

The Vector Space

[3] The Vector Space

Linear Combinations

An expression

$$\alpha_1 \mathbf{v}_1 + \cdots + \alpha_n \mathbf{v}_n$$

is a *linear combination* of the vectors $\mathbf{v}_1, \dots, \mathbf{v}_n$.

The scalars $\alpha_1, \dots, \alpha_n$ are the *coefficients* of the linear combination.

Example: One linear combination of $[2, 3.5]$ and $[4, 10]$ is

$$-5 [2, 3.5] + 2 [4, 10]$$

which is equal to $[-5 \cdot 2, -5 \cdot 3.5] + [2 \cdot 4, 2 \cdot 10]$

Another linear combination of the same vectors is

$$0 [2, 3.5] + 0 [4, 10]$$

which is equal to the zero vector $[0, 0]$.

Definition: A linear combination is *trivial* if the coefficients are all zero.

Linear Combinations: JunkCo

The JunkCo factory makes five products:



using various resources.

	metal	concrete	plastic	water	electricity
garden gnome	0	1.3	0.2	0.8	0.4
hula hoop	0	0	1.5	0.4	0.3
slinky	0.25	0	0	0.2	0.7
silly putty	0	0	0.3	0.7	0.5
salad shooter	0.15	0	0.5	0.4	0.8

For each product, there is a vector specifying how much of each resource is used per unit of product.

For making one gnome:

$$\mathbf{v}_1 = \{\text{metal}:0, \text{concrete}:1.3, \text{plastic}:0.2, \text{water}:.8, \text{electricity}:0.4\}$$

Linear Combinations: JunkCo

For making one gnome:

$$\mathbf{v}_1 = \{\text{metal:0, concrete:1.3, plastic:0.2, water:0.8, electricity:0.4}\}$$

For making one hula hoop:

$$\mathbf{v}_2 = \{\text{metal:0, concrete:0, plastic:1.5, water:0.4, electricity:0.3}\}$$

For making one slinky:

$$\mathbf{v}_3 = \{\text{metal:0.25, concrete:0, plastic:0, water:0.2, electricity:0.7}\}$$

For making one silly putty:

$$\mathbf{v}_4 = \{\text{metal:0, concrete:0, plastic:0.3, water:0.7, electricity:0.5}\}$$

For making one salad shooter:

$$\mathbf{v}_5 = \{\text{metal:1.5, concrete:0, plastic:0.5, water:0.4, electricity:0.8}\}$$

Suppose the factory chooses to make α_1 gnomes, α_2 hula hoops, α_3 slinkies, α_4 silly putties, and α_5 salad shooters.

Total resource utilization is $\mathbf{b} = \alpha_1 \mathbf{v}_1 + \alpha_2 \mathbf{v}_2 + \alpha_3 \mathbf{v}_3 + \alpha_4 \mathbf{v}_4 + \alpha_5 \mathbf{v}_5$

Linear Combinations: JunkCo: Industrial espionage

Total resource utilization is $\mathbf{b} = \alpha_1 \mathbf{v}_1 + \alpha_2 \mathbf{v}_2 + \alpha_3 \mathbf{v}_3 + \alpha_4 \mathbf{v}_4 + \alpha_5 \mathbf{v}_5$

Suppose I am spying on JunkCo.

I find out how much metal, concrete, plastic, water, and electricity are consumed by the factory.

That is, I know the vector \mathbf{b} . Can I use this knowledge to figure out how many gnomes they are making?

Computational Problem: *Expressing a given vector as a linear combination of other given vectors*

- ▶ *input:* a vector \mathbf{b} and a list $[\mathbf{v}_1, \dots, \mathbf{v}_n]$ of vectors
- ▶ *output:* a list $[\alpha_1, \dots, \alpha_n]$ of coefficients such that

$$\mathbf{b} = \alpha_1 \mathbf{v}_1 + \dots + \alpha_n \mathbf{v}_n$$


or a report that none exists.

Question: Is the solution unique?

Lights Out

Button vectors for 2×2 Lights Out:



For a given initial state vector $\mathbf{s} =$ ,

Which subset of button vectors sum to \mathbf{s} ?

Reformulate in terms of linear combinations.

Write

$$\begin{bmatrix} \bullet \\ \bullet \end{bmatrix} = \alpha_1 \begin{bmatrix} \bullet & \bullet \\ \bullet & \end{bmatrix} + \alpha_2 \begin{bmatrix} \bullet & \bullet \\ \bullet & \bullet \end{bmatrix} + \alpha_3 \begin{bmatrix} \bullet & \\ \bullet & \bullet \end{bmatrix} + \alpha_4 \begin{bmatrix} \bullet \\ \bullet & \bullet \end{bmatrix}$$

What values for $\alpha_1, \alpha_2, \alpha_3, \alpha_4$ make this equation true?

Solution: $\alpha_1 = 0, \alpha_2 = 1, \alpha_3 = 0, \alpha_4 = 0$

Solve an instance of *Lights Out*

\Rightarrow

Which set of button vectors sum to \mathbf{s} ?

