

CS-460

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Minimum Background Test

Vectors and Matrices

1. $y \cdot z = (1 \times 2) + (3 \times 3) = 11$

2. $Xy = \begin{bmatrix} 2 & 4 \\ 1 & 3 \end{bmatrix} \begin{bmatrix} 1 \\ 3 \end{bmatrix}$

$$= \begin{bmatrix} (2 \times 1) + (4 \times 3) \\ (1 \times 1) + (3 \times 3) \end{bmatrix}$$

$$= \begin{bmatrix} 14 \\ 10 \end{bmatrix}$$

3. $X = \begin{bmatrix} 2 & 4 \\ 1 & 3 \end{bmatrix}$

$$\begin{aligned} |X| &= (2 \times 3) - (1 \times 4) \\ &= 6 - 4 = 2 \neq 0. \end{aligned}$$

$\therefore X$ is non-singular, X is invertible

$$\text{adj } X = \begin{bmatrix} 3 & -4 \\ -1 & 2 \end{bmatrix}.$$

$$\begin{aligned} \text{So, } X^{-1} &= \frac{\text{adj } X}{|X|} \\ &= \frac{1}{2} \begin{bmatrix} 3 & -4 \\ -1 & 2 \end{bmatrix} \\ &= \begin{bmatrix} \frac{3}{2} & -2 \\ -\frac{1}{2} & 1 \end{bmatrix} \end{aligned}$$

4. The rank of X is 2.

Because: $[2 \ 4] \neq c[1 \ 3] \quad \forall c \in \mathbb{R}.$

i.e., both rows are linearly independent of each other

So. The rank is 2

Calculus

1. $y = x^3 + x - 5$

$$\text{So, } \frac{dy}{dx} = \frac{d}{dx} (x^3 + x - 5)$$

$$= \underline{3x^2 + 1}$$

2. $f(x_1, x_2) = x_1 \sin(x_2) e^{-x_1}$

$$\frac{\partial f}{\partial x_1} = \frac{\partial}{\partial x_1} (x_1 \sin x_2 e^{-x_1})$$

$$= \sin x_2 \left[e^{-x_1} \frac{\partial x_1}{\partial x_1} + x_1 \frac{\partial e^{-x_1}}{\partial x_1} \right]$$

$$= \sin x_2 \left[e^{-x_1} + x_1 \times -1 \times e^{-x_1} \right]$$

$$= \underline{e^{-x_1} (1 - x_1) \sin x_2}$$

$$\partial_{x_2} f = \frac{\partial}{\partial x_2} (x_1 \sin x_2 e^{-x_1})$$

$$= x_1 e^{-x_1} \frac{\partial \sin x_2}{\partial x_2}$$

$$= x_1 e^{-x_1} \cos x_2$$

$$\nabla f(x) = \begin{pmatrix} \partial_{x_1} f \\ \partial_{x_2} f \end{pmatrix}$$

$$= \begin{pmatrix} (1-x_1) e^{-x_1} \sin x_2 \\ x_1 e^{-x_1} \cos x_2 \end{pmatrix}$$

Probability and Statistics

1. The sample mean = $\frac{1+1+0+1+0}{5}$

$$\text{Sample mean } (\mu) = \frac{3}{5}$$

2. Sample variance = $\frac{\sum_{i=1}^5 (\mu - x_i)^2}{n-1}$

$$\text{Sample variance } (s^2) = \left[\left(1 - \frac{3}{5}\right)^2 + \left(1 - \frac{3}{5}\right)^2 + 0\left(-\frac{3}{5}\right)^2 + \left(1 - \frac{3}{5}\right)^2 + \left(-\frac{3}{5}\right)^2 \right] \frac{1}{4}$$

$$= \left[\left(\frac{2}{5}\right)^2 + \left(\frac{2}{5}\right)^2 + \left(\frac{3}{5}\right)^2 + \left(\frac{2}{5}\right)^2 + \left(\frac{3}{5}\right)^2 \right] \frac{1}{4}$$

$$= \frac{4+4+9+4+9}{5^2} \times \frac{1}{4}$$

$$= \frac{30}{5 \times 4 \times 5}$$

$$= \frac{3}{10}$$

3. Probability of head = $P_H = \frac{1}{2}$

Probability of tail = $P_T = \frac{1}{2}$

Probability of S = $\frac{1}{2} \times \frac{1}{2} \times \frac{1}{2} \times \frac{1}{2} \times \frac{1}{2}$

$$= \frac{1}{32}$$

4. Let the $P_t = p$ be the probability of getting tail ($P(x=1)$)
 $\therefore P_h = (1-p)$ would be the probability of getting head ($P(x=0)$)
 as $P_h + P_t = 1$

