

*Read:* Strang, Section 4.3, 4.4.

*Suggested short conceptual exercises:* Strang, Section 4.3, #12–16, 25, 26, 29. Section 4.4, #3, 4, 8, 9, 19.

1. For this problem, consider the four data points  $(t_i, b_i) = (0, 0), (1, 8), (3, 8),$  and  $(4, 20)$ . Let  $\mathbf{t} = (0, 1, 3, 4)$  be the vector of inputs and  $\mathbf{b} = (0, 8, 8, 20)$  the vector of outputs. Feel free to use a computer to solve *any* systems of equations you encounter throughout this problem.
  - (a) If there were a straight line  $b = C + Dt$  through these four points, then a certain equation  $\mathbf{A}\mathbf{x} = \mathbf{b}$  would have a solution, where  $\mathbf{x} = (C, D)$ . Write this equation in matrix form (that is, find  $\mathbf{A}$ ).
  - (b) Instead, we wish to find the “best fit” line, which means we need to solve  $\mathbf{A}\hat{\mathbf{x}} = \mathbf{p}$ , where  $\mathbf{p}$  is the projection of  $\mathbf{b}$  onto the column space of  $\mathbf{A}$ . Write down the *normal equations*  $\mathbf{A}^T\mathbf{A}\hat{\mathbf{x}} = \mathbf{A}^T\mathbf{b}$ , where  $\hat{\mathbf{x}} = (\hat{C}, \hat{D})$ , and solve for  $\hat{\mathbf{x}}$ .
  - (c) Check that  $\mathbf{e} = \mathbf{b} - \mathbf{p}$  is orthogonal to both columns of  $\mathbf{A}$ . Compute  $\|\mathbf{e}\|$ , which is the shortest distance from  $\mathbf{b}$  to the column space of  $\mathbf{A}$ . Sketch a diagram of  $\mathbf{e}, \mathbf{b}, \mathbf{p}$ , and the orthogonal subspaces  $C(\mathbf{A})$  and  $N(\mathbf{A}^T)$  to illustrate this.
  - (d) Plot the four data points in  $\mathbb{R}^2$  (on the  $tb$ -plane) and sketch the best fit line through them that you just found. Clearly mark what the vectors  $\mathbf{b} = (b_1, b_2, b_3, b_4), \mathbf{e} = (e_1, e_2, e_3, e_4)$ , and  $\mathbf{p} = (p_1, p_2, p_3, p_4)$  represent.
  - (e) Write down  $E := \|\mathbf{A}\mathbf{x} - \mathbf{b}\|^2$  as a sum of four squares—the last one is  $(C + 4D - 20)^2$ , and compute  $\partial E/\partial C$  and  $\partial E/\partial D$ . Set these derivatives equal to zero, and obtain scalars of the normal equations  $\mathbf{A}^T\mathbf{A}\hat{\mathbf{x}} = \mathbf{A}^T\mathbf{b}$ .
  - (f) The method above found the best fit degree-1 polynomial (line). Now, find the best fit degree-0 polynomial (horizontal line)  $b = C$ . Note that this will be a  $4 \times 1$  system instead of a  $4 \times 2$  system. Compute the vectors  $\mathbf{p}$  and  $\mathbf{e}$ , and the (squared) error  $\|\mathbf{e}\|^2$ .
  - (g) Find the best fit parabola (degree-2 polynomial)  $b = C + Dt + Et^2$ . On a new set of axes, plot the four data points and this parabola. Compute the vectors  $\mathbf{p}$  and  $\mathbf{e}$ , and the (squared) error  $\|\mathbf{e}\|^2$ .
  - (h) Find the best fit cubic (degree-3 polynomial)  $b = C + Dt + Et^2 + Ft^3$ . On a new set of axes, plot the four data points and this cubic. Compute the vectors  $\mathbf{p}$  and  $\mathbf{e}$ , and the (squared) error  $\|\mathbf{e}\|^2$ .
2. In this problem we will prove that orthonormal vectors are linearly independent two different ways.
  - (a) Vector proof: First, suppose that  $c_1\mathbf{q}_1 + c_2\mathbf{q}_2 + \cdots + c_k\mathbf{q}_k = \mathbf{0}$ . Show that each  $c_i = 0$ . [*Hint:* Start by multiplying both sides of the equation by  $\mathbf{q}_i^T$ .]
  - (b) Matrix proof: Let  $\mathbf{Q}$  be the matrix whose columns are the  $\mathbf{q}_i$ 's. Show that if  $\mathbf{Q}\mathbf{x} = \mathbf{0}$ , then  $\mathbf{x} = \mathbf{0}$ . [*Hint:* Since  $\mathbf{Q}$  need not be square, you cannot assume  $\mathbf{Q}^{-1}$  exists, but  $\mathbf{Q}^T$  of course will.]

