

Honor code: I pledge on my honor that: I have completed all steps in the below homework on my own, I have not used any unauthorized materials while completing this homework, and I have not given anyone else access to my homework.



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1. (1 point) Have you read and understood the honor code?

Solution: Yes

Concept: Projection

2. (2 points) Consider a matrix A and a vector \mathbf{b} which does not lie in the column space of A . Let \mathbf{p} be the projection of \mathbf{b} on to the column space of A . If $A = \begin{bmatrix} 1 & 3 \\ 0 & 2 \\ 2 & 0 \\ 3 & 1 \end{bmatrix}$ and

$$\mathbf{p} = \begin{bmatrix} 7 \\ 4 \\ 2 \\ 5 \end{bmatrix}, \text{ find } \mathbf{b}.$$

Solution:

We are given A and p . Since A has 2 columns and both are independent and has 3 rows, column space of A is a plane in 3D space. Therefore, any vector which starts at origin and ends at a point on the line normal to the plane and passing through the point at the end of p will be projected onto the plane as p .

Therefore, $b = p + ke$ where e is a unit vector perpendicular to the column space plane and $k \in R$.

Therefore to find a normal to column space plane, it should be perpendicular to all column vectors of A , i.e. $A^T e = 0$, if $e = ([e_1 \ e_2 \ e_3 \ e_4])^T$, solutions to $A^T e = 0$ is the null space of A^T .

Getting row reduced echelon form of A^T , $\begin{bmatrix} 1 & 0 & 2 & 3 \\ 3 & 2 & 0 & 1 \end{bmatrix}$, $R2 = R2 - 3R1$

$$\Rightarrow \begin{bmatrix} 1 & 0 & 2 & 3 \\ 0 & 2 & -6 & -8 \end{bmatrix}, R2 = \frac{1}{2}R2 \Rightarrow \begin{bmatrix} 1 & 0 & 2 & 3 \\ 0 & 1 & -3 & -4 \end{bmatrix}$$

Hence, null space of A^T has basis vectors, $\begin{bmatrix} -2 \\ 3 \\ 1 \\ 0 \end{bmatrix}$ and $\begin{bmatrix} -3 \\ 4 \\ 0 \\ 1 \end{bmatrix}$

Hence any linear combination of these two vectors can be taken as the vector e .

$$e = \begin{bmatrix} -2p - 3q \\ 3p + 4q \\ p \\ q \end{bmatrix} \text{ and so, } b = p + e = \begin{bmatrix} 7 - 2p - 3q \\ 4 + 3p + 4q \\ 2 + p \\ 5 + q \end{bmatrix} \text{ where } p, q \in R.$$

Example, $\begin{bmatrix} 7 \\ 4 \\ 2 \\ 5 \end{bmatrix}$.

3. (2 points) Consider the following statement: Two vectors \mathbf{b}_1 and \mathbf{b}_2 cannot have the same projection \mathbf{p} on the column space of A .

(a) Give one example where the above statement is True.

Solution:

If we consider, $A = I = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$, $A = A^\top = A^{-1} = I$.

Hence, $P = A(A^\top A)^{-1}A^\top = I(II)^{-1}I = I(I)^{-1}I = I(I)I = I$.

For any vector b , its projection on column space of A is $p = Pb = Ib = b$.

Hence for any two vectors b_1 and b_2 , their projections on column space of A will be b_1 and b_2 itself. Hence if they are two different vectors, their projections will also be different.

Hence the statement is True for this A .

- (b) Give one example where the above statement is False.

Solution:

If we consider, $A = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$, $A^\top = [1 \ 0]$,

$A^\top A = [1 \ 0] \begin{bmatrix} 1 \\ 0 \end{bmatrix} = [1]$, Therefore, $(A^\top A)^{-1} = [-1]$.

Hence, $P = A(A^\top A)^{-1}A^\top = \begin{bmatrix} 1 \\ 0 \end{bmatrix} [-1] [1 \ 0] = \begin{bmatrix} -1 \\ 0 \end{bmatrix} [1 \ 0] = \begin{bmatrix} -1 & 0 \\ 0 & 0 \end{bmatrix}$.

Therefore, for any vector b , its projection p is, $Pb = p$.

If we take $b_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$ and $b_2 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$, their projections are,

$p_1 = \begin{bmatrix} -1 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} -1 \\ 0 \end{bmatrix}$ and $p_2 = \begin{bmatrix} -1 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} -1 \\ 0 \end{bmatrix}$

Clearly there are 2 vectors such that their projections are same on column space of A . Hence the statement is False for this A .

