## Homework #7

due Tuesday, October 6, at 11:59pm

1. Find all least-squares solutions  $\hat{x}$  of each of the following systems of equations Ax = b, and compute the projection  $\hat{b}$  of b onto Col(A) and the minimum value of  $||A\widehat{x}-b||$ .

$$\mathbf{a)} \begin{pmatrix} 1 & 1 \\ 1 & 0 \\ 0 & 2 \end{pmatrix} x = \begin{pmatrix} 1 \\ 4 \\ 3 \end{pmatrix}$$

**a)** 
$$\begin{pmatrix} 1 & 1 \\ 1 & 0 \\ 0 & 2 \end{pmatrix} x = \begin{pmatrix} 1 \\ 4 \\ 3 \end{pmatrix}$$
 **b)**  $\begin{pmatrix} 1 & 2 & 1 \\ -1 & 1 & 0 \\ 2 & 2 & -1 \\ 4 & 3 & 0 \end{pmatrix} x = \begin{pmatrix} -1 \\ -1 \\ -1 \\ 7 \end{pmatrix}$ 

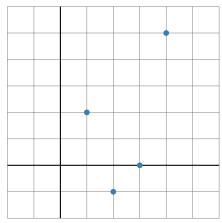
c) 
$$\begin{pmatrix} 2 & 2 & -1 \\ -4 & -5 & 5 \\ 6 & 1 & 12 \end{pmatrix} x = \begin{pmatrix} -6 \\ -24 \\ -3 \end{pmatrix}$$
 d)  $\begin{pmatrix} 3 & 0 \\ 1 & -2 \\ 3 & 1 \end{pmatrix} x = \begin{pmatrix} 9 \\ 7 \\ 7 \end{pmatrix}$ 

$$\mathbf{d}) \begin{pmatrix} 3 & 0 \\ 1 & -2 \\ 3 & 1 \end{pmatrix} x = \begin{pmatrix} 9 \\ 7 \\ 7 \end{pmatrix}$$

**2.** Consider the data points

$$p_1 = \begin{pmatrix} 1 \\ 2 \end{pmatrix}$$
  $p_2 = \begin{pmatrix} 2 \\ -1 \end{pmatrix}$   $p_3 = \begin{pmatrix} 3 \\ 0 \end{pmatrix}$   $p_4 = \begin{pmatrix} 4 \\ 5 \end{pmatrix}$ .

a) Find the best-fit line y = Cx + D through these four points, and draw it on the grid below.



- **b)** For each data point  $p_i = (a_i, b_i)$ , draw the error bar from  $(a_i, y(a_i))$  to  $(a_i, b_i)$ .
- c) What is the minimum value of  $\sum_{i=1}^{4} (b_i y(a_i))^2$ ? How do you know?
- **d)** Verify that the vector

$$(2-y(1), -1-y(2), 0-y(3), 5-y(4))$$

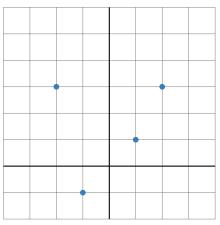
is orthogonal to (1,2,3,4) and (1,1,1,1), and explain why this is necessary.

e) Find the best-fit horizontal line y = D through these four points. Verify that D is the average of the y-values of the data points  $p_1, p_2, p_3, p_4$ .

**3.** Consider the data points

$$p_1 = \begin{pmatrix} -2 \\ 3 \end{pmatrix} \qquad p_2 = \begin{pmatrix} -1 \\ -1 \end{pmatrix} \qquad p_3 = \begin{pmatrix} 1 \\ 1 \end{pmatrix} \qquad p_4 = \begin{pmatrix} 2 \\ 3 \end{pmatrix}.$$

a) Find the best-fit parabola  $y = Cx^2 + Dx + E$  through these four points, and draw it on the grid below.



- **b)** For each data point  $p_i = (a_i, b_i)$ , draw the error bar from  $(a_i, y(a_i))$  to  $(a_i, b_i)$ .
- c) What is the minimum value of  $\sum_{i=1}^{4} (b_i y(a_i))^2$ ? How do you know?
- **4.** Consider the following data points:

$$p_1 = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \qquad p_2 = \begin{pmatrix} 1 \\ -1 \\ -2 \end{pmatrix} \qquad p_3 = \begin{pmatrix} -1 \\ 1 \\ 3 \end{pmatrix} \qquad p_4 = \begin{pmatrix} -1 \\ -1 \\ 1 \end{pmatrix}.$$

- a) Find the best-fit plane z = Cx + Dy + E through these four points.
- b) What least-squares quantity is being minimized, and why?
- **5.** Consider the data points  $p_1, \ldots, p_8$ :

$$\begin{pmatrix}1\\3\end{pmatrix},\begin{pmatrix}3\\1.5\end{pmatrix},\begin{pmatrix}2.5\\0\end{pmatrix},\begin{pmatrix}1\\-1\end{pmatrix},\begin{pmatrix}-.5\\-1\end{pmatrix},\begin{pmatrix}-2\\0\end{pmatrix},\begin{pmatrix}-2.5\\2\end{pmatrix},\begin{pmatrix}-1.5\\3.5\end{pmatrix}.$$

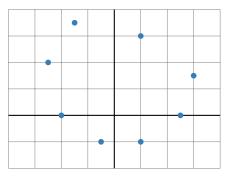
In this problem, I recommend using linalg.js to do the computations. To solve a normal equation  $A^TAx = A^T(1,2,3)$ , you would use something like

A.transpose.mult(A).solve(A.transpose.apply(vec(1,2,3)))

a) Find the best-fit ellipse

$$x^2 + By^2 + Cxy + Dx + Ey + F = 0$$

through these data points, and draw it on the grid below. (You can use a computer graphing program.)



b) What least-squares quantity is being minimized?

