

Least Square Approximation:

→ Well defined linear system has equal number of variables & equations.

Example:

$$\begin{aligned}x_1 + 2x_2 + x_3 &= 0 \\x_1 - 9x_2 + 7x_3 &= 2 \\2x_1 + 3x_2 + 5x_3 &= 5\end{aligned}$$

$$A \cdot x = b$$

$$\rightarrow \begin{bmatrix} 1 & 2 & 3 \\ 1 & -9 & 7 \\ 2 & 3 & 5 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 2 \\ 5 \end{bmatrix}$$

square matrix
of $\boxed{n} \times n$; ; ;
 $n \times \boxed{1} \xrightarrow{\text{ }} n \times 1$

→ If we have a system where number of equations > number of variables, it is called an over-determined system.

How do we solve over-determined system?

Example:

$$\begin{aligned}x_1 + 2x_2 + x_3 &= 0 \\x_1 - 9x_2 + 7x_3 &= 2 \\x_1 + 3x_2 + 5x_3 &= 4 \rightarrow \\2x_1 + 11x_2 - 9x_3 &= 5 \\9x_1 + x_2 - x_3 &= 7\end{aligned}$$

$$\begin{bmatrix} 1 & 2 & 1 \\ 1 & -9 & 7 \\ 1 & 3 & 5 \\ 2 & 11 & -9 \\ 9 & 1 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 2 \\ 4 \\ 5 \\ 7 \end{bmatrix}$$

$$\boxed{m} \times n ; ; ; n \times \boxed{1} ; m \times 1$$

$$\textcircled{A} \cdot x = b$$

→ Least square approximation method is a way to find an approximate solution of an over-determined system.

over-determined

$$\begin{matrix} \downarrow \\ A \cdot x = b \end{matrix}$$
$$\begin{matrix} \downarrow \\ (m \times n) \quad (n \times 1) \quad (m \times 1) \end{matrix}$$

How to solve such problems?

→ multiply A^T on both hand sides.

$$\begin{matrix} A^T \quad A \quad | \quad x \\ \downarrow \quad \downarrow \quad \downarrow \\ (n \times m) \quad (m \times n) \quad (n \times 1) \end{matrix} = \begin{matrix} A^T \cdot b \\ \downarrow \\ (n \times m) \quad (m \times 1) \end{matrix}$$
$$\begin{matrix} \downarrow \\ (n \times n) \quad (n \times 1) \end{matrix} \quad \downarrow \quad (n \times 1)$$

Example

From polynomial chapter:

If we had $\underbrace{(n+1)}$ nodes, we calculated the values of $\underbrace{(n+1)}$ coefficients
 x_0, x_1, \dots, x_n a_0, a_1, \dots, a_n

using vandermonde matrix.

Well-defined system

$$\rightarrow \begin{bmatrix} 1 & x_0^1 & x_0^2 & \dots & x_0^n \\ 1 & x_1^1 & x_1^2 & \dots & x_1^n \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_n^1 & x_n^2 & \dots & x_n^n \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \\ \vdots \\ a_n \end{bmatrix} = \begin{bmatrix} f(x_0) \\ f(x_1) \\ \vdots \\ f(x_n) \end{bmatrix}$$

But now, let's say

we have $\underbrace{(m+1)}$ nodes, but we need to calculate $\underbrace{(n+1)}$ coefficients
 x_0, x_1, \dots, x_m a_0, a_1, \dots, a_n

[Remember $m > n$]

Over-determined system

$$\rightarrow \begin{bmatrix} 1 & x_0^1 & x_0^2 & \dots & x_0^n \\ 1 & x_1^1 & x_1^2 & \dots & x_1^n \\ 1 & x_2^1 & x_2^2 & \dots & x_2^n \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_m^1 & x_m^2 & \dots & x_m^n \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \\ \vdots \\ a_n \end{bmatrix} = \begin{bmatrix} f(x_0) \\ f(x_1) \\ f(x_2) \\ \vdots \\ f(x_m) \end{bmatrix}$$