- (c) Based on the above examples, state the generic condition under which the above statement will be True or False.

Solution:

The condition is False except when the matrix A is invertible. This is because if A is invertible, A^\top is invertible and the projection matrix can be rewritten as, $p = Pb = A(A^\top A)^{-1}A^\top b = AA^{-1}(A^\top)^{-1}A^\top b = I Ib = Ib = b. \Rightarrow p = b$.

Hence for every vector b , its projection is itself and hence for any two different vectors b_1 and b_2 , their projections can never be the same.

4. (2 points) (a) Find the projection matrix P_1 that projects onto the line through $\mathbf{a} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$ and also the matrix P_2 that projects onto the line perpendicular to \mathbf{a} .

Solution:

For P_1 ,

for line passing through $a = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$, we consider $A = a$ and hence,

$$P_1 = A(A^T A)^{-1} A^T = \begin{bmatrix} 1 \\ 2 \end{bmatrix} ([1 \ 2] \begin{bmatrix} 1 \\ 2 \end{bmatrix})^{-1} [1 \ 2] = \begin{bmatrix} 1 \\ 2 \end{bmatrix} ([5])^{-1} [1 \ 2] = \begin{bmatrix} 1 \\ 2 \end{bmatrix} ([\frac{1}{5}]) [1 \ 2] = \frac{1}{5} \begin{bmatrix} 1 \\ 2 \end{bmatrix} [1 \ 2] = \frac{1}{5} \begin{bmatrix} 1 & 2 \\ 2 & 4 \end{bmatrix} \text{ Hence, } P_1 = \frac{1}{5} \begin{bmatrix} 1 & 2 \\ 2 & 4 \end{bmatrix}.$$

For P_2 ,

for line perpendicular to $a = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$, if a_p is a vector along that perpendicular line,

$$a_p = \begin{bmatrix} 2 \\ -1 \end{bmatrix} \text{ as } a^T \cdot a_p = 1 * 2 - 2 * 1 = 0.$$

we consider $A = a_p$ and hence,

$$P_2 = A(A^T A)^{-1} A^T = \begin{bmatrix} 2 \\ -1 \end{bmatrix} ([2 \ -1] \begin{bmatrix} 2 \\ -1 \end{bmatrix})^{-1} [2 \ -1] = \begin{bmatrix} 2 \\ -1 \end{bmatrix} ([5])^{-1} [2 \ -1] = \begin{bmatrix} 2 \\ -1 \end{bmatrix} ([\frac{1}{5}]) [2 \ -1] = \frac{1}{5} \begin{bmatrix} 2 \\ -1 \end{bmatrix} [2 \ -1] = \frac{1}{5} \begin{bmatrix} 4 & -2 \\ -2 & 1 \end{bmatrix} \text{ Hence, } P_2 = \frac{1}{5} \begin{bmatrix} 4 & -2 \\ -2 & 1 \end{bmatrix}.$$

- (b) Compute $P_1 + P_2$ and $P_1 P_2$ and explain the result.

Solution:

$$P_1 = \frac{1}{5} \begin{bmatrix} 1 & 2 \\ 2 & 4 \end{bmatrix} \text{ and } P_2 = \frac{1}{5} \begin{bmatrix} 4 & -2 \\ -2 & 1 \end{bmatrix}$$

$$P_1 + P_2 = \frac{1}{5} \begin{bmatrix} 5 & 0 \\ 0 & 5 \end{bmatrix} = I$$

Projection of b along a line and perpendicular to the same line produces 2 perpendicular vectors whose sum gives back b . If the projections are p_{p1} and p_{p2} , we can write, $b = p_{p1} + p_{p2}$.

$$\Rightarrow b = (P_1 b) + (P_2 b) \Rightarrow b = (P_1 + P_2)b.$$

Hence we get $P_1 + P_2$ as I .

$$P_1 P_2 = \frac{1}{5} \begin{bmatrix} 1 & 2 \\ 2 & 4 \end{bmatrix} \frac{1}{5} \begin{bmatrix} 4 & -2 \\ -2 & 1 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$$

Here, when we do $P_1 P_2$, we are simply projecting the columns of P_2 onto the subspace corresponding to P_1 , i.e. along the line through vector a .

But as P_2 corresponds to projecting onto line perpendicular to vector a , columns of P_2 will be along the perpendicular line and hence will be perpendicular to the vector a .