\Rightarrow

Find subset of $GF(2)$ vectors $\mathbf{v}_1, \dots, \mathbf{v}_n$ whose sum equals \mathbf{s}

\Rightarrow

Express \mathbf{s} as a linear combination of $\mathbf{v}_1, \dots, \mathbf{v}_n$

Lights Out

We can solve the puzzle if we have an algorithm for

Computational Problem: *Expressing a given vector as a linear combination of other given vectors*

Span

Definition: The set of all linear combinations of some vectors $\mathbf{v}_1, \dots, \mathbf{v}_n$ is called the *span* of these vectors

Written Span $\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$.

Span: Attacking the authentication scheme

If Eve knows the password satisfies

$$\begin{aligned}\mathbf{a}_1 \cdot \mathbf{x} &= \beta_1 \\ &\vdots \\ \mathbf{a}_m \cdot \mathbf{x} &= \beta_m\end{aligned}$$

Then she can calculate right response to any challenge in Span $\{\mathbf{a}_1, \dots, \mathbf{a}_m\}$:

Proof: Suppose $\mathbf{a} = \alpha_1 \mathbf{a}_1 + \dots + \alpha_m \mathbf{a}_m$. Then

$$\begin{aligned}\mathbf{a} \cdot \mathbf{x} &= (\alpha_1 \mathbf{a}_1 + \dots + \alpha_m \mathbf{a}_m) \cdot \mathbf{x} \\ &= \alpha_1 \mathbf{a}_1 \cdot \mathbf{x} + \dots + \alpha_m \mathbf{a}_m \cdot \mathbf{x} && \text{by distributivity} \\ &= \alpha_1 (\mathbf{a}_1 \cdot \mathbf{x}) + \dots + \alpha_m (\mathbf{a}_m \cdot \mathbf{x}) && \text{by homogeneity} \\ &= \alpha_1 \beta_1 + \dots + \alpha_m \beta_m\end{aligned}$$

Question: Any others? Answer will come later.

Span: $GF(2)$ vectors

Quiz: How many vectors are in $\text{Span} \{[1, 1], [0, 1]\}$ over the field $GF(2)$?

Answer: The linear combinations are

$$0 [1, 1] + 0 [0, 1] = [0, 0]$$

$$0 [1, 1] + 1 [0, 1] = [0, 1]$$

$$1 [1, 1] + 0 [0, 1] = [1, 1]$$

$$1 [1, 1] + 1 [0, 1] = [1, 0]$$

Thus there are four vectors in the span.

Span: $GF(2)$ vectors

Question: How many vectors in Span $\{[1, 1]\}$ over $GF(2)$?

Answer: The linear combinations are

$$0 [1, 1] = [0, 0]$$

$$1 [1, 1] = [1, 1]$$

Thus there are two vectors in the span.

Question: How many vectors in Span $\{\}$?

Answer: Only one: the zero vector

Question: How many vectors in Span $\{[2, 3]\}$ over \mathbb{R} ?

Answer: An infinite number: $\{\alpha [2, 3] : \alpha \in \mathbb{R}\}$

Forms the line through the origin and $(2, 3)$.

Generators

Definition: Let \mathcal{V} be a set of vectors. If $\mathbf{v}_1, \dots, \mathbf{v}_n$ are vectors such that $\mathcal{V} = \text{Span} \{ \mathbf{v}_1, \dots, \mathbf{v}_n \}$ then

- ▶ we say $\{ \mathbf{v}_1, \dots, \mathbf{v}_n \}$ is a *generating set* for \mathcal{V} ;
- ▶ we refer to the vectors $\mathbf{v}_1, \dots, \mathbf{v}_n$ as *generators* for \mathcal{V} .

Example: $\{ [3, 0, 0], [0, 2, 0], [0, 0, 1] \}$ is a generating set for \mathbb{R}^3 .

Proof: Must show two things:

1. Every linear combination is a vector in \mathbb{R}^3 .
2. Every vector in \mathbb{R}^3 is a linear combination.

First statement is easy: every linear combination of 3-vectors over \mathbb{R} is a 3-vector over \mathbb{R} , and \mathbb{R}^3 contains all 3-vectors over \mathbb{R} .

Proof of second statement: Let $[x, y, z]$ be any vector in \mathbb{R}^3 . I must show it is a linear combination of my three vectors....

$$[x, y, z] = (x/3) [3, 0, 0] + (y/2) [0, 2, 0] + z [0, 0, 1]$$

Generators

Claim: Another generating set for \mathbb{R}^3 is $\{[1, 0, 0], [1, 1, 0], [1, 1, 1]\}$

Another way to prove that every vector in \mathbb{R}^3 is in the span:

- ▶ We already know $\mathbb{R}^3 = \text{Span} \{[3, 0, 0], [0, 2, 0], [0, 0, 1]\}$,
- ▶ so just show $[3, 0, 0]$, $[0, 2, 0]$, and $[0, 0, 1]$ are in $\text{Span} \{[1, 0, 0], [1, 1, 0], [1, 1, 1]\}$

$$[3, 0, 0] = 3[1, 0, 0]$$

$$[0, 2, 0] = -2[1, 0, 0] + 2[1, 1, 0]$$

$$[0, 0, 1] = \cancel{-1[1, 0, 0]} - 1[1, 1, 0] + 1[1, 1, 1]$$

Why is that sufficient?

- ▶ We already know any vector in \mathbb{R}^3 can be written as a linear combination of the old vectors.
- ▶ We know each old vector can be written as a linear combination of the new vectors.
- ▶ We can convert *a linear combination of linear combination of new vectors* into *a linear combination of new vectors*.