Example:

$$a_0 + a_1 x$$

number of coefficients = 2

We want to fit a straight line through the following nodes

$$x_0 = -3$$

$$x_1 = 0$$

$$x_2 = 6$$

$$f(x_0) = 0$$

$$f(x_1) = 0$$

$$f(x_2) = 2$$

number of nodes = 3

$$P_1(x_0) = a_0 + a_1(x_0) = f(x_0) \rightarrow a_0 + a_1(-3) = 0$$

$$P_1(x_1) = a_0 + a_1(x_1) = f(x_1) \rightarrow a_0 + a_1(0) = 0$$

$$P_1(x_2) = a_0 + a_1(x_2) = f(x_2) \rightarrow a_0 + a_1(6) = 2$$

$$\begin{bmatrix} 1 & -3 \\ 1 & 0 \\ 1 & 6 \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 2 \end{bmatrix}$$

$$A \cdot x = b$$

Multiplying A^T on both sides.

$$\left[\begin{array}{ccc} 1 & 1 & 1 \\ -3 & 0 & 6 \end{array} \right] \left[\begin{array}{cc} 1 & -3 \\ 1 & 0 \\ 1 & 6 \end{array} \right] \left[\begin{array}{c} a_0 \\ a_1 \end{array} \right] = \left[\begin{array}{ccc} 1 & 1 & 1 \\ -3 & 0 & 6 \end{array} \right] \left[\begin{array}{c} 0 \\ 0 \\ 2 \end{array} \right]$$
$$\left\{ \begin{array}{c} \\ \end{array} \right. \quad \left\{ \begin{array}{c} \\ \end{array} \right.$$
$$\left[\begin{array}{cc} 3 & 3 \\ 3 & 45 \end{array} \right] \left[\begin{array}{c} a_0 \\ a_1 \end{array} \right] = \left[\begin{array}{c} 2 \\ 12 \end{array} \right]$$

Now apply Gaussian elimination/LU/inverse method to find the values of a_0 and a_1 .

Applying inverse method:

$$\left[\begin{array}{c} a_0 \\ a_1 \end{array} \right] = \left[\begin{array}{cc} 3 & 3 \\ 3 & 45 \end{array} \right]^{-1} \left[\begin{array}{c} 2 \\ 12 \end{array} \right]$$

$$= \left[\begin{array}{c} 3/7 \\ 5/21 \end{array} \right]$$

$$\therefore a_0 = 3/7, \quad a_1 = 5/21$$

$$\therefore P_1(x) = \frac{3}{7} + \frac{5}{21}x$$

Orthogonality:

→ To understand orthogonality, we need to understand vector dot product / inner product first.

→ Vector dot product returns a scalar value (a number)

2 types of notations → matrix notation → $x^T \cdot y$
 → vector notation → $\vec{x} \cdot \vec{y}$

$$x = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \quad y = \begin{bmatrix} 4 \\ 5 \\ 6 \end{bmatrix}$$

$$\textcircled{1} \text{ Matrix Notation} = x^T \cdot y$$

$$= \begin{bmatrix} 1 & 2 & 3 \end{bmatrix} \begin{bmatrix} 4 \\ 5 \\ 6 \end{bmatrix}$$

shape \rightarrow (1×3) | (3×1)

=  returns a scalar/number

② Vector Notation = $\vec{x} \cdot \vec{y}$

$$= (1 \times 4) + (2 \times 5) + (3 \times 6)$$

↳ returns a scalar / number

Inner product with itself = ℓ_2 -norm

e.g. $\vec{x} \cdot \vec{x}$ or $x^T x$

Dot product:

$$a = \begin{bmatrix} a_1 \\ a_2 \\ a_3 \end{bmatrix} \quad b = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}$$

$$a \cdot b = a_1 b_1 + a_2 b_2 + a_3 b_3$$

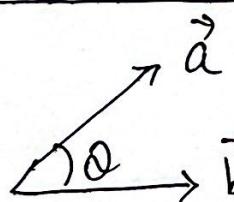
↳ returns a scalar/number.