Hence projection of every column of P_2 onto line through vector a will 0 as they are perpendicular.

Hence we get a 0 matrix.

Concept: Dot product of vectors

5. (1 point) For all the vectors $u, v \in \mathbb{R}^n$, $u^T v \leq \|u\|_2 \|v\|_2$.
Prove the statement if true, or give counterexample if false.

Solution:

$u^T v$ is same as dot product, $u \cdot v$.

Also, $u \cdot v = \|u\|_2 \|v\|_2 \cos \theta$ where θ is the angle between the two vectors u and v .

Since $\cos \theta$ always lies in the range $[-1, 1]$,

$\|u\|_2 \|v\|_2 \cos \theta$ lies in range $[-\|u\|_2 \|v\|_2, \|u\|_2 \|v\|_2]$.

Therefore, $u^T v$ lies in range $[-\|u\|_2 \|v\|_2, \|u\|_2 \|v\|_2]$.

Therefore, $u^T v \leq \|u\|_2 \|v\|_2$ for all vectors $u, v \in \mathbb{R}^n$.

Hence Proved.

Concept: Vector norms

6. (1 point) The L_p -norm of a vector $\mathbf{x} = [x_1, x_2, \dots, x_n] \in \mathbb{R}^n$ is defined as:

$$\|\mathbf{x}\|_p = (|x_1|^p + |x_2|^p + |x_3|^p + \dots + |x_n|^p)^{\frac{1}{p}}$$

(a) Prove that $\|\mathbf{x}\|_\infty = \max_{1 \leq i \leq n} |x_i|$

Solution:

$$\|\mathbf{x}\|_\infty = \lim_{p \rightarrow +\infty} \|\mathbf{x}\|_p = \lim_{p \rightarrow \infty} (|x_1|^p + \dots + |x_n|^p)^{\frac{1}{p}}$$

Suppose $\max_{1 \leq i \leq n} |x_i| = |x_j|$, Is we take out that term in the summation,

$$= \lim_{p \rightarrow \infty} (|x_j|^p \left(\frac{|x_1|^p}{|x_j|^p} + \dots + 1 + \dots + \frac{|x_n|^p}{|x_j|^p} \right))^{\frac{1}{p}}.$$

$$= \lim_{p \rightarrow \infty} (|x_j|^p \left(\left(\frac{|x_1|}{|x_j|} \right)^p + \dots + 1 + \dots + \left(\frac{|x_n|}{|x_j|} \right)^p \right))^{\frac{1}{p}}.$$

Since x_j is the largest element, $\frac{x_i}{x_j} < 1$ for $1 \leq i \leq n$ and $i \neq j$.

$$\text{Therefore, as } p \rightarrow \infty, \left(\frac{|x_1|}{|x_j|} \right)^p \rightarrow 0 \Rightarrow \lim_{p \rightarrow \infty} (|x_j|^p (0 + \dots + 1 + \dots + 0))^{\frac{1}{p}}.$$

$$= \lim_{p \rightarrow \infty} (|x_j|^p (1))^{\frac{1}{p}} = \lim_{p \rightarrow \infty} (|x_j|^p)^{\frac{1}{p}} = \lim_{p \rightarrow \infty} x_j = x_j = \max_{1 \leq i \leq n} |x_i|.$$

Hence proved that, $\|\mathbf{x}\|_\infty = \max_{1 \leq i \leq n} |x_i|$.

(b) True or False (explain with reason): $\|\mathbf{x}\|_0$ is a norm.

Solution:

FALSE

Since the $\|\mathbf{x}\|_0$ norm simply counts the number of non-zero entries in \mathbf{x} , suppose we scale \mathbf{x} by a factor $k \in \mathbb{R} - \{0\}$.

Since scaling by a non-zero value cannot make a non-zero value zero or vice versa, $\|k\mathbf{x}\|_0 = \|\mathbf{x}\|_0$ as both will have same number of non-zero entries.

This violates the property of norms which says that $\|k\mathbf{x}\| = k\|\mathbf{x}\|$.

Hence $\|\mathbf{x}\|_0$ is not a norm.

Concept: Orthogonal/Orthonormal vectors and matrices

7. (1 point) Consider the following questions:

(a) Construct a 2×2 Orthonormal matrix, such that none of its entries are real.

