

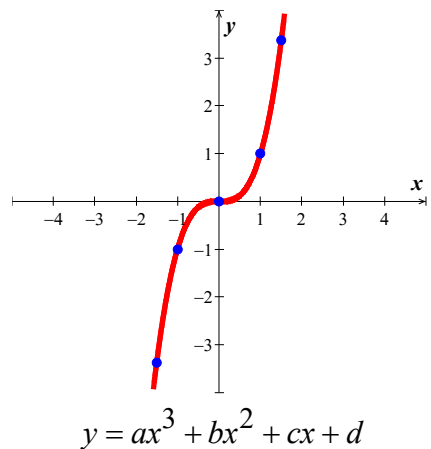
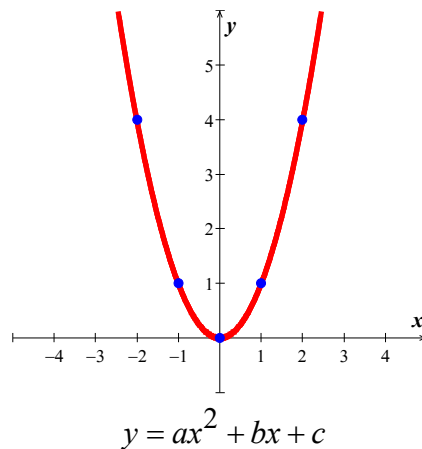
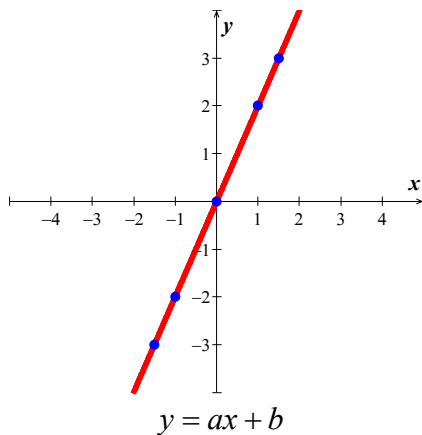
## Section 3.5 – Least Squares Analysis

The use to **best** fit data, we will use results about orthogonal projections in inner product spaces to obtain a technique for fitting a line or other polynomial.

### Fitting a Curve to Data

The common problem is to obtain a mathematical relationship between 2 variables  $x$  and  $y$  by **fitting** a curve to points in the  $xy$ -plane.

Some possibility of fitting the data



### Least Squares Fit of a Straight Line

Recall that a system of equations  $A\vec{x} = \vec{y}$  is called inconsistent if it does not have a solution. Suppose we want to fit a straight line  $y = mx + b$  to the determined points  $(x_1, y_1), \dots, (x_n, y_n)$

If the data points were collinear, the line would pass through all  $n$  points and the unknown coefficients  $m$  and  $b$  would satisfy the equations

$$\begin{array}{l} y_1 = mx_1 + b \\ y_2 = mx_2 + b \\ \vdots \\ y_n = mx_n + b \end{array} \Rightarrow \begin{bmatrix} x_1 & 1 \\ x_2 & 1 \\ \vdots & \vdots \\ x_n & 1 \end{bmatrix} \begin{bmatrix} m \\ b \end{bmatrix} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}$$

$A \quad \vec{x} = \vec{y}$

The problem is to find  $m$  and  $b$  that minimize the errors in some sense.

## Least Square Problem

Given a linear system  $A\vec{x} = \vec{y}$  of  $m$  equations in  $n$  unknowns, find a vector  $\vec{x}$  that minimizes  $\|\vec{y} - A\vec{x}\|$  with respect to the Euclidean inner product on  $\mathbb{R}^m$ . We call such as  $\vec{x}$  a least squares solution of the system, we call  $\vec{y} - A\vec{x}$  the least squares error vectors, and we call  $\|\vec{y} - A\vec{x}\|$  the least squares error.

$$A\mathbf{x} = \begin{pmatrix} e_1 \\ e_2 \\ \vdots \\ e_m \end{pmatrix}$$

The term “*least square solution*” results from the fact the minimizing  $\|\vec{y} - A\vec{x}\| = e_1^2 + e_2^2 + \dots + e_m^2$

### Example

Find the sums of squares of the errors of (2, 4), (4, 8), (6, 6)

#### Solution

$$4 = 2m + b \Rightarrow 4 - 2m - b = e_1$$

$$8 = 4m + b \Rightarrow 8 - 4m - b = e_2$$

$$6 = 6m + b \Rightarrow 6 - 6m - b = e_3$$

$$e_1^2 + e_2^2 + \dots + e_m^2 = (4 - 2m - b)^2 + (8 - 4m - b)^2 + (6 - 6m - b)^2$$

The least squares problem for this example to find the values  $m$  and  $b$  for which  $e_1^2 + e_2^2 + \dots + e_m^2$  is a minimum.

