a line.

Thus, all 3 vectors must lie in the same plane.

The converse holds as well. So $\{\vec{u}, \vec{v}, \vec{w}\}$ is lin. dependent if and only if $\vec{u}, \vec{v}, \vec{w}$ all lie in the same plane.
i.e. if $\text{span}\{\vec{u}, \vec{v}, \vec{w}\}$ is a plane or a line.

Matrices + linear independence

A an $m \times n$ matrix with columns $\vec{c}_1, \dots, \vec{c}_n$.

If $\vec{x} = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$ then

$$A \vec{x} = [\vec{c}_1 \dots \vec{c}_n] \vec{x} = x_1 \vec{c}_1 + x_2 \vec{c}_2 + \dots + x_n \vec{c}_n.$$

If we set $A \vec{x} = \vec{0}$, there are nontrivial solutions \Leftrightarrow
 $\vec{c}_1, \dots, \vec{c}_n$ are linearly dependent.

If instead we set $A \vec{x} = \vec{b}$, then there is a solution if and only if $\vec{b} = x_1 \vec{c}_1 + \dots + x_n \vec{c}_n$ for some x_1, \dots, x_n ,
i.e. \vec{b} is in $\text{span}\{\vec{c}_1, \dots, \vec{c}_n\}$.

We can summarize these in the following theorem:

Theorem: If $A = [\vec{c}_1 \dots \vec{c}_n]$ is an $m \times n$ matrix, then

- 1.) $\{\vec{c}_1, \vec{c}_2, \dots, \vec{c}_n\}$ is linearly independent if and only if $A\vec{x} = \vec{0}$ has no nontrivial solutions.
- 2.) $\mathbb{R}^m = \text{span}\{\vec{c}_1, \dots, \vec{c}_n\}$ if and only if $A\vec{x} = \vec{b}$ has a solution for every \vec{b} in \mathbb{R}^m .

For square matrices, we can relate these conditions to invertibility:

Theorem: If A is an $n \times n$ matrix, the following are equivalent:

- 1.) A is invertible.
- 2.) The columns of A are linearly independent.
- 3.) The columns of A span \mathbb{R}^n .
- 4.) The rows of A are linearly independent.
- 5.) The rows of A span \mathbb{R}^n .

Now we have a simpler way to check linear indep. if we have n vectors in \mathbb{R}^n :

Ex: Are $\begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}, \begin{bmatrix} 3 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} -1 \\ -1 \\ 1 \end{bmatrix}$ linearly indep.?

$$\text{Set } A = \begin{bmatrix} 1 & 3 & -1 \\ 2 & 0 & -1 \\ 0 & 1 & 1 \end{bmatrix}.$$

$$\begin{aligned} \det A &= -1 \begin{vmatrix} 1 & -1 \\ 2 & -1 \end{vmatrix} + 1 \begin{vmatrix} 1 & 3 \\ 2 & 0 \end{vmatrix} \\ &= -1(-1 + 2) + 1(0 - 6) \\ &= -1(1) + 1(-6) = -1 - 6 = -7 \end{aligned}$$

so A is invertible, so they are linearly independent, and they span \mathbb{R}^3 .

Dimension

Very important theorem: If U is a subspace of \mathbb{R}^n , and U is spanned by m vectors, then if U contains k linearly independent vectors, then $k \leq m$.

A maximum collection of linearly indep. vectors in a subspace is called a basis. More precisely:

Def: If U is a subspace of \mathbb{R}^n , a set of vectors $\{\vec{x}_1, \vec{x}_2, \dots, \vec{x}_m\}$ in U is a basis if:

- 1.) $\{\vec{x}_1, \vec{x}_2, \dots, \vec{x}_m\}$ is linearly independent, and
- 2.) $U = \text{span}\{\vec{x}_1, \vec{x}_2, \dots, \vec{x}_m\}$

It turns out that any two bases for a subspace will have the same # of vectors.

i.e. if $\{\vec{y}_1, \dots, \vec{y}_k\}$ is another basis for U , then $\text{span}\{\vec{y}_1, \dots, \vec{y}_k\} = U$ and $\{\vec{x}_1, \dots, \vec{x}_m\}$ is a set of linearly independent vectors in U , so $m \leq k$.

By the same argument, $k \leq m$. Thus $m = k$.

This number is called the dimension of U :

Def: If U is a subspace of \mathbb{R}^n and $\{\vec{x}_1, \dots, \vec{x}_m\}$ is a basis of U , then the dimension of U is m , denoted $\dim U = m$.

Ex: If $\{\vec{e}_1, \dots, \vec{e}_n\}$ is the standard basis in \mathbb{R}^n , then $\text{span}\{\vec{e}_1, \dots, \vec{e}_n\} = \mathbb{R}^n$ and it's linearly independent, so $\{\vec{e}_1, \dots, \vec{e}_n\}$ is indeed a basis, and $\dim \mathbb{R}^n = n$.

Ex: We saw above that any invertible $n \times n$ matrix has lin. indep rows (and cols) that span \mathbb{R}^n . Thus, the rows (or cols) of any invertible matrix forms a basis for \mathbb{R}^n .

Note: A key takeaway here is that there is more than one possible basis for each subspace, but the

dimension will always be the same.

Ex: Let $U = \left\{ \begin{bmatrix} s \\ s+t \\ s \end{bmatrix} \mid s, t \text{ in } \mathbb{R} \right\}$.