So, the probability of the Sample S be P_s .

$$P_s = \prod_{i=1}^5 p^{x_i} (1-p)^{(1-x_i)}$$

$$= p^{\sum x_i} (1-p)^{5 - \sum x_i}$$

Taking the log on both side;

$$\log(P_s) = \left(\sum_{i=1}^5 x_i \right) \log p + \left(5 - \sum_{i=1}^5 x_i \right) \log(1-p)$$

$$= 3 \log p + 2 \log(1-p)$$

$$\frac{d(\log P_s)}{dp} = \frac{3}{p} + \frac{2}{1-p} \cdot (-1)$$

$$= \frac{3}{p} - \frac{2}{1-p}$$

As we are taking the maximum probability of S , maximize the probability of $\log P_s$. So take LHS = 0

$$0 = \frac{3}{p} - \frac{2}{1-p}$$

$$\frac{3}{p} = \frac{2}{1-p}$$

$$3(1-p) = 2p$$

$$3 - 3p = 2p$$

$$8p = 3$$

$$p = \frac{3}{8}$$

So the value that maximizes the probability of sample S should have $p(x=1) = \frac{3}{5}$ & $p(x=0) = \frac{2}{5}$

5.

• What is $p(Z=T \text{ AND } y=b)$?

sol:- 0.1

• What is $p(Z=T | y=b)$

$$\text{sol; } p(Z=T | y=b) = \frac{p(Z=T \text{ AND } y=b)}{p(y=b)}$$

$$= \frac{0.1}{0.1+0.5} = \frac{10}{25}$$

$$= 0.4$$

Big - O Notation

1. $f(n) = \ln(n)$ & $g(n) = \lg(n)$.

Here \ln & \lg are base with e & 2 respectively

i.e., $f(n) = \frac{\log n}{\log e}$ & $g(n) = \frac{\log n}{\log 2}$

i.e., both the functions are equivalent. So;

$$f(n) = O(g(n)) \text{ \& } g(n) = O(f(n))$$

2. $f(n) = 3^n$ & $g(n) = n^{100}$.

$f(n)$ grows more rapidly as n becomes large

i.e., $g(n) = O(f(n))$.

3. $f(n) = 3^n$, $g(n) = 2^n$

~~g(n)~~ $f(n)$ grows rapidly as n becomes large

So, $g(n) = O(f(n))$

4. $f(n) = 1000n^2 + 2000n + 4000$, $g(n) = 3n^2$

$f(n) = O(g(n))$, since $g(n)$ grows rapidly as n becomes large

Medium Background Test

Divide and Conquer

Algorithm;

Find - transitions (i, j)

let $mid = i + \text{floor}((j-i)/2)$

let $a = \text{array}[mid]$ and $b = \text{array}[mid+1]$

if $(a == b == 1)$ return find-transitions (i, mid)

else if $(a == b == 0)$ return find transitions $(mid+1, j)$

else return (last occurrence of 0)

=

For each recursive call we know that the array $[i] == 0$

& array $[j] == 1$. When we stop, we know that

$a == 0$ & $b == 1$, so ~~we~~ array $[mid]$ is the last

only with 0

The running time can be analyzed by the recurrence

$T(n) = T(n/2) + O(1)$, $T(n) = C$ for $n \leq 4$, which solves

to $O(\log n)$. The algorithm is based on the idea of

binary search tree

Probability and Random Variables

Probability

(a) $P(A \cup B) = P(A \cap (B \cap A^c))$

Sol - False

(b) $P(A \cup B) = P(A) + P(B) - P(A \cap B)$

Sol - True

(c) $P(A) = P(A \cap B) + P(A^c \cap B)$

Sol - False

(d) $P(A|B) = P(B|A)$

Sol - False

(e) $P(A_1 \cap A_2 \cap A_3) = P(A_3 | A_2 \cap A_1) P(A_2 | A_1) P(A_1)$

Sol - True

Discrete and Continuous Distribution

Multivariate Gaussian - $\frac{1}{\sqrt{(2\pi)^d |\Sigma|}} \exp\left(-\frac{1}{2} (x - \mu)^T \Sigma^{-1} (x - \mu)\right)$

