

## Math 218D Problem Session

Week 7

### 1. Orthogonal matrices

A *orthogonal matrix* is a *square* matrix  $Q$  whose columns form an *orthonormal* set. Alternately, it is a square matrix  $Q$  such that  $Q^T Q = I_n$ .

- a) Is  $\begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}$  an orthogonal matrix?
- b) Is  $\begin{pmatrix} 2 & 1 \\ -2 & 1 \end{pmatrix}$  an orthogonal matrix?
- c) Is  $\begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$  an orthogonal matrix?

## 2. Rotation and reflection

A rotation matrix  $R_\theta = \begin{pmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{pmatrix}$  is an example of an orthogonal matrix.

- a) Confirm that  $R_\theta$  is an orthogonal matrix by checking  $R_\theta^T R_\theta = I_2$ .
- b) Draw the vectors  $R_{\pi/6} \begin{pmatrix} 1 \\ 0 \end{pmatrix}$  and  $R_{\pi/6} \begin{pmatrix} 1 \\ 1 \end{pmatrix}$ .
- c) Using dot products, compute the angle between the rotated vectors  $R_{\pi/6} \begin{pmatrix} 1 \\ 0 \end{pmatrix}$  and  $R_{\pi/6} \begin{pmatrix} 1 \\ 1 \end{pmatrix}$ . Confirm that this is the same as the angle between the two vectors  $\begin{pmatrix} 1 \\ 0 \end{pmatrix}$  and  $\begin{pmatrix} 1 \\ 1 \end{pmatrix}$ . **This is an example of a general phenomenon: multiplying by an orthogonal matrix preserves angles and lengths.**

Consider a line  $L = \text{Span}\{v\} \subset \mathbf{R}^3$ , and the orthogonal complement plane  $V = L^\perp$ . The reflection matrix for reflection across  $V$  is the orthogonal matrix

$$Q = I_3 - 2P_L,$$

where  $P_L$  is the projection matrix for  $L$ .

- d) When  $L = \text{Span}\{(0, 1, 0)\}$ , compute the reflection matrix  $Q$ . Draw the line  $L$  and the plane  $V$ . Compute and draw the vector  $(1, 1, 0)$ , the projection  $P_L(1, 1, 0)$ , and the reflection  $Q(1, 1, 0)$ .
- e) Confirm that *any* reflection matrix  $Q = I_3 - 2P_L$  is an orthogonal matrix by showing that  $Q^T Q = (I_3 - 2P_L)^T (I_3 - 2P_L)$  equals  $I_3$ .  
**Hint:** Remember that  $P_L^2 = P_L$  and  $P_L^T = P_L$ .

### 3. Gram-Schmidt and QR

The purpose of the Gram-Schmidt process is to replace a basis  $\{v_1, \dots, v_k\}$  of a subspace  $V \subset \mathbf{R}^n$  with an **orthogonal basis** of  $V$  (a basis whose vectors are an orthogonal set).

The vectors  $v_1 = \begin{pmatrix} 1 \\ 2 \\ -2 \end{pmatrix}$ ,  $v_2 = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$  are a basis for a plane  $V \subset \mathbf{R}^3$ . Set

$$u_1 = v_1, \quad u_2 = v_2 - \frac{u_1 \cdot v_2}{u_1 \cdot u_1} u_1.$$

These two vectors are the output of the Gram-Schmidt process.

a) Compute  $\frac{u_1}{\|u_1\|}$  and  $\frac{u_2}{\|u_2\|}$ , and confirm that  $\left\{ \frac{u_1}{\|u_1\|}, \frac{u_2}{\|u_2\|} \right\}$  is an orthonormal set of vectors (you need to compute 3 dot products).

b) We can find the QR decomposition of  $A = \begin{pmatrix} 1 & 1 \\ 2 & 1 \\ -2 & 1 \end{pmatrix}$  by setting

$$Q = \begin{pmatrix} \left| \frac{u_1}{\|u_1\|} \right. & \left| \frac{u_2}{\|u_2\|} \right. \\ \left. \right| & \left. \right| \end{pmatrix},$$

a  $3 \times 2$  matrix. Now,  $A = QR$  for some upper-triangular matrix  $R$ , and you saw a formula for  $R$  in lecture. Here is another way to find  $R$ :

$$R = Q^T A.$$

Use this to compute  $R$ , and confirm that  $A = QR$  by multiplying  $Q$  times  $R$ .

