

CS5691: Pattern recognition and machine learning

Quiz - 1

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I. Multiple Choice Questions

1. Suppose X is uniformly distributed over $[0, 5]$ and Y is uniformly distributed over $[0, 4]$. If X and Y are independent, then $\mathbb{P}(\max(X, Y) > 3)$ is

- (a) $\frac{9}{20}$
- (b) $\frac{1}{20}$
- (c) $\frac{11}{20}$
- (d) 1

Solution: (c). Use $\mathbb{P}(\max(X, Y) > 3) = 1 - \mathbb{P}(\max(X, Y) \leq 3) = 1 - \mathbb{P}(X \leq 3)\mathbb{P}(Y \leq 3)$, and plug in the values for $\mathbb{P}(X \leq 3)$, $\mathbb{P}(Y \leq 3)$ use their distributional information to arrive at the answer.

2. Let $X_i, i = 1, \dots, 4$ be independent Bernoulli r.v.s each with mean $p = 0.1$ and let

$$S = \sum_{i=1}^4 X_i. \text{ Then,}$$

- (a) $\mathbb{E}(X_1 | S = 2) = 0.1.$
- (b) $\mathbb{E}(X_1 | S = 2) = 0.5.$
- (c) $\mathbb{E}(X_1 | S = 2) = 0.25.$
- (d) $\mathbb{E}(X_1 | S = 2) = 0.75.$

Solution: (b) since

$$\mathbb{E}(X_1 | S = 2) = \mathbb{P}(X_1 = 1 | S = 2) = \frac{\mathbb{P}(X_1 = 1, S = 2)}{\mathbb{P}(S = 2)} = \frac{p \times 3p(1-p)^2}{\binom{4}{2}p^2(1-p)^2} = 0.5.$$

3. Let $v_1 = \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}$, $v_2 = \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix}$, $v_3 = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$, and $v_4 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$. Let $C(A)$ and $N(A)$ denote the column and null space, respectively of any matrix A . Then, which of the following statements is **false**?

- (a) $v_1, v_2 \in C(A)$, and $v_3 \in N(A)$ for some matrix A .
- (b) $v_1, v_2 \in C(A)$, and $v_3, v_4 \in N(A)$ for some matrix A .
- (c) $v_1 \in C(A)$, and $v_3, v_4 \in N(A)$ for some matrix A .
- (d) $v_1 \in C(A)$, and $v_3 \in N(A)$ for some matrix A .

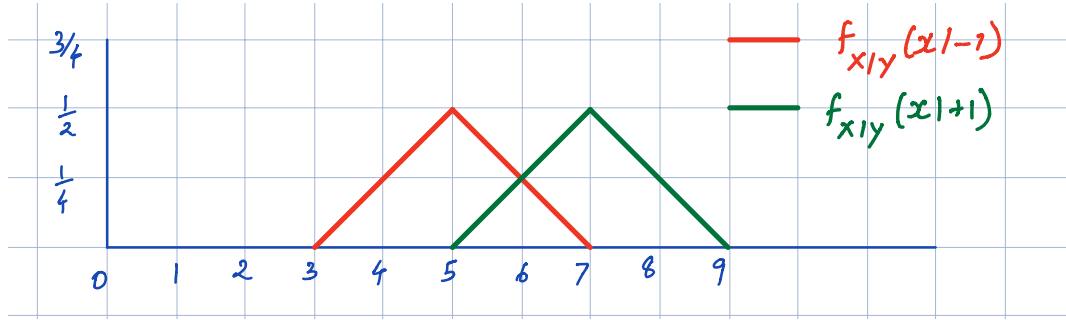
Solution: (b) since the dimensions of the column space and null space add up to 4, while using the rank-nullity theorem, these dimensions can only sum up to 3.

4. Let $Z = (X, Y)$ be a bivariate normal random variable. Then, which of the following statements is **false**?

- (a) X and Y are independent if and only if they are uncorrelated.
- (b) $X + Y$ is univariate normal.
- (c) $Y | X = x$ is distributed as a normal random variable.
- (d) $X + Y$ and $X - Y$ are independent.

Solution: (d)

5. Let $P(Y = -1) = \frac{1}{3}$, and $P(Y = +1) = \frac{2}{3}$. The class-conditionals $P(X|Y)$ are given by the graph below. The Bayes classifier is then given by which option below? (Triangle on left is the class conditional for $Y = -1$).