Length / magnitude of a vector:

$$a = \begin{bmatrix} a_1 \\ a_2 \\ a_3 \end{bmatrix}$$

$$|a| = \sqrt{(a_1)^2 + (a_2)^2 + (a_3)^2}$$

Dot product (Second approach):



$$a \cdot b = |a| |b| \cos \theta$$

$$\cos \theta = \frac{a \cdot b}{|a| |b|}$$

Orthogonal Vectors:

If \angle between 2 vectors = 90° or $\frac{\pi}{2}$

In other words, if 2 vectors are perpendicular to each other,
the vectors are orthogonal.

For orthogonal vectors, their dot product is 0. Because:

$$\vec{a} \cdot \vec{b} = |\vec{a}| |\vec{b}| \cos\left(\frac{\pi}{2}\right) \rightarrow 0$$

$$\therefore \vec{a} \cdot \vec{b} = 0$$

$$\text{or } \vec{a}^T \vec{b} = 0$$

Let's consider a set of vector S.

$$S = \{ \vec{a}, \vec{b}, \vec{c} \}$$

Set S is an orthogonal set if

$$\vec{a} \cdot \vec{b} = 0$$

$$\vec{b} \cdot \vec{c} = 0$$

$$\vec{c} \cdot \vec{a} = 0$$

i.e. each vector is perpendicular to each other.

Orthonormality:

- If ① the vectors are orthogonal (dot product=0)
 ② the length of the vectors = 1 (unit vectors)

Then the vectors are orthonormal.

$$\vec{a} = \begin{bmatrix} 4 \\ 2 \\ 1 \end{bmatrix} \quad b = \begin{bmatrix} 1 \\ -3 \\ 2 \end{bmatrix}$$

Checking if orthogonal:

$$\vec{a} \cdot \vec{b} = (4 \times 1) + (2 \times -3) + (1 \times 2) = 0 \quad [\because \text{orthogonal}]$$

Converting into orthonormal (by making length = 1):

$$\hat{a} = \frac{\vec{a}}{|\vec{a}|} = \frac{1}{\sqrt{21}} \begin{bmatrix} 4 \\ 2 \\ 1 \end{bmatrix} = \begin{bmatrix} 4/\sqrt{21} \\ 2/\sqrt{21} \\ 1/\sqrt{21} \end{bmatrix} \rightarrow |\hat{a}| = \sqrt{\left(\frac{4}{\sqrt{21}}\right)^2 + \left(\frac{2}{\sqrt{21}}\right)^2 + \left(\frac{1}{\sqrt{21}}\right)^2} = 1$$

$$\hat{b} = \frac{\vec{b}}{|\vec{b}|} = \frac{1}{\sqrt{14}} \begin{bmatrix} 1 \\ -3 \\ 2 \end{bmatrix} = \begin{bmatrix} 1/\sqrt{14} \\ -3/\sqrt{14} \\ 2/\sqrt{14} \end{bmatrix} \rightarrow |\hat{b}| = \sqrt{\left(\frac{1}{\sqrt{14}}\right)^2 + \left(\frac{-3}{\sqrt{14}}\right)^2 + \left(\frac{2}{\sqrt{14}}\right)^2} = 1$$

→ Process of making vectors into unit vectors (length=1) is called normalization

→ By doing so, we only change the magnitude, not the direction.

→ Hence, they are still orthogonal

→ Since they are orthogonal and has length=1 (unit vectors), the vectors are orthonormal.

Example:

Consider the set of vectors, S :

$$S = \left\{ \frac{1}{\sqrt{5}} (2, 1)^T, \frac{1}{\sqrt{5}} (1, -2)^T \right\}$$

Show if the set S is orthonormal or not.

