

# Linear Algebra II

Math 232

instructor:

John  
Terilla

# Orthogonal Projectors and the Minimization problem

May 8,  
2020

In a normed vector space  
 $(V, \|\cdot\|)$  one often faces  
the following problem:

Problem: Given a subspace  $W \subseteq V$   
and a vector  $v \in V \setminus W$ ,  
Find the vector in  $W$  that  
is closest to  $v$ . That is,  
find  $\operatorname{argmin}_{w \in W} \|v - w\|$ .

This problem shows up in  
many places. For example, in  
approximation of functions or in

fitting data to a model.

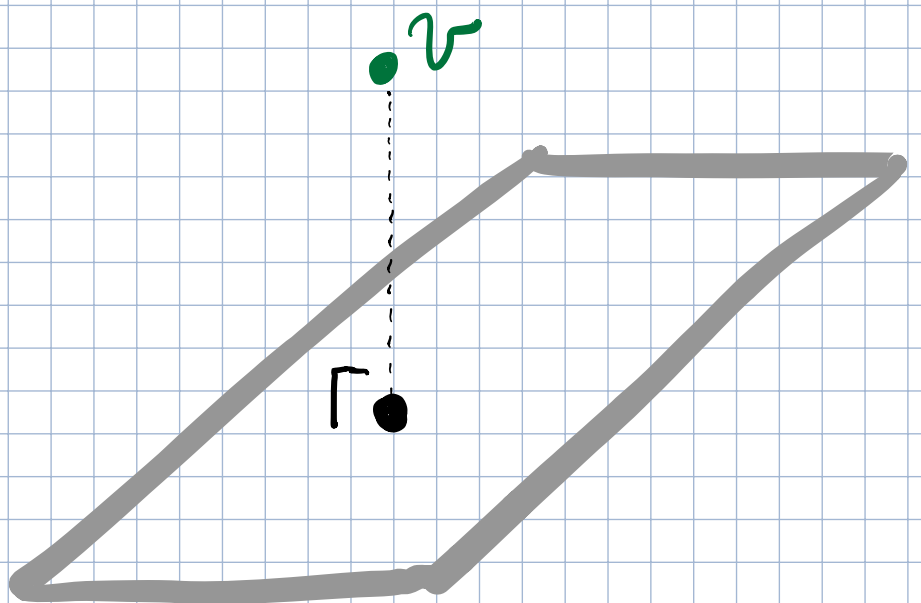
Imagine you'd like to find a polynomial  $p \in \mathcal{P}_5(\mathbb{R})$  that is closest to  $\cos(t)$ . First, you need to make "closest" a precise concept, say by equipping the space  $V = \{ \text{continuous functions } [-1, 1] \rightarrow \mathbb{R} \}$  with a norm.

Then, you're trying to find the vector  $p \in \mathcal{P}_5(\mathbb{R}) \subseteq V$  that minimizes  $\| \cos t - p \|$ .

Now, in general this minimization problem is hard.

However, if  $\|\cdot\|$  comes from an inner product, the problem is easy. The reason is that you have orthogonal projection!

Here's a picture:



I'm going to present this a little differently than Axler since we'll be interested in the case when we have an "almost inner product" defined on  $V$  that restricts to an inner product on a finite dimensional subspace  $W$ .

For example, if  $V = \{ \text{continuous functions } [0, 10] \rightarrow \mathbb{R} \}$  and

$$\langle f, g \rangle := f(0)g(0) + f(2)g(2) + f(4)g(4) + f(6)g(6) + f(8)g(8) + f(10)g(10)$$

then  $\langle, \rangle$  isn't quite an inner

product since it's possible that  $\langle f, f \rangle = 0$  even though  $f \neq 0$ .

For example

$$f(x) = \frac{x(x-2)(x-4)(x-6)(x-8)(x-10)\sin(x)}{e^{5x^2+1} + \sqrt{x^2 + 5x^4 + 1}}$$

Nonetheless, if we define  $\|f\|$

by  $\|f\| = \sqrt{\langle f, f \rangle}$ , then we

have  $\|f\| \geq 0$

$$\|\alpha f\| = |\alpha| \|f\|$$

$$\|f+g\| \leq \|f\| + \|g\|$$

it just may happen that

$\|f\| = 0$  for nonzero  $f \in V$ .

Moreover, if we define distance by  $d(f, g)$  by

$d(f, g) := \|f - g\|$  then we have

$$d(f, g) \geq 0$$

$$d(f, g) = d(g, f)$$

$$d(f, g) \leq d(f, h) + d(h, g)$$

it just may happen that  $d(f, g) = 0$  when  $f \neq g$ .

