# Vectors MATH 3512, BCIT

Matrix Methods and Statistics for Geomatics

October 8, 2018

The projection  $u_H$  of a vector u onto a hyperplane H is the vector in the hyperplane that is "most similar" to u. The formal definition for  $u_H$  requires that

- $\mathbf{0}$  u is in H
- $(u u_H)$  is orthogonal to all basis vectors of H

**Example 1: Finding a Projection.** Let H be the line spanned by  $\vec{v} = (-1, 1)^{\mathsf{T}}$  in  $\mathbb{R}^2$ . What is the projection  $\vec{w}$  of  $\vec{u} = (3, -2)^{\mathsf{T}}$ ?



Let  $\vec{w} = (w_1, w_2)^{\mathsf{T}}$ . Then (1)  $\vec{u} - \vec{w}$  is orthogonal to  $\vec{v}$  and (2)  $\vec{w} = \alpha \vec{v}$  for some  $\alpha \in \mathbb{R}$ .

$$\begin{array}{rcl}
 w_1 & - & w_2 & = & 5 \\
 w_1 & + & w_2 & = & 0
 \end{array}$$
 (1)

Cramer's rule tells us that  $\vec{w} = (2.5, -2.5)^{\mathsf{T}}$ .

Let  $u = (u_1, ..., u_n)^{\mathsf{T}}$  be a vector and H be a k-dimensional hyperplane in the vector space  $\mathbb{R}^n$ . Let  $x_1, ..., x_k$  be a basis for H. Then it is true for all vectors v in the hyperplane that

$$||u - v|| \ge ||u - u_H||$$
 (2)

Proof: use the theorem of Pythagoras for

$$||u - v||^2 = ||u - u_H||^2 + ||u_H - v||^2 \ge ||u - u_H||^2$$
 (3)

The claim follows. It illustrates what I mean when I say that  $u_H$  is the vector in H that is most similar to u.

**Example 2: Finding Another Projection.** What is the projection of  $\vec{u} = (5, 2, 10)^{\mathsf{T}}$  onto the plane T characterized by 2x + y + 3z = 0?

First we find two linearly independent vectors in H to form a basis of H, for example  $\vec{v_1} = (1,1,-1)^{\mathsf{T}}$  and  $\vec{v_2} = (0,-3,1)^{\mathsf{T}}$ . The conditions

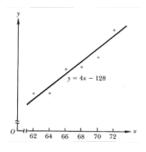
- $\mathbf{0}$   $u_H \in T$
- ②  $(u u_H) \perp v_1$
- **③**  $(u u_H) ⊥ v_2$

give us the system of linear equations

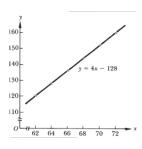
$$\begin{bmatrix} 2 & 1 & 3 \\ -1 & 1 & 1 \\ 0 & 3 & -1 \end{bmatrix} \cdot \begin{bmatrix} \hat{x} \\ \hat{y} \\ \hat{z} \end{bmatrix} = \begin{bmatrix} 0 \\ 3 \\ -4 \end{bmatrix} \tag{4}$$

for which the solution is  $u_H = (\hat{x}, \hat{y}, \hat{z})^{\mathsf{T}} = (-1, -1, 1)^{\mathsf{T}}$ .

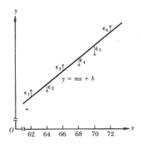
## Least Squares Method



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#### End of Lesson

Next Lesson: TBA