

CHAPTER


VI: NORM

and

ORTHO-

GONALITY

This chapter introduces some geometry to vectors, norm or length and orthogonality;

"orthogonal" means perpendicular or "making a right angle"  ;

the "square corner" is a traditional drawing.

Orthogonality will be seen to be particularly useful and fundamental,

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enabling us to minimize
bad things such as errors
in measurements or models,
and to (sometimes) make
much easier descriptions
such as linear combinations.

Here is a purely
algebraic spin on this
chapter. Algebra with
vectors so far consists only
of addition and multiplication
by real numbers.
Conspicuously missing is

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multiplication of one
vector by another. This
chapter will present two
ways to do this: the
dot product in Section
B through E and the
cross product in Section F.

SECTION VIA: NORM

DEFINITION 6.1

Suppose $\vec{x} \equiv (x_1, x_2, \dots, x_n)$
is a vector in \mathbb{R}^n .

The norm or length
or magnitude of \vec{x} is

$$\|\vec{x}\| \equiv \sqrt{x_1^2 + x_2^2 + \dots + x_n^2}$$

Example 6.2

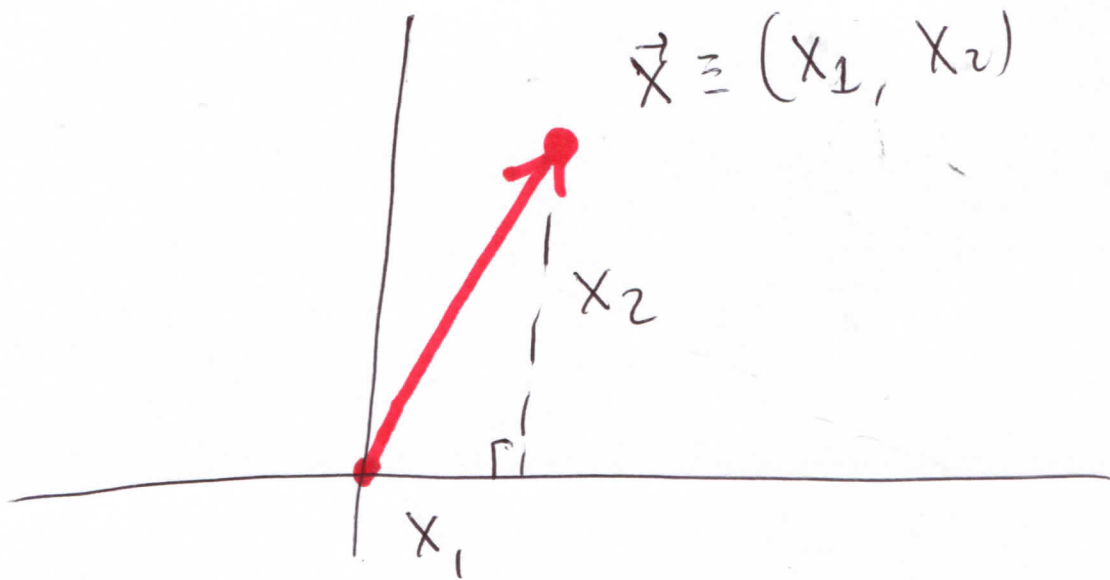
$$\| (1, -2, 0, \pi) \| =$$

$$\sqrt{1^2 + (-2)^2 + 0^2 + \pi^2} =$$

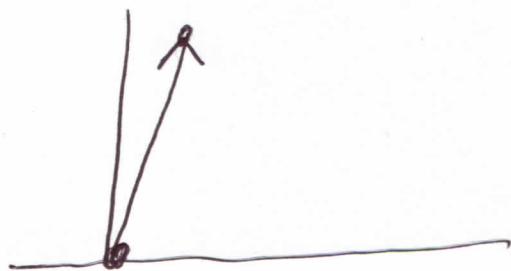
$$\sqrt{5 + \pi^2}.$$

REMARKS 6.3

For $n=2$, $\|\vec{x}\|$ is (literally) the length of any directed line segment representing \vec{x} , by the Pythagorean Theorem.



Intuitively, vectors differ from real numbers (sometimes called scalars) in that they have both magnitude and direction. Velocity, such as 50 miles per hour North by Northwest,



is a vector, while speed, such as 50 miles per hour, is a scalar.

If we wish to focus on direction only, it is natural to use vectors whose norm is one.