Note that  $\|f\| = \sqrt{\langle f, f \rangle}$

where

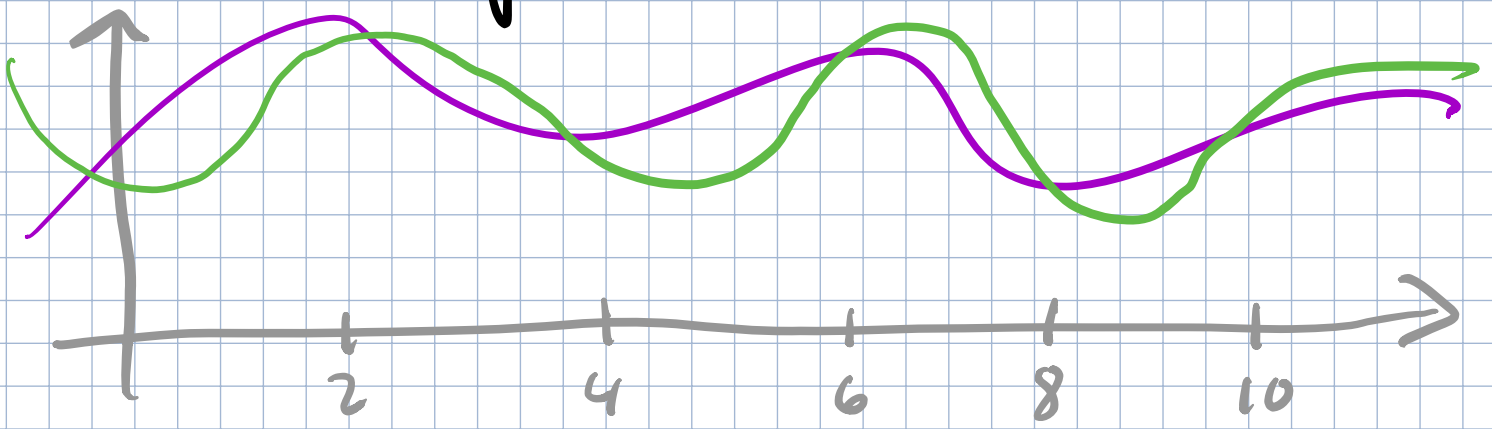


$$\langle f, g \rangle := f(0)g(0) + f(2)g(2) + f(4)g(4) \\ + f(6)g(6) + f(8)g(8) + f(10)g(10)$$

Captures a useful idea of distance:

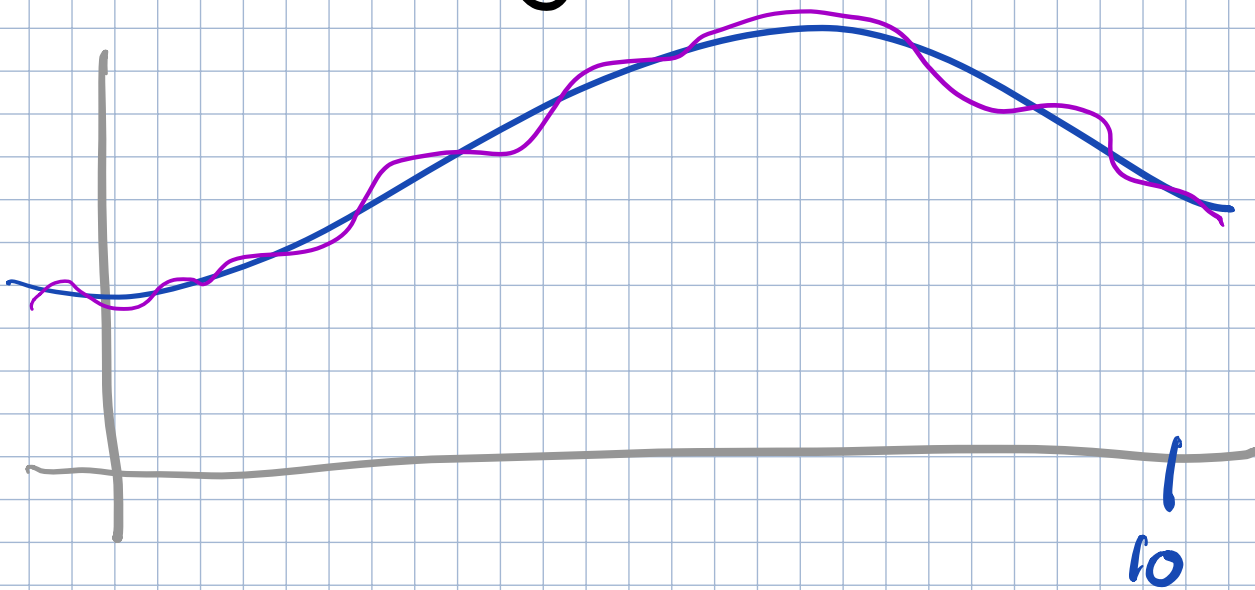
$d(f, g)$  is small  $\Leftrightarrow \langle f-g, f-g \rangle$   
is small  $\Leftrightarrow (f(0)-g(0))^2 +$   
 $(f(2)-g(2))^2 + \dots + (f(10)-g(10))^2$  is  
small  $\Leftrightarrow f(x) \approx g(x)$  when  
 $x = 0, 2, 4, 6, 8, 10$ .