Solution:

$$\begin{aligned} S &= \left\{ \frac{1}{\sqrt{5}} \begin{bmatrix} 2 \\ 1 \end{bmatrix}, \frac{1}{\sqrt{5}} \begin{bmatrix} 1 \\ -2 \end{bmatrix} \right\} \\ &= \left\{ \begin{bmatrix} \frac{2}{\sqrt{5}} \\ \frac{1}{\sqrt{5}} \end{bmatrix}, \begin{bmatrix} \frac{1}{\sqrt{5}} \\ -\frac{2}{\sqrt{5}} \end{bmatrix} \right\} \\ &\quad \downarrow \qquad \downarrow \\ &\quad \vec{u}_1 \qquad \vec{u}_2 \end{aligned}$$

$$\vec{u}_1 \cdot \vec{u}_2 = \left(\frac{2}{\sqrt{5}} \times \frac{1}{\sqrt{5}} \right) + \left(\frac{1}{\sqrt{5}} \times -\frac{2}{\sqrt{5}} \right) = 0 \quad [\therefore \text{orthogonal}]$$

$$|\vec{u}_1| = \sqrt{\left(\frac{2}{\sqrt{5}}\right)^2 + \left(\frac{1}{\sqrt{5}}\right)^2} = 1 \quad [\therefore \text{orthonormal}]$$

$$|\vec{u}_2| = \sqrt{\left(\frac{1}{\sqrt{5}}\right)^2 + \left(-\frac{2}{\sqrt{5}}\right)^2} = 1$$

\therefore Yes, the set of vectors are orthonormal.

Theorem:

Orthogonal / orthonormal matrices are matrices in which the column vectors form an ~~orthogonal~~ ^[Orthonormal] set (each column vector has length one, and is orthogonal to all other column vectors).

For square orthonormal matrices, the inverse is simply the transpose.

Properties: $Q^{-1} = Q^T$

$$Q Q^T = I$$

$$Q^T Q = I$$

Example:

$$Q = \begin{bmatrix} \vec{a} \\ \vec{b} \end{bmatrix}$$

$$\vec{a} = \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix} \quad \vec{b} = \begin{bmatrix} 1/\sqrt{2} \\ -1/\sqrt{2} \end{bmatrix}$$

$$\vec{a} \cdot \vec{b} = (1/\sqrt{2} \times 1/\sqrt{2}) + (1/\sqrt{2} \times -1/\sqrt{2}) = 0 \quad [\because \text{orthogonal}]$$

$$|\vec{a}| = \sqrt{\left(\frac{1}{\sqrt{2}}\right)^2 + \left(\frac{1}{\sqrt{2}}\right)^2} = 1 \quad [\because \text{orthonormal}]$$

$$|\vec{b}| = \sqrt{\left(\frac{1}{\sqrt{2}}\right)^2 + \left(-\frac{1}{\sqrt{2}}\right)^2} = 1$$

→ This makes the columns of the matrix orthonormal to each other.

- ∴ The matrix Q will have the properties :
 - ① $Q^{-1} = Q^T$
 - ② $Q Q^T = I$
 - ③ $Q^T Q = I$

Gram - Schmidt Process:

Lets have a basis (u_1, u_2, \dots, u_n) from a vector space. Gram - Schmidt process takes the basis (u_1, u_2, \dots, u_n) and forms a new orthogonal basis (p_1, p_2, \dots, p_n) . We can later transform these orthogonal basis into orthonormal basis (q_1, q_2, \dots, q_n) .

Original basis : u_1, u_2, u_3

↓
Gram Schmidt Process

Orthogonal basis : p_1, p_2, p_3

↓ normalization

Orthonormal basis : q_1, q_2, q_3

$$1) p_1 = u_1$$

$$2) p_2 = u_2 - \frac{u_2 \cdot p_1}{p_1 \cdot p_1} p_1$$

$$3) p_3 = u_3 - \frac{u_3 \cdot p_1}{p_1 \cdot p_1} p_1 - \frac{u_3 \cdot p_2}{p_2 \cdot p_2} p_2$$

Example:

$$u_1 = \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix} \quad u_2 = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} \quad u_3 = \begin{bmatrix} 1 \\ 1 \\ 2 \end{bmatrix}$$

$$1) \quad p_1 = u_1 = \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}$$

$$2) \quad p_2 = u_2 - \frac{u_2 \cdot p_1}{p_1 \cdot p_1} p_1$$

$$= \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} - \frac{(1 \times 1) + (0 \times -1) + (1 \times 1)}{(1 \times 1) + (-1 \times -1) + (1 \times 1)} \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}$$