Generators

We can convert a linear combination of linear combination of new vectors into a linear combination of new vectors.

- ▶ Write $[x, y, z]$ as a linear combination of the old vectors:

$$[x, y, z] = (x/3) [3, 0, 0] + (y/2) [0, 2, 0] + z [0, 0, 1]$$

- ▶ Replace each old vector with an equivalent linear combination of the new vectors:

$$[x, y, z] = (x/3) \left(3 [1, 0, 0] \right) + (y/2) \left(-2 [1, 0, 0] + 2 [1, 1, 0] \right) + z \left(\cancel{-1 [1, 0, 0]} - 1 [1, 1, 0] + 1 [1, 1, 1] \right)$$

- ▶ Multiply through, using distributivity and associativity:

$$[x, y, z] = x [1, 0, 0] - y [1, 0, 0] + y [1, 1, 0] \cancel{- z [1, 0, 0]} - z [1, 1, 0] + z [1, 1, 1]$$

- ▶ Collect like terms, using distributivity:

$$[x, y, z] = (x - y \cancel{- z}) [1, 0, 0] + (y - z) [1, 1, 0] + z [1, 1, 1]$$

Generators

Question: How to write each of the old vectors $[3, 0, 0]$, $[0, 2, 0]$, and $[0, 0, 1]$ as a linear combination of new vectors $[2, 0, 1]$, $[1, 0, 2]$, $[2, 2, 2]$, and $[0, 1, 0]$?

Answer:

$$[3, 0, 0] = 2 [2, 0, 1] - 1 [1, 0, 2] + 0 [2, 2, 2]$$

$$[0, 2, 0] = -\frac{2}{3} [2, 0, 1] - \frac{2}{3} [1, 0, 2] + 1 [2, 2, 2]$$

$$[0, 0, 1] = -\frac{1}{3} [2, 0, 1] + \frac{2}{3} [1, 0, 2] + 0 [2, 2, 2]$$

Standard generators

Writing $[x, y, z]$ as a linear combination of the vectors $[3, 0, 0]$, $[0, 2, 0]$, and $[0, 0, 1]$ is simple.

$$[x, y, z] = (x/3) [3, 0, 0] + (y/2) [0, 2, 0] + z [0, 0, 1]$$

Even simpler if instead we use $[1, 0, 0]$, $[0, 1, 0]$, and $[0, 0, 1]$:

$$[x, y, z] = x [1, 0, 0] + y [0, 1, 0] + z [0, 0, 1]$$

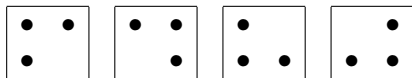
These are called *standard generators* for \mathbb{R}^3 .

Written $\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3$

Standard generators

Question: Can 2×2 *Lights Out* be solved from every starting configuration?

Equivalent to asking whether the 2×2 button vectors



are generators for $GF(2)^D$, where $D = \{(0,0), (0,1), (1,0), (1,1)\}$.

Yes! For proof, we show that each standard generator can be written as a linear combination of the button vectors:

$$\begin{array}{l} \begin{array}{|c|} \hline \bullet \\ \hline \end{array} = 1 \begin{array}{|c|c|} \hline \bullet & \bullet \\ \hline \end{array} + 1 \begin{array}{|c|c|} \hline \bullet & \bullet \\ \hline \end{array} + 1 \begin{array}{|c|c|} \hline \bullet & \bullet \\ \hline \end{array} + 0 \begin{array}{|c|c|} \hline \bullet & \bullet \\ \hline \end{array} \\ \begin{array}{|c|} \hline \bullet \\ \hline \end{array} = 1 \begin{array}{|c|c|} \hline \bullet & \bullet \\ \hline \end{array} + 1 \begin{array}{|c|c|} \hline \bullet & \bullet \\ \hline \end{array} + 0 \begin{array}{|c|c|} \hline \bullet & \bullet \\ \hline \end{array} + 1 \begin{array}{|c|c|} \hline \bullet & \bullet \\ \hline \end{array} \\ \begin{array}{|c|} \hline \bullet \\ \hline \end{array} = 1 \begin{array}{|c|c|} \hline \bullet & \bullet \\ \hline \end{array} + 0 \begin{array}{|c|c|} \hline \bullet & \bullet \\ \hline \end{array} + 1 \begin{array}{|c|c|} \hline \bullet & \bullet \\ \hline \end{array} + 1 \begin{array}{|c|c|} \hline \bullet & \bullet \\ \hline \end{array} \\ \begin{array}{|c|} \hline \bullet \\ \hline \end{array} = 0 \begin{array}{|c|c|} \hline \bullet & \bullet \\ \hline \end{array} + 1 \begin{array}{|c|c|} \hline \bullet & \bullet \\ \hline \end{array} + 1 \begin{array}{|c|c|} \hline \bullet & \bullet \\ \hline \end{array} + 1 \begin{array}{|c|c|} \hline \bullet & \bullet \\ \hline \end{array} \end{array}$$

Geometry of sets of vectors: span of vectors over \mathbb{R}

Span of a single nonzero vector \mathbf{v} :

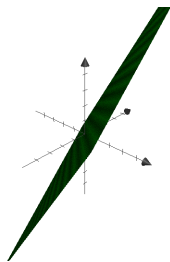
$$\text{Span } \{\mathbf{v}\} = \{\alpha \mathbf{v} : \alpha \in \mathbb{R}\}$$

This is the line through the origin and \mathbf{v} . *One-dimensional*

Span of the empty set: just the origin. *Zero-dimensional*

Span $\{[1, 2], [3, 4]\}$: all points in the plane. *Two-dimensional*

Span of two 3-vectors? Span $\{[1, 0, 1.65], [0, 1, 1]\}$ is a plane in three dimensions:



Two-dimensional

Geometry of sets of vectors: span of vectors over \mathbb{R}

Is the span of k vectors always k -dimensional?