Bernoulli - $p^x (1-p)^{1-x}$

Uniform - $\frac{1}{b-a}$ when $a \leq x \leq b$; 0 otherwise

Binomial - $\binom{n}{x} p^x (1-p)^{n-x}$

Mean, Variance and Entropy

(a) It's given that $\text{Var}(X) = E[(X - E(X))^2]$

$$(X - E(X))^2 = X^2 - 2X E(X) + (E(X))^2$$

$$E[(X - E(X))^2] = E[X^2] - 2E[X E(X)] + (E[X])^2$$

Given $\Rightarrow \text{Var } X = E[X^2] - 2(E[X])^2 + E[X]^2$

$$\text{Var } X = (E[X])^2 - E[X^2]$$

=

(b) mean = p , variance = $p(1-p)$

$$\text{entropy} = -(1-p) \log(1-p) - p \log p$$

Law of Large Numbers and Central Limit Theorem

(a) Due to the law of large numbers, if a fair die is rolled the number of times 3 shows up should be close to 1000.

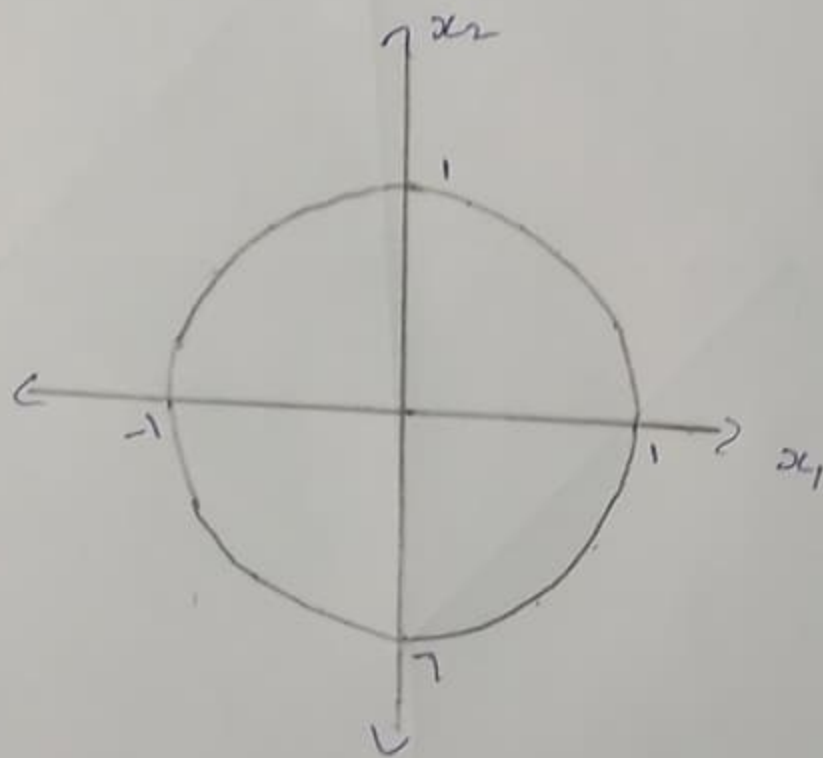
(b) The LHS should tend to the RHS as $n \rightarrow \infty$ due to Central Limit Theorem.

