

Very well done,
Yaroslavo!

grade: 20 pts

Linear Algebra

Home assignment 2: Orthogonality

Solutions are by Yaroslava Lochman.

Problem 1 (Parallel and orthogonal planes; 2pt). Determine whether the given planes are:

2 pts

(a) parallel:

(i) $4x - y + 2z = 5$ and $7x - 3y + 4z = 8$;

(ii) $x - 4y - 3z - 2 = 0$ and $3x - 12y - 9z - 7 = 0$.


(b) perpendicular:

(i) $3x - y + z = 0$ and $x + 2z = -1$;

(ii) $x - 2y + 3z = 4$ and $-2x + 5y + 4z = -1$.

Solution to the problem 1.

$$Ax + By + Cz = D \Rightarrow n = (A, B, C)^T - \text{the normal vector of the plane}$$

The planes are parallel \Leftrightarrow the normal vectors are parallel $\Leftrightarrow \frac{A_1}{A_2} = \frac{B_1}{B_2} = \frac{C_1}{C_2}$. 


The planes are perpendicular \Leftrightarrow the normal vectors are perpendicular $\Leftrightarrow \langle n_1, n_2 \rangle = 0 \Leftrightarrow A_1A_2 + B_1B_2 + C_1C_2 = 0$

(a) (parallel)

(i)

$$4x - y + 2z = 5 \Rightarrow n_1 = (4, -1, 2)^T$$

$$7x - 3y + 4z = 8 \Rightarrow n_2 = (7, -3, 4)^T$$


$$\frac{4}{7} \neq \frac{-1}{-3} \neq \frac{2}{4}$$


Hence the planes are not parallel.

(ii)

$$x - 4y - 3z - 2 = 0 \Rightarrow n_1 = (1, -4, -3)^T$$


$$3x - 12y - 9z - 7 = 0 \Rightarrow n_2 = (3, -12, -9)^T$$

$$\frac{3}{1} = \frac{-12}{-4} = \frac{-9}{-3} = 3$$


Hence the planes are parallel.

(b) (perpendicular)

(i) For $3x - y + z = 0$ and $x + 2z = -1$ we have

$$\langle n_1, n_2 \rangle = 3 + 0 + 2 = 5 \neq 0$$


Hence the planes are not perpendicular.

(ii) For $x - 2y + 3z = 4$ and $-2x + 5y + 4z = -1$ we have

$$\langle n_1, n_2 \rangle = -2 - 10 + 12 = 0 \quad \checkmark$$

Hence the planes are perpendicular.

Problem 2 (Orthogonal complement; 3pt). (a) Let W be the plane in \mathbb{R}^3 given by the equation $x - 2y - 3z = 0$. Find parametric equations for W^\perp .

3 pts

(b) Let W be the line in \mathbb{R}^3 with parametric equations $x = 2t, y = -5t, z = 4t$. Find an equation for W^\perp .

(c) Let W be the intersection of the two planes $x + y + z = 0$ and $x - y + z = 0$ in \mathbb{R}^3 . Find an equation for W^\perp .

Solution to the problem 2.

(a)

$$W = \{(x \ y \ z)^\top \in \mathbb{R}^3 \mid x - 2y - 3z = 0\} \quad \mathbf{n} = (1 \ -2 \ -3)^\top - \text{normal vector of the plane}$$

$$\text{So } \mathbf{x} \in W \Rightarrow \langle \mathbf{x}, \mathbf{n} \rangle = 0$$

$$\mathbf{y} \in W^\perp \Rightarrow \mathbf{y} \perp W \Rightarrow \langle \mathbf{y}, \mathbf{x} \rangle = 0 \ \forall \mathbf{x} \in W \Rightarrow \mathbf{y} \parallel \mathbf{n}$$

So:

$$W^\perp = \left\{ (x \ y \ z)^\top \in \mathbb{R}^3 \mid \begin{array}{l} x = t \\ y = -2t \\ z = -3t \end{array} \ t \in \mathbb{R} \right\} = \text{ls}\{\mathbf{n}\} \quad \checkmark$$

The parametric equations for W^\perp are $x = t, y = -2t, z = -3t$.

(b)

$$W = \left\{ (x \ y \ z)^\top \in \mathbb{R}^3 \mid \begin{array}{l} x = 2t \\ y = -5t \\ z = 4t \end{array} \ t \in \mathbb{R} \right\} \quad \mathbf{v} = \begin{pmatrix} 2 \\ -5 \\ 4 \end{pmatrix} - \text{direction vector of the line}$$

$$\mathbf{y} \in W^\perp \Rightarrow \langle \mathbf{y}, \mathbf{x} \rangle = 0 \ \forall \mathbf{x} \in W \Rightarrow \mathbf{y} \perp \mathbf{v} \Rightarrow W^\perp \text{ is a plane with the normal vector } \mathbf{v}$$

So:

$$W^\perp = \{(x \ y \ z)^\top \in \mathbb{R}^3 \mid 2x - 5y + 4z = 0\} \quad \checkmark$$

The equation for W^\perp is $2x - 5y + 4z = 0$.

(c)

$$W = \left\{ (x \ y \ z)^\top \in \mathbb{R}^3 \mid \begin{array}{l} x + y + z = 0 \\ x - y + z = 0 \end{array} \right\}$$

$\mathbf{n}_1 = (1 \ 1 \ 1)^\top$ - normal vector of the 1st plane.

$\mathbf{n}_2 = (1 \ -1 \ 1)^\top$ - normal vector of the 2nd plane.

$$W = \{\mathbf{x} \in \mathbb{R}^3 \mid \langle \mathbf{x}, \mathbf{n}_1 \rangle = 0 \wedge \langle \mathbf{x}, \mathbf{n}_2 \rangle = 0\} - \text{is a line}$$

We have: $\mathbf{x} \perp \mathbf{n}_1, \mathbf{x} \perp \mathbf{n}_2$. Let $\mathbf{y} \in \text{ls}\{\mathbf{n}_1, \mathbf{n}_2\} \Rightarrow \langle \mathbf{y}, \mathbf{x} \rangle = \langle \alpha_1 \mathbf{n}_1 + \alpha_2 \mathbf{n}_2, \mathbf{x} \rangle = \alpha_1 \cdot 0 + \alpha_2 \cdot 0 = 0$
 $\Rightarrow \mathbf{x} \perp \text{ls}\{\mathbf{n}_1, \mathbf{n}_2\}$. Hence:

$$W^\perp = \text{ls}\{\mathbf{n}_1, \mathbf{n}_2\}$$

To get an equation for the plane we need to compute its normal \mathbf{n} (which is concurrently a direction vector of the line W). We can do it using the cross product:

$$\mathbf{n} = \mathbf{n}_1 \times \mathbf{n}_2 = \begin{vmatrix} e_1 & e_2 & e_3 \\ 1 & 1 & 1 \\ 1 & -1 & 1 \end{vmatrix} = \begin{pmatrix} 2 \\ 0 \\ -2 \end{pmatrix} \quad \checkmark$$

So the equation for W^\perp is $2x - 2z = 0$.

Problem 3 (Distance from a point; 4pt). (a) Find the distance from the point $P = (1, 1, 0)$ to the line $\frac{x-1}{2} = \frac{y-2}{1} = \frac{z+1}{2}$.

4 pts

(b) Let π be a plane given by the equation $ax + by + cz + d = 0$ and $P(x_0, y_0, z_0)$ be a point outside it. Prove that the distance from P to π is given by the formula

$$\frac{|ax_0 + by_0 + cz_0 + d|}{\sqrt{a^2 + b^2 + c^2}}.$$

Hint: if Q is the point on π realizing the distance, then \overrightarrow{PQ} is collinear to $\mathbf{n} = (a, b, c)$ (why?). Take now any point Q' on π and find a projection of $\overrightarrow{PQ'}$ onto direction \mathbf{n}

(c) Find the distance between the point $P = (1, 0, 1)$ and the plane $2x + 2y - z = 2$.

Solution to the problem 3.