No.

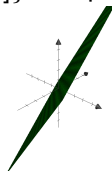
- ▶ Span $\{[0, 0]\}$ is 0-dimensional.
- ▶ Span $\{[1, 3], [2, 6]\}$ is 1-dimensional.
- ▶ Span $\{[1, 0, 0], [0, 1, 0], [1, 1, 0]\}$ is 2-dimensional.

Fundamental Question: How can we predict the dimensionality of the span of some vectors?

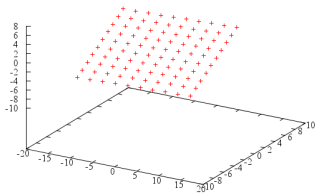
Geometry of sets of vectors: span of vectors over \mathbb{R}

Span of two 3-vectors? Span $\{[1, 0, 1.65], [0, 1, 1]\}$ is a plane in three dimensions:

Two-dimensional



Useful for plotting the plane

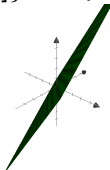


$$\begin{aligned} & \{\alpha [1, 0, 1.65] + \beta [0, 1, 1] : \\ & \alpha \in \{-5, -4, \dots, 3, 4\}, \\ & \beta \in \{-5, -4, \dots, 3, 4\}\} \end{aligned}$$

Geometry of sets of vectors: span of vectors over \mathbb{R}

Span of two 3-vectors? Span $\{[1, 0, 1.65], [0, 1, 1]\}$ is a plane in three dimensions:

Two-dimensional



Perhaps a more familiar way to specify a plane:

$$\{(x, y, z) : ax + by + cz = 0\}$$

Using dot-product, we could rewrite as

$$\{[x, y, z] : [a, b, c] \cdot [x, y, z] = 0\}$$

Set of vectors satisfying a linear equation with right-hand side *zero*.

We can similarly specify a line in three dimensions:

$$\{[x, y, z] : \mathbf{a}_1 \cdot [x, y, z] = 0, \mathbf{a}_2 \cdot [x, y, z] = 0\}$$

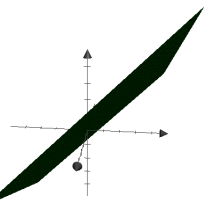
Two ways to represent a geometric object (line, plane, etc.) containing the origin:

- ▶ Span of some vectors
- ▶ Solution set of some system of linear equations with zero right-hand sides

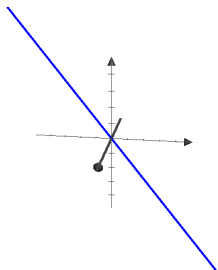
Geometry of sets of vectors: Two representations

Two ways to represent a geometric object (line, plane, etc.) containing the origin:

- ▶ Span of some vectors
- ▶ Solution set of some system of linear equations with zero right-hand sides



$$\text{Span} \{[4, -1, 1], [0, 1, 1]\} \quad \{[x, y, z] : [1, 2, -2] \cdot [x, y, z] = 0\}$$



$$\text{Span} \{[1, 2, -2]\} \quad \{[x, y, z] : \begin{aligned} &[4, -1, 1] \cdot [x, y, z] = 0, \\ &[0, 1, 1] \cdot [x, y, z] = 0 \end{aligned}\}$$

Geometry of sets of vectors: Two representations

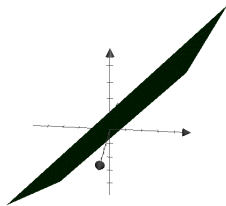
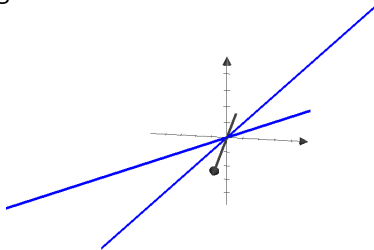
Two ways to represent a geometric object (line, plane, etc.) containing the origin:

- ▶ Span of some vectors
- ▶ Solution set of some system of linear equations with zero right-hand sides

Each representation has its uses.

Suppose you want to find the plane containing two given lines

- ▶ First line is $\text{Span} \{[4, -1, 1]\}$.
 - ▶ Second line is $\text{Span} \{[0, 1, 1]\}$.
-
- ▶ The plane containing these two lines is $\text{Span} \{[4, -1, 1], [0, 1, 1]\}$



Geometry of sets of vectors: Two representations

Two ways to represent a geometric object (line, plane, etc.) containing the origin:

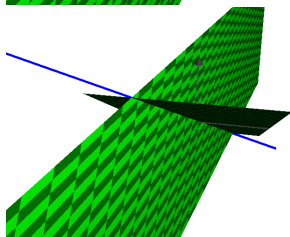
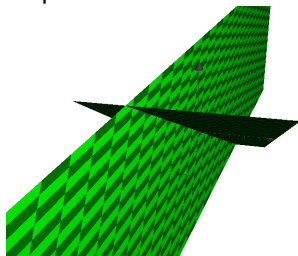
- ▶ Span of some vectors
- ▶ Solution set of some system of linear equations with zero right-hand sides

Each representation has its uses.