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

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

$$\begin{array}{r} F = 2/3 \\ F = 0 \\ F = - \end{array}$$

$$= \begin{bmatrix} 1/3 \\ 2/3 \\ 1/3 \end{bmatrix}$$

$$③ P_3 = U_3 - \frac{U_3 \cdot P_1}{P_1 \cdot P_1} P_1 - \frac{U_3 \cdot P_2}{P_2 \cdot P_2} P_2$$

$$= \begin{bmatrix} 1 \\ 1 \\ 2 \end{bmatrix} - \frac{(1 \times 1) + (1 \times -1) + (2 \times 1)}{(1 \times 1) + (-1 \times -1) + (1 \times 1)} \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix} - \frac{\left(1 \times \frac{1}{3}\right) + \left(1 \times \frac{2}{3}\right) + \left(2 \times \frac{1}{3}\right)}{\left(\frac{1}{3} \times \frac{1}{3}\right) + \left(\frac{2}{3} \times \frac{2}{3}\right) + \left(\frac{1}{3} \times \frac{1}{3}\right)} \begin{bmatrix} 1/3 \\ 2/3 \\ 1/3 \end{bmatrix}$$

$$= \begin{bmatrix} 1 \\ 1 \\ 2 \end{bmatrix} - \frac{2}{3} \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix} - \frac{5}{2} \begin{bmatrix} 1/3 \\ 2/3 \\ 1/3 \end{bmatrix}$$

$$= \begin{bmatrix} 1 \\ 1 \\ 2 \end{bmatrix} - \begin{bmatrix} 2/3 \\ -2/3 \\ 2/3 \end{bmatrix} - \begin{bmatrix} 5/6 \\ 5/3 \\ 5/6 \end{bmatrix}$$

$$= \begin{bmatrix} -1/2 \\ 0 \\ 1/2 \end{bmatrix}$$

P_1, P_2, P_3 form an orthogonal basis

If we want orthonormal basis, we can divide each vectors by its length.
(normalization).

$$\textcircled{1} \quad q_1 = \frac{P_1}{|P_1|} = \frac{1}{\sqrt{3}} \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix} = \begin{bmatrix} 1/\sqrt{3} \\ -1/\sqrt{3} \\ 1/\sqrt{3} \end{bmatrix}$$

$$\textcircled{2} \quad q_2 = \frac{P_2}{|P_2|} = \frac{1}{\sqrt{6}/3} \begin{bmatrix} 1/3 \\ 2/3 \\ 1/3 \end{bmatrix} = \begin{bmatrix} \sqrt{6}/6 \\ \sqrt{6}/3 \\ \sqrt{6}/6 \end{bmatrix}$$

$$\textcircled{3} \quad q_3 = \frac{P_3}{|P_3|} = \frac{1}{\sqrt{2}/2} \begin{bmatrix} -1/2 \\ 0 \\ 1/2 \end{bmatrix} = \begin{bmatrix} -\sqrt{2}/2 \\ 0 \\ \sqrt{2}/2 \end{bmatrix}$$

QR Decomposition:

Any real $(m \times n)$ matrix "A" with $m > n$ can be written in the form :

$$A = QR$$

Where Q is a $(m \times n)$ matrix with orthonormal columns

R is an upper triangular matrix of shape $(n \times n)$

Orthonormal matrix. Therefore has the properties:

$$\begin{matrix} A &= & Q & R \\ && \downarrow & \\ && Q & \\ && \downarrow & \\ m \times n & & m \times n & n \times n \end{matrix}$$

$$Q Q^T = I_{nn}$$

$$Q^T Q = I_{nn}$$

→ Multiplying Q^T on both sides.

$$Q^T A = \boxed{Q^T Q} R$$

$$Q^T A = \boxed{I} R$$

$$Q^T A = R$$

$$\boxed{R = Q^T A}$$

Example of QR Decomposition:

$$A = QR$$

Starting with A

$$A = \begin{bmatrix} u_{11} & u_{12} & \dots & u_{1n} \\ u_{21} & u_{22} & \dots & u_{2n} \\ \vdots & \vdots & & \vdots \\ u_{m1} & u_{m2} & \dots & u_{mn} \end{bmatrix}$$

