18.06.02: 'Vectors'

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Wednesday 05 February 2016



Lines vs. vectors

A vector is a list of real numbers

$$\vec{a} = \begin{pmatrix} a_1 \\ a_2 \\ \vdots \\ a_n \end{pmatrix}$$

We draw this vector as an arrow pointing from the point (0, 0, ..., 0) to the point $(a_1, a_2, ..., a_n)$:

$$(a_1, a_2, \dots, a_n)$$
 (0, 0, ..., 0)



By extending this vector, we obtain a line:

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This is the unique line between the origin and (a_1, a_2, \dots, a_n) .

Question. How can we write a formula for a parametrization of this line?



$$\lambda(t) = (ta_1, ta_2, \dots, ta_n).$$

Note that there's one exception to this: the zero vector

$$\vec{0} := \left(\begin{array}{c} 0 \\ 0 \\ \vdots \\ 0 \end{array} \right)$$

doesn't give us much of a line. This is the *only* special vector. We'll return to this.



Two vectors may determine the same line:

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Question. When does this happen?



If there's a real number r such that

$$\begin{pmatrix} a_1 \\ a_2 \\ \vdots \\ a_n \end{pmatrix} = \begin{pmatrix} rb_1 \\ rb_2 \\ \vdots \\ rb_n \end{pmatrix},$$

then we say that \vec{a} is a *scalar multiple* of \vec{b} , and we just write $\vec{a} = r\vec{b}$. In this case, as long as \vec{a} and \vec{b} are nonzero, they determine the same line.

So lines through the origin are the "same thing" as nonzero vectors, up to scaling.



${\bf R}^1$ and ${\bf R}^2$

We write \mathbf{R}^n for the collection of all vectors

$$\left(\begin{array}{c}
a_1\\
a_2\\
\vdots\\
a_n
\end{array}\right)$$

Thus \mathbf{R}^1 is simply the set of real numbers, but we regard those real numbers as arrows from the origin to the number as it sits on the number line:



Note that if $a \in \mathbb{R}^1$ is a nonzero real number, then any real number is a scalar mutiple of *a*. In other words, \mathbb{R}^1 is 1-dimensional.

How about \mathbb{R}^2 ? This is the set of vectors $\begin{pmatrix} a_1 \\ a_2 \end{pmatrix}$. This is the set of vectors that point from the origin to points on the plane.

A nonzero vector $\vec{a} \in \mathbf{R}^2$ specifies a line through the origin. Two vectors $\vec{a}, \vec{b} \in \mathbf{R}^2$ specify two *distinct* lines through the origin if and only if there exists no real number *r* such that $\vec{a} = r\vec{b}$:





Now remember that we translated each line along the other to end up with a parallelogram. Let's do that here in such a way that the vectors meet head to tail:

We thus get a new vector that points at the new vertex we created.



Now let's clear out the lines spanned by these vectors, and gaze lovingly at the vectors themselves:

What we've drawn there is a way of taking two vectors \vec{a} and \vec{b} , translating \vec{a} along \vec{b} and translating \vec{b} along \vec{a} to make a parallelogram. The diagonal vector of that parallelogram is

$$\vec{a} + \vec{b} = \begin{pmatrix} a_1 \\ a_2 \end{pmatrix} + \begin{pmatrix} b_1 \\ b_2 \end{pmatrix} = \begin{pmatrix} a_1 + b_1 \\ a_2 + b_2 \end{pmatrix}$$

This is vector addition.



The fact that you can get that diagonal by translating \vec{a} along \vec{b} or by translating \vec{b} along \vec{a} is way of visualizing the commutativity of vector summation. It turns out that *all* the algebraic properties of vectors have pictures to go along with them.

Question. What does associativity look like? adding with $\vec{0}$? the formation of negatives? the distribution of scalar multiplication over vector addition?



\mathbf{R}^3

Of course \mathbb{R}^3 is the set of all vectors $\begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix}$. Consider two vectors $\vec{a}, \vec{b} \in \mathbb{R}^3$

such that there exists no real number r such that $\vec{a} = r\vec{b}$. These two vectors define a plane: the vectors that can be writen as a *linear combination* $s\vec{a} + t\vec{b}$; we call this plane the *span* of \vec{a} and \vec{b} . So a vector \vec{c} does not lie in the *span* of \vec{a} and \vec{b} if and only if it can't be written as a linear combination of of \vec{a} and \vec{b} .