(a)

$$P = (1, 1, 0) \quad l: \frac{x-1}{2} = \frac{y-2}{1} = \frac{z+1}{2} \quad \mathbf{v} = (2 \ 1 \ 2)^\top - \text{direction vector}$$

Let $O = (1, 2, -1)$. This point lies on the line. Then $\overrightarrow{OP} = (0 \ -1 \ 1)^\top$. Let P' be the point on the line realizing distance. Hence the distance can be found as:

$$\rho = |\overrightarrow{PP'}| = \sqrt{|\overrightarrow{OP}|^2 - |\overrightarrow{OP'}|^2}$$

$|\overrightarrow{PP'}|$ is the projection of \overrightarrow{OP} on the line:

$$|\overrightarrow{OP'}|^2 = pr_{\mathbf{v}}^2 \overrightarrow{OP} = \frac{\langle \overrightarrow{OP}, \mathbf{v} \rangle^2}{\|\mathbf{v}\|^2} = \frac{(0 - 1 + 2)^2}{2^2 + 1^2 + 2^2} = \frac{1}{9}$$

So:

$$|\overrightarrow{PP'}| = \sqrt{2 - \frac{1}{9}} = \frac{\sqrt{17}}{3} \quad \checkmark$$

(b)

$$P = (x_0, y_0, z_0) \quad \pi : ax + by + cz + d = 0 \quad \mathbf{n} = (a \ b \ c)^\top - \text{normal vector}$$

Let Q be the point on π realizing the distance. Since the distance is the shortest:

$$\overrightarrow{PQ} \perp \overrightarrow{QQ'} \quad \forall Q' \in \pi \quad \text{or} \quad \overrightarrow{PQ} \perp \pi \quad \checkmark$$

($Q \in \pi$ minimizes $|PQ| \Leftrightarrow PQ \perp \pi$ – had been concluded using the Pythagorean theorem).
Therefore \overrightarrow{PQ} is collinear to the normal \mathbf{n} . So the distance $\rho = |\overrightarrow{PQ}|$ is an absolute value
(since the angle between vectors can be acute or obtuse) of the projection of \overrightarrow{PQ} onto \mathbf{n} :

$$Q' = (x_q \ y_q \ z_q)^\top \in \pi \Rightarrow ax_q + by_q + cz_q + d = 0 \quad \checkmark$$

$$\overrightarrow{PQ'} = (x_q - x_0 \ y_q - y_0 \ z_q - z_0)^\top$$

$$\Rightarrow \rho = |pr_n \overrightarrow{PQ'}| = \frac{|\langle \overrightarrow{PQ'}, \mathbf{n} \rangle|}{\|\mathbf{n}\|} = \frac{\left| \left\langle \begin{pmatrix} x_q - x_0 \\ y_q - y_0 \\ z_q - z_0 \end{pmatrix}, \begin{pmatrix} a \\ b \\ c \end{pmatrix} \right\rangle \right|}{\sqrt{a^2 + b^2 + c^2}} = \frac{|-d - ax_0 - by_0 - cz_0|}{\sqrt{a^2 + b^2 + c^2}}$$

So:

$$\rho = \frac{|ax_0 + by_0 + cz_0 + d|}{\sqrt{a^2 + b^2 + c^2}}. \quad \checkmark$$

(c)

$$P = (1, 0, 1) \quad \pi : 2x + 2y - z - 2 = 0$$

$$\rho = \frac{|ax_0 + by_0 + cz_0 + d|}{\sqrt{a^2 + b^2 + c^2}} = \frac{|2 + 2 \cdot 0 - 1 + -2|}{\sqrt{2^2 + 2^2 + (-1)^2}} = \frac{1}{\sqrt{9}} = \frac{1}{3} \quad \checkmark$$

Problem 4 (Cross product; 4pt). (a) For any two vectors $\mathbf{u} = (u_1, u_2, u_3)^\top$ and $\mathbf{v} = (v_1, v_2, v_3)^\top$, their **vector product**, or **cross product** $\mathbf{u} \times \mathbf{v}$ is the vector $\mathbf{w} = (w_1, w_2, w_3)^\top$ with entries

4 pts

$$w_1 = \begin{vmatrix} u_2 & u_3 \\ v_2 & v_3 \end{vmatrix}, \quad w_2 = -\begin{vmatrix} u_1 & u_3 \\ v_1 & v_3 \end{vmatrix}, \quad w_3 = \begin{vmatrix} u_1 & u_2 \\ v_1 & v_2 \end{vmatrix}.$$

Prove that \mathbf{w} is orthogonal to both \mathbf{u} and \mathbf{v} in the sense that $\mathbf{w}^\top \mathbf{u} = \mathbf{w}^\top \mathbf{v} = 0$.

Hint: these products are cofactor expansions of some 3×3 matrices

(b) Assume that $\mathbf{u}_1, \dots, \mathbf{u}_{n-1}$ are linearly independent vectors in \mathbb{R}^n . Find a formula analogous to that in part (a) for a vector that is orthogonal to the subspace spanned by $\mathbf{u}_1, \dots, \mathbf{u}_{n-1}$.

Solution to the problem 4. .

(a)

$$\mathbf{w} = \mathbf{u} \times \mathbf{v} = \begin{vmatrix} e_1 & e_2 & e_3 \\ u_1 & u_2 & u_3 \\ v_1 & v_2 & v_3 \end{vmatrix}$$

$$\mathbf{w}^\top \mathbf{u} = \begin{vmatrix} u_1 & u_2 & u_3 \\ u_1 & u_2 & u_3 \\ v_1 & v_2 & v_3 \end{vmatrix} = 0 \quad \text{since the 1}^{st} \text{ and 2}^{nd} \text{ rows are equal (therefore dependent)}$$

The same with \mathbf{v} :



$$\mathbf{w}^\top \mathbf{v} = \begin{vmatrix} v_1 & v_2 & v_3 \\ u_1 & u_2 & u_3 \\ v_1 & v_2 & v_3 \end{vmatrix} = 0 \quad \text{since the 1}^{st} \text{ and 3}^{rd} \text{ rows are equal (therefore dependent)}$$

(b) Let $\mathbf{u}_i = (u_i^1, \dots, u_i^n)^\top$. Then the answer is:

$$\mathbf{w} = \begin{vmatrix} e_1 & \cdots & e_n \\ u_1^1 & \cdots & u_1^n \\ \vdots & & \vdots \\ u_{n-1}^1 & \cdots & u_{n-1}^n \end{vmatrix}$$

One can see that \mathbf{w} is orthogonal to all \mathbf{u}_i :

$$\mathbf{w}^\top \mathbf{u}_i = \begin{vmatrix} u_i^1 & \cdots & u_i^n \\ u_1^1 & \cdots & u_1^n \\ \vdots & & \vdots \\ u_{n-1}^1 & \cdots & u_{n-1}^n \end{vmatrix} = 0 \quad \forall i \in \overline{1, n} \quad (\text{because of the two equal rows})$$

Therefore it is orthogonal to $\text{ls}\{\mathbf{u}_1, \dots, \mathbf{u}_{n-1}\}$:

$$\forall \mathbf{x} \in \text{ls}\{\mathbf{u}_1, \dots, \mathbf{u}_{n-1}\} \quad \mathbf{w}^\top \mathbf{x} = \mathbf{w}^\top \sum_{i=1}^{n-1} \alpha_i \mathbf{u}_i = \sum_{i=1}^{n-1} \alpha_i \mathbf{w}^\top \mathbf{u}_i = 0$$



Problem 5 (Orthogonal matrices; 4pt). (a) If Q_1 and Q_2 are orthogonal matrices, show that Q_1^{-1} and $Q_1 Q_2$ are orthogonal as well.

5 pts

(b) Prove that an orthogonal matrix that is also upper-triangular must be diagonal.

Solution to the problem 5. .

(a) Q_1 and Q_2 are orthogonal $\Leftrightarrow Q_1^\top = Q_1^{-1}$, $Q_2^\top = Q_2^{-1}$. So:

$$1. (Q_1^{-1})^\top = (Q_1^\top)^\top = Q_1 = (Q_1^{-1})^{-1}$$

Hence Q_1^{-1} is orthogonal.

$$2. (Q_1 Q_2)^\top = Q_2^\top Q_1^\top = Q_2^{-1} Q_1^{-1} = (Q_2 Q_1)^{-1}$$

Hence $Q_1 Q_2$ is orthogonal.

(b) Let $Q = (Q_1 \ Q_2 \ \cdots \ Q_n)$ and $Q^\top Q = I$:

$$\begin{pmatrix} Q_1^\top \\ Q_2^\top \\ \vdots \\ Q_n^\top \end{pmatrix} (Q_1 \ Q_2 \ \cdots \ Q_n) = I$$

$$\Rightarrow \begin{cases} Q_i^\top Q_i = 1 & \forall i \in \overline{1, n} \\ Q_i^\top Q_j = 0 & \forall i, j \in \overline{1, n} \ i \neq j \end{cases}$$

or: Q^{-1} is also upper-triangular; but Q^\top is lower triangular :)

$$\Rightarrow Q_1^\top Q_1 = Q_{11}^2 = 1 \Rightarrow Q_{11} = \pm 1 \Rightarrow Q_1^\top Q_j = \pm Q_{1j} = 0 \ \forall j \neq 1$$

$$\Rightarrow Q_2^\top Q_2 = Q_{12}^2 + Q_{22}^2 = 0 + Q_{22}^2 = 1 \Rightarrow Q_{22} = \pm 1 \Rightarrow Q_2^\top Q_j = \pm Q_{2j} = 0 \ \forall j \neq 2$$

and so on. Hence Q is diagonal and moreover:

$$Q = \begin{pmatrix} \pm 1 & 0 & \cdots & 0 \\ 0 & \pm 1 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \pm 1 \end{pmatrix} \checkmark$$

Problem 6 (Projection matrices; 3pt). For the vectors $\mathbf{a}_1 = (1, 0, 1)$, $\mathbf{a}_2 = (0, 1, 2)$ and $\mathbf{b} = (-1, 2, 1)$

3 pts

(a) find the matrix of the orthogonal projection P_W onto the plane $W := \text{ls}\{\mathbf{a}_1, \mathbf{a}_2\}$;

(b) find the matrix of the orthogonal projection P_{W^\perp} onto the line W^\perp ;

(c) find the components of the vector \mathbf{b} with respect to the decomposition $\mathbb{R}^3 = W \oplus W^\perp$.

