

Least-Square-Problem

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6-4 Orthogonal Projections Viewed as Approximations

- If P is a point in 3-dimensional space and W is a plane through the origin, then the point Q in W closest to P is obtained by dropping a perpendicular from P to W .

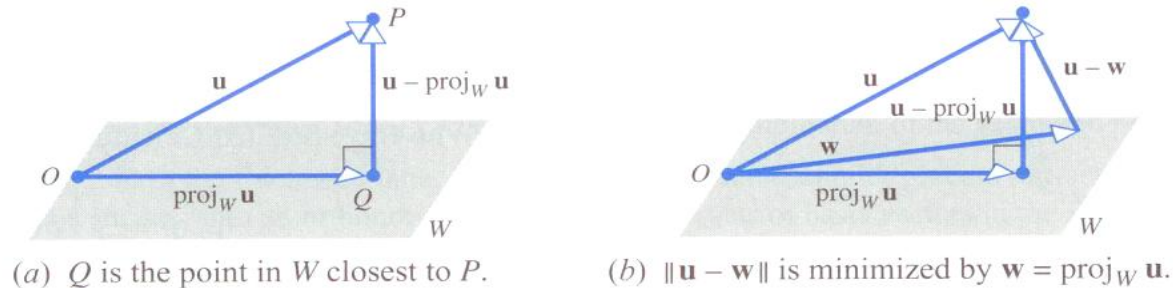


Figure 6.4.1

- If we let $\mathbf{u} = \overrightarrow{OP}$, the distance between P and W is given by $\|\mathbf{u} - \text{proj}_W \mathbf{u}\|$.
- In other words, among all vectors \mathbf{w} in W the vector $\mathbf{w} = \text{proj}_W \mathbf{u}$ minimize the distance $\|\mathbf{u} - \mathbf{w}\|$.

6-4 Best Approximation

- Remark
 - Suppose \mathbf{u} is a vector that we would like to approximate by a vector in W .
 - Any approximation \mathbf{w} will result in an “error vector” $\mathbf{u} - \mathbf{w}$ which, unless \mathbf{u} is in W , cannot be made equal to $\mathbf{0}$.
 - However, by choosing $\mathbf{w} = \text{proj}_W \mathbf{u}$ we can make the length of the error vector $||\mathbf{u} - \mathbf{w}|| = ||\mathbf{u} - \text{proj}_W \mathbf{u}||$ as small as possible.
 - Thus, we can describe $\text{proj}_W \mathbf{u}$ as the “*best approximation*” to \mathbf{u} by the vectors in W .

Theorem 6.4.1

(Best Approximation Theorem)

- If W is a finite-dimensional subspace of an inner product space V , and if \mathbf{u} is a vector in V ,
 - then $\text{proj}_W \mathbf{u}$ is the best approximation to \mathbf{u} from W in the sense that
$$|| \mathbf{u} - \text{proj}_W \mathbf{u} || < || \mathbf{u} - \mathbf{w} ||$$
for every vector \mathbf{w} in W that is different from $\text{proj}_W \mathbf{u}$.

6-4 Least Square Problem

- Given a linear system $A\mathbf{x} = \mathbf{b}$ of m equations in n unknowns
 - find a vector \mathbf{x} , if possible, that minimize $\|A\mathbf{x} - \mathbf{b}\|$ with respect to the Euclidean inner product on \mathbb{R}^m .
 - Such a vector is called a least squares solution of $A\mathbf{x} = \mathbf{b}$.

