

## § 6.1 Inner Products

20 Apr 2017

So far we have dealt only with notions pertaining to vector space structure (are two vectors parallel, is one vector in the span of some others, etc.). However, to really do geometry we need to get angle into the mix. Let's do it.

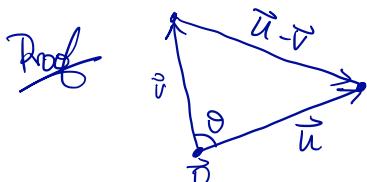
Def<sup>n</sup> If  $\vec{u}, \vec{v}$  are vectors in  $\mathbb{R}^n$ , the inner product of  $\vec{u}$  and  $\vec{v}$  is the scalar  $\vec{u}^T \vec{v}$ .  
a.k.a. "dot product"

Proposition  $\vec{u} \cdot \vec{u} = |\vec{u}|^2$

Proof  $\vec{u} = \begin{bmatrix} u_1 \\ \vdots \\ u_n \end{bmatrix} \Rightarrow \vec{u} \cdot \vec{u} = u_1^2 + \dots + u_n^2 = |\vec{u}|^2$ .

will cut straight to the chase:

Proposition  $\vec{u} \cdot \vec{v} = |\vec{u}| |\vec{v}| \cos \theta$ , where  $\theta$  is the angle between  $\vec{u}$  and  $\vec{v}$ .



Law of cosines:  $|\vec{u} - \vec{v}|^2 = |\vec{u}|^2 + |\vec{v}|^2 - 2|\vec{u}||\vec{v}|\cos\theta$   
Linearity:  $(\vec{u} - \vec{v}) \cdot (\vec{u} - \vec{v}) = |\vec{u}|^2 + |\vec{v}|^2 - 2\vec{u} \cdot \vec{v}$   
solving for the two highlighted expressions shows us they're the same.

Corollary  $\vec{u} \cdot \vec{v} = 0$  if & only if  $\vec{u}, \vec{v}$  are perpendicular.

Let's pause to reflect on this new development. In a vector space where we only have access to the vector space operations, we have no notion of perpendicularity. We require this additional structure, the inner product, for that.

Theorem The inner product satisfies

$$(*) \quad \vec{u} \cdot \vec{v} = \vec{v} \cdot \vec{u} \quad (\text{commutative})$$

$$(*) \quad (\vec{u} + \vec{v}) \cdot \vec{w} = \vec{u} \cdot \vec{w} + \vec{v} \cdot \vec{w} \quad (\text{distributive})$$

$$(*) \quad (c\vec{u}) \cdot \vec{v} = c(\vec{u} \cdot \vec{v}) \quad (\text{associative with scalar product})$$

$$(*) \quad \vec{u} \cdot \vec{u} \geq 0, \text{ with } \vec{u} \cdot \vec{u} = 0 \text{ iff } \vec{u} = 0. \quad (\text{positive})$$

Proof The first three follow from the corresponding matrix properties, since  $\vec{u} \cdot \vec{v} = \vec{u}^T \vec{v}$ . The last follows from  $\vec{u} \cdot \vec{u} = |\vec{u}|^2$ .

## Unit Vectors

Example Find the vector whose length is 1 and which points in the same direction as  $\vec{u} = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$ .

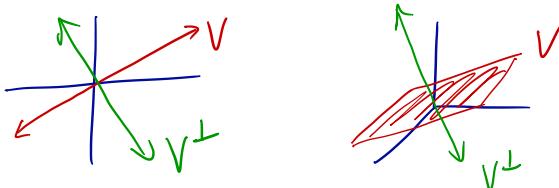
Solution If we scale  $\vec{u}$  down by its length, then that new vector will have length 1:

$$\frac{\vec{u}}{\|\vec{u}\|} = \frac{[1 \ 2 \ 3]^T}{\sqrt{1^2 + 2^2 + 3^2}} = \begin{bmatrix} \frac{1}{\sqrt{14}} \\ \frac{2}{\sqrt{14}} \\ \frac{3}{\sqrt{14}} \end{bmatrix}.$$

We say  $\frac{\vec{u}}{\|\vec{u}\|} = \begin{bmatrix} \frac{1}{\sqrt{14}} \\ \frac{2}{\sqrt{14}} \\ \frac{3}{\sqrt{14}} \end{bmatrix}$  is the unit vector of  $\vec{u}$ .

