

1. (JPE, September 1997) Let

$$A = \begin{bmatrix} 3 & 3 \\ 0 & 4 \\ 4 & -1 \end{bmatrix}, \quad \text{and} \quad b = \begin{bmatrix} 2 \\ -2 \\ 1 \end{bmatrix}$$

Use the Gram-Schmidt process to find an orthonormal basis for the column space of  $A$ . Factor  $A$  into a product  $QR$  where  $Q \in \mathbb{R}^{3 \times 2}$  has an orthonormal set of column vectors and  $R \in \mathbb{R}^{2 \times 2}$  is upper triangular. Solve the least squares problem  $Ax = b$ . Compute the norm of the residual vector,  $\|r\|$ .

2. (JPE, May 1998). Given the data (0,1), (3,4) and (6,5), use a QR factorization technique to find the best least squares fit by a linear function. Also, solve the problem via the system of normal equations.

3. (JPE, September 2009). Let

$$A = \begin{bmatrix} 1 & 1 \\ 1 & -2 \\ 1 & 3 \\ 1 & 0 \end{bmatrix}$$

- (a) Find a reduced QR factorization of  $A$  by the Gram-Schmidt process.

- (b) Use the QR factorization from (a) to find the least squares fit by a linear function for (1, -2), (-2, 0), (3, 2) and (0, 3).

4. (JPE, May 2008). Consider the linear least squares problem  $\min_x \|Ax - b\|_2^2$  with

$$A = \begin{bmatrix} 2 \\ 1 \\ 2 \end{bmatrix} \quad \text{and} \quad b = \begin{bmatrix} 5 \\ -1 \\ 0 \end{bmatrix}$$

- (a) Solve the above least squares problem using normal equations.

- (b) Compute the reduced singular value decomposition of  $A$ , then use it to solve the above least squares problem.