Solution to the problem 6. .

(a)

$$A = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 2 \end{pmatrix}$$

$$P_W = A(A^\top A)^{-1}A^\top$$

$$A^\top A = \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 2 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 2 \end{pmatrix} = \begin{pmatrix} 2 & 2 \\ 2 & 5 \end{pmatrix}$$

$$(A^\top A)^{-1} = \frac{1}{6} \begin{pmatrix} 5 & -2 \\ -2 & 2 \end{pmatrix}$$

An easier way to get P_W :
it is $I - P_{W^\perp}$

$$P_W = \frac{1}{6} \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 2 \end{pmatrix} \begin{pmatrix} 5 & -2 \\ -2 & 2 \end{pmatrix} \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 2 \end{pmatrix} = \frac{1}{6} \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 2 \end{pmatrix} \begin{pmatrix} 5 & -2 & 1 \\ -2 & 2 & 2 \end{pmatrix} = \frac{1}{6} \begin{pmatrix} 5 & -2 & 1 \\ -2 & 2 & 2 \\ 1 & 2 & 5 \end{pmatrix} \checkmark$$

(b)

$$W^\perp = \text{ls}\{\mathbf{a}\}$$

where:

$$\mathbf{a} = \mathbf{a}_1 \times \mathbf{a}_2 = \begin{pmatrix} i & j & k \\ 1 & 0 & 1 \\ 0 & 1 & 2 \end{pmatrix} = -i - 2j + k = (-1, -2, 1)^\top$$

$$P_{W^\perp} = \frac{\mathbf{a}\mathbf{a}^\top}{\mathbf{a}^\top\mathbf{a}} = \frac{1}{6} \begin{pmatrix} 1 & 2 & -1 \\ 2 & 4 & -2 \\ -1 & -2 & 1 \end{pmatrix} \quad \checkmark$$

(c)

$$\mathbf{b} = P_W \mathbf{b} + P_{W^\perp} \mathbf{b}$$

$$P_W \mathbf{b} = \frac{1}{6} \begin{pmatrix} 5 & -2 & 1 \\ -2 & 2 & 2 \\ 1 & 2 & 5 \end{pmatrix} \begin{pmatrix} -1 \\ 2 \\ 1 \end{pmatrix} = \frac{1}{6} \begin{pmatrix} -8 \\ 8 \\ 8 \end{pmatrix} = \begin{pmatrix} -4/3 \\ 4/3 \\ 4/3 \end{pmatrix}$$

$$P_{W^\perp} \mathbf{b} = \frac{1}{6} \begin{pmatrix} 1 & 2 & -1 \\ 2 & 4 & -2 \\ -1 & -2 & 1 \end{pmatrix} \begin{pmatrix} -1 \\ 2 \\ 1 \end{pmatrix} = \frac{1}{6} \begin{pmatrix} 2 \\ 4 \\ -2 \end{pmatrix} = \begin{pmatrix} 1/3 \\ 2/3 \\ -1/3 \end{pmatrix}$$

Indeed:

$$P_W \mathbf{b} + P_{W^\perp} \mathbf{b} = \begin{pmatrix} -4/3 \\ 4/3 \\ 4/3 \end{pmatrix} + \begin{pmatrix} 1/3 \\ 2/3 \\ -1/3 \end{pmatrix} = \begin{pmatrix} -1 \\ 2 \\ 1 \end{pmatrix} = \mathbf{b} \quad \checkmark$$

Problem 7 (Least squares solution; 4pt). Is there any value of s for which $x_1 = 1$ and $x_2 = 2$ is the least squares solution of the linear system below? Explain your reasoning.

4 pts

$$x_1 - x_2 = 1,$$

$$2x_1 + 3x_2 = 1,$$

$$4x_1 + 5x_2 = s.$$

A very important observation!

Solution to the problem 7. Let


$$A = \begin{pmatrix} 1 & -1 \\ 2 & 3 \\ 4 & 5 \end{pmatrix} = (\mathbf{a}_1 \quad \mathbf{a}_2) \quad \mathbf{b} = \begin{pmatrix} 1 \\ 1 \\ s \end{pmatrix}$$

Let $\mathbf{x}^* = \begin{pmatrix} 1 \\ 2 \end{pmatrix}$ be the least squares solution of the linear system $A\mathbf{x} = \mathbf{b}$. Therefore $\mathbf{e} = \mathbf{b} - A\mathbf{x}^*$ should be orthogonal to $C(A) \Leftrightarrow \langle \mathbf{e}, \mathbf{a}_i \rangle = 0 \quad i = 1, 2$

$$\mathbf{e} = \mathbf{b} - A\mathbf{x}^* = \begin{pmatrix} 1 \\ 1 \\ s \end{pmatrix} - \begin{pmatrix} 1 & -1 \\ 2 & 3 \\ 4 & 5 \end{pmatrix} \begin{pmatrix} 1 \\ 2 \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \\ s \end{pmatrix} - \begin{pmatrix} -1 \\ 8 \\ 14 \end{pmatrix} = \begin{pmatrix} 2 \\ -7 \\ s - 14 \end{pmatrix}$$

$$\begin{cases} \langle \mathbf{e}, \mathbf{a}_1 \rangle = \left\langle \begin{pmatrix} 2 \\ -7 \\ s-14 \end{pmatrix}, \begin{pmatrix} 1 \\ 2 \\ 4 \end{pmatrix} \right\rangle = 2 - 14 + 4s - 48 = 4s - 60 = 0 \Rightarrow s = 15 \\ \langle \mathbf{e}, \mathbf{a}_2 \rangle = \left\langle \begin{pmatrix} 2 \\ -7 \\ s-14 \end{pmatrix}, \begin{pmatrix} -1 \\ 3 \\ 5 \end{pmatrix} \right\rangle = -2 - 21 + 5s - 70 = 5s - 93 = 0 \Rightarrow s = 18.6 \end{cases}$$

$$\Rightarrow s \in \emptyset$$

Hence there is no value of s for which \mathbf{x}^* is the least squares solution of $A\mathbf{x} = \mathbf{b}$. 

Problem 8 (Regression; 6pt). (a) Find the least squares straight line fit to the four points $(0, 1)$, $(2, 0)$, $(3, 1)$, and $(3, 2)$.

6 pts

- (b) Find the quadratic polynomial that best fits the four points $(2, 0)$, $(3, -10)$, $(5, -48)$, and $(6, -76)$.
- (c) Find the cubic polynomial that best fits the five points $(-1, -14)$, $(0, -5)$, $(1, -4)$, $(2, 1)$, and $(3, 22)$.

Hint: the numbers are chosen so that $A^T A$ can easily be inverted. If, however, this is not so, ask Python or anybody else for a help.

Solution to the problem 8. .

(a)

$$A = \begin{pmatrix} 1 & 0 \\ 1 & 2 \\ 1 & 3 \\ 1 & 3 \end{pmatrix} \quad b = \begin{pmatrix} 1 \\ 0 \\ 1 \\ 2 \end{pmatrix}$$

I would calculate $A^T b$ first :)

The first column of A corresponds to the y-intercept, and the second – to the 1-degree term.

$$A^T A = \begin{pmatrix} 4 & 8 \\ 8 & 22 \end{pmatrix} \quad (A^T A)^{-1} = \frac{1}{12} \begin{pmatrix} 11 & -4 \\ -4 & 2 \end{pmatrix} \quad (A^T A)^{-1} A^T = \frac{1}{12} \begin{pmatrix} 11 & 3 & -1 & -1 \\ -4 & 0 & 2 & 2 \end{pmatrix}$$

$$x = (A^T A)^{-1} A^T b = \frac{1}{12} \begin{pmatrix} 8 \\ 2 \end{pmatrix} = \begin{pmatrix} 2/3 \\ 1/6 \end{pmatrix} \quad \checkmark$$

So, the quadratic polynomial $y = \frac{1}{6}x + \frac{2}{3}$ best fits the points.