$u_1 \quad u_2 \quad \dots \quad u_n$

convert u into orthogonal vectors p (Gram-Smidt process)

convert p into orthonormal vectors q (normalization)

$$A = \begin{bmatrix} 3 \\ 6 \\ 0 \end{bmatrix} \quad \begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix}$$

$u_1 \quad u_2$

$$p_1 = u_1 = \begin{bmatrix} 3 \\ 6 \\ 0 \end{bmatrix}$$

$$P_2 = U_2 - \frac{U_2 \cdot P_1}{P_1 \cdot P_1} P_1$$

$$= \begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix} - \frac{\begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix} \cdot \begin{bmatrix} 3 \\ 6 \\ 0 \end{bmatrix}}{\begin{bmatrix} 3 \\ 6 \\ 0 \end{bmatrix} \cdot \begin{bmatrix} 3 \\ 6 \\ 0 \end{bmatrix}} \cdot \begin{bmatrix} 3 \\ 6 \\ 0 \end{bmatrix}$$

$$= \begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix} - \frac{(1 \times 3) + (2 \times 6) + (2 \times 0)}{(3 \times 3) + (6 \times 6) + (0 \times 0)} \begin{bmatrix} 3 \\ 6 \\ 0 \end{bmatrix}$$

$$= \begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix} - \frac{1}{3} \begin{bmatrix} 3 \\ 6 \\ 0 \end{bmatrix}$$

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

$$= \begin{bmatrix} 0 \\ 0 \\ 2 \end{bmatrix}$$

$$\therefore P_1 = \begin{bmatrix} 3 \\ 6 \\ 0 \end{bmatrix} \quad P_2 = \begin{bmatrix} 0 \\ 0 \\ 2 \end{bmatrix}$$

P_1 and P_2 are orthogonal.

→ Now convert P_1 and P_2 into unit vectors. We will call the unit vectors q_1 and q_2 .

$$q_1 = \frac{P_1}{|P_1|} = \frac{1}{3\sqrt{5}} \begin{bmatrix} 3 \\ 6 \\ 0 \end{bmatrix} = \begin{bmatrix} \sqrt{5}/5 \\ 2\sqrt{5}/5 \\ 0 \end{bmatrix}$$

$$q_2 = \frac{P_2}{|P_2|} = \frac{1}{2} \begin{bmatrix} 0 \\ 0 \\ 2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

$$\therefore q_1 = \begin{bmatrix} \sqrt{5}/5 \\ 2\sqrt{5}/5 \\ 0 \end{bmatrix} \quad q_2 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

q_1 and q_2 are orthonormal

$$\therefore Q = \begin{bmatrix} \sqrt{5}/5 & 0 \\ 2\sqrt{5}/5 & 0 \\ 0 & 1 \end{bmatrix} \quad A = \begin{bmatrix} 3 & 1 \\ 6 & 2 \\ 0 & 2 \end{bmatrix}$$

$$R = Q^T A$$

$$= \begin{bmatrix} \sqrt{5}/5 & 2\sqrt{5}/5 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 3 & 1 \\ 6 & 2 \\ 0 & 2 \end{bmatrix}$$

$$= \begin{bmatrix} 3\sqrt{5} & \sqrt{5} \\ 0 & 2 \end{bmatrix}$$

$$A = Q R$$

$$\begin{bmatrix} 3 & 1 \\ 6 & 2 \\ 0 & 2 \end{bmatrix} = \begin{bmatrix} \sqrt{5}/5 & 0 \\ 2\sqrt{5}/5 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 3\sqrt{5} & \sqrt{5} \\ 0 & 2 \end{bmatrix}$$

Check if $Q \cdot R = A$ in your calculator.