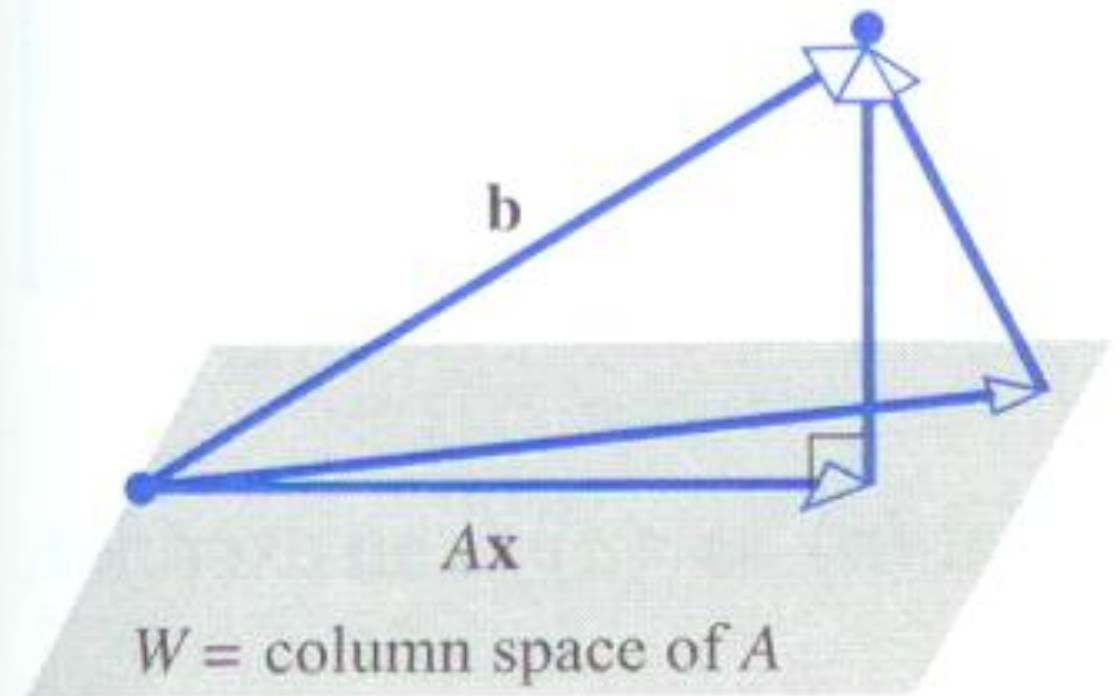


Figure 6.4.2 A least squares solution \mathbf{x} produces the

One way to find a least squares solution of $A\mathbf{x} = \mathbf{b}$ is to calculate the orthogonal projection $\text{proj}_W \mathbf{b}$ on the column space W of A and then solve the equation

$$A\mathbf{x} = \text{proj}_W \mathbf{b} \quad (2)$$

However, we can avoid calculating the projection by rewriting (2) as

$$\mathbf{b} - A\mathbf{x} = \mathbf{b} - \text{proj}_W \mathbf{b}$$

and then multiplying both sides of this equation by A^T to obtain

$$A^T(\mathbf{b} - A\mathbf{x}) = A^T(\mathbf{b} - \text{proj}_W \mathbf{b}) \quad (3)$$

Since $\mathbf{b} - \text{proj}_W \mathbf{b}$ is the component of \mathbf{b} that is orthogonal to the column space of A , it follows from Theorem 4.8.7(b) that this vector lies in the null space of A^T , and hence that

$$A^T (\mathbf{b} - \text{proj}_W \mathbf{b}) = \mathbf{0}$$

Thus, (3) simplifies to

$$A^T (\mathbf{b} - A\mathbf{x}) = \mathbf{0}$$

which we can rewrite as

$$A^T A\mathbf{x} = A^T \mathbf{b} \tag{4}$$

This is called the *normal equation* or the *normal system* associated with $A\mathbf{x} = \mathbf{b}$. When viewed as a linear system, the individual equations are called the *normal equations* associated with $A\mathbf{x} = \mathbf{b}$.

Theorem 6.4.2

- For *any* linear system $A\mathbf{x} = \mathbf{b}$, the associated **normal system**

$$A^T A \mathbf{x} = A^T \mathbf{b}$$

is *consistent*, and all solutions of the normal system are least squares solutions of $A\mathbf{x} = \mathbf{b}$.

Moreover, if W is the column space of A , and \mathbf{x} is any least squares solution of $A\mathbf{x} = \mathbf{b}$, then the orthogonal projection of \mathbf{b} on W is

$$\text{proj}_W \mathbf{b} = A\mathbf{x}$$

(or you can treat it as $A\mathbf{x} - \text{proj}_W \mathbf{b} = \mathbf{0}$)

Theorem 6.4.3

- If A is an $m \times n$ matrix, then the following are equivalent.
 - A has linearly independent column vectors.
 - $A^T A$ is invertible.

Theorem 6.4.4

- If A is an $m \times n$ matrix with linearly independent column vectors,
 - then for every $m \times 1$ matrix \mathbf{b} , the linear system $A\mathbf{x} = \mathbf{b}$ has a unique least squares solution.

- This solution is given by

$$\mathbf{x} = (A^T A)^{-1} A^T \mathbf{b}$$

- Moreover, if W is the column space of A , then the orthogonal projection of \mathbf{b} on W is

$$\text{proj}_W \mathbf{b} = A\mathbf{x} = A(A^T A)^{-1} A^T \mathbf{b}$$

6-4 Example 1 (Least Squares Solution)

- Find the least squares solution of the linear system $A\mathbf{x} = \mathbf{b}$ given by

$$x_1 - x_2 = 4$$

$$3x_1 + 2x_2 = 1$$

$$-2x_1 + 4x_2 = 3$$

and find the orthogonal projection of \mathbf{b} on the column space of A .

- Solution:

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

- Observe that A has linearly independent column vectors, so we know in advance that there is a unique least squares solution.