Solution:

If we consider the matrix, $A = \begin{bmatrix} \sqrt{\frac{3}{2}} + i & \sqrt{\frac{3}{2}} - i \\ -\sqrt{\frac{3}{2}} + i & \sqrt{\frac{3}{2}} + i \end{bmatrix}$

Dot product of columns is, $\begin{bmatrix} \sqrt{\frac{3}{2}} + i \\ -\sqrt{\frac{3}{2}} + i \end{bmatrix} \cdot \begin{bmatrix} \sqrt{\frac{3}{2}} - i \\ \sqrt{\frac{3}{2}} + i \end{bmatrix}$

$$= (\sqrt{\frac{3}{2}} + i)(\sqrt{\frac{3}{2}} - i) + (-\sqrt{\frac{3}{2}} + i)(\sqrt{\frac{3}{2}} + i) = (\sqrt{\frac{3}{2}} + i)(\sqrt{\frac{3}{2}} - i) - (\sqrt{\frac{3}{2}} + i)(\sqrt{\frac{3}{2}} - i) = 0$$

Hence the columns are orthogonal.

The norms of the columns are,

$$\begin{aligned} \|A_1\|_2 &= \sqrt{(\sqrt{\frac{3}{2}} + i)^2 + (-\sqrt{\frac{3}{2}} + i)^2} = \sqrt{\frac{3}{2} - 1 + 2\frac{3}{2}i + \frac{3}{2} - 1 - 2\frac{3}{2}i} \\ &= \sqrt{3 - 1 - 1} = 1 \end{aligned}$$

$$\begin{aligned} \|A_2\|_2 &= \sqrt{(\sqrt{\frac{3}{2}} - i)^2 + (\sqrt{\frac{3}{2}} + i)^2} = \sqrt{\frac{3}{2} - 1 - 2\frac{3}{2}i + \frac{3}{2} - 1 + 2\frac{3}{2}i} \\ &= \sqrt{3 - 1 - 1} = 1 \end{aligned}$$

Since the norms of the columns are 1 and their dot product is 0, the columns of A are orthonormal.

Hence A is a orthogonal matrix.

- (b) Now, construct a 4×4 Orthogonal matrix, such that all its entries are +1, -1, +2 or -2.

Solution:

Since allowed values are 1, -1, 2, -2, the minimum possible $\|A_i\|_2$ (norm of i th column) is when the column has all elements 1 or -1. Minimum norm is $\sqrt{(\pm 1)^2 + (\pm 1)^2 + (\pm 1)^2 + (\pm 1)^2} = 2$.

Hence norm of a column with elements from 1, -1, 2, -2 cannot be 1.

Hence a 4×4 matrix orthogonal matrix cannot be constructed using these values since it is not possible to have a column with norm of 1 which is necessary condition for all columns in a orthogonal matrix.

8. (1 point) Consider the vectors $\mathbf{a} = \begin{bmatrix} 4 \\ 6 \\ 2 \\ 5 \end{bmatrix}$ and $\mathbf{b} = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 1 \end{bmatrix}$

- (a) What multiple of \mathbf{a} is closest to \mathbf{b} ?

Solution:

The closest multiple of \mathbf{a} to \mathbf{b} is the projection of \mathbf{b} on \mathbf{a} .

$$\text{Projection of } \mathbf{b} \text{ on } \mathbf{a} \text{ is, } p = \frac{\mathbf{a}\mathbf{a}^\top}{\mathbf{a}^\top\mathbf{a}}\mathbf{b} = \frac{1}{\begin{bmatrix} 4 & 6 & 2 & 5 \end{bmatrix} \begin{bmatrix} 4 \\ 6 \\ 2 \\ 5 \end{bmatrix}} \begin{bmatrix} 4 \\ 6 \\ 2 \\ 5 \end{bmatrix} \begin{bmatrix} 4 & 6 & 2 & 5 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ 0 \\ 1 \end{bmatrix}$$

$$= \frac{1}{81} \begin{bmatrix} 16 & 24 & 8 & 20 \\ 24 & 36 & 12 & 30 \\ 8 & 12 & 4 & 10 \\ 20 & 30 & 10 & 25 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ 0 \\ 1 \end{bmatrix} = \frac{1}{81} \begin{bmatrix} 16 + 20 \\ 24 + 30 \\ 8 + 10 \\ 20 + 25 \end{bmatrix} = \frac{1}{81} \begin{bmatrix} 36 \\ 54 \\ 18 \\ 45 \end{bmatrix} = \frac{9}{81} \begin{bmatrix} 4 \\ 6 \\ 2 \\ 5 \end{bmatrix} = \frac{1}{9}\mathbf{a}$$

Therefore the multiple of \mathbf{a} closest to \mathbf{b} is $\frac{1}{9}\mathbf{a}$.