(b)

$$A = \begin{pmatrix} 1 & 2 & 4 \\ 1 & 3 & 9 \\ 1 & 5 & 25 \\ 1 & 6 & 36 \end{pmatrix} \quad b = \begin{pmatrix} 0 \\ -10 \\ -48 \\ -76 \end{pmatrix} \quad \checkmark$$

The first column of A corresponds to the y-intercept, the second – to the 1-degree term, the third – to the 2-degree term.

$$A^T A = \begin{pmatrix} 4 & 16 & 74 \\ 16 & 74 & 376 \\ 74 & 376 & 2018 \end{pmatrix} \quad (A^T A)^{-1} = \frac{1}{90} \begin{pmatrix} 1989 & -1116 & 135 \\ -1116 & 649 & -80 \\ 135 & -80 & 10 \end{pmatrix}$$

$$(A^T A)^{-1} A^T = \frac{1}{90} \begin{pmatrix} 297 & -144 & -216 & 153 \\ -138 & 111 & 129 & -102 \\ 15 & -15 & -15 & 15 \end{pmatrix}$$

$$x = (A^T A)^{-1} A^T b = \frac{1}{90} \begin{pmatrix} 180 \\ 450 \\ -270 \end{pmatrix} = \begin{pmatrix} 2 \\ 5 \\ -3 \end{pmatrix} \quad \checkmark$$

actually, that is an exact fit :)

So, the quadratic polynomial $y = 3x^2 + 5x + 2$ best fits the points.

(c)

$$A = \begin{pmatrix} 1 & -1 & 1 & -1 \\ 1 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 \\ 1 & 2 & 4 & 8 \\ 1 & 3 & 9 & 27 \end{pmatrix} \quad b = \begin{pmatrix} -14 \\ -5 \\ -4 \\ 1 \\ 22 \end{pmatrix} \quad \checkmark$$

The first column of A corresponds to the y-intercept, the second – to the 1-degree term, the third – to the 2-degree term, the fourth – to the 3-degree term.

$$A^T A = \begin{pmatrix} 5 & 5 & 15 & 35 \\ 5 & 15 & 35 & 99 \\ 15 & 35 & 99 & 275 \\ 35 & 99 & 275 & 795 \end{pmatrix} \quad (A^T A)^{-1} = \frac{1}{2520} \begin{pmatrix} 1944 & -60 & -1440 & 420 \\ -60 & 1000 & -150 & -70 \\ -1440 & -150 & 1755 & -525 \\ 420 & -70 & -525 & 175 \end{pmatrix}$$

$$(A^T A)^{-1} A^T = \frac{1}{2520} \begin{pmatrix} 144 & 1944 & 864 & -576 & 144 \\ -1140 & -60 & 720 & 780 & -300 \\ 990 & -1440 & -360 & 1080 & -270 \\ -210 & 420 & 0 & -420 & 210 \end{pmatrix}$$

$$x = (A^T A)^{-1} A^T b = \frac{1}{2520} \begin{pmatrix} -12600 \\ 7560 \\ -10080 \\ 5040 \end{pmatrix} = \begin{pmatrix} -5 \\ 3 \\ -4 \\ 2 \end{pmatrix} \quad \checkmark$$

again, it is actually exact fit :)

So, the cubic polynomial $y = 2x^3 - 4x^2 + 3x - 5$ best fits the points.

4 pts

Problem 9 (Least square solution; 5pt). Assume \mathbf{u}_1 and \mathbf{u}_2 are two orthogonal vectors in \mathbb{R}^n and set $\mathbf{a}_1 = \mathbf{u}_1$, $\mathbf{a}_2 = \mathbf{u}_1 + \varepsilon \mathbf{u}_2$ for $\varepsilon > 0$. Let also A be the matrix with columns \mathbf{a}_1 and \mathbf{a}_2 and \mathbf{b} a vector linearly independent of \mathbf{a}_1 and \mathbf{a}_2 . In this problem, we discuss the least square solution to the system $A\mathbf{x} = \mathbf{b}$ as $\varepsilon \rightarrow 0$.

- Calculate the matrix $A^T A$, its inverse, and then $\hat{\mathbf{x}} = (A^T A)^{-1} A^T \mathbf{b}$ explicitly. Show that $\hat{\mathbf{x}}$ explodes as $\varepsilon \rightarrow 0$.
- Calculate the projection $A\hat{\mathbf{x}}$ of \mathbf{b} onto $\text{col}(A)$ and check that it does not depend on $\varepsilon > 0$. Explain the result.

Solution to the problem 9. .

(a)

$$\begin{aligned}
 A &= (\mathbf{a}_1 \quad \mathbf{a}_2) = (\mathbf{u}_1 \quad \mathbf{u}_1 + \varepsilon \mathbf{u}_2) & A^\top &= \begin{pmatrix} \mathbf{a}_1^\top \\ \mathbf{a}_2^\top \end{pmatrix} = \begin{pmatrix} \mathbf{u}_1^\top \\ \mathbf{u}_1^\top + \varepsilon \mathbf{u}_2^\top \end{pmatrix} \\
 A^\top A &= \begin{pmatrix} \mathbf{a}_1^\top \\ \mathbf{a}_2^\top \end{pmatrix} (\mathbf{a}_1 \quad \mathbf{a}_2) = \begin{pmatrix} \mathbf{u}_1^\top \\ \mathbf{u}_1^\top + \varepsilon \mathbf{u}_2^\top \end{pmatrix} (\mathbf{u}_1 \quad \mathbf{u}_1 + \varepsilon \mathbf{u}_2) = \begin{pmatrix} \mathbf{u}_1^\top \mathbf{u}_1 & \mathbf{u}_1^\top (\mathbf{u}_1 + \varepsilon \mathbf{u}_2) \\ (\mathbf{u}_1^\top + \varepsilon \mathbf{u}_2^\top) \mathbf{u}_1 & (\mathbf{u}_1^\top + \varepsilon \mathbf{u}_2^\top) (\mathbf{u}_1 + \varepsilon \mathbf{u}_2) \end{pmatrix} = \\
 &= \begin{pmatrix} \mathbf{u}_1^\top \mathbf{u}_1 & \mathbf{u}_1^\top \mathbf{u}_1 + \varepsilon \mathbf{u}_1^\top \mathbf{u}_2 \\ \mathbf{u}_1^\top \mathbf{u}_1 + \varepsilon \mathbf{u}_2^\top \mathbf{u}_1 & \mathbf{u}_1^\top \mathbf{u}_1 + \varepsilon (\mathbf{u}_2^\top \mathbf{u}_1 + \mathbf{u}_1^\top \mathbf{u}_2) + \varepsilon^2 \mathbf{u}_2^\top \mathbf{u}_2 \end{pmatrix}
 \end{aligned}$$

Since $\langle \mathbf{u}_1, \mathbf{u}_2 \rangle = 0$:

$$A^\top A = \begin{pmatrix} \|\mathbf{u}_1\|^2 & 0 \\ 0 & \|\mathbf{u}_1\|^2 + \varepsilon^2 \|\mathbf{u}_2\|^2 \end{pmatrix}$$

$$\det A^\top A = \|\mathbf{u}_1\|^2 (\|\mathbf{u}_1\|^2 + \varepsilon^2 \|\mathbf{u}_2\|^2) - 0 = \varepsilon^2 \|\mathbf{u}_1\|^2 \|\mathbf{u}_2\|^2$$

$$(A^\top A)^{-1} = \frac{1}{\varepsilon^2 \|\mathbf{u}_1\|^2 \|\mathbf{u}_2\|^2} \begin{pmatrix} \|\mathbf{u}_1\|^2 + \varepsilon^2 \|\mathbf{u}_2\|^2 & 0 \\ 0 & \|\mathbf{u}_1\|^2 \end{pmatrix} = \frac{1}{\varepsilon^2 \|\mathbf{u}_2\|^2} \begin{pmatrix} 1 & -1 \\ -1 & 1 \end{pmatrix} + \frac{1}{\|\mathbf{u}_1\|^2} \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$$

$$\begin{aligned}
 (A^\top A)^{-1} A^\top &= \frac{1}{\varepsilon^2 \|\mathbf{u}_2\|^2} \begin{pmatrix} 1 & -1 \\ -1 & 1 \end{pmatrix} \begin{pmatrix} \mathbf{u}_1^\top \\ \mathbf{u}_1^\top + \varepsilon \mathbf{u}_2^\top \end{pmatrix} + \frac{1}{\|\mathbf{u}_1\|^2} \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} \mathbf{u}_1^\top \\ \mathbf{u}_1^\top + \varepsilon \mathbf{u}_2^\top \end{pmatrix} = \\
 &= \frac{1}{\varepsilon^2 \|\mathbf{u}_2\|^2} \begin{pmatrix} \mathbf{u}_1^\top - \mathbf{u}_1^\top - \varepsilon \mathbf{u}_2^\top \\ -\mathbf{u}_1^\top + \mathbf{u}_1^\top + \varepsilon \mathbf{u}_2^\top \end{pmatrix} + \frac{1}{\|\mathbf{u}_1\|^2} \begin{pmatrix} \mathbf{u}_1^\top \\ 0 \end{pmatrix} = \frac{1}{\varepsilon \|\mathbf{u}_2\|^2} \begin{pmatrix} -\mathbf{u}_2^\top \\ \mathbf{u}_2^\top \end{pmatrix} + \frac{1}{\|\mathbf{u}_1\|^2} \begin{pmatrix} \mathbf{u}_1^\top \\ 0 \end{pmatrix} = \\
 &= \frac{1}{\varepsilon \|\mathbf{u}_2\|^2} \begin{pmatrix} -1 \\ 1 \end{pmatrix} \mathbf{u}_2^\top + \frac{1}{\|\mathbf{u}_1\|^2} \begin{pmatrix} 1 \\ 0 \end{pmatrix} \mathbf{u}_1^\top
 \end{aligned}$$

