

## Assignment 6

**Hand-in Evaluation Deadline: 5:00 pm, December 9th**  
**In-class Evaluation: L1: 2:40 pm - 2:50 pm, December 13th**  
**L2: 9:40 am - 9:50 am, December 13th**

The material in lectures may differ between  $\{L1, L2\}$  on the one hand and  $\{L3, L4\}$  on the other, and therefore the homework assignment will differ for  $\{L1, L2\}$  and  $\{L3, L4\}$ .

It is therefore **not** advisable to go to lecture L3 or L4 for the in-class homework evaluation if you attend L1 or L2!

1. (a) Consider the least squares problem  $A\mathbf{x} = \mathbf{b}$ , where

$$A = \begin{bmatrix} 1 & 2 \\ 2 & -1 \\ 1 & 1 \\ -1 & 0 \end{bmatrix} \text{ and } \mathbf{b} = \begin{bmatrix} 4 \\ 1 \\ 0 \\ -2 \end{bmatrix}.$$

- Is there a unique least squares solution to  $A\mathbf{x} = \mathbf{b}$ ? Explain.
- Find all least squares solutions to  $A\mathbf{x} = \mathbf{b}$ .
- Find the orthogonal projection  $\hat{\mathbf{b}}$  of  $\mathbf{b}$  onto  $\text{Col } A$ .
- Find the orthogonal projection  $\tilde{\mathbf{b}}$  of  $\mathbf{b}$  onto  $\text{Null } A^T$ .

- (b) Let  $B = \begin{bmatrix} 1 & 2 & 3 \\ 2 & -1 & 1 \\ 1 & 1 & 2 \\ -1 & 0 & -1 \end{bmatrix} = [\mathbf{a}_1, \mathbf{a}_2, \mathbf{a}_1 + \mathbf{a}_2]$ , where  $\mathbf{a}_i$  are the columns of the matrix  $A$  above, and  $\mathbf{b}$  is as defined above.

- Is there a unique least squares solution to  $B\mathbf{x} = \mathbf{b}$ ? Explain.
- Find all least squares solutions to  $B\mathbf{x} = \mathbf{b}$ .
- Find the orthogonal projection  $\hat{\mathbf{b}}$  of  $\mathbf{b}$  onto  $\text{Col } B$ .
- Find the orthogonal projection  $\tilde{\mathbf{b}}$  of  $\mathbf{b}$  onto  $\text{Null } B^T$ .

- (c) Let  $C = \begin{bmatrix} 1 & 2 & 4 \\ 2 & -1 & 1 \\ 1 & 1 & 0 \\ -1 & 0 & -2 \end{bmatrix} = [\mathbf{a}_1, \mathbf{a}_2, \mathbf{b}]$ , where  $\mathbf{b}$  is as above.

- Is there a unique least squares solution to  $C\mathbf{x} = \mathbf{b}$ ? Explain.
- Find all least squares solutions to  $C\mathbf{x} = \mathbf{b}$ .
- Find the orthogonal projection  $\hat{\mathbf{b}}$  of  $\mathbf{b}$  onto  $\text{Col } C$ .
- Find the orthogonal projection  $\tilde{\mathbf{b}}$  of  $\mathbf{b}$  onto  $\text{Null } C^T$ .

2. Note that  $\begin{bmatrix} 1 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ -1 \end{bmatrix} = 1 + 0 - 1 = 0$ .

Find the orthogonal projection of  $\mathbf{x} = \begin{bmatrix} 3 \\ -1 \\ -2 \end{bmatrix}$  on  $\text{Span} \left\{ \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ -1 \end{bmatrix} \right\}$ .

3. (a) Find the best least squares fit by a linear function to the data
- |     |    |   |   |   |
|-----|----|---|---|---|
| $x$ | -1 | 0 | 1 | 2 |
| $y$ | 0  | 1 | 3 | 9 |
- (b) Plot your linear function from part (a) along with the data on a coordinate system.
- (c) Find the best least squares fit by a quadratic polynomial.
- (d) Sketch your quadratic function from part (c) along with the data on a coordinate system.
4. (a) Find the projection matrix for projecting a vector on  $\text{Span}\{\mathbf{a}\}$  for a nonzero vector  $\mathbf{a}$ .
- (b) How does this compare with the expression in Theorem 21.3? Explain why this makes sense.
5. Let  $P$  be the projection matrix for finding the orthogonal projection on  $\text{Col } A$ . Show that  $P^2 = P$ , and explain why that makes sense intuitively.

6. Let  $A = \begin{bmatrix} 2 & 1 \\ 1 & 1 \\ 2 & 1 \end{bmatrix}$  and  $\mathbf{b} = \begin{bmatrix} 12 \\ 6 \\ 18 \end{bmatrix}$ .

- (a) Use the Gram-Schmidt Process to find an orthonormal basis for the column space of  $A$ .
- (b) Find a  $QR$ -factorization of  $A$ .
- (c) Solve the least squares problem  $A\mathbf{x} = \mathbf{b}$ .
7. Let  $U$  and  $V$  be  $n \times n$  (square) matrices with orthonormal columns. Show that  $UV$  has orthonormal columns.
8. Let  $A$  be an  $m \times n$  matrix of rank  $n$  and let  $\mathbf{b} \in \mathbb{R}^m$ . Show that if  $Q$  and  $R$  are the matrices derived from applying the Gram-Schmidt Process to the column vectors of  $A$  and

$$\hat{\mathbf{b}} = c_1 \mathbf{q}_1 + c_2 \mathbf{q}_2 + \cdots + c_n \mathbf{q}_n$$

is the orthogonal projection of  $\mathbf{b}$  onto  $\text{Col } A$ , then

- (a)  $\mathbf{c} = Q^T \mathbf{b}$
- (b)  $\hat{\mathbf{b}} = QQ^T \mathbf{b}$
- (c)  $QQ^T = A(A^T A)^{-1} A^T$ .
9. Let  $S$  be a subspace of  $\mathbb{R}^n$ .
- Show that the mapping  $\mathbf{x} \mapsto \hat{\mathbf{x}}$ , where  $\hat{\mathbf{x}}$  is the orthogonal projection of  $\mathbf{x}$  onto  $S$ , is a linear transformation.

10. Prove that  $(S^\perp)^\perp = S$  for any subspace  $S$  of  $\mathbb{R}^n$ .
11. Find the eigenvectors and corresponding eigenspaces of  $A = \begin{bmatrix} 6 & -4 \\ 3 & -1 \end{bmatrix}$ .
12. Show that the eigenvalues of a triangular matrix are the diagonal elements of the matrix.
13. Let  $A$  be an  $n \times n$  matrix. Prove that  $A$  is singular if and only if 0 is an eigenvalue of  $A$ .
14. Let  $T$  be transformation that rotates points in  $\mathbb{R}^3$  about some line through the origin.  $T$  is a linear transformation (you do not have to prove this). Without writing  $A$ , find an eigenvalue of  $A$  and describe the corresponding eigenspace.
15. Suppose  $Q$  and  $R$  are  $n \times n$  matrices, and  $Q$  is invertible. Show that the matrices  $A = QR$  and  $B = RQ$  are similar.  
(This is the basis for the *QR-algorithm*, a numerically stable method for computing the eigenvalues and eigenvectors of a matrix: You iteratively find the *QR*-decomposition of  $A$ , then redefine  $A$  to be  $RQ$ , and repeat.)