DEFINITION 6.4

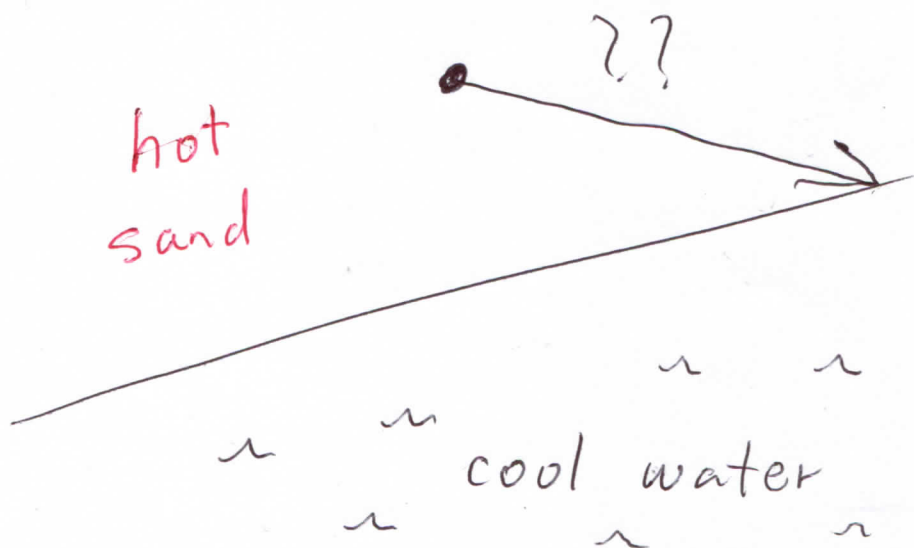
A **unit vector** is a vector whose norm is one.

SECTION VI B: ORTHOGONALITY

and

DOT PRODUCT

Consider the following
scenario.

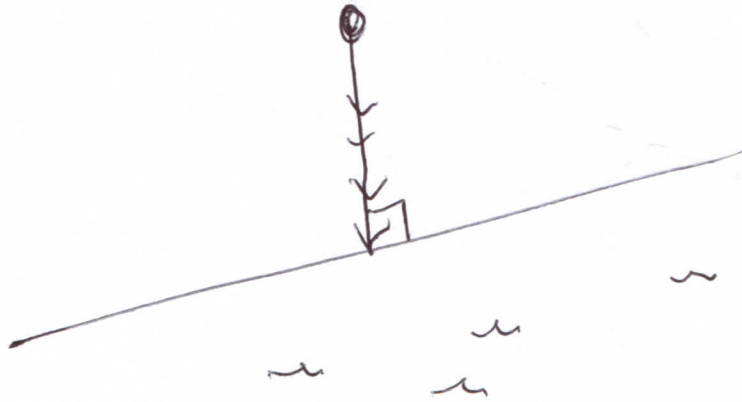



You're on the beach,
with hot sand burning your
feet. You rush to the cool
water to cool your feet,
with your path indicated by
the arrow with question marks.

If this were a lecture,
we would now ask the class
"Why is that path dumb?"

The answer to that
insensitive question is that,
assuming we want to reach
the water as quickly as,

possible, our path should look like the following:



The geometry here is that your path should make a right angle () with the shoreline, to minimize distance.

TERMINOLOGY 6.5

" $\vec{a} \perp \vec{b}$ " means the vector

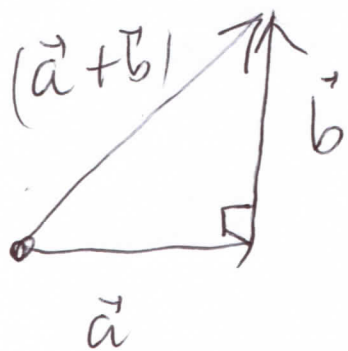
\vec{a} is orthogonal, or

perpendicular, to the vector
 \vec{b} .

We'd like to characterize
orthogonality algebraically.

Motivation comes from the

Pythagorean Theorem

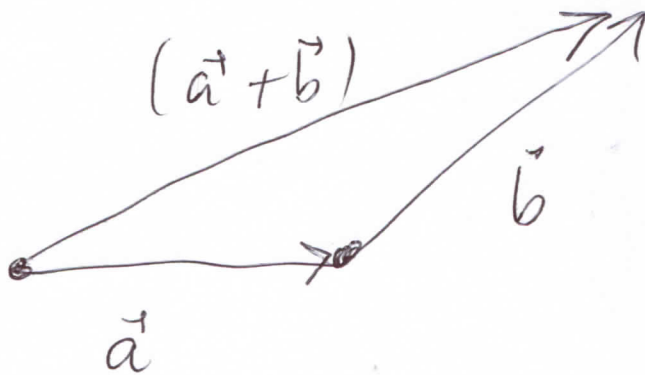


$$\|\vec{a} + \vec{b}\|^2 = \|\vec{a}\|^2 + \|\vec{b}\|^2$$

$$\underline{\text{iff}} \quad \vec{a} \perp \vec{b}$$

The following calculation could be done with vectors in \mathbb{R}^n , for any n , but we will do it for $n=2$, for simplicity.