When \mathbf{b} is orthogonal to \mathbf{u}_2 , we have:

$$\hat{\mathbf{x}} = \frac{1}{\varepsilon \|\mathbf{u}_2\|^2} \begin{pmatrix} -1 \\ 1 \end{pmatrix} \mathbf{u}_2^\top \mathbf{b} + \frac{1}{\|\mathbf{u}_1\|^2} \begin{pmatrix} 1 \\ 0 \end{pmatrix} \mathbf{u}_1^\top \mathbf{b} = \frac{1}{\|\mathbf{u}_1\|^2} \begin{pmatrix} 1 \\ 0 \end{pmatrix} \mathbf{u}_1^\top \mathbf{b}$$

that doesn't depend on ε . In other cases the component $\frac{1}{\varepsilon \|\mathbf{u}_2\|^2} \begin{pmatrix} -1 \\ 1 \end{pmatrix} \mathbf{u}_2^\top \mathbf{b} \rightarrow \infty$ as $\varepsilon \rightarrow 0$ which leads to $\hat{\mathbf{x}}$ exploding as $\varepsilon \rightarrow 0$.

(b) The projection matrix:

$$\begin{aligned}
 P &= \frac{1}{\varepsilon \|\mathbf{u}_2\|^2} (\mathbf{u}_1 \quad \mathbf{u}_1 + \varepsilon \mathbf{u}_2) \begin{pmatrix} -1 \\ 1 \end{pmatrix} \mathbf{u}_2^\top + \frac{1}{\|\mathbf{u}_1\|^2} (\mathbf{u}_1 \quad \mathbf{u}_1 + \varepsilon \mathbf{u}_2) \begin{pmatrix} 1 \\ 0 \end{pmatrix} \mathbf{u}_1^\top = \\
 &= \frac{1}{\varepsilon \|\mathbf{u}_2\|^2} \varepsilon \mathbf{u}_2 \mathbf{u}_2^\top + \frac{1}{\|\mathbf{u}_1\|^2} \mathbf{u}_1 \mathbf{u}_1^\top = \frac{\mathbf{u}_2 \mathbf{u}_2^\top}{\|\mathbf{u}_2\|^2} + \frac{\mathbf{u}_1 \mathbf{u}_1^\top}{\|\mathbf{u}_1\|^2}
 \end{aligned}$$

does not depend on ε so neither does the projection $P\mathbf{b}$.

because $P\mathbf{b} = A\mathbf{x}$ is the projection onto $\text{Col}(A)$, which is independent of ε . On the other hand, \mathbf{x} gives coordinates of that $P\mathbf{b}$ in the basis \mathbf{a}_1 and \mathbf{a}_2 , which "worsens" as $\varepsilon \rightarrow 0$ and degenerates at $\varepsilon = 0$

3 pts

Problem 10 (Gram–Schmidt; 3pt). Use the Gram–Schmidt process to transform the basis $\mathbf{u}_1, \dots, \mathbf{u}_k$ into an orthonormal basis.

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

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

Solution to the problem 10. .

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

$$\mathbf{w}_1 = \mathbf{u}_1 = \begin{pmatrix} 1 \\ 3 \end{pmatrix} \quad \mathbf{w}_2 = \mathbf{u}_2 - \frac{\mathbf{w}_1 \mathbf{w}_1^\top}{\mathbf{w}_1^\top \mathbf{w}_1} \mathbf{u}_2$$

$$\frac{\mathbf{w}_1 \mathbf{w}_1^\top}{\mathbf{w}_1^\top \mathbf{w}_1} = \frac{1}{10} \begin{pmatrix} 1 \\ 3 \end{pmatrix} \begin{pmatrix} 1 & 3 \end{pmatrix} = \frac{1}{10} \begin{pmatrix} 1 & 3 \\ 3 & 9 \end{pmatrix} \quad \checkmark$$

$$\mathbf{w}_2 = \begin{pmatrix} 2 \\ -2 \end{pmatrix} - \frac{1}{10} \begin{pmatrix} 1 & 3 \\ 3 & 9 \end{pmatrix} \begin{pmatrix} 2 \\ -2 \end{pmatrix} = \frac{1}{10} \begin{pmatrix} 20 - 2 + 6 \\ -20 - 6 + 18 \end{pmatrix} = \frac{1}{5} \begin{pmatrix} 12 \\ -4 \end{pmatrix} \quad \checkmark$$

$$\hat{\mathbf{w}}_1 = \frac{\mathbf{w}_1}{\|\mathbf{w}_1\|_1} = \frac{1}{\sqrt{10}} \begin{pmatrix} 1 \\ 3 \end{pmatrix}$$

$$\hat{\mathbf{w}}_2 = \frac{\mathbf{w}_2}{\|\mathbf{w}_2\|_2} = \frac{1}{\sqrt{10}} \begin{pmatrix} 3 \\ -1 \end{pmatrix}$$

So the answer is $\hat{\mathbf{w}}_1 = \begin{pmatrix} 1/\sqrt{10} \\ 3/\sqrt{10} \end{pmatrix}, \hat{\mathbf{w}}_2 = \begin{pmatrix} 3/\sqrt{10} \\ -1/\sqrt{10} \end{pmatrix} \quad \checkmark$

it is easier to calculate $\mathbf{w}_1^\top \mathbf{u}_2$ first

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

$$\mathbf{w}_1 = \mathbf{u}_1 = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} \quad \mathbf{w}_2 = \mathbf{u}_2 - \frac{\mathbf{w}_1 \mathbf{w}_1^\top}{\mathbf{w}_1^\top \mathbf{w}_1} \mathbf{u}_2 \quad \mathbf{w}_3 = \mathbf{u}_3 - \frac{\mathbf{w}_1 \mathbf{w}_1^\top}{\mathbf{w}_1^\top \mathbf{w}_1} \mathbf{u}_3 - \frac{\mathbf{w}_2 \mathbf{w}_2^\top}{\mathbf{w}_2^\top \mathbf{w}_2} \mathbf{u}_3$$

$$\frac{\mathbf{w}_1 \mathbf{w}_1^\top}{\mathbf{w}_1^\top \mathbf{w}_1} = \frac{1}{2} \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} \begin{pmatrix} 1 & 0 & 1 \end{pmatrix} = \frac{1}{2} \begin{pmatrix} 1 & 0 & 1 \\ 0 & 0 & 0 \\ 1 & 0 & 1 \end{pmatrix} \quad \checkmark$$

$$\mathbf{w}_2 = \begin{pmatrix} 1 \\ 3 \\ -2 \end{pmatrix} - \frac{1}{2} \begin{pmatrix} 1 & 0 & 1 \\ 0 & 0 & 0 \\ 1 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ 3 \\ -2 \end{pmatrix} = \frac{1}{2} \begin{pmatrix} 2 - 1 + 2 \\ 6 - 0 \\ -4 - 1 + 2 \end{pmatrix} = \frac{1}{2} \begin{pmatrix} 3 \\ 6 \\ -3 \end{pmatrix} \quad \checkmark$$

$$\frac{\mathbf{w}_2 \mathbf{w}_2^\top}{\mathbf{w}_2^\top \mathbf{w}_2} = \frac{1}{54} \begin{pmatrix} 3 \\ 6 \\ -3 \end{pmatrix} \begin{pmatrix} 3 & 6 & -3 \end{pmatrix} = \frac{1}{54} \begin{pmatrix} 9 & 18 & -9 \\ 18 & 36 & -18 \\ -9 & -18 & 9 \end{pmatrix} = \frac{1}{6} \begin{pmatrix} 1 & 2 & -1 \\ 2 & 4 & -2 \\ -1 & -2 & 1 \end{pmatrix}$$

