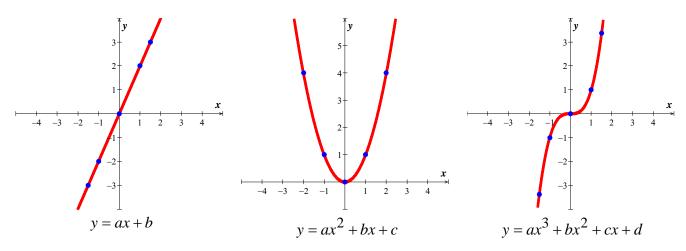
# Section 3.5 – Least Squares Analysis

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  $(x_1, y_1), ..., (x_n, y_n)$ 

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

$$y_{1} = mx_{1} + b$$

$$y_{2} = mx_{2} + b$$

$$\vdots \quad \vdots \quad \vdots$$

$$y_{n} = mx_{n} + b$$

$$\Rightarrow \begin{bmatrix} x_{1} & 1 \\ x_{2} & 1 \\ \vdots & \vdots \\ x_{n} & 1 \end{bmatrix} \begin{bmatrix} m \\ b \end{bmatrix} = \begin{bmatrix} y_{1} \\ y_{2} \\ \vdots \\ y_{n} \end{bmatrix}$$

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

The problem is to find m and b that minimize the errors is 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.

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

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

### **Example**

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

#### **Solution**

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

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

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

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

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

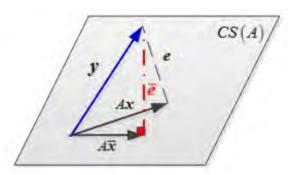
#### **Theorem**

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 is  $\mathbb{R}^m$ 

 $\vec{y}$  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  $\vec{v}$  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  $\vec{V}$ , then  $proj_{CS(A)} \vec{y}$  is the best approximation to  $\vec{y}$  from CS(A) is the sense that

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

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

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

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

If the columns of A are linearly independent, then  $A^TA$  is invertible so has a unique solution  $\overline{x}$ . This solution is often expressed theoretically as

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

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

# **Proof**

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

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

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

$$(\vec{y} - A\overline{x}) \cdot A\vec{w} = 0$$

$$\vec{w}$$
 in  $\mathbb{R}^{I}$ 

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

$$A^T \left( \vec{y} - A \overline{x} \right) = 0$$

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

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

#### **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

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

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

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

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

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

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

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

$$= \binom{-3.12}{607}$$

**Or** 

$$m = \frac{\begin{vmatrix} 111,970 & 250 \\ 2,255 & 5 \end{vmatrix}}{\begin{vmatrix} 12,750 & 250 \\ 250 & 5 \end{vmatrix}}$$
$$= \frac{-3,900}{1,250}$$
$$= -\frac{78}{25}$$

$$b = \frac{\begin{vmatrix} 12,750 & 111,970 \\ 250 & 2,255 \end{vmatrix}}{1,250}$$
$$= \frac{758,750}{1,250}$$
$$= \frac{607}{1}$$

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

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

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

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

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

$$\begin{pmatrix} m \\ b \end{pmatrix} = \frac{1}{285} \begin{pmatrix} 21 & 3 \\ 3 & 14 \end{pmatrix} \begin{pmatrix} 1 \\ 10 \end{pmatrix}$$

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

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

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

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

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

≈ 4.556

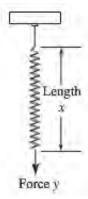
# **Exercises** Section 3.5 – Least Squares Analysis

(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.**  $\vec{u} = (-3, -3, 8, 9); \quad \vec{v}_1 = (3, 1, 0, 1), \quad \vec{v}_2 = (1, 2, 1, 1), \quad \vec{v}_3 = (-1, 0, 2, -1)$
- **9.**  $\vec{u} = (6, 3, 9, 6); \quad \vec{v}_1 = (2, 1, 1, 1), \quad \vec{v}_2 = (1, 0, 1, 1), \quad \vec{v}_3 = (-2, -1, 0, -1)$
- **10.**  $\vec{u} = (-2, 0, 2, 4); \quad v_1 = (1, 1, 3, 0), \quad \vec{v}_2 = (-2, -1, -2, 1), \quad \vec{v}_3 = (-3, -1, 1, 3)$
- 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, y = -t, z = 4t
  - a) Find a basis for W.
  - b) Find the standard matrix for the orthogonal projection on W.
  - c) Use the matrix in part (b) to find the orthogonal projection of a point  $P_0(x_0, y_0, z_0)$  on W.
  - d) 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, y = t, z = t And the line m given by the equations x = s, y = 2s 1, 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,
  - a) If A is an  $m \times n$  matrix, then  $A^T A$  is a square matrix.
  - b) If  $A^T A$  is invertible, then A is invertible.
  - c) If A is invertible, then  $A^T A$  is invertible.
  - d) If  $A\vec{x} = \vec{b}$  is a consistent linear system, then  $A^T A \vec{x} = A^T \vec{b}$  is also consistent.
  - e) If  $A\vec{x} = \vec{b}$  is an inconsistent linear system, then  $A^T A \vec{x} = A^T \vec{b}$  is also inconsistent.
  - f) Every linear system has a least squares solution.
  - g) Every linear system has a unique least squares solution.
  - h) If A is an  $m \times n$  matrix with linearly independent columns and  $\vec{b}$  is in  $R^m$ , 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  $(r, \upsilon)$  of a comet satisfies an equation of the form

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

Where  $\beta$  is a constant and e is the eccentricity of the orbit, with  $0 \le e < 1$  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.

Determine the type of orbit, and predict where the orbit will be when v = 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:

- 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).