## Theorem

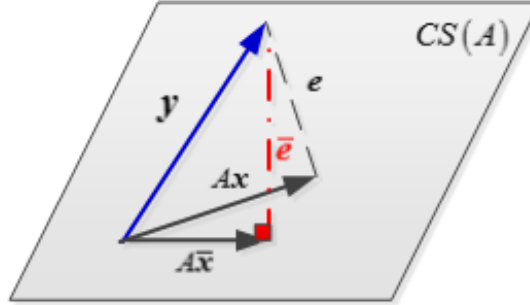
If  $A$  is an  $m \times n$  matrix, the equation  $A\vec{x} = \vec{y}$  has a solution if and only if  $\vec{y}$  is in the column space of  $A$ .

$$\vec{y} - A\vec{x} = \vec{e}$$

$A\vec{x}$  is a vector that is in the column space of  $A$ . For this  $A$  the column space is a plane in  $\mathbb{R}^m$

$\vec{y}$  is a vector, not in the column space of  $A$  (otherwise  $A\vec{x} = \vec{y}$  has an exact solution)

$\vec{e}$  is the error vector, the difference between  $\vec{y}$  and  $A\vec{x}$



The length  $\|\vec{e}\|$  is a *minimum* exactly when  $\vec{e} \perp CS(A)$

## Best Approximation Theorem

If  $CS(A)$  is a finite dimensional subspace of an inner product space, and if  $\vec{y}$  is a vector in  $V$ , then

$proj_{CS(A)} \vec{y}$  is the best approximation to  $\vec{y}$  from  $CS(A)$  in the sense that

$$\left\| \vec{y} - proj_{CS(A)} \vec{y} \right\| < \left\| \vec{y} - \vec{w} \right\|$$

For every vector  $\vec{w}$  in  $CS(A)$  that is different from  $proj_{CS(A)} \vec{y}$

## Theorem

For every linear system  $A\vec{x} = \vec{y}$ , the associated normal system

$$A^T A \vec{x} = A^T \vec{y}$$

is consistent, and all solutions are least squares solutions of  $A\vec{x} = \vec{y}$

If the columns of  $A$  are linearly independent, then  $A^T A$  is invertible so has a unique solution  $\vec{x}$ .

This solution is often expressed theoretically as

$$\left( A^T A \right)^{-1} A^T A \vec{x} = \left( A^T A \right)^{-1} A^T \vec{y}$$

$$\bar{x} = \left( A^T A \right)^{-1} A^T \vec{y}$$

### **Proof**

Let the vector  $\bar{x}$  is a least squares solution to  $A\bar{x} = \vec{y} \Leftrightarrow (\vec{y} - A\bar{x}) \perp CS(A)$

$$(\vec{y} - A\bar{x}) \cdot \vec{z} = 0 \quad \vec{z} \text{ in } CS(A) \quad \& \quad \vec{z} = A\vec{w}$$

$$(\vec{y} - A\bar{x}) \cdot A\vec{w} = 0 \quad \vec{w} \text{ in } \mathbb{R}^n$$

$$A^T (\vec{y} - A\bar{x}) \cdot \vec{w} = 0$$

$$A^T (\vec{y} - A\bar{x}) = 0$$

$$A^T \vec{y} - A^T A\bar{x} = 0$$

$$A^T \vec{y} = A^T A\bar{x}$$

### **Theorem**

If  $A$  is an  $m \times n$  matrix, then the following are equivalent

- a)  $A$  has linearly independent column vectors.
- b)  $A^T A$  is invertible.

### **Example**

Find the equation of the line that best fits the given points in the least-squares sense.