Example 2

$$x_0 = -3$$

$$f(x_0) = 0$$

$$x_1 = 0$$

$$f(x_1) = 0$$

$$x_2 = 6$$

$$f(x_2) = 2$$

$$A \cdot x = b$$
$$\begin{bmatrix} 1 & -3 \\ 1 & 0 \\ 1 & 6 \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 2 \end{bmatrix}$$

$\downarrow \quad \downarrow$

$$u_1 \quad u_2$$

$$p_1 = u_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

$$p_2 = u_2 - \frac{u_2 \cdot p_1}{p_1 \cdot p_1} p_1$$

$$= \begin{bmatrix} -3 \\ 0 \\ 6 \end{bmatrix} - \frac{(-3 \times 1) + (0 \times 1) + (6 \times 1)}{(1 \times 1) + (1 \times 1) + (1 \times 1)} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

$$= \begin{bmatrix} -3 \\ 0 \\ 6 \end{bmatrix} - 1 \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

$$= \begin{bmatrix} -4 \\ -1 \\ 5 \end{bmatrix}$$

$$P_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

$$P_2 = \begin{bmatrix} -4 \\ -1 \\ 5 \end{bmatrix}$$

P_1 and P_2 are orthogonal vectors.

$$q_1 = \frac{P_1}{|P_1|} = \frac{1}{\sqrt{3}} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 1/\sqrt{3} \\ 1/\sqrt{3} \\ 1/\sqrt{3} \end{bmatrix}$$

$$q_2 = \frac{P_2}{|P_2|} = \frac{1}{\sqrt{42}} \begin{bmatrix} -4 \\ -1 \\ 5 \end{bmatrix} = \begin{bmatrix} -4/\sqrt{42} \\ -1/\sqrt{42} \\ 5/\sqrt{42} \end{bmatrix}$$

$$\therefore Q = \begin{bmatrix} 1/\sqrt{3} & -4/\sqrt{42} \\ 1/\sqrt{3} & -1/\sqrt{42} \\ 1/\sqrt{3} & 5/\sqrt{42} \end{bmatrix}$$

→ matrix Q has orthonormal columns.

$$R = Q^T A$$

$$= \begin{bmatrix} 1/\sqrt{3} & 1/\sqrt{3} & 1/\sqrt{3} \\ -4/\sqrt{42} & -1/\sqrt{42} & 5/\sqrt{42} \end{bmatrix} \begin{bmatrix} 1 & -3 \\ 1 & 0 \\ 1 & 6 \end{bmatrix}$$

$$R = \begin{bmatrix} \sqrt{3} & \sqrt{3} \\ 0 & \sqrt{42} \end{bmatrix}$$

$$A x = b$$

Applying Least square approximation method by applying A^T on both sides.

$$\begin{matrix} A^T & A & x & = & A^T b \\ \downarrow & \downarrow & \downarrow & & \downarrow \\ (QR)^T & (QR) & x & = & (QR)^T b \end{matrix}$$

$$R^T \boxed{Q^T Q} R x = R^T Q^T b$$

$$\downarrow I_{nn} \quad [\text{because } Q = \text{orthonormal matrix}]$$

$$\cancel{R^T} R x = \cancel{R^T} Q^T b$$

$$\boxed{\cancel{R^T} R x = Q^T b}$$

$$R_2 = Q^T b$$

$$\begin{bmatrix} \sqrt{3} & \sqrt{3} \\ 0 & \sqrt{42} \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \end{bmatrix} = \begin{bmatrix} 1/\sqrt{3} & 1/\sqrt{3} & 1/\sqrt{3} \\ -4/\sqrt{42} & -1/\sqrt{42} & 5/\sqrt{42} \end{bmatrix} \begin{bmatrix} 0 \\ 0 \\ 2 \end{bmatrix}$$

$$\begin{bmatrix} \sqrt{3} & \sqrt{3} \\ 0 & \sqrt{42} \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \end{bmatrix} = \begin{bmatrix} 2/\sqrt{3} \\ 5\sqrt{2}/\sqrt{21} \end{bmatrix}$$

$$\begin{bmatrix} a_0 \\ a_1 \end{bmatrix} = \begin{bmatrix} 3/7 \\ 5/21 \end{bmatrix}$$