## STAT 512: Statistical Inference

Autumn 2022

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Quiz Session 2: Independence and Conditional Independence

Some problems in this notes are selected from the STAT 512 lecture notes [Perlman, 2020] written by *Prof. Michael Perlman* (MDP).

**Problem 1** (Simpson's Paradox; Example 3.3 in MDP). Given three events A, B, C, prove or disprove the following implication:

$$\begin{cases} P(A|B,C) > P(A|B^c,C), \\ P(A|B,C^c) > P(A|B^c,C^c) \end{cases} \xrightarrow{?} P(A|B) > P(A|B^c).$$

$$(1)$$

If the above implication does not hold, could we add some assumptions to make it hold?

## Solution.

The implication (1) does not hold in general. Here is a counterexample.

Physics	Accept	Reject	
Female	60	40	P(A F,Ph)=0.6
Male	50	50	P(A M,Ph)=0.5
English	Accept	Reject	
Female	250	750	P(A F,En)=0.25
Male	20	80	P(A M,En)=0.2
Total	Accept	Reject	
Female	310	790	P(A F) = 0.28
Male	70	130	P(A M) = 0.35

One could verify that

$$\begin{array}{lll} P(A|F,Ph) &> & P(A|M,Ph), \\ P(A|F,En) &> & P(A|M,En), \end{array}$$

but

P(A|F) < P(A|M).

A key reasoning of the Simpson's paradox is that most female students applied to English, where the acceptance rate is considerably low; see also Figure 1.



Figure 1: Explanation of the Simpson's paradox.

If B is independent with C, then the implication (1) does hold.

Proof. Note that

$$\begin{array}{lll} P(A|B) &=& P(A|B,C)P(C|B) + P(A|B,C^{c})P(C^{c}|B), \\ P(A|B^{c}) &=& P(A|B^{c},C)P(C|B^{c}) + P(A|B^{c},C^{c})P(C^{c}|B^{c}). \end{array}$$

Given the independence between B and C, it follows that

$$P(A|B) = P(A|B,C)P(C) + P(A|B,C^{c})P(C^{c}), P(A|B^{c}) = P(A|B^{c},C)P(C) + P(A|B^{c},C^{c})P(C^{c}),$$

and hence

$$P(A|B) - P(A|B^{c}) = P(C) \underbrace{\left[P(A|B,C) - P(A|B^{c},C)\right]}_{>0} + P(C^{c}) \underbrace{\left[P(A|B,C^{c}) - P(A|B^{c},C^{c})\right]}_{>0}$$
  
> 0.

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**Problem 2** (Uniform on the Unit Disk; MDP Example 1.12). Let (X, Y) follow a uniform distribution on the unit disk  $D = \{(x, y) \in \mathbb{R}^2 : x^2 + y^2 \leq 1\}$ .

- (a) Show that X and Y are not independent. (Notes: Try not to calculate the marginal distributions of X and Y at this point.)
- (b) Verify that X and Y are uncorrelated.
- (c) Calculate the marginal probability density functions (PDF) of X and Y.
- (d) Find the conditional PDF of Y|X.
- (e) Show that  $\frac{Y}{\sqrt{1-X^2}}$  is independent of X.
- (f) Consider representing the random vector (X, Y) in polar coordinates as  $(R, \Theta)$ , where  $R = \sqrt{X^2 + Y^2}$ and  $\Theta = \arctan(Y/X)$ . Verify that R and  $\Theta$  are independent.



(a) The sign information of  $\mathbb{E}(XY)$ . (b) Conditional distribution of Y|X.

Figure 2: Unit disk D and other graphical illustrations.

 $(g^*)$  Let  $S = \frac{Y}{\sqrt{1-X^2}}$  and  $T = \frac{X}{\sqrt{1-Y^2}}$ . Prove or disprove: S and T are independent.

 $(h^{**})$  Find the joint PDF f(s,t) and cumulative distribution function  $F(s,t) = \mathbb{P}(S \le s, T \le t)$  of (S,T).

## Solution.

(a) Notice that the range of (X, Y) is the unit disk D, while the marginal ranges of X and Y are both [-1, 1]. Given that  $D \neq [-1, 1] \times [-1, 1]$ , we conclude that X and Y are not independent.

(b) By the symmetry of D, we know that  $\mathbb{E}(XY) = \mathbb{E}(X) = \mathbb{E}(Y) = 0$ ; see also Figure 2a. Hence, X and Y are uncorrelated, i.e., has no linear trend.