For any vector $\vec{a} \equiv (a_1, a_2)$ and $\vec{b} \equiv (b_1, b_2)$,



$$\|\vec{a} + \vec{b}\|^2 =$$

$$\|(a_1 + b_1, a_2 + b_2)\|^2 =$$

$$(a_1 + b_1)^2 + (a_2 + b_2)^2 =$$

$$(a_1^2 + 2a_1b_1 + b_1^2) +$$

$$(a_2^2 + 2a_2b_2 + b_2^2) =$$

$$(a_1^2 + a_2^2) + (b_1^2 + b_2^2)$$

$$+ 2(a_1b_1 + a_2b_2) =$$

$$\|\vec{a}\|^2 + \|\vec{b}\|^2 + \underbrace{2(a_1b_1 + a_2b_2)}_{\uparrow}$$

GIVE THIS a name

NOTICE that the Pythagorean Theorem holds precisely when that last quantity equals zero.

This motivates the following definition.

DEFINITION 6.6

The dot or inner or scalar product is

(In \mathbb{R}^2)

$$(a_1, a_2) \bullet (b_1, b_2)$$

$$\equiv (a_1 b_1 + a_2 b_2).$$

Example 6.7

$$\begin{aligned}(1, 2) \cdot (3, -4) &= 1 \cdot 3 + 2(-4) \\ &= -5.\end{aligned}$$

DEFINITION 6.8

For any natural number n ,
if $\vec{a} \equiv (a_1, a_2, \dots, a_n)$ and
 $\vec{b} \equiv (b_1, b_2, \dots, b_n)$, then

$$\vec{a} \cdot \vec{b} \equiv$$

$$(a_1 b_1 + a_2 b_2 + \dots + a_n b_n)$$

is the dot or inner or
scalar product
of \vec{a} and \vec{b} .

Example 6.9

$$(1, 1, 5, -3) \cdot (4, 7, 0, 2) = \\ (1 \cdot 4 + 1 \cdot 7 + 5 \cdot 0 + (-3) \cdot 2) = 5$$

Our calculations before
Definition 6.6, generalized
to arbitrary n , motivate
the following definition.

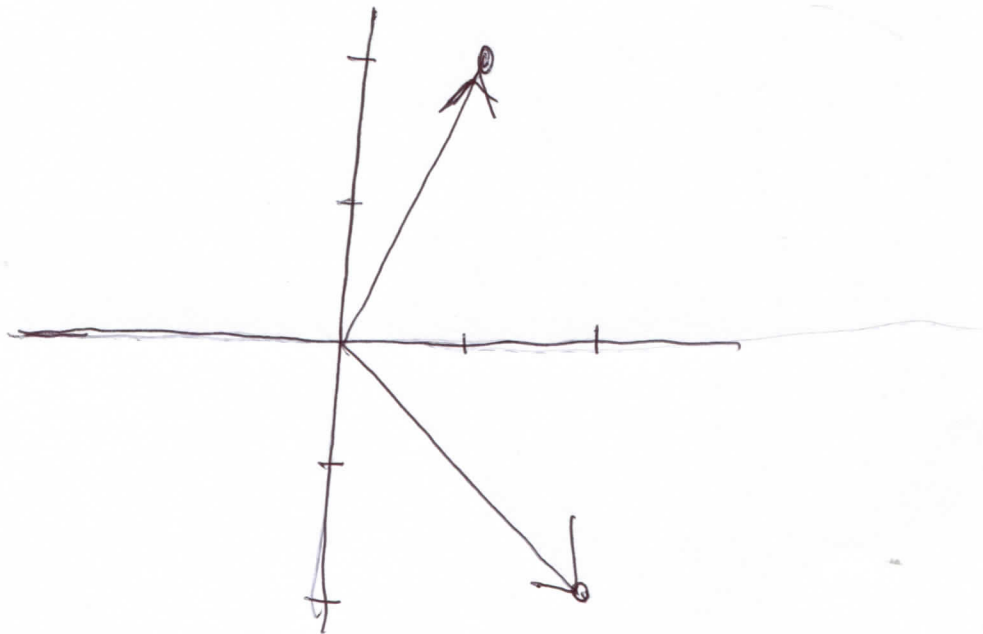
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DEFINITION 6.10

Two vectors \vec{a}, \vec{b} in \mathbb{R}^n are orthogonal, $\vec{a} \perp \vec{b}$, if $\vec{a} \cdot \vec{b} = 0$.