$$\mathbf{w}_3 = \begin{pmatrix} 0 \\ 2 \\ 1 \end{pmatrix} - \frac{1}{2} \begin{pmatrix} 1 & 0 & 1 \\ 0 & 0 & 0 \\ 1 & 0 & 1 \end{pmatrix} \begin{pmatrix} 0 \\ 2 \\ 1 \end{pmatrix} - \frac{1}{6} \begin{pmatrix} 1 & 2 & -1 \\ 2 & 4 & -2 \\ -1 & -2 & 1 \end{pmatrix} \begin{pmatrix} 0 \\ 2 \\ 1 \end{pmatrix} = \frac{1}{6} \begin{pmatrix} 0 - 3 - 4 + 1 \\ 12 - 0 - 8 + 2 \\ 6 - 3 + 4 - 1 \end{pmatrix} = \begin{pmatrix} -1 \\ 1 \\ 1 \end{pmatrix} \quad \checkmark$$

$$\hat{\mathbf{w}}_1 = \frac{\mathbf{w}_1}{\|\mathbf{w}\|_1} = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}$$

$$\hat{\mathbf{w}}_2 = \frac{\mathbf{w}_2}{\|\mathbf{w}\|_2} = \frac{1}{3\sqrt{6}} \begin{pmatrix} 3 \\ 6 \\ -3 \end{pmatrix}$$

$$\hat{\mathbf{w}}_3 = \frac{\mathbf{w}_3}{\|\mathbf{w}\|_3} = \frac{1}{\sqrt{3}} \begin{pmatrix} -1 \\ 1 \\ 1 \end{pmatrix}$$

So the answer is $\hat{\mathbf{w}}_1 = \begin{pmatrix} 1/\sqrt{2} \\ 0 \\ 1/\sqrt{2} \end{pmatrix}$, $\hat{\mathbf{w}}_2 = \begin{pmatrix} 1/\sqrt{6} \\ 2/\sqrt{6} \\ -1/\sqrt{6} \end{pmatrix}$, $\hat{\mathbf{w}}_3 = \begin{pmatrix} -1/\sqrt{3} \\ 1/\sqrt{3} \\ 1/\sqrt{3} \end{pmatrix}$. ✓

Problem 11 (QR; 5pt). Find the QR -decomposition of the matrices below using the Gram–Schmidt algorithm:

5 pts

$$(a) \begin{pmatrix} 1 & -1 \\ 2 & 3 \end{pmatrix}; \quad (b) \begin{pmatrix} 1 & 2 \\ 0 & 1 \\ 1 & 4 \end{pmatrix}; \quad (c) \begin{pmatrix} 1 & 0 & 2 \\ 0 & 1 & 1 \\ 2 & 0 & 1 \end{pmatrix}$$

Solution to the problem 11. .

(a)

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

$$q_1 = \begin{pmatrix} 1 \\ 2 \end{pmatrix} \quad \frac{q_1 q_1^\top}{q_1^\top q_1} = \frac{1}{5} \begin{pmatrix} 1 & 2 \\ 2 & 4 \end{pmatrix}$$

$$q_2 = \begin{pmatrix} -1 \\ 3 \end{pmatrix} - \frac{1}{5} \begin{pmatrix} 1 & 2 \\ 2 & 4 \end{pmatrix} \begin{pmatrix} -1 \\ 3 \end{pmatrix} = \frac{1}{5} \begin{pmatrix} -5 + 1 - 6 \\ 15 + 2 - 12 \end{pmatrix} = \begin{pmatrix} -2 \\ 1 \end{pmatrix}$$

$$\hat{q}_1 = \begin{pmatrix} 1/\sqrt{5} \\ 2/\sqrt{5} \end{pmatrix} \quad \hat{q}_2 = \begin{pmatrix} -2/\sqrt{5} \\ 1/\sqrt{5} \end{pmatrix} \quad Q = \begin{pmatrix} 1/\sqrt{5} & -2/\sqrt{5} \\ 2/\sqrt{5} & 1/\sqrt{5} \end{pmatrix}$$

$$R = \begin{pmatrix} \hat{q}_1^\top a_1 & \hat{q}_1^\top a_2 \\ 0 & \hat{q}_2^\top a_2 \end{pmatrix} = \begin{pmatrix} 5/\sqrt{5} & 5/\sqrt{5} \\ 0 & 5/\sqrt{5} \end{pmatrix}$$

$$\Rightarrow A = \begin{pmatrix} 1 & -1 \\ 2 & 3 \end{pmatrix} = \begin{pmatrix} 1/\sqrt{5} & -2/\sqrt{5} \\ 2/\sqrt{5} & 1/\sqrt{5} \end{pmatrix} \begin{pmatrix} \sqrt{5} & \sqrt{5} \\ 0 & \sqrt{5} \end{pmatrix} \quad \checkmark$$

(b)

$$A = \begin{pmatrix} 1 & 2 \\ 0 & 1 \\ 1 & 4 \end{pmatrix}$$

$$q_1 = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} \quad \frac{q_1 q_1^\top}{q_1^\top q_1} = \frac{1}{2} \begin{pmatrix} 1 & 0 & 1 \\ 0 & 0 & 0 \\ 1 & 0 & 1 \end{pmatrix}$$

$$q_2 = \begin{pmatrix} 2 \\ 1 \\ 4 \end{pmatrix} - \frac{1}{2} \begin{pmatrix} 1 & 0 & 1 \\ 0 & 0 & 0 \\ 1 & 0 & 1 \end{pmatrix} \begin{pmatrix} 2 \\ 1 \\ 4 \end{pmatrix} = \frac{1}{2} \begin{pmatrix} 4-2-4 \\ 2-0 \\ 8-2-4 \end{pmatrix} = \begin{pmatrix} -1 \\ 1 \\ 1 \end{pmatrix} \quad \checkmark$$

$$\hat{q}_1 = \begin{pmatrix} 1/\sqrt{2} \\ 0 \\ 1/\sqrt{2} \end{pmatrix} \quad \hat{q}_2 = \begin{pmatrix} -1/\sqrt{3} \\ 1/\sqrt{3} \\ 1/\sqrt{3} \end{pmatrix} \quad Q = \begin{pmatrix} 1/\sqrt{2} & -1/\sqrt{3} \\ 0 & 1/\sqrt{3} \\ 1/\sqrt{2} & 1/\sqrt{3} \end{pmatrix} \quad \checkmark$$

$$R = \begin{pmatrix} \hat{q}_1^\top a_1 & \hat{q}_1^\top a_2 \\ 0 & \hat{q}_2^\top a_2 \end{pmatrix} = \begin{pmatrix} 2/\sqrt{2} & 6/\sqrt{2} \\ 0 & 3/\sqrt{3} \end{pmatrix}$$

$$\Rightarrow A = \begin{pmatrix} 1 & 2 \\ 0 & 1 \\ 1 & 4 \end{pmatrix} = \begin{pmatrix} 1/\sqrt{2} & -1/\sqrt{3} \\ 0 & 1/\sqrt{3} \\ 1/\sqrt{2} & 1/\sqrt{3} \end{pmatrix} \begin{pmatrix} \sqrt{2} & 3\sqrt{2} \\ 0 & \sqrt{3} \end{pmatrix} \quad \checkmark$$

(c)

$$A = \begin{pmatrix} 1 & 0 & 2 \\ 0 & 1 & 1 \\ 2 & 0 & 1 \end{pmatrix}$$

$$q_1 = \begin{pmatrix} 1 \\ 0 \\ 2 \end{pmatrix} \quad \frac{q_1 q_1^\top}{q_1^\top q_1} = \frac{1}{5} \begin{pmatrix} 1 & 0 & 2 \\ 0 & 0 & 0 \\ 2 & 0 & 4 \end{pmatrix} \quad \checkmark$$

$$q_2 = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} - \frac{1}{5} \begin{pmatrix} 1 & 0 & 2 \\ 0 & 0 & 0 \\ 2 & 0 & 4 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} \quad \checkmark \quad \frac{q_2 q_2^\top}{q_2^\top q_2} = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

$$q_3 = \begin{pmatrix} 2 \\ 1 \\ 1 \end{pmatrix} - -\frac{1}{5} \begin{pmatrix} 1 & 0 & 2 \\ 0 & 0 & 0 \\ 2 & 0 & 4 \end{pmatrix} \begin{pmatrix} 2 \\ 1 \\ 1 \end{pmatrix} - \begin{pmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} 2 \\ 1 \\ 1 \end{pmatrix} = \frac{1}{5} \begin{pmatrix} 10-2-2 \\ 5-5-0 \\ 5-4-4 \end{pmatrix} = \frac{1}{5} \begin{pmatrix} 6 \\ 0 \\ -3 \end{pmatrix} \quad \checkmark$$