**Note:** The method of finding  $R$  given in lecture is much faster, as it involves only book-keeping your work from finding  $Q$ .

c) Explain why this formula for  $R$  worked, i.e. why  $A = QR$  had to imply that  $Q^T A = R$ .

**Hint:** Multiply both sides of  $A = QR$  by  $Q^T$ . What does  $Q^T Q$  always equal, for a matrix  $Q$  with orthonormal columns?

#### 4. Least squares

We want to find the line  $y = Cx + D$  which best fits the data points  $(1, 3), (2, 2), (-2, 1)$  (in the least-squares sense). If there were a line which was an exact fit, the coefficients  $C$  and  $D$  would solve the equation

$$A \begin{pmatrix} C \\ D \end{pmatrix} = \begin{pmatrix} 3 \\ 2 \\ 1 \end{pmatrix}, \text{ where } A = \begin{pmatrix} 1 & 1 \\ 2 & 1 \\ -2 & 1 \end{pmatrix}.$$

But there is no solution to this, as these 3 data points are not collinear. Instead, we'll find the *least-squares solution*  $\hat{x} = \begin{pmatrix} C \\ D \end{pmatrix}$ , i.e. the solution to

$$A^T A \hat{x} = A^T \begin{pmatrix} 3 \\ 2 \\ 1 \end{pmatrix}.$$

a) Compute  $A^T A$  and  $A^T \begin{pmatrix} 3 \\ 2 \\ 1 \end{pmatrix}$ , and solve for the least-squares solutions  $\hat{x} = \begin{pmatrix} C \\ D \end{pmatrix}$ .

b) Plot the data points and the least-squares line  $y = Cx + D$ .

c) What do the numbers in the vector  $A\hat{x}$  mean?

d) Compute the *error*  $\left\| A\hat{x} - \begin{pmatrix} 3 \\ 2 \\ 1 \end{pmatrix} \right\|^2$ .

e) You already found the *QR* decomposition for this matrix  $A$  in problem 2. Solve the equation

$$R\hat{x} = Q^T \begin{pmatrix} 3 \\ 2 \\ 1 \end{pmatrix},$$

and confirm that this  $\hat{x}$  is the same vector you found in part a).

## 5. Another Gram–Schmidt

- a) Apply the Gram–Schmidt process to the vectors  $v_1 = (1, 1, 0)$ ,  $v_2 = (1, 0, 1)$ ,  $v_3 = (0, 1, 1)$  to obtain an orthogonal set  $u_1, u_2, u_3$ .

(Recall that  $u_1 = v_1$ ,  $u_2 = v_2 - \frac{u_1 \cdot v_2}{u_1 \cdot u_1} u_1$ ,  $u_3 = v_3 - \frac{u_1 \cdot v_3}{u_1 \cdot u_1} u_1 - \frac{u_2 \cdot v_3}{u_2 \cdot u_2} u_2$ .)

- b) Find the QR decomposition of  $A = \begin{pmatrix} 1 & 1 & 0 \\ 1 & 0 & 1 \\ 0 & 1 & 1 \end{pmatrix}$ .

- c) Consider the vector  $b = (1, 1, 1)$ . Since  $\{u_1, u_2, u_3\}$  is a basis for  $\mathbf{R}^3$ , there are scalars  $x_1, x_2, x_3$  such that  $b = x_1 u_1 + x_2 u_2 + x_3 u_3$ . Solve for these scalars by taking the dot product of this equation with each of  $u_1, u_2, u_3$ , giving 3 equations

$$b \cdot u_i = (x_1 u_1 + x_2 u_2 + x_3 u_3) \cdot u_i \text{ for } i = 1, 2, 3.$$

(These equations simplify dramatically when you compute the dot products.)

- d) Explain how you could instead solve for these scalars using the formula  $QQ^T = P_{\mathbf{R}^3} = I_3$ .

**Hint:** First,  $Q(Q^T b) = b$ . Second,  $Q \begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix} = (a_1 \frac{u_1}{\|u_1\|} + a_2 \frac{u_2}{\|u_2\|} + a_3 \frac{u_3}{\|u_3\|})$ .