Example 6.11 In building or standing, orthogonality is essential for stability or comfort. Staring at vectors may not be sufficient to determine orthogonality, even if we try very hard to make a good picture.

For example, is
 $(1, 2) \perp (2, -2)$? Here's
a picture.



Even if we trusted our art skills, we might have trouble deciding if that's a right angle between those vectors.

Algebra relieves us of uncertainty. Take the dot product:

$$(1, 2) \cdot (2, -2) = 2 - 4 = -2;$$

all we care is that the dot product is not zero, therefore the vectors are not orthogonal.

DOT PRODUCT PROPERTIES 6.12

For n a natural number,
 \vec{a} , \vec{b} , and \vec{c} are in \mathbb{R}^n and
 α is a real number.

$$(a) \|\vec{a}\|^2 = \vec{a} \cdot \vec{a}$$

$$(b) \|\vec{a} + \vec{b}\|^2 = \|\vec{a}\|^2 + \|\vec{b}\|^2 + 2(\vec{a} \cdot \vec{b})$$

(c) (Pythagorean Theorem)

$$\|\vec{a} + \vec{b}\|^2 = \|\vec{a}\|^2 + \|\vec{b}\|^2$$

if and only if $\vec{a} \perp \vec{b}$.

$$(d) \vec{a} \cdot (\vec{b} + \vec{c}) = (\vec{a} \cdot \vec{b}) + (\vec{a} \cdot \vec{c})$$

$$(e) (\alpha \vec{a}) \cdot \vec{b} = \alpha (\vec{a} \cdot \vec{b})$$

$$(f) (\vec{a} \cdot \vec{b}) = (\vec{b} \cdot \vec{a})$$

$$(g) \vec{a} \cdot \vec{a} = 0 \text{ if and only if } \vec{a} = \vec{0}.$$

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Let's return to the hot-sand-induced motivation at the beginning of this section.

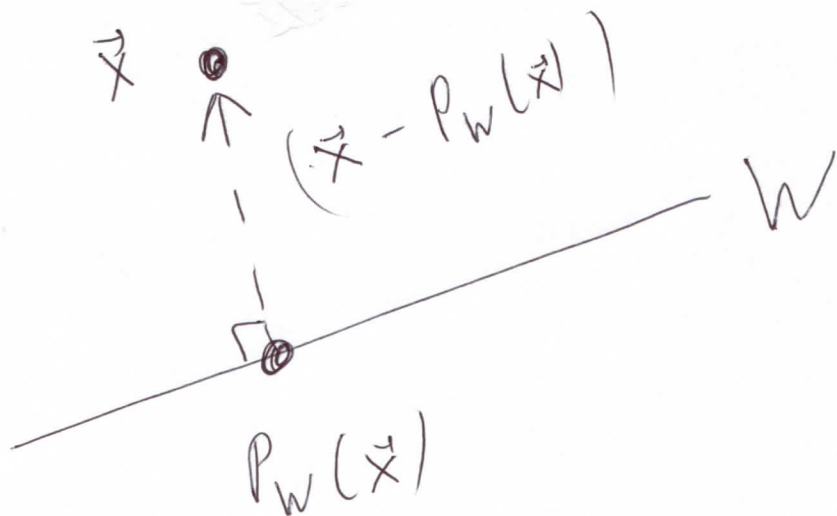
DEFINITION 6.13

If \vec{x} is in \mathbb{R}^n and W is a subspace of \mathbb{R}^n , then the (orthogonal) projection of \vec{x} onto W , denoted $P_W(\vec{x})$, is a vector in W such that

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$(\vec{x} - P_W(\vec{x})) \perp W$; that is

$(\vec{x} - P_W(\vec{x})) \perp \vec{w}$, for all \vec{w}
in W



Intuitively, W is reality,
 \vec{x} is a measurement (mistakes
were made), and $P_W(\vec{x})$ is the
possible value closest to \vec{x} .

But let's not take that last assertion on faith.