- (b) Find orthonormal vectors \mathbf{q}_1 and \mathbf{q}_2 that lie in the plane formed by \mathbf{a} and \mathbf{b} ?

Solution:

$$q_1 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 0 \\ 0 \\ 1 \end{bmatrix}. \quad \|q_1\|_2 = \frac{1}{2}(1+1) = 1 \text{ and since } q_1 \text{ is a multiple of } b, \text{ it lies on the}$$

plane formed by a and b. Since any vector on the plane formed by a and b is a

$$\text{linear combination of a and b, } q_2 = \begin{bmatrix} 4x+y \\ 6x \\ 2x \\ 5x+y \end{bmatrix}. \text{ Since } q_1 \text{ and } q_2 \text{ are perpendicular}$$

(orthogonal), $q_1 \cdot q_2 = 0$.

$$\text{Therefore, } \frac{1}{\sqrt{2}}(4x+y+5x+y) = 0 \Rightarrow 9x+2y = 0 \Rightarrow y = \frac{-9x}{2}$$

If we take x as 2, y is -9. Therefore

$$\begin{bmatrix} 8-9 \\ 12 \\ 4 \\ 10-9 \end{bmatrix} = \begin{bmatrix} -1 \\ 12 \\ 4 \\ 1 \end{bmatrix}$$

Since q_2 is orthonormal, it must have $\|q_2\|_2 = 1$, $q_2 = \frac{1}{\sqrt{1+144+16+1}} \begin{bmatrix} -1 \\ 12 \\ 4 \\ 1 \end{bmatrix} =$

$$\frac{1}{\sqrt{162}} \begin{bmatrix} -1 \\ 12 \\ 4 \\ 1 \end{bmatrix} = \frac{1}{9\sqrt{2}} \begin{bmatrix} -1 \\ 12 \\ 4 \\ 1 \end{bmatrix}$$

$$q_1 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 0 \\ 0 \\ 1 \end{bmatrix} \text{ and } q_2 = \frac{1}{9\sqrt{2}} \begin{bmatrix} -1 \\ 12 \\ 4 \\ 1 \end{bmatrix}$$

9. (1 point) True or False : If A is Unitary matrix then A^2 must be an Unitary matrix. Prove the statement if True, or give counterexample if false.

Solution:

TRUE

Given that A is a unitary matrix, i.e. $A\bar{A}^\top = \bar{A}^\top A = I$.

$$A = \bar{A}^\top$$

Also we know that for any two matrices A and B, $\overline{AB}^\top = (\overline{AB})^\top$ as taking conjugate after multiplying two complex values / matrices is same as taking conjugate and then multiplying.

$$= \bar{B}^\top \bar{A}^\top.$$

Therefore, $A^2 = AA$ and hence, $(A^2)(\overline{A^2})^\top = AA\bar{A}^\top \bar{A}^\top = A(\bar{A}^\top \bar{A}^\top)A = AIA = A$.

Since A is unitary, $A\bar{A}^\top = I$, $\Rightarrow A(I)\bar{A}^\top = A\bar{A}^\top = I$.

Therefore, $(A^2)(\overline{A^2})^\top = I$ and hence A^2 is a unitary matrix.

10. (1 point) If Q is an orthogonal matrix, show that for any two vectors x and y of the proper dimension :

$$\|Qx - Qy\| = \|x - y\|$$

Solution:

Since Q is a matrix, $\|Qx - Qy\| = \|Q(x - y)\|$.

Since Q is an orthogonal matrix, $Q^{-1} = Q^\top$.

Now, if we take, $x - y = z$, $\|Qz\|^2 = (Qz)^\top Qz = z^\top Q^\top Qz = z^\top Q^{-1}Qz = z^\top (I)z = z^\top z = \|z\|^2$.

Therefore, $\|Qz\|^2 = \|z\|^2$ and since $\|Qz\|$ and $\|z\|$ are always positive, $\|Qz\| = \|z\|$.

Hence, $\|Qx - Qy\| = \|x - y\|$.

Concept: Determinants

11. (2 points) Let A be a $n \times n$ matrix such that $A[i][j] = \begin{cases} 1 & i - j = 1 \text{ OR } i = j \\ -1 & j - i = 1 \\ 0 & \text{otherwise} \end{cases}$

Prove $|A_n| = |A_{n-1}| + |A_{n-2}|$.