[Hint: you can't see it on the graph above, but you can see it on this demo.] *Remark:* Carl Friedrich Gauss (1777–1865), arguably the greatest mathematician since antiquity, kept food on the table by doing astronomical calculations. He invented much of the linear algebra you are learning in order to compute the trajectories of celestial bodies. Essentially performing the calculations in this problem, he correctly predicted the (elliptical) orbit of the asteroid Ceres as it passed behind the sun in 1801.

- **6.** Suppose that  $\hat{x}$  is a vector such that  $A\hat{x} = (1, 1, -1, -1)$ . Explain why  $\hat{x}$  is not a least-squares solution of Ax = (1, 1, 1, 1).
- 7. For each set of vectors, decide if they are orthogonal, orthonormal, or neither; then compute  $Q^TQ$ , where Q is the matrix with the vectors as columns.

a) 
$$\left\{ \begin{pmatrix} 1\\2\\-1 \end{pmatrix}, \begin{pmatrix} -2\\1\\0 \end{pmatrix} \right\}$$
 b)  $\left\{ \frac{1}{2} \begin{pmatrix} 1\\1\\1\\1 \end{pmatrix}, \frac{1}{2} \begin{pmatrix} 1\\1\\-1\\-1 \end{pmatrix}, \frac{1}{\sqrt{2}} \begin{pmatrix} 0\\0\\1\\-1 \end{pmatrix} \right\}$  c)  $\left\{ \frac{1}{3} \begin{pmatrix} 1\\2\\2 \end{pmatrix}, \frac{1}{3} \begin{pmatrix} -2\\1\\2 \end{pmatrix} \right\}$  d)  $\left\{ \frac{1}{3} \begin{pmatrix} 1\\2\\2 \end{pmatrix}, \frac{1}{\sqrt{2}} \begin{pmatrix} 0\\1\\-1 \end{pmatrix}, \frac{1}{\sqrt{18}} \begin{pmatrix} -4\\1\\1 \end{pmatrix} \right\}$ 

**8.** The following subspaces V are given as the span of an *orthogonal* set of vectors. For each subspace V and vector b, compute the orthogonal projection  $b_V$  using the

projection formula.

a) 
$$V = \operatorname{Span} \left\{ \begin{pmatrix} 0 \\ 1 \\ 2 \end{pmatrix}, \begin{pmatrix} 2 \\ -2 \\ 1 \end{pmatrix} \right\}$$

$$b = \begin{pmatrix} 3 \\ 8 \\ 1 \end{pmatrix}$$
b) 
$$V = \operatorname{Span} \left\{ \begin{pmatrix} 1 \\ 2 \\ 1 \end{pmatrix}, \begin{pmatrix} -1 \\ 1 \\ -1 \end{pmatrix} \right\}$$

$$c) \quad V = \operatorname{Span} \left\{ \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \\ -1 \\ -1 \end{pmatrix}, \begin{pmatrix} 1 \\ -1 \\ 1 \\ -1 \end{pmatrix} \right\}$$

$$d) \quad V = \operatorname{Span} \left\{ \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \end{pmatrix} \right\}$$

$$b = \begin{pmatrix} 4 \\ 2 \\ -4 \\ 2 \end{pmatrix}$$

$$d) \quad V = \operatorname{Span} \left\{ \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \right\}$$

$$b = \begin{pmatrix} 7 \\ -3 \\ 2 \end{pmatrix}$$

$$e) \quad V = \operatorname{Span} \left\{ \begin{pmatrix} 1 \\ 2 \\ 2 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ -1 \end{pmatrix}, \begin{pmatrix} -4 \\ 1 \\ 1 \end{pmatrix} \right\}$$

$$b = \begin{pmatrix} 9 \\ -2 \\ 3 \end{pmatrix}$$

- **9.** For each subspace V of Problem 8, scale the spanning vectors to find an *orthonormal* basis of V, and compute the projection matrix  $P_V$  using the formula  $P_V = QQ^T$ . (Your answers should be exact, in terms of square roots.)
- **10.** Suppose that  $\{u_1, u_2, \dots, u_n\}$  is an orthonormal basis of  $\mathbb{R}^n$ . Use the projection formula to explain why

$$I_n = u_1 u_1^T + u_2 u_2^T + \dots + u_n u_n^T.$$

**11.** Suppose that  $\{u_1, u_2, \dots, u_n\}$  is an orthogonal set of vectors. Give a direct proof that  $\{u_1, u_2, \dots, u_n\}$  is linearly independent by taking the dot product of the vector equation

$$x_1u_1 + x_2u_2 + \dots + x_nu_n = 0$$

with  $u_i$ .