Here's a picture:

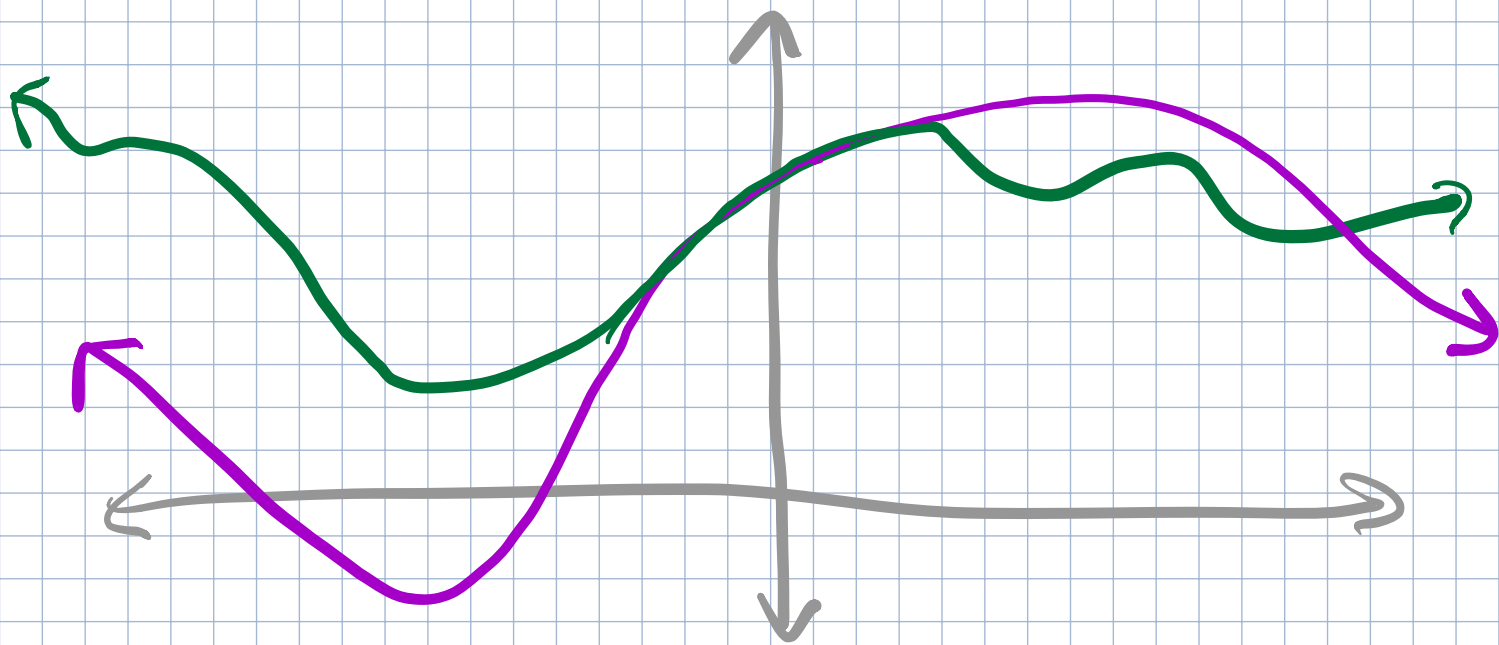


Other choices for "almost" inner products on  $V$  that capture reasonable notions of distance are

$$\langle f, g \rangle = \int_0^{l_0} f(x)g(x)dx$$



Also,  $\langle f, g \rangle = f(0)g(0) + f'(0)g'(0) + f''(0)g''(0) + f'''(0)g'''(0)$



Etc...

