

Homework 7

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Due: 2023/12/03 10:59pm

1. (a) Show the proof of general Bayes' Rule (four cases).

	Y discrete	Y continuous
X discrete	$P(Y = y X = x) = \frac{P(X=x Y=y)P(Y=y)}{P(X=x)}$	$f_Y(y X = x) = \frac{P(X=x Y=y)f_Y(y)}{P(X=x)}$
X continuous	$P(Y = y X = x) = \frac{f_X(x Y=y)P(Y=y)}{f_X(x)}$	$f_{Y X}(y x) = \frac{f_{X Y}(x y)f_Y(y)}{f_X(x)}$

- (b) Show the proof of general LOTP (four cases).

	Y discrete	Y continuous
X discrete	$P(X = x) = \sum_y P(X = x Y = y)P(Y = y)$	$P(X = x) = \int_{-\infty}^{\infty} P(X = x Y = y)f_Y(y)dy$
X continuous	$f_X(x) = \sum_y f_X(x Y = y)P(Y = y)$	$f_X(x) = \int_{-\infty}^{\infty} f_{X Y}(x y)f_Y(y)dy$

2. A chicken lays a $\text{Pois}(\lambda)$ number N of eggs. Each egg hatches a chick with probability p , independently. Let X be the number which hatch, and Y be the number which do NOT hatch.
- Find the joint PMF of N, X, Y . Are they independent?
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 - Find the correlation between N (the number of eggs) and X (the number of eggs which hatch). Simplify; your final answer should work out to a simple function of p (the λ should cancel out).
3. Let X and Y be i.i.d. $\text{Expo}(\lambda)$, and $T = X + Y$.
- Find the conditional CDF of T given $X = x$. Be sure to specify where it is zero.
 - Find the conditional PDF $f_{T|X}(t | x)$, and verify that it is a valid PDF.
 - Find the conditional PDF $f_{X|T}(x | t)$, and verify that it is a valid PDF.

- (d) In class we have shown that the marginal PDF of T is $f_T(t) = \lambda^2 t e^{-\lambda t}$, for $t > 0$. Give a short alternative proof of this fact, based on the previous parts and Bayes' rule.
4. Let U_1, U_2, U_3 be i.i.d. $\text{Unif}(0, 1)$, and let $L = \min(U_1, U_2, U_3)$, $M = \max(U_1, U_2, U_3)$.
- (a) Find the marginal CDF and marginal PDF of M , and the joint CDF and joint PDF of L, M . Hint: For the latter, start by considering $P(L \geq l, M \leq m)$.
- (b) Find the conditional PDF of M given L .
5. Let X, Y, Z be r.v.s such that $X \sim \mathcal{N}(0, 1)$ and conditional on $X = x$, Y and Z are i.i.d. $\mathcal{N}(x, 1)$.
- (a) Find the joint PDF of X, Y, Z .
- (b) By definition, Y and Z are conditionally independent given X . Discuss intuitively whether or not Y and Z are also unconditionally independent.
- (c) Find the joint PDF of Y and Z . You can leave your answer as an integral, though the integral can be done with some algebra (such as completing the square) and facts about the Normal distribution.
6. This problem explores a visual interpretation of covariance. Data are collected for $n \geq 2$ individuals, where for each individual two variables are measured (e.g., height and weight). Assume independence across individuals (e.g., person 1's variables give no information about the other people), but not within individuals (e.g., a person's height and weight may be correlated).

Let $(x_1, y_1), \dots, (x_n, y_n)$ be the n data points. The data are considered here as fixed, known numbers – they are the observed values after performing an experiment. Imagine plotting all the points (x_i, y_i) in the plane, and drawing the rectangle determined by each pair of points. For example, the points $(1, 3)$ and $(4, 6)$ determine the rectangle with vertices $(1, 3), (1, 6), (4, 6), (4, 3)$.

