

## Lecture 8: Orthogonal projections

We will now come back to our original aim: Given a vector space  $V$ , a subspace  $W$ , and  $v \in V$ .

Find the vector  $w \in W$  which is closest to  $v$ .

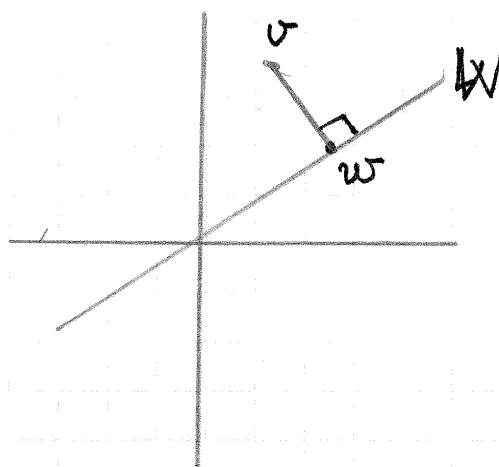
First let us clarify what "closest to" means. The tool to measure distance is the norm, so we want  $\|v - w\|$  to be as small as possible. Thus our problem is:

Find a vector  $w \in W$  such that

$$\|v - w\| \leq \|v - w'\|$$

for all  $w' \in W$ .

Now, let us recall that if  $W = \mathbb{R}w$  is a line, then



The vector  $w$  on the line  $W$  is the one with the property that  $v - w \perp W$ .

We will start by showing that this will always be the case.

Theorem Let  $V$  be a vector space with inner product  $\langle \cdot, \cdot \rangle$ . Let  $W \subset V$  be a subspace and  $v \in V$ . If  $v - w \perp W$  then  $\|v - w\| \leq \|v - u\|$  for all  $u \in W$  and  $\|v - w\| = \|v - u\|$  if and only if  $w = u$ . Thus  $w$  is the member of  $W$  closest to  $v$ .

Proof. First we remark, that  $\|v - w\| \leq \|v - u\|$  if and only if  $\|v - w\|^2 \leq \|v - u\|^2$ . Now we simply calculate

$$\begin{aligned} \|v - u\|^2 &= \| (v - w) + (w - u) \|^2 \\ &= \|v - w\|^2 + \|w - u\|^2 \quad \leftarrow \text{because } v - w \perp W \text{ and } w - u \in W \end{aligned}$$

$$(*) \quad \geq \|v - w\|^2 \quad \leftarrow \text{because } \|w - u\|^2 \geq 0.$$

$$\text{So } \|v - u\| \geq \|v - w\|. \text{ If } \|v - u\|^2 = \|v - w\|^2,$$

then we see - using  $(*)$  - that  $\|w - u\|^2 = 0$ , or

$w = u$ . As  $\|v - w\| = \|v - u\|$  if  $u = w$ , we have

shown that the statement is correct.  $\square$

Theorem Let  $V$  be a vector space with inner product  $(\cdot, \cdot)$ . Let  $W \subset V$  be a subspace, and  $v \in V$ . If  $w \in W$  is closest to  $v$  then  $v - w \perp W$ .

Proof: We know that  $\|v - w\|^2 \leq \|v - u\|^2$  for all  $u \in W$ . Therefore the function  $f: \mathbb{R} \rightarrow \mathbb{R}$

$$F(t) = \|v - w + tx\|^2 \quad (x \in W)$$

has a minimum at  $t=0$ . We have

$$\begin{aligned} F(t) &= (v - w + tx, v - w + tx) \\ &= (v - w, v - w) + t(v - w, x) \\ &\quad + t(x, v - w) + t^2(x, x) \\ &= \|v - w\|^2 + 2t(v - w, x) + t^2\|x\|^2. \end{aligned}$$

Therefore

$$0 = F'(0) = 2(v - w, x).$$

As  $x \in W$  was arbitrary, it follows that  $v - w \perp W$   $\blacksquare$

Our task is now to construct the vector  $w$  such that  $v - w \perp W$ . The idea, how to do that comes from the Gram-Schmidt.