Suppose you want to find the intersection of two given planes:

- ▶ First plane is
 $\{[x, y, z] : [4, -1, 1] \cdot [x, y, z] = 0\}.$
- ▶ Second plane is
 $\{[x, y, z] : [0, 1, 1] \cdot [x, y, z] = 0\}.$

- ▶ The intersection is $\{[x, y, z] :$
 $[4, -1, 1] \cdot [x, y, z] = 0, [0, 1, 1] \cdot [x, y, z] = 0\}$



Two representations: What's common?

Subset of \mathbb{F}^D that satisfies three properties:

Property V1 Subset contains the zero vector $\mathbf{0}$

Property V2 If subset contains \mathbf{v} then it contains $\alpha \mathbf{v}$ for every scalar α

Property V3 If subset contains \mathbf{u} and \mathbf{v} then it contains $\mathbf{u} + \mathbf{v}$

Span $\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$ satisfies

- ▶ Property V1 because

$$0 \mathbf{v}_1 + \dots + 0 \mathbf{v}_n$$

- ▶ Property V2 because

$$\text{if } \mathbf{v} = \beta_1 \mathbf{v}_1 + \dots + \beta_n \mathbf{v}_n \text{ then } \alpha \mathbf{v} = \alpha \beta_1 \mathbf{v}_1 + \dots + \alpha \beta_n \mathbf{v}_n$$

- ▶ Property V3 because

$$\begin{aligned} &\text{if } \mathbf{u} = \alpha_1 \mathbf{v}_1 + \dots + \alpha_n \mathbf{v}_n \\ &\text{and } \mathbf{v} = \beta_1 \mathbf{v}_1 + \dots + \beta_n \mathbf{v}_n \\ &\text{then } \mathbf{u} + \mathbf{v} = (\alpha_1 + \beta_1) \mathbf{v}_1 + \dots + (\alpha_n + \beta_n) \mathbf{v}_n \end{aligned}$$

Abstract vector spaces

In traditional, abstract approach to linear algebra:

- ▶ We don't define vectors as sequences $[1,2,3]$ or even functions $\{a:1, b:2, c:3\}$.
- ▶ We define a vector space over a field \mathbb{F} to be any set \mathcal{V} that is equipped with
 - ▶ an *addition* operation, and
 - ▶ a *scalar-multiplication* operation

satisfying certain axioms (e.g. commutative and distributive laws) and Properties V1, V2, V3.

Abstract approach has the advantage that it avoids committing to specific structure for vectors.

I avoid abstract approach in this class because more concrete notion of vectors is helpful in developing intuition.

Geometric objects that exclude the origin

How to represent a line that does *not* contain the origin?

Start with a line that *does* contain the origin.

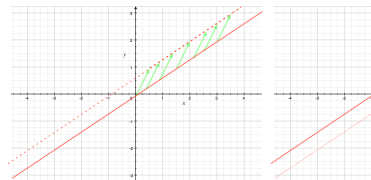
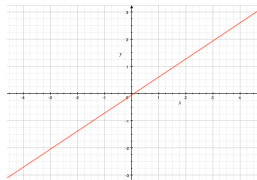
We know that points of such a line form a vector space \mathcal{V} .

Translate the line by adding a vector \mathbf{c} to every vector in \mathcal{V} :

$$\{\mathbf{c} + \mathbf{v} : \mathbf{v} \in \mathcal{V}\}$$

(abbreviated $\mathbf{c} + \mathcal{V}$)

Result is line through \mathbf{c} instead of through origin.



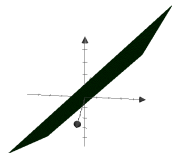
Geometric objects that exclude the origin

How to represent a plane that does *not* contain the origin?



Start with a plane that *does* contain the origin.

We know that points of such a plane form a vector space \mathcal{V} .

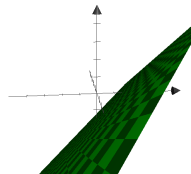


Translate it by adding a vector \mathbf{c} to every vector in \mathcal{V}

$$\{\mathbf{c} + \mathbf{v} : \mathbf{v} \in \mathcal{V}\}$$

(abbreviated $\mathbf{c} + \mathcal{V}$)

▶ Result is plane containing \mathbf{c} .



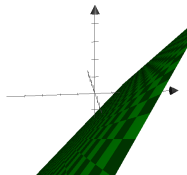
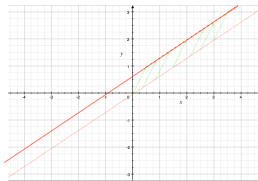
Affine space

Definition: If \mathbf{c} is a vector and \mathcal{V} is a vector space then

$$\mathbf{c} + \mathcal{V}$$

is called an *affine space*.

Examples: A plane or a line not necessarily containing the origin.



Affine space and affine combination

Example: The plane containing $\mathbf{u}_1 = [3, 0, 0]$, $\mathbf{u}_2 = [-3, 1, -1]$, and $\mathbf{u}_3 = [1, -1, 1]$.