Solution:

Here, $i = j$ represents all the diagonal elements of A.

$i - j = 1 \Rightarrow i = j + 1$ represents elements $A[2][1], A[3][2], \dots, A[j+1][j]$. These are the elements just below the diagonal elements of A.

$j - i = 1 \Rightarrow i = j - 1$ represents elements $A[1][2], A[2][3], \dots, A[j-1][j]$. These are the elements just above the diagonal elements of A.

Therefore, A is of the form, $A =$

$$\begin{bmatrix} 1 & -1 & 0 & 0 & \dots & 0 & 0 & 0 \\ 1 & 1 & -1 & 0 & \dots & 0 & 0 & 0 \\ 0 & 1 & 1 & -1 & \dots & 0 & 0 & 0 \\ 0 & 0 & 1 & 1 & \dots & 0 & 0 & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & 0 & 0 & \dots & 1 & -1 & 0 \\ 0 & 0 & 0 & 0 & \dots & 1 & 1 & -1 \\ 0 & 0 & 0 & 0 & \dots & 0 & 1 & 1 \end{bmatrix}$$

Using $\det(A)$ formula, $\det(A_n) = (-1)^{1+1}(1)(\det(M_{11})) + (-1)^{2+1}(-1)(\det(M_{12}))$, (other terms are not taken as they are multiplied with 0) where M_{ij} is the matrix formed by removing i th row and j th column from A_n .

$$\det(A_n) = \det(M_{11}) - (-1)\det(M_{12}) = \det(M_{11}) + \det(M_{12})$$

$$M_{11} = \begin{bmatrix} 1 & -1 & 0 & \dots & 0 & 0 & 0 \\ 1 & 1 & -1 & \dots & 0 & 0 & 0 \\ 0 & 1 & 1 & \dots & 0 & 0 & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & 0 & \dots & 1 & -1 & 0 \\ 0 & 0 & 0 & \dots & 1 & 1 & -1 \\ 0 & 0 & 0 & \dots & 0 & 1 & 1 \end{bmatrix} \quad M_{12} = \begin{bmatrix} 1 & -1 & 0 & \dots & 0 & 0 & 0 \\ 0 & 1 & -1 & \dots & 0 & 0 & 0 \\ 0 & 1 & 1 & \dots & 0 & 0 & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & 0 & \dots & 1 & -1 & 0 \\ 0 & 0 & 0 & \dots & 1 & 1 & -1 \\ 0 & 0 & 0 & \dots & 0 & 1 & 1 \end{bmatrix}$$

We can see clearly that M_{11} is a $(n-1) \times (n-1)$ matrix with central diagonal elements 1, elements above diagonal -1, elements below diagonal 1 and rest of the elements 0. Hence, $M_{11} = A_{n-1}$.

Also, $\det(M_{12}) = \det(m_{11}) - (-1)\det(m_{12})$ (other terms are not taken as they are multiplied with 0), where m_{ij} is the $(n-2) \times (n-2)$ matrix we get after removing i th row and j th column from M_{12} . Here, if we remove 1st row and 2nd column from M_{12} , the resulting matrix has all zeros in its first column. Hence, $\det(m_{12}) = 0$.

Hence, $\det(M_{12}) = \det(m_{11}) = \det(A_{n-2})$ as after removing 1st row and 1st column from M_{12} , we get back the same form as A with 1s in central diagonal, -1s above and 1s below it. Hence $m_{11} = A_{n-2}$ (as m_{11} is a $(n-2) \times (n-2)$ matrix).

Therefore, $\det(A_n) = \det(A_{n-1}) + \det(A_{n-2})$. Hence Proved.

12. (1 point) What is the least number of zeros in a $n \times n$ matrix that will guarantee $\det(A) = 0$. Construct such matrix for $n = 4$.
On the other hand, what is the maximum numbers of zeros in a $n \times n$ matrix that will guarantee $\det(A) \neq 0$. Construct such matrix for $n = 4$.

Solution:

For $\det(A) = 0$, it is enough to have a dependent row in A which can be written as linear combinations of other rows in A . It is not necessary that the row should have zeroes.

Hence minimum number of zeroes in a $n \times n$ matrix for $\det(A) = 0$ is 0.

$$\text{For } n = 4, A = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \end{bmatrix}$$

Here, if we do gaussian elimination, the first 3 rows become 0. Due to zero rows, $\det(A) = 0$.