Linear Algebra

Vector norms

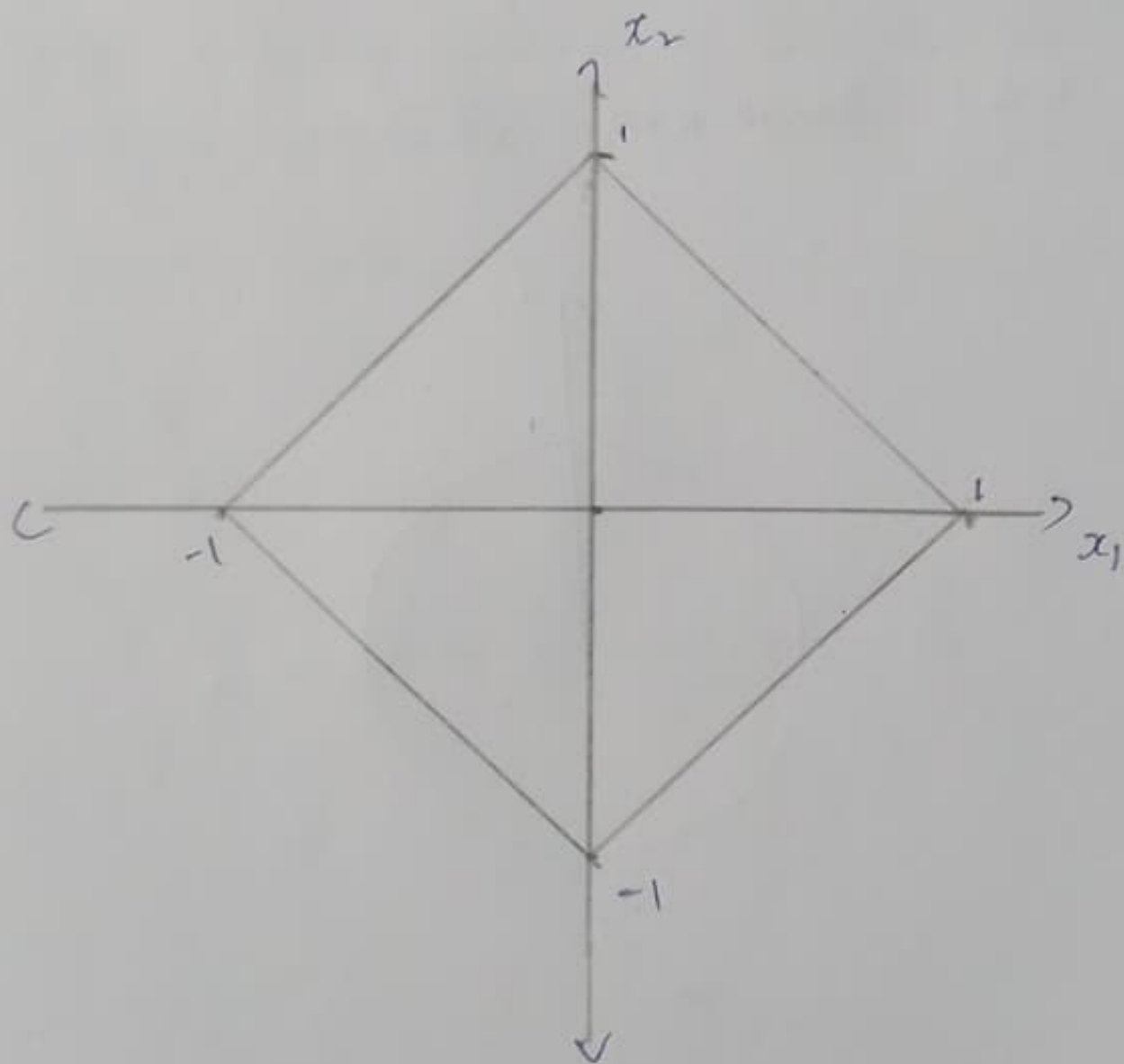
(a) $\|x\|_2 \leq 1$ (Recall $\|x\|_2 = \sqrt{x_1^2 + x_2^2}$)