(c) Notice that the joint PDF of (X, Y) is

$$f_{X,Y}(x,y) = \frac{1}{\pi} \cdot \mathbb{1}_D(x,y), \tag{2}$$

where

$$\mathbb{1}_D(x,y) = \begin{cases} 1 & \text{if } (x,y) \in D, \\ 0 & \text{otherwise.} \end{cases}$$

By definition, the marginal PDF of X is

$$f_X(x) = \frac{1}{\pi} \int_{-\sqrt{1-x^2}}^{\sqrt{1-x^2}} dy = \frac{2}{\pi} \sqrt{1-x^2} \cdot \mathbb{1}_{[-1,1]}(x).$$

Likewise, the marginal PDF of Y is  $f_Y(y) = \frac{2}{\pi} \sqrt{1-y^2} \cdot \mathbb{1}_{[-1,1]}(y).$ 

(d) By definition (see Section 1.5 in Lecture Note 1), the conditional PDF of Y|X is

$$f_{Y|X}(y|x) = \frac{f_{X,Y}(x,y)}{f_X(x)} = \frac{\mathbb{1}_D(x,y)/\pi}{2\sqrt{1-x^2} \cdot \mathbb{1}_{[-1,1]}(x)/\pi} = \frac{1}{2\sqrt{1-x^2}} \cdot \mathbb{1}_{\left[-\sqrt{1-x^2},\sqrt{1-x^2}\right]}(y).$$

**Remark 1.** It is worth mentioning that the domains/supports of the conditional PDF  $f_{Y|X}(y|x)$  and marginal PDF  $f_Y(y)$  are different! The former one,  $\left[-\sqrt{1-x^2}, \sqrt{1-x^2}\right]$  depends on the choice of X = x, while the latter one, [-1, 1], is independent of X. Furthermore, given that  $f_{Y|X}(y|x) \neq f_Y(y)$ , it provides another way to verify the dependency between X and Y. (e) By (d), we know that the conditional distribution Y|X = x is Uniform  $\left[-\sqrt{1-x^2}, \sqrt{1-x^2}\right]$ ; see also Figure 2b. Thus,  $\frac{Y}{\sqrt{1-X^2}}|X$  follows the distribution Uniform [-1,1], which in turn shows that  $\frac{Y}{\sqrt{1-X^2}}$  is independent of X.

(f) By the uniformity of (X, Y) on D, the joint range of  $(R, \Theta)$  is  $[0, 1] \times [0, 2\pi]$ , which is the cross product of the marginal ranges of R and  $\Theta$ . Moreover, the joint cumulative distribution function (CDF) of  $(R, \Theta)$  is

$$F_{R,\Theta}(r,\theta) \equiv \mathbb{P}\left(0 \le R \le r, 0 \le \Theta \le \theta\right)$$
$$= \frac{\theta r^2/2}{\pi}$$
$$= r^2 \cdot \frac{\theta}{2\pi}$$
$$= \mathbb{P}(0 \le R \le r) \cdot \mathbb{P}\left(0 \le \Theta \le \theta\right),$$

yielding the product of marginal CDFs of R and  $\Theta$ . Hence, R and  $\Theta$  are independent.

**Remark 2.** One can also leverage the Jacobian method (see Section 8.2 in Lecture Note 8) to directly compute the joint PDF of  $(R, \Theta)$  as:

$$\begin{split} f_{R,\Theta}(r,\theta) &= f_{X,Y}(r\cos\theta, r\sin\theta) \\ &= \frac{1}{\pi} \cdot \mathbbm{1}_D(r\cos\theta, r\sin\theta) \cdot \left| \frac{\partial(x,y)}{\partial(r,\theta)} \right| \\ &= \frac{1}{\pi} \cdot \mathbbm{1}_{\{[0,1] \times [0,2\pi]\}}(r,\theta) \cdot \left| \det \begin{pmatrix} \cos\theta & -r\sin\theta\\ \sin\theta & r\cos\theta \end{pmatrix} \right| \\ &= \frac{r}{\pi} \cdot \mathbbm{1}_{\{[0,1] \times [0,2\pi]\}}(r,\theta) \\ &= 2r \mathbbm{1}_{[0,1]}(r) \cdot \frac{1}{2\pi} \mathbbm{1}_{[0,2\pi]}(\theta), \end{split}$$

which is the product of two marginal PDFs  $f_R(r)$  and  $f_{\Theta}(\theta)$ .