$$\hat{q}_1 = \begin{pmatrix} 1/\sqrt{5} \\ 0 \\ 2/\sqrt{5} \end{pmatrix} \quad \hat{q}_2 = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} \quad \hat{q}_3 = \begin{pmatrix} 2/\sqrt{5} \\ 0 \\ -1/\sqrt{5} \end{pmatrix} \quad Q = \begin{pmatrix} 1/\sqrt{5} & 0 & 2/\sqrt{5} \\ 0 & 1 & 0 \\ 2/\sqrt{5} & 0 & -1/\sqrt{5} \end{pmatrix}$$

$$R = \begin{pmatrix} \hat{q}_1^\top a_1 & \hat{q}_1^\top a_2 & \hat{q}_1^\top a_3 \\ 0 & \hat{q}_2^\top a_2 & \hat{q}_2^\top a_3 \\ 0 & 0 & \hat{q}_3^\top a_3 \end{pmatrix} = \begin{pmatrix} 5/\sqrt{5} & 0 & 4/\sqrt{5} \\ 0 & 1 & 1 \\ 0 & 0 & 3/\sqrt{5} \end{pmatrix}$$

$$\Rightarrow A = \begin{pmatrix} 1 & 0 & 2 \\ 0 & 1 & 1 \\ 2 & 0 & 1 \end{pmatrix} = \begin{pmatrix} 1/\sqrt{5} & 0 & 2/\sqrt{5} \\ 0 & 1 & 0 \\ 2/\sqrt{5} & 0 & -1/\sqrt{5} \end{pmatrix} \begin{pmatrix} \sqrt{5} & 0 & 4/\sqrt{5} \\ 0 & 1 & 1 \\ 0 & 0 & 3/\sqrt{5} \end{pmatrix} \quad \checkmark$$

Problem 12 (Householder reflection and QR; 7 pts). (a) Find the unit vector $\mathbf{u} \in \mathbb{R}^2$ such that the Householder reflection $Q_{\mathbf{u}} := I - 2\mathbf{u}\mathbf{u}^\top$ maps the vector $(1, 2)^\top$ onto a vector collinear to $(1, 0)^\top$

7 pts

(b) explain how $Q_{\mathbf{u}}$ helps to derive the QR factorization of the matrix (a) of Problem 11.

- (c) Find the QR -factorization of matrices in (b) and (c) of Problem 11 using the Householder reflections approach.

Solution to the problem 12. .

(a)

$$\mathbf{x} = \begin{pmatrix} 1 \\ 2 \end{pmatrix} \quad Q_{\mathbf{u}}\mathbf{x} = \|\mathbf{x}\| \begin{pmatrix} 1 \\ 0 \end{pmatrix} \quad Q_{\mathbf{u}} = I - 2\mathbf{u}\mathbf{u}^T \quad (\|\mathbf{u}\| = 1)$$

To get \mathbf{u} we need to subtract $\begin{pmatrix} 1 \\ 0 \end{pmatrix}$ from \mathbf{x} and normalize the vector:

$$\hat{\mathbf{u}} = \begin{pmatrix} 1 \\ 2 \end{pmatrix} - \sqrt{5} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 1 - \sqrt{5} \\ 2 \end{pmatrix} \quad \mathbf{u} = \frac{\hat{\mathbf{u}}}{\|\hat{\mathbf{u}}\|} = \frac{1}{\sqrt{10 - 2\sqrt{5}}} \begin{pmatrix} 1 - \sqrt{5} \\ 2 \end{pmatrix} \quad \checkmark$$

$$\begin{aligned} \mathbf{u}\mathbf{u}^T &= \frac{1}{10 - 2\sqrt{5}} \begin{pmatrix} 1 - \sqrt{5} \\ 2 \end{pmatrix} \begin{pmatrix} 1 - \sqrt{5} & 2 \end{pmatrix} = \frac{1}{10 - 2\sqrt{5}} \begin{pmatrix} 6 - 2\sqrt{5} & 2 - 2\sqrt{5} \\ 2 - 2\sqrt{5} & 4 \end{pmatrix} = \\ &= \frac{1}{5 - \sqrt{5}} \begin{pmatrix} 3 - \sqrt{5} & 1 - \sqrt{5} \\ 1 - \sqrt{5} & 2 \end{pmatrix} = \frac{5 + \sqrt{5}}{20} \begin{pmatrix} 3 - \sqrt{5} & 1 - \sqrt{5} \\ 1 - \sqrt{5} & 2 \end{pmatrix} = \frac{1}{10} \begin{pmatrix} 5 - \sqrt{5} & -2\sqrt{5} \\ -2\sqrt{5} & 5 + \sqrt{5} \end{pmatrix} \quad \checkmark \\ \Rightarrow Q_{\mathbf{u}} &= \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} - \frac{1}{5} \begin{pmatrix} 5 - \sqrt{5} & -2\sqrt{5} \\ -2\sqrt{5} & 5 + \sqrt{5} \end{pmatrix} = \frac{1}{5} \begin{pmatrix} \sqrt{5} & 2\sqrt{5} \\ 2\sqrt{5} & -\sqrt{5} \end{pmatrix} = \frac{\sqrt{5}}{5} \begin{pmatrix} 1 & 2 \\ 2 & -1 \end{pmatrix} \quad \checkmark \end{aligned}$$

And we can check:

$$Q_{\mathbf{u}}\mathbf{x} = \frac{\sqrt{5}}{5} \begin{pmatrix} 1 & 2 \\ 2 & -1 \end{pmatrix} \begin{pmatrix} 1 \\ 2 \end{pmatrix} = \sqrt{5} \begin{pmatrix} 1 \\ 0 \end{pmatrix} \quad \|\mathbf{x}\| = \sqrt{5} = \|\mathbf{x}\| \quad \checkmark$$

So the answer is:

$$\mathbf{u} = \frac{1}{\sqrt{10 - 2\sqrt{5}}} \begin{pmatrix} 1 - \sqrt{5} \\ 2 \end{pmatrix} \quad \checkmark$$

(b)

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

Since $\mathbf{x} = \mathbf{a}_1$, $Q_{\mathbf{u}}$ is thus the result of the first iteration of QR factorization using the Householder reflections and:

$$Q_{\mathbf{u}}A = \begin{pmatrix} \sqrt{5} & \sqrt{5} \\ 0 & -\sqrt{5} \end{pmatrix} \quad \checkmark$$

give us the first row $(\sqrt{5} \quad \sqrt{5})$ of the R matrix .

the first column $(\sqrt{5}; 0)^T$

- (c) QR -factorization of Problem 11 (b), (c) using the Householder reflections approach:
11. (b)

$$A = \begin{pmatrix} 1 & 2 \\ 0 & 1 \\ 1 & 4 \end{pmatrix} \quad \mathbf{a}_1 = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}$$

$$\hat{\mathbf{u}}_1 = \mathbf{a}_1 - \|\mathbf{a}_1\| \mathbf{e}_1 = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} - \sqrt{2} \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 1 - \sqrt{2} \\ 0 \\ 1 \end{pmatrix} \quad \mathbf{u}_1 = \frac{1}{\sqrt{4 - 2\sqrt{2}}} \begin{pmatrix} 1 - \sqrt{2} \\ 0 \\ 1 \end{pmatrix}$$

$$\begin{aligned}
\mathbf{u}_1 \mathbf{u}_1^\top &= \frac{1}{4-2\sqrt{2}} \begin{pmatrix} 1-\sqrt{2} \\ 0 \\ 1 \end{pmatrix} (1-\sqrt{2} \quad 0 \quad 1) = \frac{1}{4-2\sqrt{2}} \begin{pmatrix} 3-2\sqrt{2} & 0 & 1-\sqrt{2} \\ 0 & 0 & 0 \\ 1-\sqrt{2} & 0 & 1 \end{pmatrix} = \\
&= \frac{4+2\sqrt{2}}{8} \begin{pmatrix} 3-2\sqrt{2} & 0 & 1-\sqrt{2} \\ 0 & 0 & 0 \\ 1-\sqrt{2} & 0 & 1 \end{pmatrix} = \frac{1}{4} \begin{pmatrix} 2-\sqrt{2} & 0 & -\sqrt{2} \\ 0 & 0 & 0 \\ -\sqrt{2} & 0 & 2+\sqrt{2} \end{pmatrix} \\
&\Rightarrow Q_{\mathbf{u}_1} = \frac{1}{2} \begin{pmatrix} \sqrt{2} & 0 & \sqrt{2} \\ 0 & 2 & 0 \\ \sqrt{2} & 0 & -\sqrt{2} \end{pmatrix}
\end{aligned}$$