- **12.** Give an example of each of the following, or explain why no such example exists.
  - a) A matrix Q with orthonormal columns, but  $QQ^T \neq I_n$ .
  - b) Two nonzero orthogonal vectors that are linearly dependent.
  - **c)** An orthonormal basis for the plane x + y + z = 0.

**13.** Use the Gram–Schmidt process to find orthogonal bases of the following subspaces.

a) Span 
$$\left\{ \begin{pmatrix} 1\\1\\0 \end{pmatrix}, \begin{pmatrix} 1\\0\\2 \end{pmatrix} \right\}$$
 b) Span  $\left\{ \begin{pmatrix} 1\\-1\\2\\-2 \end{pmatrix}, \begin{pmatrix} 1\\-3\\2\\-1 \end{pmatrix}, \begin{pmatrix} 2\\2\\4\\-1 \end{pmatrix} \right\}$ 

c) Span 
$$\left\{ \begin{pmatrix} 1 \\ -2 \\ 3 \end{pmatrix}, \begin{pmatrix} 2 \\ -5 \\ 1 \end{pmatrix}, \begin{pmatrix} -1 \\ 5 \\ 12 \end{pmatrix} \right\}$$
 d) Nul  $\begin{pmatrix} 1 & 2 & -1 & 4 \\ 3 & 6 & -3 & 12 \end{pmatrix}$ 

**14.** Consider the subspace V = Col(A), where

$$A = \begin{pmatrix} 1 & 1 & 2 \\ -1 & -3 & 2 \\ 2 & 2 & 4 \\ -2 & -1 & -1 \end{pmatrix}.$$

Find an orthonormal basis  $\{u_1, u_2, u_3, u_4\}$  of  $\mathbb{R}^4$  such that  $\{u_1, u_2, u_3\}$  is a basis for V. Your answer should be exact, in terms of square roots.

**15.** For each of the following matrices A and vectors b, find the QR decomposition of A, and find the least-squares solution of Ax = b by back-substitution in  $R\widehat{x} = Q^T b$ . Your answers should be exact, in terms of square roots.

**a)** 
$$A = \begin{pmatrix} 1 & 1 \\ 1 & 0 \\ 0 & 2 \end{pmatrix}, \ b = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$$
 **b)**  $A = \begin{pmatrix} 1 & 1 & 2 \\ -1 & -3 & 2 \\ 2 & 2 & 4 \\ -2 & -1 & -1 \end{pmatrix}, \ b = \begin{pmatrix} 1 \\ 0 \\ 1 \\ 0 \end{pmatrix}$ 

- **16.** In this problem, we use a *QR* decomposition to *quickly* compute the best-fit parabola with specified *y*-values at x = -2, -1, 1, 2, as in Problem 3.
  - a) Compute the matrix A such that the least squares solution of  $A(C,D,E) = (b_1,b_2,b_3,b_4)$  gives the coefficients of the parabola  $y = Cx^2 + Dx + E$  that best fits the data points  $(-2,b_1)$ ,  $(-1,b_2)$ ,  $(1,b_3)$ ,  $(2,b_4)$ . (Presumably you computed this in Problem 3.)
  - **b)** Find the *QR* decomposition of *A*.
  - c) Find the best-fit parabola through the points (-2,3), (-1,-1), (1,1), (2,3) by back-substitution in  $R\hat{x} = Q^T b$ . You should get the same answer as in Problem 3.

Note that we can now repeat part **c**) with new *y*-values in  $O(n^2)$  time.

- **17.** An *orthogonal* matrix is a square matrix with ortho*normal* columns.<sup>1</sup>
  - **a)** If *Q* is an orthogonal matrix, show that *Q* is invertible, that  $Q^{-1} = Q^T$ , and that  $Q^{-1}$  is orthogonal.

<sup>&</sup>lt;sup>1</sup>I am not responsible for this terminology.

- **b)** If  $Q_1$  and  $Q_2$  are orthogonal matrices of the same size, show that  $Q_1Q_2$  is orthogonal.
- **18.** Decide if each statement is true or false, and explain why.
  - a) A least-squares solution  $\hat{x}$  of Ax = b is a solution of  $Ax = b_{Col(A)}$ .
  - **b)** Any solution of  $A^TAx = A^Tb$  is a least-squares solution of Ax = b.
  - **c)** If *A* has full column rank, then Ax = b has exactly one least-squares solution for every *b*.
  - **d)** If Ax = b has at least one least-squares solution for every b, then A has full row rank.
  - e) A matrix with orthogonal columns has full row rank.
  - **f)** If  $\{v_1, \dots, v_n\}$  is a linearly independent set of vectors, then it is orthogonal.
  - g) If Q has orthonormal columns, then the distance from x to y equals the distance from Qx to Qy.
  - **h)** If A = QR is a QR-factorization of a matrix A, then the rows of Q form an orthonormal basis for Col(A).