In all these cases, what we have is a symmetric, nonnegative bilinear pairing  $\langle, \rangle$  on a real vector space  $V$ , meaning a bilinear function  $\langle, \rangle: V \times V \rightarrow \mathbb{R}$

Satisfying

$$\langle v, w \rangle = \langle w, v \rangle$$

$$\langle v+v', w \rangle = \langle v, w \rangle + \langle v', w \rangle$$

$$\langle \alpha v, w \rangle = \alpha \langle v, w \rangle$$

$$\langle v, v \rangle \geq 0$$

Not necessarily  
assuming  
 $\langle v, v \rangle = 0 \Rightarrow v = 0$

If  $W \subseteq V$  is a finite dimensional subspace on which  $\langle, \rangle$  is nondegenerate

meaning  $w \in W$  and  $\langle w, w \rangle = 0 \Rightarrow w = 0$ , then we have

an orthogonal projection

operator

$$P: V \rightarrow W$$

that solves the minimization problem. That is, for any  $v \in V$ , we have

$$\|v - P v\| \leq \|v - w\|$$

for any  $w \in W$ .

$P$  has a simple formula:

$$P v := \langle v, e_1 \rangle e_1 + \langle v, e_2 \rangle e_2 + \dots + \langle v, e_r \rangle e_r$$

where  $e_1, e_2, \dots, e_r$  is any orthonormal basis for  $W$ .

What follows is a proof that

- The definition of  $P: V \rightarrow W$  does not depend on the choice of basis  $e_1, e_2, \dots, e_n$  of  $W$
- $P_W$  solves the minimization problem.

To get started let  $B = e_1, \dots, e_n$  be an orthonormal basis for  $W$  and let

$$P_B: V \rightarrow W$$
$$v \mapsto \langle v, e_1 \rangle e_1 + \dots + \langle v, e_n \rangle e_n$$

Note that  $P_B|_W = \text{id}_W$  and so

$$P_B^2 = P_B.$$

Therefore  $V = W \oplus \text{null}(P_B)$

Since every vector  $v \in V$  can be written  $v = P_B v + v - P_B v$  and

$$W \cap \text{null}(P_B) = \{0\}.$$

Check:  $P_B(v - P_B v) = P_B v - P_B^2 v$

$$= P_B v - P_B v = 0 \quad \text{So } v - P_B v \in$$

$\text{null}(P_B)$ . Also,  $w \in W \cap \text{null} P_B$

$$\Rightarrow P_B w = 0 \Rightarrow \text{id}_B w = 0 \Rightarrow w = 0$$

So  $W \cap \text{null} P_B = \{0\}$ .

Define  $W^\perp = \{v \in V : \langle v, w \rangle = 0$   
for all  $w \in W\}$ .

Claim:  $W^\perp = \text{null } P_B$ .

Proof: Suppose  $v \in W^\perp$  and notice  
that  $P_B v = \sum_{i=1}^n \langle v, e_i \rangle e_i = 0$  since  
each  $\langle v, e_i \rangle = 0$ .

On the other hand, suppose  
 $v \in \text{null } (P_B)$ . Then  $P_B v = 0 \Rightarrow$

$$0 = \langle P_B v, P_B v \rangle = \sum_{i=1}^n \langle v, e_i \rangle^2 \Rightarrow$$
$$\langle v, e_i \rangle = 0 \quad \forall i \Rightarrow v \in W^\perp.$$





Conclusion:  $P_B$  doesn't depend  
on the basis  $B$ . In fact,  
 $P_B$  is simply the map  
 $W \oplus W^\perp \xrightarrow{P} W$ .

---

Note, we have the Pythagorean  
theorem: If  $\langle v, w \rangle = 0$  then  
 $\|v+w\|^2 = \|v\|^2 + \|w\|^2$  since  
 $\|v+w\|^2 = \langle v+w, v+w \rangle$   
 $= \langle v, v \rangle + \langle v, w \rangle + \langle w, v \rangle + \langle w, w \rangle$   
 $= \|v\|^2 + \|w\|^2$

---

Let's return to  $P: V \rightarrow W$ .

Theorem: For all  $v \in V, w \in W$   
 $\|v - Pv\| \leq \|v - w\|$ .

Proof: Let  $v \in V$  and  $w \in W$ .

$$\|v - Pv\|^2 \leq \|v - Pv\|^2 + \|Pv - w\|^2$$

$$\Rightarrow \|v - Pv + Pv - w\|^2 = \|v - w\|^2$$

↳ This equality follows from

$v - Pv \in W^\perp$  and  $Pv - w \in W$ .