For $\det(A) \neq 0$, A should have all rows as independent rows. Hence to get maximum number of zeroes in A and for it to still have all rows independent, we can make every row have all elements 0 except for 1 element such that the position of the non-zero element in each row is unique.

If we take the value of the non-zero element as 1, we simply get the identity matrix.

Since for a $n \times n$ identity matrix has all elements 0 except along the diagonal, it has number of zeros = (total number of elements in A) - (number of non-zero elements in A). Number of zeroes = $(n \times n) - (n) = n(n-1)$.

This is the maximum zeroes possible as to get even 1 more zero, we need to make one of the non-zero elements as 0. This causes its corresponding row to become a zero row which makes $\det(A)$ to 0.

Hence maximum number of zeroes in a $n \times n$ matrix for $\det(A) \neq 0$ is $n(n-1)$.

$$\text{For } n = 4, A = I = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

Here, since it is a identity matrix, it is invertible and hence $\det(A) \neq 0$ and it has 12 zeros.

13. (1 point) This question is about properties 9 and 10 of determinants.

(a) Prove that $\det(AB) = \det(A)\det(B)$

Solution:

Since any matrix can either be invertible or be non-invertible, there are 4 possibilities for the matrices A and B, and if we prove that the equality holds for all the cases then it is proved for any A and B.

(i) B is not invertible and A can be invertible or not-invertible.

I.e. $\det(B) = 0$. Since B is not invertible, B must have a non-trivial (non-zero) vector x such that $Bx = 0$.

Hence if we compute the multiplication of the same non-trivial vector with AB, $(AB)x = A(Bx) = A0 = 0$. Hence there is a non-trivial solution to $(AB)x = 0$ and therefore AB must also be non-invertible. Hence, $\det(AB) = 0$. Hence, $LHS = \det(AB) = 0 = \det(A) * 0 = \det(A) * \det(B) = RHS$. Hence proved for this case when B is not invertible.

(ii) B is invertible and A is not invertible

I.e. $\det(B) \neq 0$ and $\det(A) = 0$.

Since A is not invertible, there exists a non-trivial vector x such that $Ax = 0$.

Also since B is invertible, B^{-1} exists. For a vector y , such that $B^{-1}y = 0$, y must be a trivial vector (zero vector) as B^{-1} is invertible. Hence, for the non-trivial (non-zero) vector x , $B^{-1}x$ is also non-trivial as for $B^{-1}x$ to be trivial/zero vector, x must also be trivial which is not the case.

If we take the non-trivial vector $B^{-1}x$, Hence in $(AB)(B^{-1}x) = A(BB^{-1})x = Ax = 0$. I.e. AB has a non-trivial solution for $(AB)y = 0$ and hence AB is not invertible and therefore, $\det(AB) = 0$. In the formula, $LHS = \det(AB) = 0$, $RHS = \det(A)\det(B) = 0 * \det(B) = 0$

$LHS = RHS$ and hence the equality holds.

(iii) B is invertible and A is invertible.

I.e. $\det(B) \neq 0$ and $\det(A) \neq 0$. Therefore, we can replace A and B by a product of elementary matrices, $A = E_{a1}...E_{am}$ and $B = E_{b1}...E_{bn}$.

From properties of determinants, we know that $\det(E_1M) = \det(E_1)\det(M)$ if E_1 is a elementary matrix. Hence, $\det(A) = \det(E_{a1}...E_{am}) = \det(E_{a1})... \det(E_{am})$ and $\det(B) = \det(E_{b1}...E_{bn}) = \det(E_{b1})... \det(E_{bn})$.

$RHS = \det(A)\det(B) = \det(E_{a1})... \det(E_{am})\det(E_{b1})... \det(E_{bn})$, using the above mentioned property, $= \det(E_{a1}...E_{am}E_{b1}...E_{bn}) = \det(AB) = LHS$. Hence proved for this case.

Since the equality holds for all the cases, it is proved.

- (b) Prove that $\det(A^\top) = \det(A)$

Solution:

We can prove using mathematical induction,

If we consider $n \times n$ matrix A ,

For the base case, $n=1$, since it is a single value in the matrix, $A = A^\top$ and hence. $\det(A_1) = \det(A_1^\top)$.

Induction Step, we assume that for $n=k-1$, $\det(A_{k-1}) = \det(A_{k-1}^\top)$.