(40, 482), (45, 467), (50, 452), (55, 432), (60, 421)

### **Solution**

Let  $y = mx + b$  be the equation of the line that best fits the given points. Then

$$\begin{pmatrix} 40 & 1 \\ 45 & 1 \\ 50 & 1 \\ 55 & 1 \\ 60 & 1 \end{pmatrix} \begin{pmatrix} m \\ b \end{pmatrix} = \begin{pmatrix} 482 \\ 467 \\ 452 \\ 432 \\ 421 \end{pmatrix}$$

$$\text{Where } A = \begin{pmatrix} 40 & 1 \\ 45 & 1 \\ 50 & 1 \\ 55 & 1 \\ 60 & 1 \end{pmatrix} \quad \mathbf{x} = \begin{pmatrix} m \\ b \end{pmatrix} \quad \mathbf{y} = \begin{pmatrix} 482 \\ 467 \\ 452 \\ 432 \\ 421 \end{pmatrix}$$

Using the normal equation formula:  $A^T Ax = A^T y$

$$\begin{pmatrix} 40 & 45 & 50 & 55 & 60 \\ 1 & 1 & 1 & 1 & 1 \end{pmatrix} \begin{pmatrix} 40 & 1 \\ 45 & 1 \\ 50 & 1 \\ 55 & 1 \\ 60 & 1 \end{pmatrix} \begin{pmatrix} m \\ b \end{pmatrix} = \begin{pmatrix} 40 & 45 & 50 & 55 & 60 \\ 1 & 1 & 1 & 1 & 1 \end{pmatrix} \begin{pmatrix} 482 \\ 467 \\ 452 \\ 432 \\ 421 \end{pmatrix}$$

$$\begin{pmatrix} 12,750 & 250 \\ 250 & 5 \end{pmatrix} \begin{pmatrix} m \\ b \end{pmatrix} = \begin{pmatrix} 111,970 \\ 2,255 \end{pmatrix}$$

$$X = A^{-1}B$$

$$\begin{pmatrix} m \\ b \end{pmatrix} = \frac{1}{1250} \begin{pmatrix} 5 & -250 \\ -250 & 12,750 \end{pmatrix} \begin{pmatrix} 111,970 \\ 2,255 \end{pmatrix}$$

$$= \begin{pmatrix} -3.12 \\ 607 \end{pmatrix}$$

**Or**

$$m = \frac{\begin{vmatrix} 111,970 & 250 \\ 2,255 & 5 \end{vmatrix}}{\begin{vmatrix} 12,750 & 250 \\ 250 & 5 \end{vmatrix}}$$

$$= \frac{-3,900}{1,250}$$

$$= -\frac{78}{25}$$

$$b = \frac{\begin{vmatrix} 12,750 & 111,970 \\ 250 & 2,255 \end{vmatrix}}{1,250}$$

$$= \frac{758,750}{1,250}$$

$$= 607$$

$$\text{Thus, } y = -\frac{78}{25}x + 607 \quad \text{or} \quad y = -3.12x + 607$$

### Example

Given the system equation: 
$$\begin{cases} x_1 - x_2 = 4 \\ 3x_1 + 2x_2 = 1 \\ -2x_1 + 4x_2 = 3 \end{cases}$$

- a) Find the least-squares solution of the linear system  $A\vec{x} = \vec{y}$
- b) Find the orthogonal projection of  $\vec{y}$  on the column space of  $A$
- c) Find the **error vector** and the **error**

### Solution

$$a) \quad A = \begin{pmatrix} 1 & -1 \\ 3 & 2 \\ -2 & 4 \end{pmatrix} \quad \vec{x} = \begin{pmatrix} m \\ b \end{pmatrix} \quad \vec{y} = \begin{pmatrix} 4 \\ 1 \\ 3 \end{pmatrix}$$

$$A^T A \vec{x} = A^T \vec{y}$$

$$\begin{pmatrix} 1 & 3 & -2 \\ -1 & 2 & 4 \end{pmatrix} \begin{pmatrix} 1 & -1 \\ 3 & 2 \\ -2 & 4 \end{pmatrix} \begin{pmatrix} m \\ b \end{pmatrix} = \begin{pmatrix} 1 & 3 & -2 \\ -1 & 2 & 4 \end{pmatrix} \begin{pmatrix} 4 \\ 1 \\ 3 \end{pmatrix}$$