(g) We claim that S and T are not independent. To prove this, it suffices to show that

$$\mathbb{P}(S \le 1/2, T \le 1/2) \neq \mathbb{P}(S \le 1/2) \cdot \mathbb{P}(T \le 1/2).$$
(3)

By (e), we know that S and T follow Uniform [-1, 1]. Hence,

$$\mathbb{P}(S \le 1/2) = \mathbb{P}(T \le 1/2) = \int_{-1}^{1/2} \frac{1}{2} ds = \frac{3}{4}.$$

On the other hand,

$$\mathbb{P}(S \le 1/2, T \le 1/2) = \mathbb{P}\left(-1 \le \frac{Y}{\sqrt{1-X^2}} \le \frac{1}{2}, -1 \le \frac{X}{\sqrt{1-Y^2}} \le \frac{1}{2}\right)$$
$$= \mathbb{P}(A_1 \cup A_2 \cup A_3 \cup A_4),$$

where, referring to Figure 3,

$$A_{1} = \{X \leq 0, Y \leq 0\},\$$

$$A_{2} = \{X < 0, Y > 0, X^{2} + 4Y^{2} \leq 1\},\$$

$$A_{3} = \{X > 0, Y < 0, 4X^{2} + Y^{2} \leq 1\},\$$

$$A_{4} = \{X \geq 0, Y \geq 0, X^{2} + 4Y^{2} \leq 1, 4X^{2} + Y^{2} \leq 1\},\$$

and they are mutually exclusive.



Figure 3: Graphical display of the probability  $\mathbb{P}(S \leq 1/2, T \leq 1/2)$ .

Given the uniformity of (X, Y) on D, we know that  $\mathbb{P}(A_1) = \frac{1}{4}$ . In addition, using the property that the area of an ellipse  $\frac{x^2}{a^2} + \frac{y^2}{b^2} = 1$  is  $\pi ab$  (which can be proved by integrating  $\int_{-a}^{a} 2b\sqrt{1 - x^2/a^2}dx$ ), we have that

$$\mathbb{P}(A_2) = \mathbb{P}(A_3) = \frac{\frac{1}{4} \cdot \pi \cdot \frac{1}{2}}{\pi} = \frac{1}{8}.$$

(Or, one can resort to Cavalieri's principle to argue that the area of  $A_2 + A_3$  is equal to 1/4 of the area of D.) Finally, we are not going to directly compute  $\mathbb{P}(A_4)$ . Instead, according to (3), we intend to show that

$$\mathbb{P}(S \le 1/2, T \le 1/2) = \sum_{i=1}^{4} \mathbb{P}(A_i) = \frac{1}{4} + 2 \times \frac{1}{8} + \mathbb{P}(A_4) = \frac{1}{2} + \mathbb{P}(A_4)$$
$$> \mathbb{P}(S \le 1/2) \cdot \mathbb{P}(T \le 1/2) = \frac{9}{16}.$$

Notice that two ellipses  $4x^2 + y^2 = 1$  and  $x^2 + 4y^2 = 1$  intersects at  $\left(\frac{1}{\sqrt{5}}, -\frac{1}{\sqrt{5}}\right)$  in the first quadrant. By the shape of the circular boundaries of  $A_4$ , it contains the square with lower left vertex (0,0) and upper right vertex  $\left(\frac{1}{\sqrt{5}}, -\frac{1}{\sqrt{5}}\right)$ , whose probability is  $\frac{1}{5\pi}$ . Therefore,

$$\mathbb{P}(S \le 1/2, T \le 1/2) = \frac{1}{2} + \mathbb{P}(A_4) \ge \frac{1}{2} + \frac{1}{5\pi} > \frac{9}{16}$$

It completes our proof for the claim that S and T are not independent.

(h) Based on the Jacobian method (Theorem 8.1 in Lecture Note 8), we take  $s = \frac{y}{\sqrt{1-x^2}}$ ,  $t = \frac{x}{\sqrt{1-y^2}}$ , and calculate the determinant

$$\left|\frac{\partial(s,t)}{\partial(x,y)}\right| = \left|\frac{\frac{\partial s}{\partial x}}{\frac{\partial t}{\partial x}} \frac{\frac{\partial s}{\partial y}}{\frac{\partial t}{\partial y}}\right| = \left|\frac{\frac{xy}{(1-x^2)^{\frac{3}{2}}}}{\frac{1}{\sqrt{1-y^2}}} \frac{\frac{1}{\sqrt{1-x^2}}}{\frac{xy}{(1-y^2)^{\frac{3}{2}}}}\right| = \frac{1-x^2-y^2}{(1-x^2)^{\frac{3}{2}}(1-y^2)^{\frac{3}{2}}}$$