Want to express this plane as $\mathbf{u}_1 + \mathcal{V}$
where \mathcal{V} is the span of two vectors
(a plane containing the origin)

Let $\mathcal{V} = \text{Span} \{\mathbf{a}, \mathbf{b}\}$ where

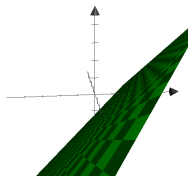
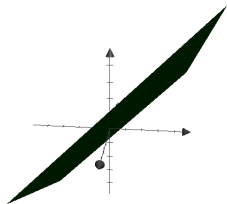
$$\mathbf{a} = \mathbf{u}_2 - \mathbf{u}_1 \text{ and } \mathbf{b} = \mathbf{u}_3 - \mathbf{u}_1$$

Since $\mathbf{u}_1 + \mathcal{V}$ is a translation of a plane, it is also a plane.

- ▶ $\text{Span} \{\mathbf{a}, \mathbf{b}\}$ contains $\mathbf{0}$, so $\mathbf{u}_1 + \text{Span} \{\mathbf{a}, \mathbf{b}\}$ contains \mathbf{u}_1 .
- ▶ $\text{Span} \{\mathbf{a}, \mathbf{b}\}$ contains $\mathbf{u}_2 - \mathbf{u}_1$ so $\mathbf{u}_1 + \text{Span} \{\mathbf{a}, \mathbf{b}\}$ contains \mathbf{u}_2 .
- ▶ $\text{Span} \{\mathbf{a}, \mathbf{b}\}$ contains $\mathbf{u}_3 - \mathbf{u}_1$ so $\mathbf{u}_1 + \text{Span} \{\mathbf{a}, \mathbf{b}\}$ contains \mathbf{u}_3 .

Thus the plane $\mathbf{u}_1 + \text{Span} \{\mathbf{a}, \mathbf{b}\}$ contains $\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3$.

Only one plane contains those three points, so this is that one.



Affine space and affine combination

Example: The plane containing $\mathbf{u}_1 = [3, 0, 0]$, $\mathbf{u}_2 = [-3, 1, -1]$, and $\mathbf{u}_3 = [1, -1, 1]$:

$$\mathbf{u}_1 + \text{Span} \{ \mathbf{u}_2 - \mathbf{u}_1, \mathbf{u}_3 - \mathbf{u}_1 \}$$

Cleaner way to write it?

$$\begin{aligned} \mathbf{u}_1 + \text{Span} \{ \mathbf{u}_2 - \mathbf{u}_1, \mathbf{u}_3 - \mathbf{u}_1 \} &= \{ \mathbf{u}_1 + \alpha (\mathbf{u}_2 - \mathbf{u}_1) + \beta (\mathbf{u}_3 - \mathbf{u}_1) : \alpha, \beta \in \mathbb{R} \} \\ &= \{ \mathbf{u}_1 + \alpha \mathbf{u}_2 - \alpha \mathbf{u}_1 + \beta \mathbf{u}_3 - \beta \mathbf{u}_1 : \alpha, \beta \in \mathbb{R} \} \\ &= \{ (1 - \alpha - \beta) \mathbf{u}_1 + \alpha \mathbf{u}_2 + \beta \mathbf{u}_3 : \alpha, \beta \in \mathbb{R} \} \\ &= \{ \gamma \mathbf{u}_1 + \alpha \mathbf{u}_2 + \beta \mathbf{u}_3 : \gamma + \alpha + \beta = 1 \} \end{aligned}$$

Definition: A linear combination $\gamma \mathbf{u}_1 + \alpha \mathbf{u}_2 + \beta \mathbf{u}_3$ where $\gamma + \alpha + \beta = 1$ is an *affine combination*.

Affine combination

Definition: A linear combination

$$\alpha_1 \mathbf{u}_1 + \alpha_2 \mathbf{u}_2 + \cdots + \alpha_n \mathbf{u}_n$$

where

$$\alpha_1 + \alpha_2 + \cdots + \alpha_n = 1$$

is an *affine combination*.

Definition: The set of all affine combinations of vectors $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n$ is called the *affine hull* of those vectors.

$$\text{Affine hull of } \mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n = \mathbf{u}_1 + \text{Span} \{ \mathbf{u}_2 - \mathbf{u}_1, \dots, \mathbf{u}_n - \mathbf{u}_1 \}$$

This shows that the affine hull of some vectors is an affine space..

Geometric objects not containing the origin: equations

Can express a plane as $\mathbf{u}_1 + \mathcal{V}$ or affine hull of $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n$.

More familiar way to express a plane:

The solution set of an equation $ax + by + cz = d$

In vector terms,

$$\{[x, y, z] : [a, b, c] \cdot [x, y, z] = d\}$$

In general, a geometric object (point, line, plane, ...) can be expressed as the solution set of a system of linear equations.

$$\{\mathbf{x} : \mathbf{a}_1 \cdot \mathbf{x} = \beta_1, \dots, \mathbf{a}_m \cdot \mathbf{x} = \beta_m\}$$

Conversely, is the solution set an affine space?

Consider solution set of a contradictory system of equations, e.g. $1x = 1, 2x = 1$:

- ▶ Solution set is empty....
- ▶ ...but a vector space \mathcal{V} always contains the zero vector,
- ▶ ...so an affine space $\mathbf{u}_1 + \mathcal{V}$ always contains at least one vector.

