

**Math 110**

July 16, 2018

Orthogonal Projections (SOLUTIONS)

1. Since  $V = U \oplus U^\perp$ , write  $v = u + w$  where  $u \in U$  and  $w \in U^\perp$ . By definition  $P_U(v) = u$ . Since  $(U^\perp)^\perp = U$ , then  $P_{U^\perp}(v) = w$ . Therefore  $P_U + P_{U^\perp} = \text{id}_V$ .
2. Let  $u_1 = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \end{pmatrix}$ . Then the projection of  $e_1$  onto  $U$  is

$$(e_1 \cdot u_1)u_1 = \frac{1}{2}(e_1 + e_2)$$

and the projection of  $e_2$  onto  $U$  is the same thing:

$$(e_2 \cdot u_1)u_1 = \frac{1}{2}(e_1 + e_2).$$

Since the columns of the matrix for a  $P_U$  are  $P_U(e_1)$  and  $P_U(e_2)$ , the matrix for  $P_U$  is

$$\begin{pmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} \end{pmatrix}.$$

3. Let

$$u_3 = \frac{1}{\sqrt{3}} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}.$$

Then  $u_3$  is an orthonormal basis for  $U^\perp$ . Similarly to the last question,

$$P_{U^\perp}(e_i) = (e_i \cdot u_3)u_3 = \frac{1}{3} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}.$$

Therefore the matrix for  $P_{U^\perp}$  is

$$\frac{1}{3} \begin{pmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{pmatrix}.$$

By question 1 the matrix for  $P_U$  is

$$\begin{pmatrix} 1 & & \\ & 1 & \\ & & 1 \end{pmatrix} - \frac{1}{3} \begin{pmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{pmatrix} = \begin{pmatrix} \frac{2}{3} & -\frac{1}{3} & -\frac{1}{3} \\ -\frac{1}{3} & \frac{2}{3} & -\frac{1}{3} \\ -\frac{1}{3} & -\frac{1}{3} & \frac{2}{3} \end{pmatrix}.$$

4. (a) The bilinearity properties follow from the fact that  $A$  is linear and the dot product is bilinear. The symmetry property follows from the fact that  $A$  is symmetric. Write

$$x = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

so that

$$\langle x, x \rangle = 4x_1^2 - 4x_1x_2 + 5x_2^2.$$

Complete the square to see that this is a sum of two positive things:

$$4x_1^2 - 4x_1x_2 + 5x_2^2 = (2x_1 - x_2)^2 - x_2^2 + 5x_2^2 = (2x_1 - x_2)^2 + 4x_2^2.$$

Therefore  $\langle x, x \rangle \geq 0$  for all  $x$  and if  $\langle x, x \rangle = 0$ , then  $2x_1 - x_2 = 0$  and  $x_2 = 0$  so that  $x_1 = x_2 = 0$ .

(b) The sets  $\|x\|^2 = c$  are of the form

$$4x_1^2 - 4x_1x_2 + 5x_2^2 = c$$

that is,

$$(2x_1 - x_2)^2 + 4x_2^2 = c.$$

This is a conic section and is bounded (doesn't go to infinity) since  $4x_2^2 < c$  and  $(2x_1 - x_2)^2 < c$ . Therefore it's an ellipse, though the equation might not look so familiar since it's a rotated ellipse.

The following might help to see what's going on. Suppose that you can find a symmetric matrix  $B$  such that  $A = B^2$ . Then  $B = B^*$  and so

$$x \cdot Ax = x \cdot B^2x = Bx \cdot Bx.$$

Therefore the solutions of to the equation

$$x \cdot Ax = c$$

are the same as the those of the equation

$$Bx \cdot Bx = c$$

which are the same as  $B^{-1}y$  for

$$y \cdot y = c.$$

The solutions to the last equation are just circles, and the points  $B^{-1}y$  are circles stretched by the linear map  $B^{-1}$ . Since  $A$  is symmetric with nonnegative eigenvalues, it turns out you can find  $B$  such that  $B^2 = A$ . The easy way is to diagonalize  $A$ :

$$A = P \begin{pmatrix} \lambda_1 & \\ & \lambda_2 \end{pmatrix} P^{-1}$$

for some  $\lambda_1, \lambda_2 > 0$ . Then define

$$B = P \begin{pmatrix} \sqrt{\lambda_1} & \\ & \sqrt{\lambda_2} \end{pmatrix} P^{-1}.$$

We'll cover square roots more later.

(c)  $\|e_1\| = 2$  so  $u_1 = \frac{1}{2}e_1$ . Then

$$\langle e_2, u_1 \rangle = \frac{1}{2}e_2 \cdot Ae_1 = -1$$

so

$$e_2 - \langle e_2, u_1 \rangle u_1 = e_2 + \frac{1}{2}e_1$$

so

$$u_2 = \frac{e_2 + \frac{1}{2}e_1}{2} = \frac{1}{4}e_1 + \frac{1}{2}e_2.$$

5. The vector  $x = (1, 0)$  would be a nonzero vector such that  $\langle x, x \rangle = 0$ .
6. The idea is to move  $P$  to the origin in a way that preserves distances, perform an orthogonal projection, and then move  $P$  back. Namely, let  $u$  be such that  $P + u$  passes through 0. Let  $w$  be the closest point to  $v + u$  on  $P + u$ . Then  $w - u$  is the closest point to  $v$  on  $P$ . Therefore the point on  $P$  closest to  $v$  is  $P_{P+u}(v + u) - u$ .

7. Let

$$x = \begin{pmatrix} 1 \\ 3 \\ 2 \\ 1 \\ 3 \\ 2 \\ 2 \\ 4 \\ 4 \end{pmatrix}, \quad y = \begin{pmatrix} 2 \\ 7 \\ 5 \\ 2 \\ 5 \\ 6 \\ 3 \\ 8 \\ 7 \end{pmatrix}.$$

Orthogonally project  $y$  onto  $\text{Span}(x)$ :

$$P_{\text{Span}(x)}(y) = ax.$$

The number  $a$  is the number that minimizes

$$\sum_{i=1}^9 (ax_i - y_i)^2$$

and therefore is the constant in the desired linear relation. Normalize  $x$  to get  $u$ , an orthonormal basis for  $\text{Span}(x)$ :

$$u = \frac{1}{8} \begin{pmatrix} 1 \\ 3 \\ 2 \\ 1 \\ 3 \\ 2 \\ 2 \\ 4 \\ 4 \end{pmatrix}$$

$$P_{\text{Span}(x)}(y) = (y \cdot u)u.$$

This turns out to be  $2x$ , so  $a$  indeed is 2 exactly.

8. Let  $v_i$  be the  $i$ th column of  $M$  so that  $M$  can be represented by the list of vectors  $(v_1, \dots, v_n)$  viewed as column vectors placed next to each other. Perform Gram-Schmidt on  $(v_1, \dots, v_n)$  to get  $(u_1, \dots, u_n)$ . Then  $v_k$  is in the span of  $(u_1, \dots, u_k)$  so that

$$v_k = (v_k \cdot u_1)u_1 + (v_k \cdot u_2)u_2 + \cdots + (v_k \cdot u_k)u_k.$$

Therefore

$$(v_1, \dots, v_n) = (u_1, \dots, u_n)R$$

where the  $k$ th column of  $R$  is

$$\begin{pmatrix} v_k \cdot u_1 \\ v_k \cdot u_2 \\ \vdots \\ v_k \cdot u_k \\ 0 \\ 0 \\ \vdots \\ 0 \end{pmatrix}.$$

Let  $Q$  be the matrix whose  $i$ th column is  $u_i$ . Then  $M = QR$ .