

# Inner Product Spaces, Orthogonality, Projection

## Exercises

1. Let  $X$  and  $Y$  be normed linear spaces. Let  $\{\mathbf{x}_1, \dots, \mathbf{x}_n\}$  be a basis for  $X$  and  $\{\mathbf{y}_1, \dots, \mathbf{y}_m\}$  a basis for  $Y$ . Prove that if  $x_i \perp y_j$  for all  $i \in \{1, \dots, n\}$ ,  $j \in \{1, \dots, m\}$ , then  $X$  and  $Y$  are orthogonal spaces.

Because  $\{\mathbf{x}_1, \dots, \mathbf{x}_n\}$  is a basis for  $X$ , all  $\mathbf{x} \in X$  can be expressed as linear combinations of  $\{\mathbf{x}_1, \dots, \mathbf{x}_n\}$  and similarly for  $\mathbf{y} \in Y$ . Formally, we have

$$\mathbf{x} = \sum_i \alpha_i \mathbf{x}_i$$

for all  $\mathbf{x} \in X$  and arbitrary  $\alpha$  and

$$\mathbf{y} = \sum_j \beta_j \mathbf{y}_j$$

for all  $\mathbf{y} \in Y$  and arbitrary  $\beta$ . We need to show

$$\langle \sum_i \alpha_i \mathbf{x}_i, \sum_j \beta_j \mathbf{y}_j \rangle = 0$$

By the bilinearity of the inner product, we have

$$\begin{aligned} \langle \sum_i \alpha_i \mathbf{x}_i, \sum_j \beta_j \mathbf{y}_j \rangle &= \sum_i \langle \alpha_i \mathbf{x}_i, \sum_j \beta_j \mathbf{y}_j \rangle \\ &= \sum_i \sum_j \langle \alpha_i \mathbf{x}_i, \beta_j \mathbf{y}_j \rangle \\ &= \sum_i \sum_j \alpha_i \beta_j \underbrace{\langle \mathbf{x}_i, \mathbf{y}_j \rangle}_{=0} \end{aligned}$$

■

2. Prove: If a vector  $\alpha$  is in the null space of a set of vectors  $\{\mathbf{x}_1, \dots, \mathbf{x}_n\}$ , then it is orthogonal to the space spanned by  $\{\mathbf{y}_1, \dots, \mathbf{y}_m\}$  where

$$\mathbf{y}_i = \{x_{1i}, \dots, x_{ni}\}$$

Let

$$X = \begin{bmatrix} \mathbf{x}_1 & \cdots & \mathbf{x}_n \end{bmatrix}$$

store the  $\mathbf{x}$  vectors. If  $\boldsymbol{\alpha} \in N(X)$ , then

$$X\boldsymbol{\alpha} = \mathbf{0}$$

or

$$\boldsymbol{\alpha}^T X^T = \mathbf{0}$$

Now let

$$X^T = \begin{bmatrix} \mathbf{y}_1 & \cdots & \mathbf{y}_m \end{bmatrix} = \begin{bmatrix} \mathbf{x}_1 \\ \vdots \\ \mathbf{x}_n \end{bmatrix}$$

We can represent the space spanned by these vectors with  $X^T\boldsymbol{\beta}$  with  $\boldsymbol{\beta}$  taking arbitrary linear combinations of the columns of  $X^T$ . We want to show

$$\boldsymbol{\alpha}^T X^T \boldsymbol{\beta} = \mathbf{0}$$

Because  $\boldsymbol{\alpha}^T X^T = \mathbf{0}$ , this must be the case. ■.

3. Donald Trump tweeted 100 times in April, 150 times in May, and 110 times in June.<sup>1</sup> Let  $\mathbf{b} = (100, 150, 110)$  represent the number of tweets in each month. Project  $\mathbf{b}$  onto the linear space spanned by  $\mathbf{a} = (1, 1, 1)$ . Interpret your result.

We have  $\mathbf{p} = \hat{x}\mathbf{a}$  and  $\mathbf{e} = \mathbf{b} - \mathbf{p}$ . We need  $\mathbf{e}$  to be orthogonal to  $\mathbf{a}$ , or

$$\begin{aligned} \mathbf{a}^T (\mathbf{b} - \hat{x}\mathbf{a}) &= 0 \\ \mathbf{a}^T \mathbf{b} - \hat{x}\mathbf{a}^T \mathbf{a} &= 0 \\ \mathbf{a}^T \mathbf{b} &= \hat{x}\mathbf{a}^T \mathbf{a} \\ \frac{\mathbf{a}^T \mathbf{b}}{\mathbf{a}^T \mathbf{a}} &= \hat{x} \end{aligned}$$

Substituting our values, this becomes

$$\frac{\sum_i b_i}{\sum_i 1} = \frac{100 + 150 + 110}{3} = 120$$

Notice that for  $n$  months of tweeting, this is

$$\frac{1}{n} \sum_i b_i$$

or simply the mean number of tweets.

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<sup>1</sup>Disclaimer: these data are of suspect quality.