Solving for  $x^2, y^2$ , we obtain that  $x^2 = \frac{t^2(1-s^2)}{1-s^2t^2}$  and  $y^2 = \frac{s^2(1-t^2)}{1-s^2t^2}$ . Hence, the joint PDF of (S,T) is

$$\begin{split} f(s,t) &= \frac{1}{\pi} \left| \frac{\partial(x,y)}{\partial(s,t)} \right| \cdot \mathbbm{1}_{\{x^2+y^2 \le 1\}} \\ &= \frac{1}{\pi} \cdot \frac{(1-x^2)^{\frac{3}{2}}(1-y^2)^{\frac{3}{2}}}{1-x^2-y^2} \cdot \mathbbm{1}_{\left\{\frac{t^2(1-s^2)}{1-s^2t^2} + \frac{s^2(1-t^2)}{1-s^2t^2} \le 1\right\}} \\ &= \frac{\sqrt{(1-t^2)(1-s^2)}}{\pi(1-s^2t^2)} \cdot \mathbbm{1}_{\{(1-s^2)(1-t^2) \ge 0\}} \\ &= \frac{\sqrt{(1-t^2)(1-s^2)}}{\pi(1-s^2t^2)} \cdot \mathbbm{1}_{\{(s,t)\in[-1,1]\times[-1,1]\}}. \end{split}$$

Notice that f(s,t) cannot be factored into the product of two marginal densities, so it provides another way to justify the dependence between S and T.

To compute the CDF  $F(s,t) = \int_{-1}^{s} \int_{-1}^{t} f(u,v) \, du \, dv$ , we need to consider 4 different cases regarding the signs of s and t.

**Case 1:**  $-1 \le s < 0, -1 \le t < 0$ . Then, from  $-1 < \frac{y}{\sqrt{1-x^2}} \le s, -1 < \frac{x}{\sqrt{1-y^2}} \le t$ , we know that the integral range of  $f_{X,Y}(x,y)$  on D satisfies

$$x^{2} + y^{2} \le 1$$
  $\frac{y^{2}}{s^{2}} + x^{2} \ge 1$ ,  $\frac{x^{2}}{t^{2}} + y^{2} \ge 1$ .

See also Figure 4 for a graphical illustration. Thus,

$$\begin{aligned} F(s,t) &= \int \int_{D_1} \frac{1}{\pi} \, dx dy \\ &= \frac{4}{\pi} \left[ \frac{\pi}{4} - \left( \frac{\pi t}{4} + \frac{\pi s}{4} - \int_0^{\theta_1} \int_0^1 tr \, dr d\theta - \int_{\theta_1}^{\frac{\pi}{2}} \int_0^1 sr \, dr d\theta \right) \right] \\ &= (1-t) + \frac{2(1-s)\theta_1}{\pi}, \end{aligned}$$

where  $\theta_1 = \arctan\left(\frac{s\sqrt{1-t^2}}{t\sqrt{1-s^2}}\right)$ .



Figure 4: Integrating range of  $f_{X,Y}(x,y)$  when  $-1 \le s < 0, -1 \le t < 0$ .

**Case 2:**  $0 \le s \le 1, -1 \le t < 0$ . Then, from  $-1 < \frac{y}{\sqrt{1-x^2}} \le s, -1 < \frac{x}{\sqrt{1-y^2}} \le t$ , we know that the integral

range of  $f_{X,Y}(x,y)$  on D satisfies

$$\frac{x^2}{t^2} + y^2 \ge 1$$

and

$$\frac{y^2}{s^2} + x^2 \le 1, y \ge 0 \quad \text{ or } \quad \frac{x^2}{t^2} + y^2 \le 1, x > 0.$$

See also Figure 5 for a graphical illustration. Thus,

$$\begin{split} F(s,t) &= \int \int_{D_2} \frac{1}{\pi} \, dx dy \\ &= \frac{1}{\pi} \left[ \frac{\pi s}{2} - 2 \left( \int_0^{\theta_1} \int_0^1 tr \, dr d\theta + \int_{\theta_1}^{\frac{\pi}{2}} \int_0^1 sr \, dr d\theta \right) + \frac{\pi}{2} - \frac{\pi t}{2} \right] \\ &= \frac{1}{\pi} \left[ \frac{\pi s}{2} - 2 \left( \frac{t\theta_1}{2} + \frac{s}{2} \left( \frac{\pi}{2} - \theta_1 \right) \right) + \frac{\pi}{2} - \frac{\pi t}{2} \right] \\ &= \frac{1 - t}{2} + \frac{(1 - s)\theta_1}{2\pi}, \end{split}$$

where  $\theta_1 = \arctan\left(-\frac{s\sqrt{1-t^2}}{t\sqrt{1-s^2}}\right)$ .