Once we've found such a \vec{c} , I claim that any vector $\vec{v} \in \mathbf{R}^3$ can be written as a linear combination of \vec{a} , \vec{b} , and \vec{c} . Geometrically, this means any point (v_1, v_2, v_3) lies in the span of \vec{a} , \vec{b} , and \vec{c} .



Algebraically, this means that for any $\vec{v} = \begin{pmatrix} v_1 \\ v_2 \\ v_3 \end{pmatrix}$, there is a solution (r, s, t)

to the equation $\vec{v} = r\vec{a} + s\vec{b} + t\vec{c}$. But that equation is secretly the system of linear equations

 $v_1 = ra_1 + sb_1 + tc_1;$ $v_2 = ra_2 + sb_2 + tc_2;$ $v_3 = ra_3 + sb_3 + tc_3.$

What we're saying is that this set of equations has a solution.



But there's more: the vectors \vec{a} , \vec{b} , and \vec{c} are *linearly independent* – that is, the lines they span are independent.

What this means algebraically is that none of them are zero, and there's no way to write

$$\vec{c} = s\vec{a} + t\vec{b}$$

or

$$\vec{b} = p\vec{a} + q\vec{c}$$

or

$$\vec{a} = m\vec{b} + n\vec{c}.$$



We can express this more efficiently: what it means for \vec{a} , \vec{b} , and \vec{c} to be linearly independent is that if

$$r\vec{a} + s\vec{b} + t\vec{c} = \vec{0},$$

then r = s = t = 0.

What we will learn is that for any
$$\vec{v} = \begin{pmatrix} v_1 \\ v_2 \\ v_3 \end{pmatrix}$$
, there is a *unique* solution

(r, s, t) to the equation $\vec{v} = r\vec{a} + s\vec{b} + t\vec{c}$.



Let's do an example. We'll find three independent vectors. We have to start with the nonzero vector $\vec{a} = \begin{pmatrix} 1 \\ 2 \\ 1 \end{pmatrix}$. For our next vector, we just need to select a vector that is not a multiple of \vec{a} . Here's one: $\vec{b} = \begin{pmatrix} 2 \\ 0 \\ 0 \end{pmatrix}$.



Now for the tricky bit. We want a third vector, $\vec{c} = \begin{pmatrix} c_1 \\ c_2 \\ c_3 \end{pmatrix}$, that doesn't lie

in the plane spanned by \vec{a} and \vec{b} . That's more than just making sure that \vec{c} is not a scalar multiple of \vec{a} or \vec{b} .

OK, does
$$\begin{pmatrix} -5/2\\ 15\\ 15/2 \end{pmatrix}$$
 lie in the plane spanned by $\vec{a} = \begin{pmatrix} 1\\ 2\\ 1 \end{pmatrix}$ and $\vec{b} = \begin{pmatrix} 2\\ 0\\ 0 \end{pmatrix}$?



It looks like it does!

$$\frac{15}{2} \begin{pmatrix} 1\\2\\1 \end{pmatrix} + (-5) \begin{pmatrix} 2\\0\\0 \end{pmatrix} = \begin{pmatrix} -2/5\\15\\15/2 \end{pmatrix}.$$

So how do we make sure that we get a vector that doesn't live on that plane?

Here's one that will work:
$$\begin{pmatrix} 0 \\ 0 \\ 2 \end{pmatrix}$$
. What makes me so sure that will work?



See, a linear combination of
$$\vec{a} = \begin{pmatrix} 1 \\ 2 \\ 1 \end{pmatrix}$$
 and $\vec{b} = \begin{pmatrix} 2 \\ 0 \\ 0 \end{pmatrix}$ that gives $\begin{pmatrix} 0 \\ 0 \\ 2 \end{pmatrix}$

would have to have a nonzero coefficient on \vec{a} , and that will give a nonzero second coordinate.

So we have our example:

$$\vec{a} = \begin{pmatrix} 1 \\ 2 \\ 1 \end{pmatrix}, \ \vec{b} = \begin{pmatrix} 2 \\ 0 \\ 0 \end{pmatrix}, \ \vec{c} = \begin{pmatrix} 0 \\ 0 \\ 2 \end{pmatrix}$$

are three linearly independent vectors in \mathbb{R}^3 .