$$Q_{\mathbf{u}_1} A = \frac{1}{2} \begin{pmatrix} \sqrt{2} & 0 & \sqrt{2} \\ 0 & 2 & 0 \\ \sqrt{2} & 0 & -\sqrt{2} \end{pmatrix} \begin{pmatrix} 1 & 2 \\ 0 & 1 \\ 1 & 4 \end{pmatrix} = \begin{pmatrix} \sqrt{2} & 3\sqrt{2} \\ 0 & 1 \\ 0 & -\sqrt{2} \end{pmatrix}$$

$$\mathbf{a}_2 = \begin{pmatrix} 1 \\ -\sqrt{2} \end{pmatrix} (\in \mathbb{R}^2)$$

$$\hat{\mathbf{u}}_2 = \mathbf{a}_2 - \|\mathbf{a}_2\| \mathbf{e}_1 = \begin{pmatrix} 1 \\ -\sqrt{2} \end{pmatrix} - \sqrt{3} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 1-\sqrt{3} \\ -\sqrt{2} \end{pmatrix} \quad \mathbf{u}_2 = \frac{1}{\sqrt{6-2\sqrt{3}}} \begin{pmatrix} 1-\sqrt{3} \\ -\sqrt{2} \end{pmatrix}$$

$$\begin{aligned}
\mathbf{u}_2 \mathbf{u}_2^\top &= \frac{1}{6-2\sqrt{3}} \begin{pmatrix} 1-\sqrt{3} \\ -\sqrt{2} \end{pmatrix} (1-\sqrt{3} \quad -\sqrt{2}) = \frac{1}{6-2\sqrt{3}} \begin{pmatrix} 4-2\sqrt{3} & -\sqrt{2}(1-\sqrt{3}) \\ -\sqrt{2}(1-\sqrt{3}) & 2 \end{pmatrix} = \\
&= \frac{3+\sqrt{3}}{12} \begin{pmatrix} 4-2\sqrt{3} & -\sqrt{2}(1-\sqrt{3}) \\ -\sqrt{2}(1-\sqrt{3}) & 2 \end{pmatrix} = \frac{1}{6} \begin{pmatrix} 3-\sqrt{3} & \sqrt{6} \\ \sqrt{6} & 3+\sqrt{3} \end{pmatrix}
\end{aligned}$$

$$\Rightarrow \hat{Q}_{\mathbf{u}_2} = \frac{1}{3} \begin{pmatrix} \sqrt{3} & -\sqrt{6} \\ -\sqrt{6} & -\sqrt{3} \end{pmatrix} \Rightarrow Q_{\mathbf{u}_2} = \frac{1}{3} \begin{pmatrix} 3 & 0 & 0 \\ 0 & \sqrt{3} & -\sqrt{6} \\ 0 & -\sqrt{6} & -\sqrt{3} \end{pmatrix}$$

$$Q_{\mathbf{u}_2} Q_{\mathbf{u}_1} A = \frac{1}{3} \begin{pmatrix} 3 & 0 & 0 \\ 0 & \sqrt{3} & -\sqrt{6} \\ 0 & -\sqrt{6} & -\sqrt{3} \end{pmatrix} \begin{pmatrix} \sqrt{2} & 3\sqrt{2} \\ 0 & 1 \\ 0 & -\sqrt{2} \end{pmatrix} = \begin{pmatrix} \sqrt{2} & 3\sqrt{2} \\ 0 & \sqrt{3} \\ 0 & 0 \end{pmatrix}$$

So the full QR factorization is:

$$\begin{aligned}
R &= \begin{pmatrix} \sqrt{2} & 3\sqrt{2} \\ 0 & \sqrt{3} \\ 0 & 0 \end{pmatrix} \quad Q = Q_{\mathbf{u}_1} Q_{\mathbf{u}_2} = \frac{1}{6} \begin{pmatrix} \sqrt{2} & 0 & \sqrt{2} \\ 0 & 2 & 0 \\ \sqrt{2} & 0 & -\sqrt{2} \end{pmatrix} \begin{pmatrix} 3 & 0 & 0 \\ 0 & \sqrt{3} & -\sqrt{6} \\ 0 & -\sqrt{6} & -\sqrt{3} \end{pmatrix} = \\
&= \begin{pmatrix} 1/\sqrt{2} & -1/\sqrt{3} & -1/\sqrt{6} \\ 0 & 1/\sqrt{3} & -2/\sqrt{6} \\ 1/\sqrt{2} & 1/\sqrt{3} & 1/\sqrt{6} \end{pmatrix} \quad \checkmark
\end{aligned}$$

And the reduced QR factorization is:

$$Q = \begin{pmatrix} 1/\sqrt{2} & -1/\sqrt{3} \\ 0 & 1/\sqrt{3} \\ 1/\sqrt{2} & 1/\sqrt{3} \end{pmatrix} \quad R = \begin{pmatrix} \sqrt{2} & 3\sqrt{2} \\ 0 & \sqrt{3} \end{pmatrix} \quad \checkmark$$

that matches the result of GS approach.

11. (c)

$$A = \begin{pmatrix} 1 & 0 & 2 \\ 0 & 1 & 1 \\ 2 & 0 & 1 \end{pmatrix} \quad \mathbf{a}_1 = \begin{pmatrix} 1 \\ 0 \\ 2 \end{pmatrix}$$

$$\hat{\mathbf{u}}_1 = \mathbf{a}_1 - \|\mathbf{a}_1\| \mathbf{e}_1 = \begin{pmatrix} 1 \\ 0 \\ 2 \end{pmatrix} - \sqrt{5} \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 1 - \sqrt{5} \\ 0 \\ 2 \end{pmatrix} \quad \mathbf{u}_1 = \frac{1}{\sqrt{10 - 2\sqrt{5}}} \begin{pmatrix} 1 - \sqrt{5} \\ 0 \\ 2 \end{pmatrix}$$

$$\begin{aligned} \mathbf{u}_1 \mathbf{u}_1^\top &= \frac{1}{10 - 2\sqrt{5}} \begin{pmatrix} 1 - \sqrt{5} \\ 0 \\ 2 \end{pmatrix} (1 - \sqrt{5} \quad 0 \quad 2) = \frac{1}{5 - \sqrt{5}} \begin{pmatrix} 3 - \sqrt{5} & 0 & 1 - \sqrt{5} \\ 0 & 0 & 0 \\ 1 - \sqrt{5} & 0 & 2 \end{pmatrix} = \\ &= \frac{5 + \sqrt{5}}{20} \begin{pmatrix} 3 - \sqrt{5} & 0 & 1 - \sqrt{5} \\ 0 & 0 & 0 \\ 1 - \sqrt{5} & 0 & 2 \end{pmatrix} = \frac{1}{10} \begin{pmatrix} 5 - \sqrt{5} & 0 & -2\sqrt{5} \\ 0 & 0 & 0 \\ -2\sqrt{5} & 0 & 5 + \sqrt{5} \end{pmatrix} \end{aligned}$$

$$\Rightarrow Q_{\mathbf{u}_1} = \frac{1}{5} \begin{pmatrix} \sqrt{5} & 0 & 2\sqrt{5} \\ 0 & 5 & 0 \\ 2\sqrt{5} & 0 & -\sqrt{5} \end{pmatrix}$$

$$Q_{\mathbf{u}_1} A = \frac{1}{5} \begin{pmatrix} \sqrt{5} & 0 & 2\sqrt{5} \\ 0 & 5 & 0 \\ 2\sqrt{5} & 0 & -\sqrt{5} \end{pmatrix} \begin{pmatrix} 1 & 0 & 2 \\ 0 & 1 & 1 \\ 2 & 0 & 1 \end{pmatrix} = \begin{pmatrix} \sqrt{5} & 0 & \frac{4}{5}\sqrt{5} \\ 0 & 1 & 1 \\ 0 & 0 & \frac{3}{5}\sqrt{5} \end{pmatrix}$$

$$\mathbf{a}_2 = \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \mathbf{e}_1 (\in \mathbb{R}^2) \Rightarrow Q_{\mathbf{u}_2} = I$$

$$Q_{\mathbf{u}_2} Q_{\mathbf{u}_1} A = \begin{pmatrix} \sqrt{5} & 0 & 4/\sqrt{5} \\ 0 & 1 & 1 \\ 0 & 0 & 3/\sqrt{5} \end{pmatrix}$$

$$\Rightarrow R = \begin{pmatrix} \sqrt{5} & 0 & 4/\sqrt{5} \\ 0 & 1 & 1 \\ 0 & 0 & 3/\sqrt{5} \end{pmatrix} \quad Q = Q_{\mathbf{u}_1} Q_{\mathbf{u}_2} = Q_{\mathbf{u}_1} = \begin{pmatrix} 1/\sqrt{5} & 0 & 2/\sqrt{5} \\ 0 & 1 & 0 \\ 2/\sqrt{5} & 0 & -1/\sqrt{5} \end{pmatrix} \quad \checkmark$$

that matches the result of GS approach.