Example 1

- We have

$$A^T A = \begin{bmatrix} 1 & 3 & -2 \\ -1 & 2 & 4 \end{bmatrix} \begin{bmatrix} 1 & -1 \\ 3 & 2 \\ -2 & 4 \end{bmatrix} = \begin{bmatrix} 14 & -3 \\ -3 & 21 \end{bmatrix}$$

$$A^T \mathbf{b} = \begin{bmatrix} 1 & 3 & -2 \\ -1 & 2 & 4 \end{bmatrix} \begin{bmatrix} 4 \\ 1 \\ 3 \end{bmatrix} = \begin{bmatrix} 1 \\ 10 \end{bmatrix}$$

so the normal system $A^T A \mathbf{x} = A^T \mathbf{b}$ in this case is $\begin{bmatrix} 14 & -3 \\ -3 & 21 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 10 \end{bmatrix}$

- Solving this system yields the least squares solution

$$x_1 = 17/95, x_2 = 143/285$$

- The orthogonal projection of \mathbf{b} on the column space of A is

$$A \mathbf{x} = \begin{bmatrix} 1 & -1 \\ 3 & 2 \\ -2 & 4 \end{bmatrix} \begin{bmatrix} 17/95 \\ 143/285 \end{bmatrix} = \begin{bmatrix} -92/285 \\ 439/285 \\ 94/57 \end{bmatrix}$$

Example 2

(Orthogonal Projection on a Subspace)

- Find the orthogonal projection of the vector $\mathbf{u} = (-3, -3, 8, 9)$ on the subspace of R^4 spanned by the vectors

$$\mathbf{u}_1 = (3, 1, 0, 1), \mathbf{u}_2 = (1, 2, 1, 1), \mathbf{u}_3 = (-1, 0, 2, -1)$$

- Solution:

- The subspace spanned by \mathbf{u}_1 , \mathbf{u}_2 , and \mathbf{u}_3 , is the column space of

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

- If \mathbf{u} is expressed as a column vector, we can find the orthogonal projection of \mathbf{u} on W by finding a least squares solution of the system $A\mathbf{x} = \mathbf{u}$.
- $\text{proj}_W \mathbf{u} = A\mathbf{x}$ from the least square solution.

6-4 Example 2

- From Theorem 6.4.4, the least squares solution is given by

$$\mathbf{x} = (A^T A)^{-1} A^T \mathbf{u}$$

- That is,

$$\mathbf{x} = \left(\begin{bmatrix} 3 & 1 & 0 & 1 \\ 1 & 2 & 1 & 1 \\ -1 & 0 & 2 & -1 \end{bmatrix} \begin{bmatrix} 3 & 1 & -1 \\ 1 & 2 & 0 \\ 0 & 1 & 2 \\ 1 & 1 & -1 \end{bmatrix} \right)^{-1} \begin{bmatrix} 3 & 1 & 0 & 1 \\ 1 & 2 & 1 & 1 \\ -1 & 0 & 2 & -1 \end{bmatrix} \begin{bmatrix} -3 \\ -3 \\ 8 \\ 9 \end{bmatrix} = \begin{bmatrix} -1 \\ 2 \\ 1 \end{bmatrix}$$

- Thus, $\text{proj}_W \mathbf{u} = A\mathbf{x} = [-2 \ 3 \ 4 \ 0]^T$

Assignment-1

► In Exercises 3–6, find the least squares solution of the equation $A\mathbf{x} = \mathbf{b}$. ◀

$$3. A = \begin{bmatrix} 1 & -1 \\ 2 & 3 \\ 4 & 5 \end{bmatrix}; \mathbf{b} = \begin{bmatrix} 2 \\ -1 \\ 5 \end{bmatrix}$$

$$4. A = \begin{bmatrix} 2 & -2 \\ 1 & 1 \\ 3 & 1 \end{bmatrix}; \mathbf{b} = \begin{bmatrix} 2 \\ -1 \\ 1 \end{bmatrix}$$

$$5. A = \begin{bmatrix} 1 & 0 & -1 \\ 2 & 1 & -2 \\ 1 & 1 & 0 \\ 1 & 1 & -1 \end{bmatrix}; \mathbf{b} = \begin{bmatrix} 6 \\ 0 \\ 9 \\ 3 \end{bmatrix}$$

$$6. A = \begin{bmatrix} 2 & 0 & -1 \\ 1 & -2 & 2 \\ 2 & -1 & 0 \\ 0 & 1 & -1 \end{bmatrix}; \mathbf{b} = \begin{bmatrix} 0 \\ 6 \\ 0 \\ 6 \end{bmatrix}$$

17. Find the orthogonal projection of \mathbf{u} on the subspace of R^3 spanned by the vectors \mathbf{v}_1 and \mathbf{v}_2 .

$$\mathbf{u} = (1, -6, 1); \quad \mathbf{v}_1 = (-1, 2, 1), \quad \mathbf{v}_2 = (2, 2, 4)$$

18. Find the orthogonal projection of \mathbf{u} on the subspace of R^4 spanned by the vectors \mathbf{v}_1 , \mathbf{v}_2 , and \mathbf{v}_3 .

$$\begin{aligned} \mathbf{u} &= (6, 3, 9, 6); \quad \mathbf{v}_1 = (2, 1, 1, 1), \quad \mathbf{v}_2 = (1, 0, 1, 1), \\ \mathbf{v}_3 &= (-2, -1, 0, -1) \end{aligned}$$