Now my claim is that any vector
$$\vec{v} = \begin{pmatrix} v_1 \\ v_2 \\ v_3 \end{pmatrix}$$
 can be written as a *unique* \vec{v}_1 linear combination of \vec{a} , \vec{b} , and \vec{c} . If $\vec{v} = \begin{pmatrix} 2 \\ -6 \\ 14 \end{pmatrix}$, for example, we're trying

to solve this system of linear equations:

$$2 = r + 2s + 0t;$$

-6 = 2r + 0s + 0t;
14 = r + 0s + 2t.

So we straight away get r = -3, s = 5/2, and t = 17/2.



Here is the general algebraic picture:

Definition. If $\vec{a}_1, \vec{a}_2, ..., \vec{a}_k$ are vectors of \mathbb{R}^n , then the *span* of $\vec{a}_1, \vec{a}_2, ..., \vec{a}_k$ is the set of all linear combinations

 $r_1\vec{a}_1 + r_2\vec{a}_2 + \dots + r_k\vec{a}_k.$



So
$$\vec{v} = \begin{pmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{pmatrix}$$
 lies in the span of $\vec{a}_1 = \begin{pmatrix} a_{11} \\ a_{21} \\ \vdots \\ a_{n1} \end{pmatrix}$, ..., $\vec{a}_k = \begin{pmatrix} a_{1k} \\ a_{2k} \\ \vdots \\ a_{nk} \end{pmatrix}$

if and only if the system of linear equations

$$v_{1} = r_{1}a_{11} + r_{2}a_{12} + \dots + r_{k}a_{1k};$$

$$v_{2} = r_{1}a_{21} + r_{2}a_{22} + \dots + r_{k}a_{2k};$$

$$\vdots$$

$$v_{n} = r_{1}a_{n1} + r_{2}a_{n2} + \dots + r_{k}a_{nk}.$$

has at least one solution.



Geometrically, that means that a vector \vec{v} lies in the span of $\vec{a}_1, \vec{a}_2, \dots, \vec{a}_k$ when you can build a multidimensional parallelopiped out of $\vec{a}_1, \vec{a}_2, \dots, \vec{a}_k$ to get \vec{v} pointing from the origin across the diagonal.





Figure 1: Doesn't know how to make an image of multidimensional paralellopipeds.



Definition. We say vectors $\vec{a}_1, \vec{a}_2, ..., \vec{a}_k$ of \mathbb{R}^n are *linearly independent* if we're in the following situation: any *vanishing* linear combination

$$r_1 \vec{a}_1 + r_2 \vec{a}_2 + \dots + r_k \vec{a}_k = \vec{0}$$

must be a trivial linear combination – that is, we must have

$$r_1=r_2=\cdots=r_k=0.$$



So the vectors
$$\vec{a}_1 = \begin{pmatrix} a_{11} \\ a_{21} \\ \vdots \\ a_{n1} \end{pmatrix}$$
, ..., $\vec{a}_k = \begin{pmatrix} a_{1k} \\ a_{2k} \\ \vdots \\ a_{nk} \end{pmatrix}$ are linearly indepen

dent if and only if any system of linear equations

$$v_{1} = r_{1}a_{11} + r_{2}a_{12} + \dots + r_{k}a_{1k};$$

$$v_{2} = r_{1}a_{21} + r_{2}a_{22} + \dots + r_{k}a_{2k};$$

$$\vdots$$

$$v_{n} = r_{1}a_{n1} + r_{2}a_{n2} + \dots + r_{k}a_{nk}.$$

has at most one solution.



Geometrically, that means that none of the vectors \vec{a}_i lie in the span of the remaining vectors

$$\vec{a}_1, \vec{a}_2, \dots, \vec{a}_{i-1}, \vec{a}_{i+1}, \dots, \vec{a}_k.$$





Figure 2: Cool. Can we go play in the snow now?



We'd like a more efficient way of checking whether a vector lies in a span of a certain collection of vectors and whether that collection of vectors is linearly independent.

To do this, we'll want to introduce two fundamental manipulations of vectors: forming the dot product of two vectors and combining a list of vectors into a matrix. The dot product extracts a bunch of helpful geometry, and matrices give us an efficient way to organize and conceptualize systems of linear equations.

For next week, please read \$\$1.2-2.1 of Strang.

The first problem set will be posted on Monday.