Figure 5: Integrating range of  $f_{X,Y}(x,y)$  when  $0 \le s \le 1, -1 \le t < 0$ .

**Case 3:**  $-1 \le s < 0, 0 \le t \le 1$ . Then, from  $-1 < \frac{y}{\sqrt{1-x^2}} \le s, -1 < \frac{x}{\sqrt{1-y^2}} \le t$ , we know that the integral range of  $f_{X,Y}(x,y)$  on D satisfies

$$x^2 + \frac{y^2}{s^2} > 1$$

and

$$x^{2} + y^{2} \le 1, x \le 0$$
 or  $\frac{x^{2}}{t^{2}} + y^{2} \le 1, x > 0.$ 

See also Figure 6 for a graphical illustration. Thus,

$$F(s,t) = \int \int_{D_3} \frac{1}{\pi} dx dy$$
  
=  $\frac{1}{\pi} \left[ \frac{\pi}{2} - \frac{\pi s}{2} + \frac{\pi t}{2} - 2\left(\frac{t\theta_1}{2} + \frac{s}{2}\left(\frac{\pi}{2} - \theta_1\right)\right) \right]$   
=  $\frac{1+t}{2} - s + \frac{(s-t)\theta_1}{\pi},$ 



Figure 6: Integrating range of  $f_{X,Y}(x,y)$  when  $-1 \le s < 0, 0 \le t \le 1$ .

**Case 4:**  $0 \le s \le 1, 0 \le t \le 1$ . Then, from  $-1 < \frac{y}{\sqrt{1-x^2}} \le s, -1 < \frac{x}{\sqrt{1-y^2}} \le t$ , we know that the integral range of  $f_{X,Y}(x,y)$  on D satisfies

$$x^{2} + y^{2} \le 1, y < 0$$
 or  $x^{2} + \frac{y^{2}}{s^{2}} \le 1, y \ge 0$ 

and

$$x^2 + y^2 \le 1, x < 0$$
 or  $\frac{x^2}{t^2} + y^2 \le 1, x \ge 0$ 

See also Figure 7 for a graphical illustration. Thus,

$$\begin{split} F(s,t) &= \int \int_{D_4} \frac{1}{\pi} \, dx dy \\ &= \frac{1}{\pi} \left[ \frac{\pi}{4} + \frac{\pi s}{4} + \frac{\pi t}{4} + 2\left( \frac{t\theta_1}{2} + \frac{s}{2} \left( \frac{\pi}{2} - \theta_1 \right) \right) \right] \\ &= \frac{1 + t + 2s}{4} + \frac{(s - t)\theta_1}{2\pi}, \end{split}$$

where  $\theta_1 = \arctan\left(\frac{s\sqrt{1-t^2}}{t\sqrt{1-s^2}}\right)$ .



Figure 7: Integrating range of  $f_{X,Y}(x,y)$  when  $0 \le s \le 1, 0 \le t \le 1$ .

**Problem 3** (Homework 1; Exercise 1.32 in Casella and Berger 2002). An employer is about to hire one new employee from a group of N candidates, whose future potential can be rated on a scale from 1 to N. The employer proceeds according to the following rules:

(a) Each candidate is seen in succession (in random order) and a decision is made whether to hire the candidate.

(b) Having rejected m-1 candidates (m > 1), the employer can hire the mth candidate only if the mth candidate is better than the previous m-1.

Suppose a candidate is hired on the *i*th trial. What is the probability that the best candidate was hired?

**Solution.** Let E be the event that  $i^{th}$  candidate is the best, and F be the event that  $i^{th}$  candidate is better than the previous i - 1 candidates. The probability that the best candidate was hired is

$$\mathbb{P}(E|F) = \frac{\mathbb{P}(F|E) \cdot \mathbb{P}(E)}{\mathbb{P}(F)}.$$

Here,  $\mathbb{P}(F|E) = 1$  because if the *i*<sup>th</sup> candidate is the best among all, it must be better than the previous i-1 candidates. Moreover,  $\mathbb{P}(E) = \frac{1}{N}$  and  $\mathbb{P}(F) = \frac{1}{i}$ . Thus,

$$\mathbb{P}(E|F) = \frac{i}{N}$$

## References

- G. Casella and R. Berger. *Statistical Inference*. Duxbury advanced series. Thomson Learning, 2nd ed. edition, 2002.
- M. Perlman. Probability and mathematical statistics i (stat 512 lecture notes), 2020. URL https://sites.stat.washington.edu/people/mdperlma/STAT%20512%20MDP%20Notes.pdf.