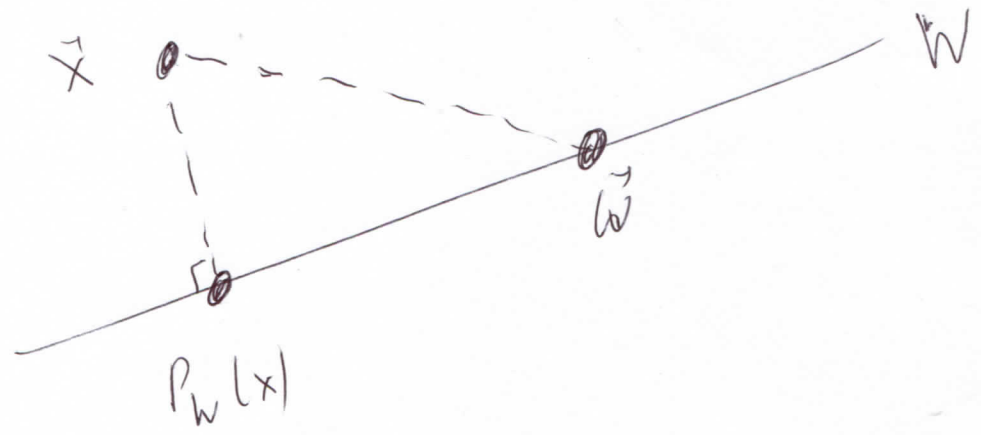
THEOREM 6.14

If \vec{x} is in \mathbb{R}^n and W is a subspace of \mathbb{R}^n , then

$$\|\vec{x} - P_W(\vec{x})\| \leq \|\vec{x} - \vec{w}\|$$

for all \vec{w} in W .

Proof:



By the Pythagorean theorem,

$$\begin{aligned}\|\vec{x} - \vec{w}\|^2 &= \|(\vec{x} - P_W(\vec{x})) + (P_W(\vec{x}) - \vec{w})\|^2 \\ &= \|\vec{x} - P_W(\vec{x})\|^2 + \|P_W(\vec{x}) - \vec{w}\|^2 \\ &\geq \|\vec{x} - P_W(\vec{x})\|^2.\end{aligned}$$

TERMINOLOGY 6.15

Theorem 6.14 explains why $P_W(\vec{x})$ is called a **best approximation** or **least-squares approximation**

of \vec{x} from W_g^c

"least-squares" refers to the norm of Section VI A.

We will focus for the rest of this section on one-dimensional projections; that is, W in Definition 6.13 equal to a one-dimensional subspace of \mathbb{R}^n .

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DEFINITION 6.16

For \vec{a}, \vec{b} vectors in \mathbb{R}^n , $\vec{b} \neq \vec{0}$,
denote by

$$\text{proj}_{\vec{b}} \equiv P_{\text{span}(\vec{b})};$$

that is, the (orthogonal)

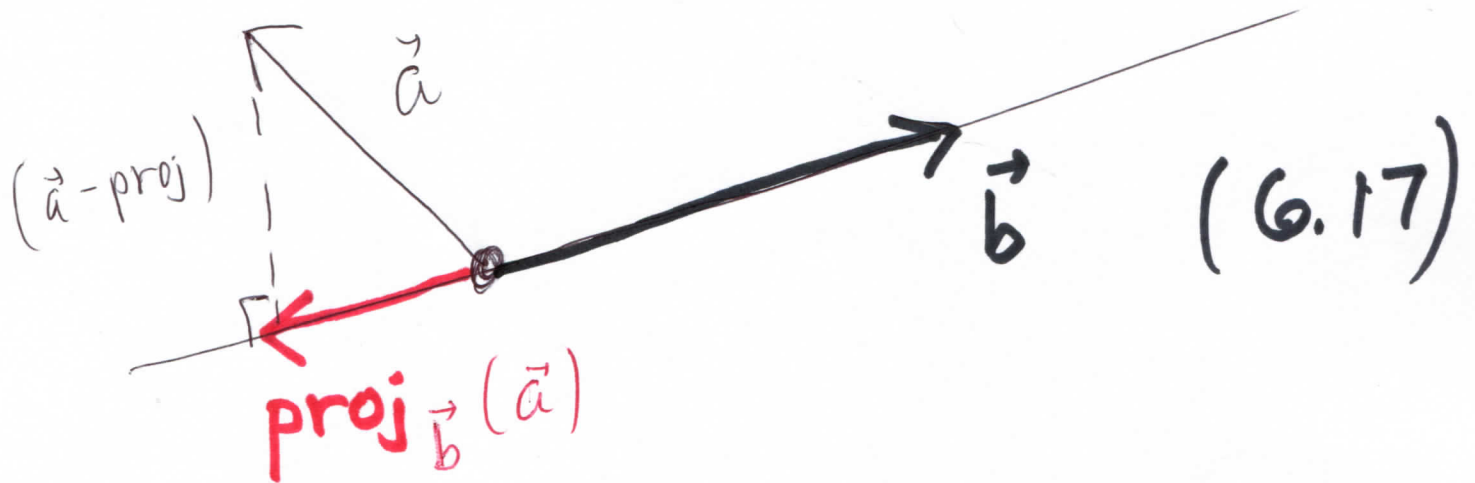
projection of \vec{a} onto \vec{b} ,

denoted $\text{proj}_{\vec{b}}(\vec{a})$

is a real multiple of \vec{b}

such that

$$(\vec{a} - \text{proj}_{\vec{b}}(\vec{a})) \perp \vec{b}.$$



This is a geometric definition; by this we mean that a picture tells the entire story.