Let  $W = \mathbb{R}u$  and  $v \in V$ . The Gram-Schmidt applied to  $u$  and  $v$  shows, that

$$v - \frac{(v, u)}{\|u\|^2} u \perp W$$

so  $w = \frac{(v, u)}{\|u\|^2} u$  is the vector (point) on the line

$W$  closest to  $v$ . What if the dimension of  $W$  is greater than one? Assume that we have found an orthogonal basis  $v_1, \dots, v_n$  for  $W$ . Applying the Gram-Schmidt to the vectors  $v_1, \dots, v_n$ , shows that

$$w - \sum_{j=1}^n \frac{(w, v_j)}{\|v_j\|^2} v_j$$

is orthogonal to each one of the vectors  $v_1, \dots, v_n$ , and hence - because the inner product is linear in each factor at the line - orthogonal to all vectors of the form  $c_1 v_1 + \dots + c_n v_n = \sum_{j=1}^n c_j v_j$ . But each vector of  $W$  can be written in this way. Thus our vector  $w$  closest to  $v$  is given by

$$w = \sum_{j=1}^n \frac{(v, v_j)}{\|v_j\|^2} v_j$$

Let us look at another motivation; We know that  $w \in W$ . Hence, if  $v_1, \dots, v_n \in W$  is a basis, we can find scalars  $c_1, \dots, c_n \in \mathbb{R}$  such that

$$w = \sum_{k=1}^n c_k v_k.$$

So what are those numbers? As  $v - w \perp v_j$  for  $j = 1, \dots, n$  and  $v_k \perp v_j$  for  $k \neq j$  we get:

$$\begin{aligned} 0 &= (v, v_j) - (w, v_j) \\ &= (v, v_j) - \sum_{k=1}^n c_k (v_k, v_j) \\ &= (v, v_j) - c_j \|v_j\|^2. \end{aligned}$$

Solving for  $c_j$  we get

$$c_j = \frac{(v, v_j)}{\|v_j\|^2}.$$

Thus

$$w = \sum_{j=1}^n \frac{(v, v_j)}{\|v_j\|^2} v_j.$$

We collect this in the following theorem:

We will now prove this directly:

Theorem Let  $V$  be a vector space with inner product  $(\cdot, \cdot)$ .

Let  $W \subset V$  be a subspace and assume that  $v_1, \dots, v_n \in W$  is an orthogonal basis for  $W$ . For  $v \in V$  let

$$w = \sum_{j=1}^n \frac{(v, v_j)}{\|v_j\|^2} v_j \in W. \text{ Then } v - w \perp W \text{ (or)}$$

$w$  is the vector in  $W$  closest to  $v$ .

Proof. We have

$$\begin{aligned} 0 &= (v - w, v_j) = (v, v_j) - (w, v_j) \\ &= (v, v_j) - \sum_{k=1}^n \frac{(v, v_k)}{\|v_k\|^2} (v_k, v_j) \\ &= (v, v_j) - \frac{(v, v_j)}{\|v_j\|^2} \|v_j\|^2 \left\{ \begin{array}{l} (v_k, v_j) = 0 \\ \text{if } k \neq j \end{array} \right. \\ &= (v, v_j) - (v, v_j) = 0 \quad \left\{ \begin{array}{l} (v_j, v_j) = \|v_j\|^2 \end{array} \right. \end{aligned}$$

Hence  $v - w \perp v_j$ . But, as we just saw, this implies, that  $v - w \perp W$  because  $v_1, \dots, v_n$  is a basis  $\square$

Let us now look at what we just did from the point of view of linear maps. What is given in the beginning is a vector space with an inner product and a subspace  $W$ . Then, for each  $v \in V$ , we associated a unique vector  $w \in W$ . Thus we got a map

$$P: V \rightarrow W, v \mapsto w.$$

We even have an explicit formula for  $P(v)$ :

Let (if possible)  $v_1, \dots, v_n$  be an orthogonal basis for  $W$ , then

$$P(v) = \sum_{k=1}^n \frac{(v, v_k)}{\|v_k\|^2} v_k$$

This shows that  $P$  is linear. We saw earlier that if  $v \in W$ , then

$$v = \sum_{k=1}^n \frac{(v, v_k)}{\|v_k\|^2} v_k$$

So  $P(v) = v$  for all  $v \in V$ . In particular we get Lemma  $P^2 = P$ .