Is this a subspace of \mathbb{R}^3 ?

Notice that $\begin{bmatrix} s \\ s+t \\ s \end{bmatrix} = s \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} + t \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$,

so since s and t can be any real #s,

$U = \text{span} \left\{ \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \right\}$, so it is a subspace, since the

span of any set of vectors is a subspace.

What is $\dim U$? (We know $\dim U \leq 2$. Why? By the very important theorem. Any basis will be lin. indep. so the number of basis elts can be at most the # of spanning vectors.)

$\{\vec{u}, \vec{v}\}$ spans U so we need to check if it is lin. indep.

Since neither \vec{u} nor \vec{v} is a scalar multiple of the other, they are linearly independent. Thus $\{\vec{u}, \vec{v}\}$ is a basis, so $\dim U = 2$.

Ex: $\dim \{0\} = 0$. There is no linearly independent collection of vectors in $\{0\}$.

Theorem: Let $U \neq \{0\}$ be a subspace of \mathbb{R}^n . Then:

- 1.) U has a basis and $\dim U \leq n$.
- 2.) Any linearly independent set in U can be enlarged to a basis of U (by adding vectors from a fixed basis of U).
- 3.) Any spanning set for U can be cut down to a basis of U (by deleting vectors).

Ex: If $\vec{u} = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$, $\vec{v} = \begin{bmatrix} 2 \\ 1 \\ 0 \end{bmatrix}$, $\{\vec{u}, \vec{v}\}$ is linearly independent in \mathbb{R}^3 . Thus, we should be able to add a vector from the standard basis to $\{\vec{u}, \vec{v}\}$ to make it a basis.

Does \vec{e}_1 work? i.e. is $\{\vec{u}, \vec{v}, \vec{e}_1\}$ linearly independent and span \mathbb{R}^3 ? If

$$A = \begin{bmatrix} 1 & 2 & 1 \\ 1 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix}, \text{ then } \det A = -1, \text{ so}$$

A is invertible, so $\{\vec{u}, \vec{v}, \vec{e}_1\}$ is a basis.

In general, if $\dim U = m$ and we have m vectors which either span U or are lin. independent, then they form a basis. That is:

Theorem: If U is a subspace of \mathbb{R}^n where $\dim U = m$, let $B = \{\vec{x}_1, \dots, \vec{x}_m\}$, m vectors in U . Then

B is linearly independent if and only if $\text{span} B = U$.

Ex: $\left\{ \begin{bmatrix} 1 \\ 2 \end{bmatrix}, \begin{bmatrix} 1 \\ -1 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right\}$ spans \mathbb{R}^2 ,

and any pair forms a basis, since each pair is lin. indep. So, e.g. $\left\{ \begin{bmatrix} 1 \\ 2 \end{bmatrix}, \begin{bmatrix} 1 \\ -1 \end{bmatrix} \right\}$ is a basis for \mathbb{R}^2 .

If one subspace is inside of another, how do their dimensions compare?

Theorem: If U and V are subspaces of \mathbb{R}^n , and $U \subseteq V$, i.e. U is contained in V , then

1.) $\dim U \leq \dim V$.

2.) If $\dim U = \dim V$, then $U = V$.

Ex: If U is a subspace of \mathbb{R}^3 , then

$$\dim U = 3 \iff U = \mathbb{R}^3,$$

$\dim U = 2 \iff U$ is a plane through the origin,

$\dim U = 1 \iff U$ is a line through the origin.

Ex: (3a) let $U = \text{span} \left\{ \begin{bmatrix} 1 \\ -1 \\ 2 \\ 0 \end{bmatrix}, \begin{bmatrix} 2 \\ 3 \\ 0 \\ 3 \end{bmatrix}, \begin{bmatrix} 1 \\ 9 \\ -6 \\ 6 \end{bmatrix} \right\}$.

What is $\dim U$? We know that $\dim U \leq 3$, and some subset of these 3 vectors form a basis, so we first check if they are lin. independent:

$$a \begin{bmatrix} 1 \\ -1 \\ 2 \\ 0 \end{bmatrix} + b \begin{bmatrix} 2 \\ 3 \\ 0 \\ 3 \end{bmatrix} + c \begin{bmatrix} 1 \\ 9 \\ -6 \\ 6 \end{bmatrix} = \vec{0}$$

$$\Rightarrow \left[\begin{array}{ccc|c} 1 & 2 & 1 & 0 \\ -1 & 3 & 9 & 0 \\ 2 & 0 & -6 & 0 \\ 0 & 3 & 6 & 0 \end{array} \right] \rightarrow \left[\begin{array}{ccc|c} 1 & 2 & 1 & 0 \\ 0 & 5 & 10 & 0 \\ 0 & -4 & -8 & 0 \\ 0 & 3 & 6 & 0 \end{array} \right]$$

$$\rightarrow \left[\begin{array}{ccc|c} 1 & 2 & 1 & 6 \\ 0 & 1 & 2 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right] \Rightarrow \text{there's a parameter, so}$$

they are not lin. independent.

However, $\left\{ \begin{bmatrix} 1 \\ -1 \\ 2 \\ 0 \end{bmatrix}, \begin{bmatrix} 2 \\ 3 \\ 0 \\ 3 \end{bmatrix} \right\}$ is lin. independent,

so it forms a basis for U , so $\dim U = 2$.

Practice problems: 5.2 : 1, 2bd, 3bcd, 4cdf, 6adf