As with using the dot product to characterize orthogonality, we would like

an algebraic formula
for $\text{proj}_{\vec{b}}(\vec{a})$.

First, the definition implies
that

$$\text{proj}_{\vec{b}}(\vec{a}) = t\vec{b}$$

for some real t . All we
need is t . From (6.17),

$$(\vec{a} - t\vec{b}) \perp \vec{b} \quad \rightarrow$$

$$0 = (\vec{a} - t\vec{b}) \cdot \vec{b} = (\vec{a} \cdot \vec{b})$$

$$- t(\vec{b} \cdot \vec{b}) \quad \rightarrow$$

$$t = \frac{(\vec{a} \cdot \vec{b})}{\vec{b} \cdot \vec{b}} = \frac{(\vec{a} \cdot \vec{b})}{\|\vec{b}\|^2}.$$

We have shown the following.

Theorem 6.18 If \vec{a} and

\vec{b} are in \mathbb{R}^n and $\vec{b} \neq \vec{0}$,

then

$$\text{proj}_{\vec{b}}(\vec{a}) = \left[\frac{(\vec{a} \cdot \vec{b})}{\|\vec{b}\|^2} \right] \vec{b}$$

DEFINITION 6.19 The

component of \vec{a} in the direction of \vec{b} is

$$\text{comp}_{\vec{b}}(\vec{a}) \equiv \frac{\vec{a} \cdot \vec{b}}{\|\vec{b}\|}$$

Example 6.20

The projection of $(1, 2, 3, 4)$
onto $(1, 0, -1, 0)$ is

$$\text{proj}_{(1, 0, -1, 0)} (1, 2, 3, 4) =$$

$$\left[\frac{(1, 2, 3, 4) \cdot (1, 0, -1, 0)}{\|(1, 0, -1, 0)\|^2} \right] (1, 0, -1, 0)$$

$$= \left(\frac{-2}{2} \right) (1, 0, -1, 0) = (-1, 0, 1, 0).$$

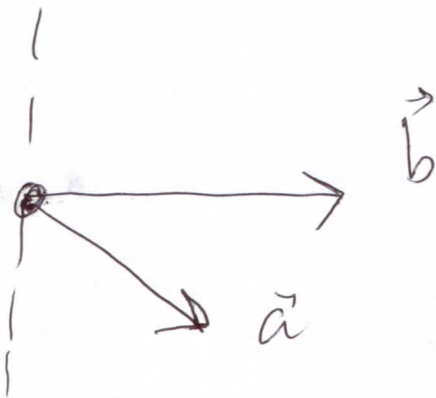
The component of $(1, 2, 3, 4)$

in the direction of $(1, 0, -1, 0)$ is

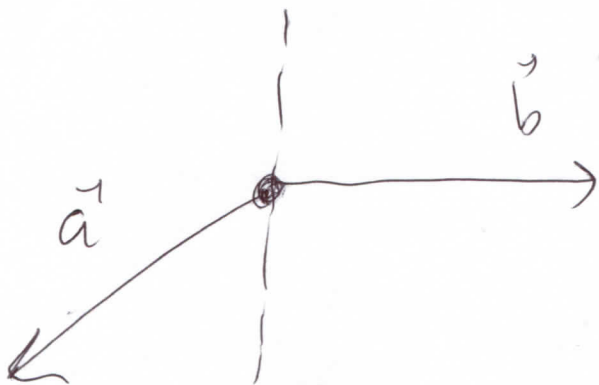
$$\text{comp}_{(1, 0, -1, 0)} (1, 2, 3, 4) = \frac{(1, 2, 3, 4) \cdot (1, 0, -1, 0)}{\|(1, 0, -1, 0)\|}$$

$$= \frac{-2}{\sqrt{2}} = -\sqrt{2}$$

Notice that projection is a vector, while component is a number. Component is almost the norm of projection, except that it could be negative.



positive
component
 $\text{comp}_b(\vec{a})$



negative
component
 $\text{comp}_b(\vec{a})$

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Physically, \vec{b} might be the direction you ride your bike, while \vec{a} is the wind velocity. The only part of \vec{a} that affects your bike speed is $\text{proj}_{\vec{b}}(\vec{a})$; if $\text{comp}_{\vec{b}}(\vec{a})$ is negative, you have some wind in your face; if $\text{comp}_{\vec{b}}(\vec{a})$ is positive, you feel a wind on your bike propelling you forward.