The map  $P$  is called the orthogonal projection onto  $W$ .

The projection part comes from  $P^2 = P$  and orthogonal from the fact, that  $v - P(v) \perp W$ .

The result of this discussion is the following:

To find the vector  $w$  closest to  $v$  we have to:

(1) Find (if possible) a basis  $u_1, \dots, u_r$  for  $W$ .

(2) If this is not an orthogonal basis, then we use

Gram-Schmidt to construct an orthonormal basis  $v_1, \dots, v_n$ .

$$(3) \text{ Then } w = \sum_{k=1}^n \frac{(v_i, v_k)}{\|v_k\|^2} v_k.$$

Example: Let  $W$  be the line  $W = \mathbb{R}(1, 2)$ .

Then  $v = (1, 2)$  is a basis (orthogonal!) for  $W$ .

It follows, that the orthogonal projection is given by

$$P(x, y) = \frac{x+2y}{5} (1, 2).$$

Let  $(x, y) = (3, 1)$ . Then (as an example)

$$P(3, 1) = (1, 2).$$

Example Let  $W$  be the line given by  $y = 3x$ .

Then  $(1, 3) \in W$  and hence  $W = \mathbb{R}(1, 3)$ . It follows

that

$$P(x, y) = \frac{x+3y}{10} (1, 3).$$

Example Let  $W$  be the plane generated by the vectors  $(1, 1, 1)$  and  $(1, 0, 1)$ . Find the orthogonal projection  $P: \mathbb{R}^3 \rightarrow W$ .

Solution: We notice first that  $((1, 1, 1), (1, 0, 1)) = 2 \neq 0$

so this is not an orthogonal basis. Using Gram-Schmidt

we get:

$$v_1 = (1, 1, 1)$$

$$v_2 = (1, 0, 1) - \frac{2}{3}(1, 1, 1) = \left(\frac{1}{3}, -\frac{2}{3}, \frac{1}{3}\right) = \frac{1}{3}(1, -2, 1).$$

To avoid fractions, we can use  $(1, -2, 1)$  instead of  $\frac{1}{3}(1, -2, 1)$ . Thus, the orthogonal projection is:

$$P(x, y, z) = \frac{x+y+z}{3} (1, 1, 1) + \frac{x-2y+z}{6} (1, -2, 1)$$

$$= \left( \frac{2x+2y+2z}{6} + \frac{x-2y+z}{6}, \frac{2x+2y+2z}{6} - 2 \frac{x-2y+z}{6}, \frac{x-2y+z}{6} \right)$$

$$\left( \frac{2x+2y+2z}{6} + \frac{x-2y+z}{6} \right)$$

$$= \left( \frac{x+z}{2}, y, \frac{x+z}{2} \right).$$

Example: Let  $W$  be the plane  $\{(x, y, z) \in \mathbb{R}^3 : x+y+2z=0\}$ .

Find the orthogonal projection  $P: \mathbb{R}^3 \rightarrow W$ .

Solution: We notice, that our first step is to find

an orthogonal basis for  $W$ . The vectors  $(1, -1, 0)$

and  $(2, 0, -1)$  are in  $W$ , but are not orthogonal.

but

$$(2, 0, -1) - \frac{3}{2}(1, -1, 0) = (1, 1, -1) \in W$$

and is orthogonal to  $(1, -1, 0)$ . So we get:

$$P(x, y, z) = \frac{x-y}{2} (1, -1, 0) + \frac{x+y-z}{3} (1, 1, -1)$$

$$= \left( \frac{5x-y-2z}{6}, \frac{-x+5y-2z}{6}, \frac{-x-y+z}{3} \right).$$