Turns out this the only exception:

Theorem: The solution set of a linear system is either empty or an affine space.

Affine spaces and linear systems

Theorem: The solution set of a linear system is either empty or an affine space.

Each linear system corresponds to a linear system with zero right-hand sides:

$$\begin{array}{rcl} \mathbf{a}_1 \cdot \mathbf{x} & = & \beta_1 \\ & \vdots & \\ \mathbf{a}_m \cdot \mathbf{x} & = & \beta_m \end{array} \quad \Longrightarrow \quad \begin{array}{rcl} \mathbf{a}_1 \cdot \mathbf{x} & = & 0 \\ & \vdots & \\ \mathbf{a}_m \cdot \mathbf{x} & = & 0 \end{array}$$

Definition:

A linear equation $\mathbf{a} \cdot \mathbf{x} = 0$ with zero right-hand side is a *homogeneous* linear equation. A system of homogeneous linear equations is called a *homogeneous* linear system.

We already know: The solution set of a homogeneous linear system is a vector space.

Lemma: Let \mathbf{u}_1 be a solution to a linear system. Then, for any other vector \mathbf{u}_2 ,
 \mathbf{u}_2 is also a solution
if and only if
 $\mathbf{u}_2 - \mathbf{u}_1$ is a solution to the corresponding homogeneous linear system.

Affine spaces and linear systems

$$\begin{array}{ccc} \mathbf{a}_1 \cdot \mathbf{x} = \beta_1 & & \mathbf{a}_1 \cdot \mathbf{x} = 0 \\ \vdots & \implies & \vdots \\ \mathbf{a}_m \cdot \mathbf{x} = \beta_m & & \mathbf{a}_m \cdot \mathbf{x} = 0 \end{array}$$

Lemma: Let \mathbf{u}_1 be a solution to a linear system. Then, for any other vector \mathbf{u}_2 , \mathbf{u}_2 is also a solution if and only if $\mathbf{u}_2 - \mathbf{u}_1$ is a solution to the corresponding homogeneous linear system.

Proof: We assume $\mathbf{a}_1 \cdot \mathbf{u}_1 = \beta_1, \dots, \mathbf{a}_m \cdot \mathbf{u}_1 = \beta_m$, so

$$\begin{array}{ccccc} \mathbf{a}_1 \cdot \mathbf{u}_2 = \beta_1 & & \mathbf{a}_1 \cdot \mathbf{u}_2 - \mathbf{a}_1 \cdot \mathbf{u}_1 = 0 & & \mathbf{a}_1 \cdot (\mathbf{u}_2 - \mathbf{u}_1) = 0 \\ \vdots & \text{iff} & \vdots & \text{iff} & \vdots \\ \mathbf{a}_m \cdot \mathbf{u}_2 = \beta_m & & \mathbf{a}_m \cdot \mathbf{u}_2 - \mathbf{a}_m \cdot \mathbf{u}_1 = 0 & & \mathbf{a}_m \cdot (\mathbf{u}_2 - \mathbf{u}_1) = 0 \end{array}$$

QED

Lemma: Let \mathbf{u}_1 be a solution to a linear system. Then, for any other vector \mathbf{u}_2 ,
 \mathbf{u}_2 is also a solution
if and only if
 $\mathbf{u}_2 - \mathbf{u}_1$ is a solution to the corresponding homogeneous linear system.

We use this lemma to prove the theorem:

Theorem: The solution set of a linear system is either empty or an affine space.

- ▶ Let \mathcal{V} = set of solutions to corresponding homogeneous linear system.
- ▶ If the linear system has no solution, its solution set is empty.
- ▶ If it does have a solution \mathbf{u}_1 then

$$\begin{aligned} \{\text{solutions to linear system}\} &= \{\mathbf{u}_2 : \mathbf{u}_2 - \mathbf{u}_1 \in \mathcal{V}\} \\ &\quad (\text{substitute } \mathbf{v} = \mathbf{u}_2 - \mathbf{u}_1) \\ &= \{\mathbf{u}_1 + \mathbf{v} : \mathbf{v} \in \mathcal{V}\} \end{aligned}$$

QED

Number of solutions to a linear system

We just proved:

If \mathbf{u}_1 is a solution to a linear system then

$$\{\text{solutions to linear system}\} = \{\mathbf{u}_1 + \mathbf{v} : \mathbf{v} \in \mathcal{V}\}$$

where $\mathcal{V} = \{\text{solutions to corresponding homogeneous linear system}\}$

Implications:

Long ago we asked: *How can we tell if a linear system has only one solution?*

Now we know: If a linear system has a solution \mathbf{u}_1 then that solution is unique if the only solution to the corresponding homogeneous linear system is $\mathbf{0}$.

Long ago we asked: How can we find the number of solutions to a linear system over $GF(2)$?

Now we know: Number of solutions either is zero or is equal to the number of solutions to the corresponding *homogeneous* linear system.