## Orthogonal complements

If  $V$  is a subspace of  $\mathbb{R}^n$ , then we define the orthogonal complement of  $V$  to be the set of vectors in  $\mathbb{R}^n$  perpendicular to every vector in  $V$ . We call this set  $V^\perp$ . E.g.:



## Facts

- (\*)  $\vec{x} \in V^\perp$  iff  $\vec{x} \perp \vec{y}$  for all  $\vec{y}$  in a spanning list  $V$ .  
(\* )  $V^\perp$  is a subspace

Theorem Let  $A$  be an  $m \times n$  matrix.

$$(i) (\text{Row } A)^\perp = \text{Nul } A, \quad (ii) (\text{Col } A)^\perp = \text{Nul } A^T$$

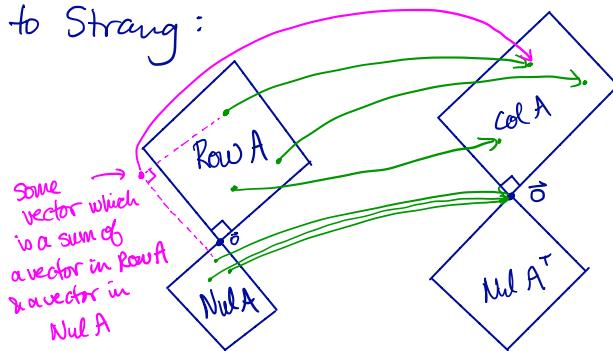
Proof  $\vec{x} \in \text{Nul } A \Rightarrow A\vec{x} = \vec{0} \Rightarrow \vec{x} \perp$  to each row of  $A$ ,

by the row-based description of the matrix-vector product.

So  $\vec{x} \in (\text{Row } A)^\perp$ . Conversely, if  $\vec{x} \in (\text{Row } A)^\perp$  then clearly  $A\vec{x} = \vec{0}$ , so  $\vec{x} \in \text{Nul } A$ . So  $(\text{Row } A)^\perp = \text{Nul } A$ .

Applying (i) to  $A^T$  gives (ii). (the concept, not my drawing)

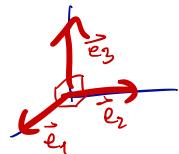
So we have the following beautiful schematic diagram,  
due to Strang:



## §6.2 Orthogonal Lists

A list  $\{\vec{v}_1, \dots, \vec{v}_n\}$  is said to be orthogonal if it's pairwise orthogonal :  $\vec{v}_i \cdot \vec{v}_j = 0$  for every pair  $(i, j)$  with  $i \neq j$ .

Example  $\{\vec{e}_1, \dots, \vec{e}_n\}$  in  $\mathbb{R}^n$  is orthogonal.



$$\{\vec{v}_1, \dots, \vec{v}_n\}$$

Prop. An orthogonal list of nonzero vectors is lin. ind.

Proof

If  $c_1\vec{v}_1 + \dots + c_n\vec{v}_n = \vec{0}$ , then

$$\vec{v}_1 \cdot (c_1\vec{v}_1 + \dots + c_n\vec{v}_n) = \vec{v}_1 \cdot \vec{0} \Rightarrow$$

$$c_1|\vec{v}_1|^2 + c_2 \cdot 0 + \dots + c_n \cdot 0 = 0 \Rightarrow$$

$$c_1|\vec{v}_1|^2 = 0.$$

So  $c_1 = 0$ . Similarly,  $c_2 = \dots = c_n = 0$ . So the list is linearly independent.

An orthogonal basis is what it says on the tin: a basis which is orthogonal.

What's great about them is how easy it is to find coordinates:

Proposition If  $\{\vec{v}_1, \dots, \vec{v}_k\}$  is an orthogonal basis of  $V \subset \mathbb{R}^n$ , then  $\vec{v} = c_1 \vec{v}_1 + \dots + c_k \vec{v}_k$  implies  $c_j = \frac{\vec{v} \cdot \vec{v}_j}{\vec{v}_j \cdot \vec{v}_j}$ .

Proof  $\vec{v} \cdot \vec{v}_j = (c_1 \vec{v}_1 + \dots + c_k \vec{v}_k) \cdot \vec{v}_j = c_j \vec{v}_j \cdot \vec{v}_j + \text{terms of zeros.}$   
so  $c_j = \frac{\vec{v} \cdot \vec{v}_j}{\vec{v}_j \cdot \vec{v}_j}$ .