REMARKS 6.21

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Results such as Theorem 6.18 and Definition 6.10, that intertwine algebra and geometry, are the best of both worlds: the precision of algebra and the intuition of geometry.

Since we have been talking so much about orthogonality, it is time to mention the other extreme, being parallel.

DEFINITIONS 6.22 ^{p. 423}

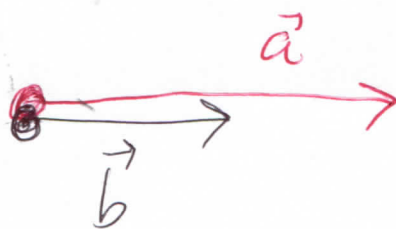
If \vec{a} and \vec{b} are two vectors in \mathbb{R}^n , \vec{a} and \vec{b} are

parallel if one vector is a multiple of the other.

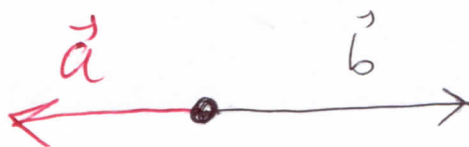
See the vector drawings before Example 1.22.

If the multiple above is positive, \vec{a} and \vec{b} **point in the same direction**; if the multiple is negative, \vec{a} and \vec{b} **point in opposite directions**.

SAME
DIRECTION



OPPOSITE
DIRECTION



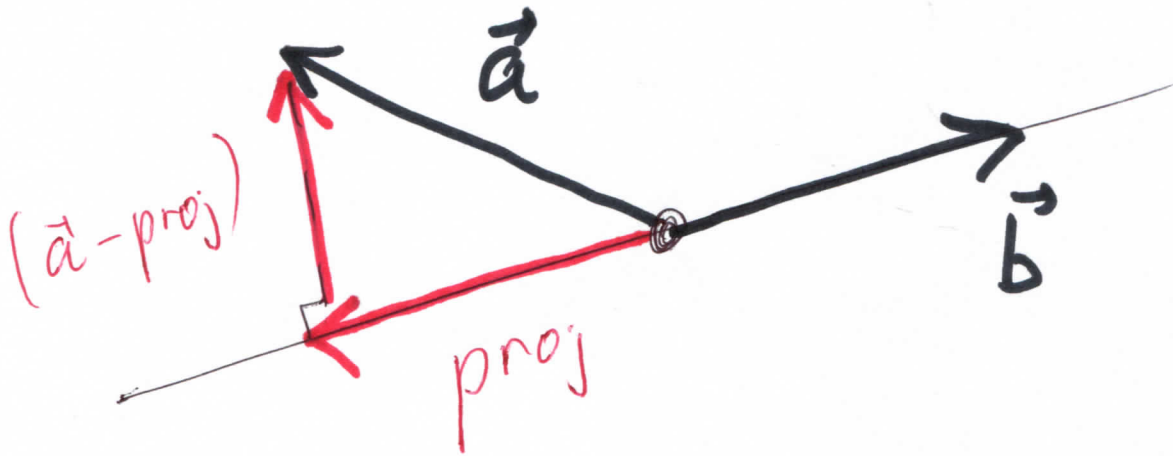
PROPOSITION 6.23

If \vec{a} and \vec{b} are in \mathbb{R}^n and $\vec{b} \neq \vec{0}$, then

$$\vec{a} = \text{proj}_{\vec{b}}(\vec{a}) + (\vec{a} - \text{proj}_{\vec{b}}(\vec{a}))$$

writes \vec{a} as a sum of two vectors, one parallel to \vec{b} , the other orthogonal to \vec{b} .

See (6.17).



Example 6.24

Write $(1, 0, -1, 2, 1)$ as a sum of two vectors, one parallel to $(0, 1, 2, 3, 4)$, the other perpendicular to $(0, 1, 2, 3, 4)$.

