

Q1 kurtosis of a sum

Proof: Since X and Y are two independent random variables with zero mean and unit variance

$$E[X] = E[Y] = 0, \quad E[X^2] = E[Y^2] = 1$$

therefore $X+Y$ is also a random variable with zero mean

$$\begin{aligned} \text{Kurt}(X+Y) &= E[(X+Y)^4] - 3(E[(X+Y)^2])^2 \\ &= E[X^4 + 4X^3Y + 6X^2Y^2 + 4XY^3 + Y^4] - 3(E[X^2 + 2XY + Y^2])^2 \end{aligned}$$

using linearity of expectation, we obtain the following

$$\begin{aligned} \text{Kurt}(X+Y) &= E[X^4] + 4E[X^3Y] + 6E[X^2Y^2] + 4E[XY^3] + E[Y^4] \\ &\quad - 3(E[X^2] + 2E[XY] + E[Y^2])^2 \end{aligned}$$

Since X and Y are independent

$$\begin{aligned} \text{Kurt}(X+Y) &= E[X^4] + 4E[X^3]E[Y] + 6E[X^2]E[Y^2] + 4E[X]E[Y^3] + E[Y^4] \\ &\quad - 3(E[X^2] + 2E[X]E[Y] + E[Y^2])^2 \end{aligned}$$

Given $E[X] = E[Y] = 0$ and $E[X^2] = E[Y^2] = 1$

$$\begin{aligned} \text{Kurt}(X+Y) &= E[X^4] + 6E[X^2]E[Y^2] + E[Y^4] - 3(E[X^2] + E[Y^2])^2 \\ &= E[X^4] + 6 + E[Y^4] - 3(1+1)^2 \\ &= E[X^4] + E[Y^4] - 6 \end{aligned}$$

$$\text{Now consider } \text{Kurt}(X) = E[X^4] - 3(E[X^2])^2 = E[X^4] - 3$$

$$\text{Kurt}(Y) = E[Y^4] - 3(E[Y^2])^2 = E[Y^4] - 3$$

$$\text{Kurt}(X) + \text{Kurt}(Y) = E[X^4] + E[Y^4] - 6 = \text{Kurt}(X+Y)$$

therefore we have shown that $\text{Kurt}(X+Y) = \text{Kurt}(X) + \text{Kurt}(Y)$



Q2 Negentropy and Mutual Information

Y_1, Y_2 potentially dependent random continuous variables.

Both with same covariance matrix Σ

$$I(Y_1; Y_2) = \iint p(y_1, y_2) \log \frac{p(y_1, y_2)}{p(y_1)p(y_2)} dy_1 dy_2^{(1)}$$

$$J(Y_1) = H(Y_g) - H(Y_1)^{(2)} \quad J(Y_2) = H(Y_g) - H(Y_2)^{(3)} \text{ where}$$

$$\text{entropy of } Y_i \text{ is } H(Y_i) = - \int p(y_i) \log p(y_i) dy_i^{(4)} \text{ for } i=1, 2$$

$$\text{entropy of a Gaussian random variable } H(Y_g) = \frac{1}{2} \log(\det(2\pi e \Sigma))^{(5)}$$

$$(a) \quad I(Y_1; Y_2) = H(Y_1) + H(Y_2) - H(Y_1, Y_2)$$

proof: The joint entropy $H(Y_1, Y_2)$ is given

$$H(Y_1, Y_2) = - \iint p(y_1, y_2) \log p(y_1, y_2) dy_1 dy_2$$

According to (4) definition of entropy

$$H(Y_1) = - \int p(y_1) \log p(y_1) dy_1$$

$$H(Y_2) = - \int p(y_2) \log p(y_2) dy_2$$

According to sum rule $p(x) = \int p(x, y) dy$

$$H(Y_1) = - \iint p(y_1, y_2) \log p(y_1) dy_1 dy_2 \quad \text{Similarly}$$

$$H(Y_2) = - \iint p(y_2, y_1) \log p(y_2) dy_2 dy_1$$

we can then show that

$$H(Y_1) + H(Y_2) - H(Y_1, Y_2)$$

$$= - \iint p(y_1, y_2) \log p(y_1) dy_1 dy_2 - \iint p(y_1, y_2) \log p(y_2) dy_1 dy_2 + \iint p(y_1, y_2) \log p(y_1, y_2) dy_1 dy_2$$

$$= \iint p(y_1, y_2) \log \frac{p(y_1, y_2)}{p(y_1)p(y_2)} dy_1 dy_2 = I(Y_1; Y_2)$$