$$\begin{pmatrix} 14 & -3 \\ -3 & 21 \end{pmatrix} \begin{pmatrix} m \\ b \end{pmatrix} = \begin{pmatrix} 1 \\ 10 \end{pmatrix}$$

$$\begin{pmatrix} m \\ b \end{pmatrix} = \frac{1}{285} \begin{pmatrix} 21 & 3 \\ 3 & 14 \end{pmatrix} \begin{pmatrix} 1 \\ 10 \end{pmatrix} \quad X = A^{-1}B$$

$$= \begin{pmatrix} \frac{51}{285} \\ \frac{143}{285} \end{pmatrix}$$

$$= \begin{pmatrix} \frac{17}{95} \\ \frac{143}{285} \end{pmatrix}$$

$$\text{Thus } y = \frac{17}{95}x + \frac{143}{285} \quad \text{or} \quad y = 0.1789x + 0.5018$$

- b) The orthogonal projection of  $\vec{y}$  on the column space of  $A$

$$A\vec{x} = \begin{pmatrix} 1 & -1 \\ 3 & 2 \\ -2 & 4 \end{pmatrix} \begin{pmatrix} \frac{17}{95} \\ \frac{143}{285} \end{pmatrix}$$

$$= \begin{pmatrix} -\frac{92}{285} \\ \frac{439}{285} \\ \frac{94}{57} \end{pmatrix}$$

$$c) \quad \vec{y} - A\vec{x} = \begin{pmatrix} 4 \\ 1 \\ 3 \end{pmatrix} - \begin{pmatrix} -\frac{92}{285} \\ \frac{439}{285} \\ \frac{94}{57} \end{pmatrix}$$

$$= \begin{pmatrix} \frac{1232}{285} \\ -\frac{154}{285} \\ \frac{4}{3} \end{pmatrix}$$

The **error**:  $\|\vec{y} - A\vec{x}\| = \sqrt{\left(\frac{1232}{285}\right)^2 + \left(-\frac{154}{285}\right)^2 + \left(\frac{4}{3}\right)^2}$   
 $\approx 4.556$

## Exercises      Section 3.5 – Least Squares Analysis

(1 – 7) Find the equation of the line that best fits the given points in the least-squares sense and find the error.

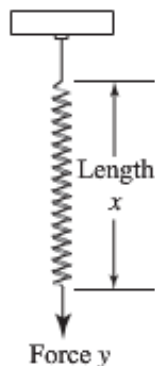
1.  $\{(0, 2), (1, 2), (2, 0)\}$
2.  $\{(1, 5), (2, 4), (3, 1), (4, 1), (5, -1)\}$
3.  $\{(0, 1), (1, 3), (2, 4), (3, 4)\}$
4.  $\{(-2, 0), (-1, 0), (0, 1), (1, 3), (2, 5)\}$
5.  $\{(2, 3), (3, 2), (5, 1), (6, 0)\}$
6.  $\{(-1, 0), (0, 1), (1, 2), (2, 4)\}$
7.  $\{(1, 0), (2, 1), (4, 2), (5, 3)\}$

(8 – 10) Find the orthogonal projection of the vector  $\vec{u}$  on the subspace of  $\mathbb{R}^4$  spanned by the vectors

8.  $\vec{u} = (-3, -3, 8, 9); \vec{v}_1 = (3, 1, 0, 1), \vec{v}_2 = (1, 2, 1, 1), \vec{v}_3 = (-1, 0, 2, -1)$
9.  $\vec{u} = (6, 3, 9, 6); \vec{v}_1 = (2, 1, 1, 1), \vec{v}_2 = (1, 0, 1, 1), \vec{v}_3 = (-2, -1, 0, -1)$
10.  $\vec{u} = (-2, 0, 2, 4); \vec{v}_1 = (1, 1, 3, 0), \vec{v}_2 = (-2, -1, -2, 1), \vec{v}_3 = (-3, -1, 1, 3)$

11. Find the standard matrix for the orthogonal projection  $P$  of  $\mathbb{R}^2$  on the line passes through the origin and makes an angle  $\theta$  with the positive  $x$ -axis.

12. Hooke's law in physics states that the length  $x$  of a uniform spring is a linear function of the force  $y$  applied to it. If we express the relationship as  $y = mx + b$ , then the coefficient  $m$  is called the spring constant.