The signed area contributed by (x_i, y_i) and (x_j, y_j) is the area of the rectangle they determine if the slope of the line between them is positive, and is the negative of the area of the rectangle they determine if the slope of the line between them is negative. (Define the signed area to be 0 if $x_i = x_j$ or $y_i = y_j$, since then the rectangle is degenerate.) So the signed area is positive if a higher x value goes with a higher y value for the pair of points, and negative otherwise. Assume that the x_i are all distinct and the y_i are all distinct.

- (a) The sample covariance of the data is defined to be

$$r = \frac{1}{n} \sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})$$

where

$$\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i \text{ and } \bar{y} = \frac{1}{n} \sum_{i=1}^n y_i$$

are the sample means. (There are differing conventions about whether to divide by $n - 1$ or n in the definition of sample covariance, but that need not concern us for this problem.)

Let (X, Y) be one of the (x_i, y_i) pairs, chosen uniformly at random. Determine precisely how $\text{Cov}(X, Y)$ is related to the sample covariance.

- (b) Let (X, Y) be as in (a), and (\tilde{X}, \tilde{Y}) be an independent draw from the same distribution. That is, (X, Y) and (\tilde{X}, \tilde{Y}) are randomly chosen from the n points, independently (so it is possible for the same point to be chosen twice).

Express the total signed area of the rectangles as a constant times $E((X - \tilde{X})(Y - \tilde{Y}))$. Then show that the sample covariance of the data is a constant times the total signed area of the rectangles.

Hint: Consider $E((X - \tilde{X})(Y - \tilde{Y}))$ in two ways: as the average signed area of the random rectangle formed by (X, Y) and (\tilde{X}, \tilde{Y}) , and using properties of expectation to relate it to $\text{Cov}(X, Y)$. For the former, consider the n^2 possibilities for which point (X, Y) is and which point (\tilde{X}, \tilde{Y}) ; note that n such choices result in degenerate rectangles.

- (c) Based on the interpretation from (b), give intuitive explanations of why for any r.v.s W_1, W_2, W_3 and constants a_1, a_2 , covariance has the following properties:
- (i) $\text{Cov}(W_1, W_2) = \text{Cov}(W_2, W_1)$;
 - (ii) $\text{Cov}(a_1 W_1, a_2 W_2) = a_1 a_2 \text{Cov}(W_1, W_2)$;
 - (iii) $\text{Cov}(W_1 + a_1, W_2 + a_2) = \text{Cov}(W_1, W_2)$;
 - (iv) $\text{Cov}(W_1, W_2 + W_3) = \text{Cov}(W_1, W_2) + \text{Cov}(W_1, W_3)$.
7. (Optional Challenging Problem) We use the notation $X \perp\!\!\!\perp Y | Z$ to represent the statement: random variables X and Y are conditionally independent given random variable Z . Now given any four continuous random variables X, Y, Z, W , show the following properties of conditional independence:
- (a) Symmetry: $X \perp\!\!\!\perp Y | Z \iff Y \perp\!\!\!\perp X | Z$.
 - (b) Decomposition: $X \perp\!\!\!\perp (Y, W) | Z \Rightarrow X \perp\!\!\!\perp Y | Z$
 - (c) Weak Union: $X \perp\!\!\!\perp (Y, W) | Z \Rightarrow X \perp\!\!\!\perp (Y, W) | (Z, W)$.
 - (d) Contract: $X \perp\!\!\!\perp Y | Z \text{ \& } X \perp\!\!\!\perp W | (Y, Z) \iff X \perp\!\!\!\perp (Y, W) | Z$.
 - (e) Intersection: For any positive joint PDF of X, Y, Z, W , $X \perp\!\!\!\perp Y | (Z, W) \text{ \& } X \perp\!\!\!\perp Z | (Y, W) \iff X \perp\!\!\!\perp (Y, Z) | W$.

In fact, these properties are found by Judea Pearl, who won 2011 Turing Award for fundamental contributions to artificial intelligence through the development of a calculus for probabilistic and causal reasoning. As Judea Pearl commented: “Exploiting conditional independence to generate fast probabilistic computations is one of the main contributions CS has made to probability theory”.