Therefore we have shown that $I(Y_1; Y_2) = H(Y_1) + H(Y_2) - H(Y_1, Y_2)$



(b) Show $I(Y_1; Y_2) = C - \sum_{i=1}^2 J(Y_i)$. for C only depends on $H(Y_g)$ and $H(Y_1, Y_2)$

Assume $I(Y_1; Y_2) = C - \sum_{i=1}^2 J(Y_i)$, we want to find C

from (a) $I(Y_1; Y_2) = H(Y_1) + H(Y_2) - H(Y_1, Y_2)$

$$\begin{aligned}\sum_{i=1}^2 J(Y_i) &= J(Y_1) + J(Y_2) \\ &= 2H(Y_g) - H(Y_1) - H(Y_2)\end{aligned}$$

$$I(Y_1; Y_2) = C - \sum_{i=1}^2 J(Y_i) = C - 2H(Y_g) + H(Y_1) + H(Y_2)$$

$$\text{Hence } C - 2H(Y_g) = H(Y_1, Y_2)$$

therefore we have shown that there exist a $C = 2H(Y_g) + H(Y_1, Y_2)$

$$\text{that make } I(Y_1; Y_2) = C - \sum_{i=1}^2 J(Y_i)$$



Q3 Under-constrained equations

$$\Phi = [4 \ 1] \quad y = 1$$

(a) Solve $\min_{x \in \mathbb{R}^2} \|x\|_2$ subject to $\Phi x = y$

① the objective function $\|x\|_2 = \sqrt{x_1^2 + x_2^2}$

the constraint equation $\Phi x = y$, i.e. $4x_1 + x_2 = 1$

Q4 Sparsity

$$(a) \quad \Phi = \begin{bmatrix} 1 & 1/\sqrt{2} & 0 & -1/\sqrt{2} \\ 0 & 1/\sqrt{2} & 1 & 1/\sqrt{2} \end{bmatrix}$$

Assume that there exist 2 distinct 1-sparse solutions x, y

denote non-zero entry at j and k

$$x = [x_1, x_2, x_3, x_4] \text{ where } x_j \neq 0$$

$$y = [y_1, y_2, y_3, y_4] \text{ where } y_k \neq 0$$

Consider j -th entry and k -th entry of Φx and Φy .

$$(\Phi x)_j = \Phi_j x_j \neq 0$$

$$(\Phi y)_k = \Phi_k y_k \neq 0$$

first and second row of Φ are orthogonal.

$$\text{Assume } x \neq y, \quad z = x + y \text{ results in } \Phi z = 2b$$

z is not a 1-sparse solution since it has x_i and y_j

there 1-sparse solution to $\Phi x = b$ is unique

and any 1-sparse solution $x \in \mathbb{R}^4$ can be recovered.

(b) Consider all possible combination of choosing 2 indices out of 4 to form distinct 2-sparse solutions.

$$\text{The combinations are } x = [x_1, x_2, 0, 0] \quad x = [x_1, 0, x_3, 0]$$

$$x = [x_1, 0, 0, x_4] \quad x = [0, x_2, x_3, 0] \quad x = [0, x_2, 0, x_4]$$

$$x = [0, 0, x_3, x_4]$$

Considering all possible choices, we can find up to 6 distinct 2-sparse solutions that satisfies $\Phi x = b$


Q5 Sparsity $\Phi = \begin{bmatrix} 1 & 1/\sqrt{2} & 0 & -1/\sqrt{2} \\ 0 & 1/\sqrt{2} & 1 & 1/\sqrt{2} \end{bmatrix}$ solution $b = [0, 3]^T$

(a) $x = [0, 0, 3, 0]^T$

x satisfies the given equation $\Phi x = b$

i.e. $\begin{bmatrix} 1 & 1/\sqrt{2} & 0 & -1/\sqrt{2} \\ 0 & 1/\sqrt{2} & 1 & 1/\sqrt{2} \end{bmatrix} \begin{bmatrix} 0 \\ 0 \\ 3 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 3 \end{bmatrix}$

x has only one non-zero element

Therefore x is a 1-sparse solution. 