- (a) $h^*(x) = \begin{cases} -1 & \text{if } x \leq 6 \\ +1 & \text{if } x > 6 \end{cases}$ (b) $h^*(x) = \begin{cases} -1 & \text{if } x \leq \frac{17}{3} \\ +1 & \text{if } x > \frac{17}{3} \end{cases}$
- (c) $h^*(x) = \begin{cases} -1 & \text{if } x \leq \frac{19}{3} \\ +1 & \text{if } x > \frac{19}{3} \end{cases}$ (d) $h^*(x) = \begin{cases} -1 & \text{if } x > 6 \\ +1 & \text{if } x \leq 6 \end{cases}$

Solution: Options (d) is clearly wrong, while option (a) would be right if the prior probabilities were $\frac{1}{2}$. Now, between options (b) and (c), observe that the former has a threshold to the left of 6, and the latter to the right of 6. Given that prior probabilities make class with label +1 more likely, the threshold for Bayes classifier would be to the left of 6, making option (b) the right choice. A more rigorous proof would involve explicit calculation of the threshold, by using the density for a triangular distribution, together with the prior probabilities.

II. A problem that requires a detailed solution

1. Let X and Y be r.v.s with the joint density given by

$$f(x, y) = \frac{1}{8\sqrt{3}\pi} \exp\left(-\frac{x^2}{6} - \frac{y^2}{24} + \frac{xy}{12} + \frac{x}{12} + \frac{y}{6} - \frac{7}{24}\right). \quad (1)$$

Answer the following:

(3+2 marks)

- (a) Find the means and variances of X and Y . Also, find the covariance between X and Y .
 (b) Find the conditional density of Y given $X = x$. Also, calculate $\mathbb{E}[Y | x]$.

Solution: Recall that the bivariate normal density $f(x, y)$ is given by

$$f(x, y) = \frac{1}{2\pi\sigma_1\sigma_2\sqrt{1-\rho^2}} \exp\left(-\frac{1}{2(1-\rho^2)} \left(\left(\frac{x-\mu_1}{\sigma_1}\right)^2 + \left(\frac{y-\mu_2}{\sigma_2}\right)^2 - 2\rho \left(\frac{x-\mu_1}{\sigma_1}\right) \left(\frac{y-\mu_2}{\sigma_2}\right) \right)\right). \quad (2)$$

From the coefficients of x^2 , y^2 and xy terms in (1) and (2), we obtain

$$\sigma_1^2(1-\rho^2) = 3, \quad \sigma_2^2(1-\rho^2) = 12, \quad \sigma_1\sigma_2(1-\rho^2) = 6\rho.$$

Further, from the factor outside the exponential, we have

$$\begin{aligned} 2\pi\sigma_1\sigma_2\sqrt{1-\rho^2} &= 8\sqrt{3}\pi \\ \Rightarrow \sigma_1^2\sigma_2^2(1-\rho^2) &= 48. \end{aligned}$$

Using $\sigma_1^2(1-\rho^2) = 3$, we obtain $3\sigma_2^2 = 48$, or $\sigma_2^2 = 16$.

Similarly, using $\sigma_2^2(1-\rho^2) = 12$, we obtain $\sigma_1^2 = 4$.

Substituting σ_1, σ_2 in $\sigma_1^2\sigma_2^2(1-\rho^2) = 48$, leads to $\rho = 1/2$, (why not $-1/2$?)

To infer the values of μ_1 , and μ_2 , substitute the values of σ_1, σ_2, ρ in (2), and equate the coefficients of x and y terms to obtain the following set of equations:

$$-2\mu_1 + \mu_2 = -1, \quad \mu_1 - \mu_2 = -2.$$

Solving the above equations, we obtain $\mu_1 = 1$, and $\mu_2 = 3$.

Thus, (X, Y) is bivariate normal with means 1, 3, variances 4, 16, and correlation coefficient $\frac{1}{2}$. Further, $Cov(X, Y) = \rho\sigma_1\sigma_2 = 4$. Now, answer to part (a) is immediate.

For part (b), let $U = \frac{X-1}{2}$, $V = Y - 3$. Then (U, V) is standard bivariate normal with correlation $\rho = \frac{1}{2}$, and from the class notes, it can be seen that

$$V | U = u \text{ is normally distributed with mean } \rho u = \frac{u}{2}, \text{ and variance } (1 - \rho^2) = \frac{3}{4}.$$

Now, $V | X = x$ is the same as $V | u = (x - 1)/2$, implying $V | X = x$ is normal with mean $\frac{x-1}{4}$, and variance $\frac{3}{4}$.

Finally, $Y | X = x$ is the same as $4V + 3 | X = x$, implying $Y | X = x$ is normal with mean $x + 2$ and variance 12.