Number of solutions: checksum function

MD5 checksums and sizes of the released files:

3c63a6d97333f4da35976b6a0755eb67	12732276	Python-3.2.2.tgz
9d763097a13a59ff53428c9e4d098a05	10743647	Python-3.2.2.tar.bz2
3720ce9460597e49264bbb63b48b946d	8923224	Python-3.2.2.tar.xz
f6001a9b2be57ecfbefa865e50698cdf	19519332	python-3.2.2-macosx10.3.dmg
8fe82d14dbb2e96a84fd6fa1985b6f73	16226426	python-3.2.2-macosx10.6.dmg
cccb03e14146f7ef82907cf12bf5883c	18241506	python-3.2.2-pdb.zip
72d11475c986182bcb0e5c91acec45bc	19940424	python-3.2.2.amd64-pdb.zip
ddeb3e3fb93ab5a900adb6f04edab21e	18542592	python-3.2.2.amd64.msi
8afb1b01e8fab738e7b234eb4fe3955c	18034688	python-3.2.2.msi

A *checksum function* maps long files to short sequences.

Idea:

- ▶ Web page shows the checksum of each file to be downloaded.
- ▶ Download the file and run the checksum function on it.
- ▶ If result does not match checksum on web page, you know the file has been corrupted.
- ▶ If random corruption occurs, how likely are you to detect it?

Impractical but instructive checksum function:

- ▶ *input*: an n -vector \mathbf{x} over $GF(2)$
- ▶ *output*: $[\mathbf{a}_1 \cdot \mathbf{x}, \mathbf{a}_2 \cdot \mathbf{x}, \dots, \mathbf{a}_{64} \cdot \mathbf{x}]$

where $\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_{64}$ are sixty-four n -vectors.

Number of solutions: checksum function

Our checksum function:

- ▶ *input*: an n -vector \mathbf{x} over $GF(2)$
- ▶ *output*: $[\mathbf{a}_1 \cdot \mathbf{x}, \mathbf{a}_2 \cdot \mathbf{x}, \dots, \mathbf{a}_{64} \cdot \mathbf{x}]$

where $\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_{64}$ are sixty-four n -vectors.

Suppose \mathbf{p} is the original file, and it is randomly corrupted during download.

What is the probability that the corruption is undetected?

The checksum of the original file is $[\beta_1, \dots, \beta_{64}] = [\mathbf{a}_1 \cdot \mathbf{p}, \dots, \mathbf{a}_{64} \cdot \mathbf{p}]$.

Suppose corrupted version is $\mathbf{p} + \mathbf{e}$.

Then checksum of corrupted file matches checksum of original if and only if

$$\begin{array}{lll} \mathbf{a}_1 \cdot (\mathbf{p} + \mathbf{e}) = \beta_1 & \text{iff} & \mathbf{a}_1 \cdot \mathbf{p} - \mathbf{a}_1 \cdot (\mathbf{p} + \mathbf{e}) = 0 & \text{iff} & \mathbf{a}_1 \cdot \mathbf{e} = 0 \\ \vdots & & \vdots & & \vdots \\ \mathbf{a}_{64} \cdot (\mathbf{p} + \mathbf{e}) = \beta_{64} & & \mathbf{a}_{64} \cdot \mathbf{p} - \mathbf{a}_{64} \cdot (\mathbf{p} + \mathbf{e}) = 0 & & \mathbf{a}_{64} \cdot \mathbf{e} = 0 \end{array}$$

iff \mathbf{e} is a solution to the homogeneous linear system $\mathbf{a}_1 \cdot \mathbf{x} = 0, \dots, \mathbf{a}_{64} \cdot \mathbf{x} = 0$.

Number of solutions: checksum function

Suppose corrupted version is $\mathbf{p} + \mathbf{e}$.

Then checksum of corrupted file matches checksum of original if and only if \mathbf{e} is a solution to homogeneous linear system

$$\begin{aligned} \mathbf{a}_1 \cdot \mathbf{x} &= 0 \\ &\vdots \\ \mathbf{a}_{64} \cdot \mathbf{x} &= 0 \end{aligned}$$

If \mathbf{e} is chosen according to the uniform distribution,

$$\begin{aligned} &\text{Probability } (\mathbf{p} + \mathbf{e} \text{ has same checksum as } \mathbf{p}) \\ &= \text{Probability } (\mathbf{e} \text{ is a solution to homogeneous linear system}) \\ &= \frac{\text{number of solutions to homogeneous linear system}}{\text{number of } n\text{-vectors}} \\ &= \frac{\text{number of solutions to homogeneous linear system}}{2^n} \end{aligned}$$

Question:

How to find out number of solutions to a homogeneous linear system over $GF(2)$?

Geometry of sets of vectors: convex hull

Earlier, we saw: The \mathbf{u} -to- \mathbf{v} line segment is

$$\{\alpha \mathbf{u} + \beta \mathbf{v} : \alpha \in \mathbb{R}, \beta \in \mathbb{R}, \alpha \geq 0, \beta \geq 0, \alpha + \beta = 1\}$$

Definition: For vectors $\mathbf{v}_1, \dots, \mathbf{v}_n$ over \mathbb{R} , a linear combination

$$\alpha_1 \mathbf{v}_1 + \dots + \alpha_n \mathbf{v}_n$$

is a *convex combination* if the coefficients are all nonnegative and they sum to 1.

- ▶ Convex hull of a single vector is a point.
- ▶ Convex hull of two vectors is a line segment.
- ▶ Convex hull of three vectors is a triangle

Convex hull of more vectors? Could be higher-dimensional... but not necessarily.

For example, a convex polygon is the convex hull of its vertices

2-Dimensional Convex Hull of 3-Vectors over \mathbb{R}