Now, for $n=k$, Using cofactor expansion of determinant,

$\det(A_k) = a_{11}\det(M_{11}) - a_{12}\det(M_{12}) + \dots + (-1)^{k+1}a_{1k}\det(M_{1k})$, where M_{ij} is the $(k-1) \times (k-1)$ matrix we get after removing i th row and j th column from A_k .

Similarly, $\det(A_k^\top) = a_{11}\det(M_{11}) - a_{21}\det(M_{21}) + \dots + (-1)^{k+1}a_{k1}\det(M_{k1}) = a_{11}\det(M_{11}^\top) + \dots + (-1)^{k+1}a_{k1}\det(M_{1k}^\top)$

Since $M_{11} \dots M_{1k}$ are $(k-1) \times (k-1)$ matrices, $\det(M_{11}) = \det(M_{11}^\top) \dots \det(M_{1k}) = \det(M_{1k}^\top)$. Hence, $\det(A_k^\top) = a_{11}\det(M_{11}) + \dots + (-1)^{k+1}a_{k1}\det(M_{1k}) = \det(A_k)$.

Hence it is shown that for $n=k$, $\det(A_k) = \det(A_k^\top)$.

Hence by induction, $\det(A) = \det(A^\top)$.

14. (1 point) Let $\mathbf{u} = \begin{bmatrix} 3 \\ 2 \end{bmatrix}$ and $\mathbf{v} = \begin{bmatrix} 6 \\ 4 \end{bmatrix}$.

- (a) Find the area of the triangle whose vertices are $\mathbf{u}, \mathbf{v}, \mathbf{u} + \mathbf{v}$

Solution:

Here by inspection, $v = 2u$. Hence $u + v = 3u$. Therefore, all 3 vectors lie on the same line (along u).

Hence area of triangle is 0.

- (b) Suppose you rotate these vectors along the origin such that the heads of vectors \mathbf{u} and \mathbf{v} trace two concentric circles, then find the area of figure trapped between circles

Solution:

By rotating these vectors along origin, the heads of the vectors trace a circle of radius = magnitude of the vector.

Hence r_u = radius of trace by rotating the u vector = $\|u\|_2 = \sqrt{3^2 + 2^2} = \sqrt{9 + 4} = \sqrt{13}$

r_v = radius of trace by rotating the v vector = $\|v\|_2 = \sqrt{6^2 + 4^2} = \sqrt{36 + 16} = \sqrt{52}$.

Hence area of figure trapper between the two circles = area of circle by v - area of circle by u = $\pi r_v^2 - \pi r_u^2 = \pi((\sqrt{52})^2 - (\sqrt{13})^2) = \pi(52 - 13) = 39\pi$
 $= 39\pi \cong 122.522 \text{ units}^2$.

15. (2 points) The determinant of the following matrix can be computed as a sum of 120 (5!) terms.

$$A = \begin{bmatrix} x & x & x & x & x \\ x & x & x & x & x \\ 0 & 0 & 0 & x & x \\ 0 & 0 & 0 & x & x \\ 0 & 0 & 0 & x & x \end{bmatrix}$$

State true or false with an appropriate explanation: All the 120 terms in the determinant will be 0.

Solution:

TRUE

Following the cofactor expansion formula for determinants, $\det(A)$ will be a linear combination of terms of the form, $a_{1j_1}a_{2j_2}\dots a_{5j_5}$ where j_1, \dots, j_5 is a permutation of $\{1, 2, 3, 4, 5\}$.

Therefore each term can be obtained by selecting one unique column for each row and multiplying the intersection elements of those rows and columns. To get a non-zero term, we should select the 5 elements in the term such that not even one of them is a 0.

If $x \neq 0$, in row 1 and 2 we can pick any column since all the elements are x .

But, the last 3 rows have only 2 non-zero elements each in them in the same last 2 columns. Hence we can avoid picking a 0 in row 4 and 5 by picking the 4th and 5th column elements respectively (elements = x).

But for the 3rd row, we cannot choose column 1, 2, 3 as their elements are 0. We also cannot pick column 4 or 5 as they have already been picked by rows 4 and 5.

Hence it is not possible to pick any permutation of $\{1, 2, 3, 4, 5\}$ such that every element in $\{a_{1j_1}, a_{2j_2}, \dots, a_{5j_5}\}$ is non zero. Hence their product will always be 0. Hence every term in the cofactor expansion of determinant will be 0.

Hence Proved.