3. For each of the following, answer either *true* (with a reason) or *false* (with a counterexample).
- If  $\mathbf{Q}$  is an orthogonal matrix, then  $\mathbf{Q}^{-1}$  is orthogonal.
  - If  $\mathbf{Q}$  is an orthogonal matrix, then  $\mathbf{Q}^T$  is orthogonal.
  - If  $\mathbf{Q}_1$  and  $\mathbf{Q}_2$  are orthogonal matrices, then  $\mathbf{Q}_1\mathbf{Q}_2$  is orthogonal.
  - If  $\mathbf{Q}$  is a matrix with orthonormal columns (need not be square), then  $\|\mathbf{Q}\mathbf{x}\| = \|\mathbf{x}\|$  for every  $\mathbf{x}$ .
4. What multiple of  $\mathbf{A} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$  should be subtracted from  $\mathbf{b} = \begin{bmatrix} 6 \\ 0 \end{bmatrix}$  to make the resulting vector  $\mathbf{B}$  orthogonal to  $\mathbf{a}$ ? Sketch a figure showing  $\mathbf{A}$ ,  $\mathbf{b}$ , and  $\mathbf{B}$ . Then normalize  $\mathbf{A}$  and  $\mathbf{B}$  to get an orthonormal set,  $\mathbf{q}_1$  and  $\mathbf{q}_2$ .
5. Let  $\mathbf{a}$ ,  $\mathbf{b}$ , and  $\mathbf{c}$  be the (independent) column vectors of the matrix

$$\mathbf{M} = \begin{bmatrix} 1 & 2 & 4 \\ 0 & 0 & 5 \\ 0 & 3 & 6 \end{bmatrix}.$$

Use the Gram-Schmidt process to produce an orthonormal basis  $\mathbf{q}_1$ ,  $\mathbf{q}_2$ , and  $\mathbf{q}_3$ . Then write  $\mathbf{M} = \mathbf{Q}\mathbf{R}$ , where  $\mathbf{Q}$  is orthogonal and  $\mathbf{R}$  is upper-triangular.

6. Recall that if  $\|\mathbf{u}\| = 1$ , then the rank-1 matrix  $\mathbf{u}\mathbf{u}^T$  is the projection matrix onto  $\mathbf{u}$ . In this case,  $\mathbf{Q} = \mathbf{I} - 2\mathbf{u}\mathbf{u}^T$  is a *reflection matrix*.
- Reflecting twice across the same axis is the identity. Verify that indeed,  $\mathbf{Q}^2 = \mathbf{I}$ .
  - Compute  $\mathbf{Q}\mathbf{u}$ , and simplify this expression as much as possible.
  - Suppose  $\mathbf{v}$  is orthogonal to  $\mathbf{u}$ . Compute  $\mathbf{Q}\mathbf{v}$ , and simplify as much as possible.
  - Describe in plain English which subspace  $\mathbf{Q}$  is reflecting across. Your answer should involve  $\mathbf{u}$ . Include a sketch.
  - Compute the reflection matrix  $\mathbf{Q}_1 = \mathbf{I} - 2\mathbf{u}_1\mathbf{u}_1^T$  where  $\mathbf{u}_1 = (0, 1)$ . Compute  $\mathbf{Q}_1\mathbf{x}_1$ , where  $\mathbf{x}_1 = (a, b)$ , and sketch the vectors  $\mathbf{u}_1$ ,  $\mathbf{x}_1$ , and  $\mathbf{Q}_1\mathbf{x}_1$  in the plane.
  - Compute the reflection matrix  $\mathbf{Q}_2 = \mathbf{I} - 2\mathbf{u}_2\mathbf{u}_2^T$  where  $\mathbf{u}_2 = (0, \sqrt{2}/2, \sqrt{2}/2)$ . Compute  $\mathbf{Q}_2\mathbf{x}_2$ , where  $\mathbf{x}_2 = (1, 1, 1)$ , and sketch the vectors  $\mathbf{u}_2$ ,  $\mathbf{x}_2$ , and  $\mathbf{Q}_2\mathbf{x}_2$  in  $\mathbb{R}^3$ .