Sol



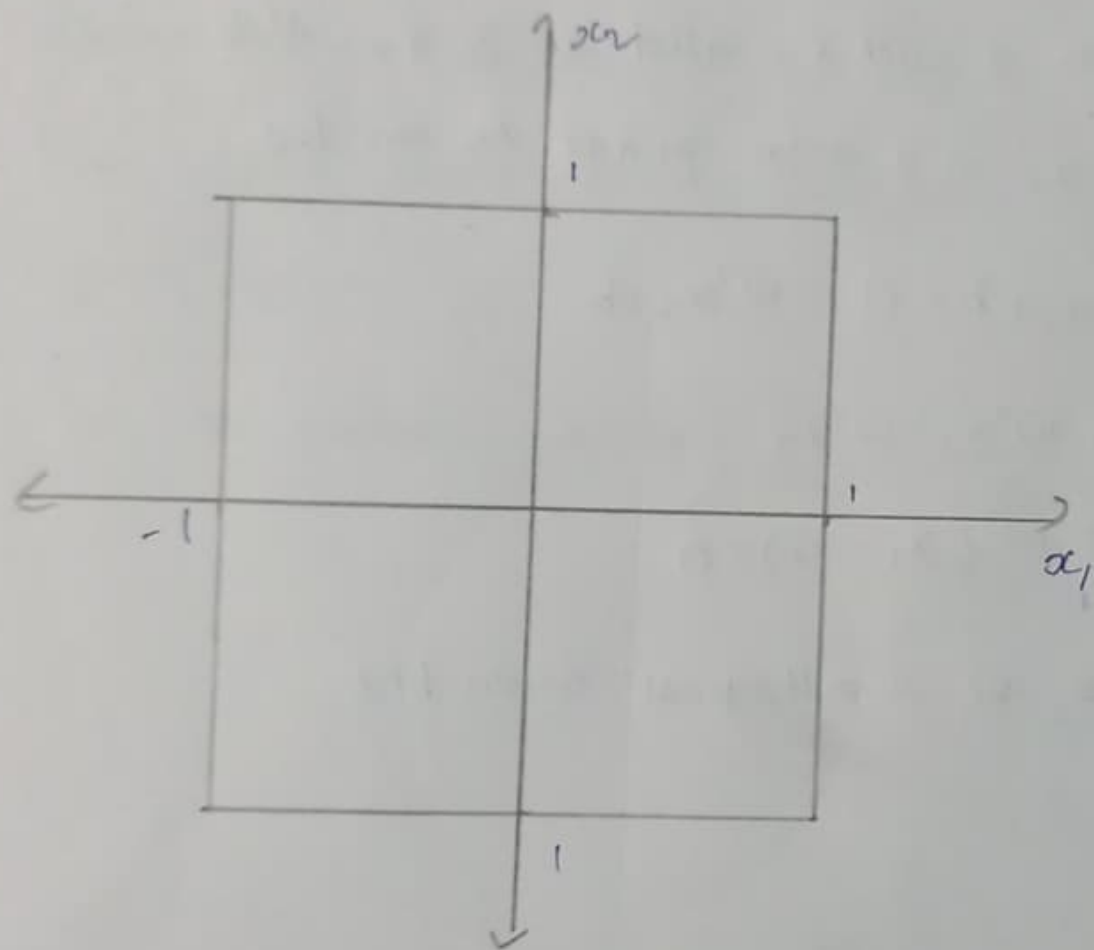
(b) $\|x\|_1 \leq 1$ (Recall $\|x\|_1 = |x_1| + |x_2|$)

Sol



(c) $\|x\|_\infty \leq 1$ (Recall $\|x\|_\infty = \max_i |x_i|$)

Sol



Geometry

(a) This line is all x such that $w^T x + b = 0$.

Consider 2 such x , called x_1 & x_2 . Note

$x_1 - x_2$ is a vector parallel to our line.

$$w^T x_1 + b = 0 = w^T x_2 + b$$

$$\Rightarrow w^T x_1 = w^T x_2$$

$$\Rightarrow w^T (x_1 - x_2) = 0$$

i.e., w is orthogonal to our line.

(b) Let x be any point on the hyperplane $w^T x + b = 0$.

The Euclidean distance from the origin to the hyperplane can be computed by projecting x onto the normal vector of the hyperplane, which is given by w .

$$\begin{aligned} \text{i.e., the distance given by } \frac{|w^T x|}{\|w\|_2} &= \frac{|b|}{\|w\|_2} \\ &= \frac{|b|}{\|w\|_2} \end{aligned}$$