Suppose a particular unstretched spring has a measured length of 6.1 inches.(i.e.,  $x = 6.1$  when  $y = 0$  ). Forces of 2 pounds, 4 pounds, and 6 pounds are then applied to the spring, and the corresponding lengths are found to be 7.6 inches, 8.7 inches, and 10.4 inches. Find the spring constant.



13. Prove: If  $A$  has a linearly independent column vectors, and if  $\vec{b}$  is orthogonal to the column space of  $A$ , then the least squares solution of  $A\vec{x} = \vec{b}$  is  $\vec{x} = \vec{0}$ .
14. Let  $A$  be an  $m \times n$  matrix with linearly independent row vectors. Find a standard matrix for the orthogonal projection of  $\mathbb{R}^n$  onto the row space of  $A$ .
15. Let  $W$  be the line with parametric equations  $x = 2t, \quad y = -t, \quad z = 4t$
- Find a basis for  $W$ .
  - Find the standard matrix for the orthogonal projection on  $W$ .
  - Use the matrix in part (b) to find the orthogonal projection of a point  $P_0(x_0, y_0, z_0)$  on  $W$ .
  - Find the distance between the point  $P_0(2, 1, -3)$  and the line  $W$ .
16. In  $\mathbb{R}^3$ , consider the line  $l$  given by the equations  $x = t, \quad y = t, \quad z = t$   
 And the line  $m$  given by the equations  $x = s, \quad y = 2s - 1, \quad z = 1$   
 Let  $P$  be the point on  $l$ , and let  $Q$  be a point on  $m$ .  
 Find the values of  $t$  and  $s$  that minimize the distance between the lines by minimizing the squared distance  $\|P - Q\|^2$
17. Determine whether the statement is true or false,
- If  $A$  is an  $m \times n$  matrix, then  $A^T A$  is a square matrix.
  - If  $A^T A$  is invertible, then  $A$  is invertible.
  - If  $A$  is invertible, then  $A^T A$  is invertible.
  - If  $A\vec{x} = \vec{b}$  is a consistent linear system, then  $A^T A\vec{x} = A^T \vec{b}$  is also consistent.
  - If  $A\vec{x} = \vec{b}$  is an inconsistent linear system, then  $A^T A\vec{x} = A^T \vec{b}$  is also inconsistent.
  - Every linear system has a least squares solution.
  - Every linear system has a unique least squares solution.
  - If  $A$  is an  $m \times n$  matrix with linearly independent columns and  $\vec{b}$  is in  $\mathbb{R}^m$ , then  $A\vec{x} = \vec{b}$  has a unique least squares solution.
18. A certain experiment produces the data  $\{(1, 1.8), (2, 2.7), (3, 3.4), (4, 3.8), (5, 3.9)\}$ .  
 Find the function that it will fit these data in the form of  $y = \beta_1 x + \beta_2 x^2$

19. According to Kepler's first law, a comet should have an ellipse, parabolic, or hyperbolic orbit (with gravitational attractions from the planets ignored). In suitable polar coordinates, the position  $(r, \nu)$  of a comet satisfies an equation of the form

$$r = \beta + e(r \cdot \cos \nu)$$

Where  $\beta$  is a constant and  $e$  is the eccentricity of the orbit, with  $0 \leq e < 1$  for an ellipse,  $e = 1$  for a parabolic, and  $e > 1$  for a hyperbola.

Suppose observations of a newly discovered comet provide the data below.

$\nu$	.88	1.10	1.42	1.77	2.14
$r$	3.00	2.30	1.65	1.25	1.01

Determine the type of orbit, and predict where the orbit will be when  $\nu = 4.6$  (*radians*)?

20. To measure the takeoff performance of an airplane, the horizontal position of the plane was measured every second, from  $t = 0$  to  $t = 12$

The position (in *feet*) were:

0, 8.8, 29.9, 62.0, 104.7, 159.1, 222.0, 294.5, 380.4, 471.1, 571.7, 686.8, and 809.2

- a) Find the least square cubic curve  $y = \beta_0 + \beta_1 t + \beta_2 t^2 + \beta_3 t^3$  for these data.
- b) Estimate the velocity of the plane when  $t = 4.5$  *sec*, using the result from part (a).