(b) $x = [x_1, x_2, x_3, x_4]^T$

$$\Phi x = \begin{bmatrix} x_1 + (1/\sqrt{2})x_2 - (1/\sqrt{2})x_4 \\ (1/\sqrt{2})x_2 + x_3 + (1/\sqrt{2})x_4 \end{bmatrix} = \begin{bmatrix} 0 \\ 3 \end{bmatrix}$$

express x_1 and x_3 in term of x_2 and x_4 gives.

$$x_1 = \left(\frac{1}{\sqrt{2}}\right)x_4 - \left(\frac{1}{\sqrt{2}}\right)x_2$$

$$x_3 = 3 - \left(\frac{1}{\sqrt{2}}\right)x_2 - \left(\frac{1}{\sqrt{2}}\right)x_4$$

$$\left(3 - \frac{1}{\sqrt{2}}x_2 - \frac{1}{\sqrt{2}}x_4\right) \left(3 - \frac{1}{\sqrt{2}}x_2 - \frac{1}{\sqrt{2}}x_4\right)$$

$$9 - \frac{3}{\sqrt{2}}x_2 - \frac{3}{\sqrt{2}}x_4 - \frac{3}{\sqrt{2}}x_2 + \frac{1}{2}x_2^2 + \frac{1}{2}x_2x_4$$

$$- \frac{3}{\sqrt{2}}x_4 + \frac{1}{2}x_2x_4 + \frac{1}{2}x_4^2$$

$$\|x\|_2^2 = x_1^2 + x_2^2 + x_3^2 + x_4^2$$

$$= \left(\frac{1}{\sqrt{2}}x_4 - \frac{1}{\sqrt{2}}x_2\right)^2 + x_2^2 + \left(3 - \frac{1}{\sqrt{2}}x_2 - \frac{1}{\sqrt{2}}x_4\right)^2 + x_4^2$$

$$= \frac{1}{2}x_4^2 - \cancel{x_2x_4} + \frac{1}{2}x_2^2 + \underline{x_2^2} + 9 - \frac{6}{\sqrt{2}}x_2 - \frac{6}{\sqrt{2}}x_4 + \cancel{x_2x_4} + \underline{\frac{1}{2}x_2^2} + \underline{\frac{1}{2}x_4^2} + x_4^2$$

$$= 2x_2^2 + 2x_4^2 + 9 - \frac{6}{\sqrt{2}}x_2 - \frac{6}{\sqrt{2}}x_4$$

$$\frac{\partial \|x\|_2^2}{\partial x_2} = 4x_2 - \frac{6}{\sqrt{2}} = 0$$

$$x_2 = \frac{3}{2\sqrt{2}}$$

$$\frac{\partial \|x\|_2^2}{\partial x_4} = 4x_4 - \frac{6}{\sqrt{2}} = 0$$

$$x_4 = \frac{3}{2\sqrt{2}}$$

$$\text{so } x_1 = \frac{1}{\sqrt{2}} \cdot \frac{3}{2\sqrt{2}} - \frac{1}{\sqrt{2}} \cdot \frac{3}{2\sqrt{2}} = 0$$

$$x_3 = 3 - \frac{1}{\sqrt{2}} \cdot \frac{3}{2\sqrt{2}} - \frac{1}{\sqrt{2}} \cdot \frac{3}{2\sqrt{2}} = 3 - \frac{3}{2} = \frac{3}{2}$$

$x = [0, \frac{3}{2\sqrt{2}}, \frac{3}{2}, \frac{3}{2\sqrt{2}}]$, the sparsity of x is 3.



(c)