SOLUTION:

$$\text{proj}_{(0,1,2,3,4)} (1, 0, -1, 2, 1) =$$

$$\left[\frac{(1, 0, -1, 2, 1) \cdot (0, 1, 2, 3, 4)}{\|(0, 1, 2, 3, 4)\|^2} \right] (0, 1, 2, 3, 4)$$

$$= \frac{8}{30} (0, 1, 2, 3, 4) = \frac{4}{15} (0, 1, 2, 3, 4);$$

$$(1, 0, -1, 2, 1) - \frac{4}{15} (0, 1, 2, 3, 4) =$$

$$\frac{1}{15} [(15, 0, -15, 30, 15) - (0, 4, 8, 12, 16)]$$

$$= \frac{1}{15} (15, -4, -23, 18, -1); \text{ thus}$$

$$(1, 0, -1, 2, 1) =$$

$$\frac{4}{15} (0, 1, 2, 3, 4) + \frac{1}{15} (15, -4, -23, 18, -1)$$

↖ ↗
orthogonal (dot product zero)

Since Proposition 6.23 is an orthogonal sum, we may apply the Pythagorean Theorem.

COROLLARY 6.25

For \vec{a}, \vec{b} as in Proposition 6.23,

$$\|\vec{a}\|^2 = \|\vec{a} - \text{proj}_{\vec{b}}(\vec{a})\|^2 + \|\text{proj}_{\vec{b}}(\vec{a})\|^2$$

The following is surprisingly useful, especially for infinite-dimensional dot products.

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CAUCHY - SCHWARZ

INEQUALITY 6.26

(In finite dimension)

For \vec{a}, \vec{b} in \mathbb{R}^n ,

$$|\vec{a} \cdot \vec{b}| \leq \|\vec{a}\| \|\vec{b}\|.$$

Proof: If $\vec{b} = \vec{0}$, both sides of the inequality are zero

If $\vec{b} \neq \vec{0}$, then by Corollary

$$6.25, \|\vec{a}\| \geq \|\text{proj}_{\vec{b}}(\vec{a})\| =$$

$$\left\| \left(\frac{\vec{a} \cdot \vec{b}}{\|\vec{b}\|^2} \right) \vec{b} \right\| = \frac{|\vec{a} \cdot \vec{b}|}{\|\vec{b}\|}.$$

REMARKS and

Example 6.27.

If armed with trigonometry, the Cauchy-Schwarz inequality allows us to calculate the measures of angles between arbitrary nontrivial vectors with dot products; see Appendix Four.

For integral powers of integers k , there are many formulas for sums of powers of k ; e.g.,

$$(1 + 2 + 3 + \dots + n) = \frac{n(n+1)}{2}, \quad p. 430$$

$$(1^2 + 2^2 + 3^2 + \dots + n^2) = \frac{n(n+1)(2n+1)}{6}, \text{ etc.}$$

For nonintegral powers, such formulas might not exist, but Cauchy-Schwarz allows us to estimate; e.g.

$$(\sqrt{1} + \sqrt{2} + \sqrt{3} + \dots + \sqrt{n}) =$$

$$(1, 1, \dots, 1) \cdot (\sqrt{1}, \sqrt{2}, \sqrt{3}, \dots, \sqrt{n})$$

$$\leq \| (1, 1, \dots, 1) \| \| (\sqrt{1}, \sqrt{2}, \dots, \sqrt{n}) \|$$

$$= \sqrt{n} \sqrt{1 + 2 + 3 + \dots + n}$$

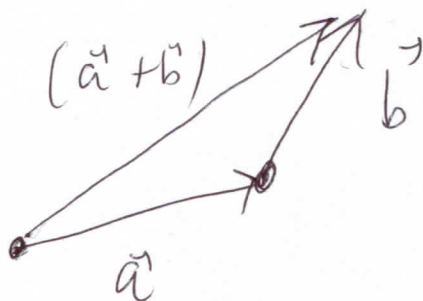
$$= \sqrt{n} \sqrt{\frac{n(n+1)}{2}} = n \frac{\sqrt{n+1}}{\sqrt{2}}.$$

Another consequence of Cauchy Schwarz is the following, which is very believable when one draws a picture of vector addition.

TRIANGLE INEQUALITY 6.28

$$\|\vec{a} + \vec{b}\| \leq \|\vec{a}\| + \|\vec{b}\|,$$

for any \vec{a}, \vec{b} in \mathbb{R}^n .



Proof: By 6.12 and
Cauchy Schwarz, in that
order,

$$\begin{aligned}\|\vec{a} + \vec{b}\|^2 &= \|\vec{a}\|^2 + \|\vec{b}\|^2 + 2(\vec{a} \cdot \vec{b}) \\ &\leq \|\vec{a}\|^2 + \|\vec{b}\|^2 + 2\|\vec{a}\|\|\vec{b}\| \\ &= (\|\vec{a}\| + \|\vec{b}\|)^2\end{aligned}$$