

Statistics 581, Problem Set 1 Solutions

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1. Let X and Y be i.i.d. $\text{Uniform}(0, 1)$ random variables. Define $U = X + Y$, $V = \max(X, Y) = X \vee Y$.

- (i) What is the range of (U, V) ?
- (ii) Find the joint density function $f_{U,V}(u, v)$ of the pair (U, V) . Are U and V independent?

Solution: (i) The range of (X, Y) is

$A = \{(x, y) : 0 \leq x \leq 1, 0 \leq y \leq 1\}$. The range of (U, V) is

$B = \{(u, v) : 0 \leq u \leq 1, u/2 \leq v \leq u\} \cup \{(u, v) : 1 < u \leq 2, u/2 \leq v \leq 1\}$.

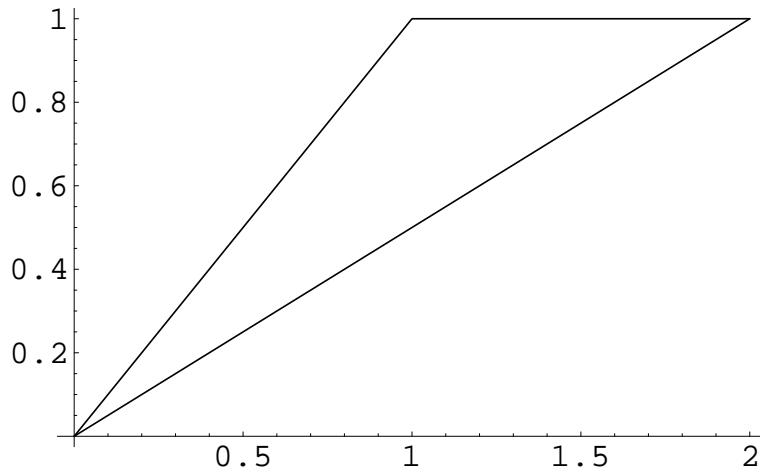


Figure 1: Range of U, V .

(ii) First solution - via Jacobians: The transformation $(X, Y) \rightarrow (U, V)$ is 2-1 and onto from A to B . On the set $x < y$, its inverse is given by $X = U - V, Y = V$; on the set $x > y$, its inverse is given by $X = V, Y = U - V$. These mappings are continuously differentiable on $B^* \equiv B \setminus \{(u, v) : v = u/2\} = B \setminus$ a null set. On B^* the Jacobian of the transformations are

$$\det \begin{pmatrix} 1 & -1 \\ 0 & 1 \end{pmatrix} = 1 \quad \text{if } x < y, \quad \det \begin{pmatrix} 0 & 1 \\ 1 & -1 \end{pmatrix} = -1 \quad \text{if } x > y. \quad (1)$$

Thus by the usual transformation of densities formula, the joint density of (U, V) is obtained from $f_{X,Y}(x, y) = 1_{[0,1]}(x)1_{[0,1]}(y)$ as follows:

$$\begin{aligned} f_{U,V}(u, v) &= f_{X,Y}(x(u, v), y(u, v)) \left| \det \frac{\partial(x, y)}{\partial(u, v)} \right| 1_{[x(u, v) < y(u, v)]} \\ &\quad + f_{X,Y}(x(u, v), y(u, v)) \left| \det \frac{\partial(x, y)}{\partial(u, v)} \right| 1_{[x(u, v) > y(u, v)]} \\ &= (1_{[0,1]}(u - v)1_{[0,1]}(v) \cdot 1 + 1_{[0,1]}(v)1_{[0,1]}(u - v) \cdot 1) 1_{[v > u/2]} \\ &= 2 \cdot 1_B(u, v). \end{aligned}$$

Thus the joint density of (U, V) is uniform on B . The random variables U and V are clearly *not* independent since the range of (U, V) is not a product set in R^2 ; moreover, the joint density of (U, V) does not factor into the product of its marginal densities. [The marginal densities are given by

$$f_U(u) = \int f_{U,V}(u, v)dv = \begin{cases} \int_{u/2}^u 2dv = u, & u \in [0, 1] \\ \int_{u/2}^1 2dv = 2 - u, & u \in (1, 2] \end{cases}$$

and

$$f_V(v) = \int f_{U,V}(u, v)du = \int_v^{2v} 2du = 2v 1_{[0,1]}(v).]$$

Second solution by direction calculation of the joint distribution function: Note that we can write

$$\begin{aligned} &P(U \leq u, V \leq v) \\ &= P(X + Y \leq u, X \vee Y \leq v) = P(X + Y \leq u, X \leq v, Y \leq v) \\ &= \begin{cases} v^2 & \text{if } 0 < v < 1, 2v \leq u, \\ v^2 - \frac{1}{2}(2v - u)^2 & \text{if } v < u < 2v, \\ \frac{1}{2}u^2 & \text{if } 0 < u < v. \end{cases} \end{aligned}$$

(This is easy by pictures!) Computing $(\partial^2/\partial u\partial v)P(U \leq u, V \leq v)$ on each of these pieces separately again yields $f_{U,V}(u, v) = 21_B(u, v)$. Also note that the marginal distribution functions of U and V are given by $F_U(u) = (1/2)u^2 1_{[0,1)}(u) + \{1 - \frac{1}{2}(2-u)^2\} 1_{[1,2)}(u)$ on $0 \leq u < 2$ and $F_V(v) = v^2$ for $0 < v < 1$.

2. Prove part (ii) of Proposition 1.1: There exists a minimal field, σ -field, and monotone class generated by any class of subsets of Ω .

Solution: Let \mathcal{C} be a class of subsets of Ω . By part (i) of Proposition 1.1, the intersection of any family of fields, σ -fields, or monotone classes, are again fields, σ -fields, or monotone classes, respectively. Consider the collection $\{\mathcal{A}_\lambda\}$, of all fields containing \mathcal{C} . This collection is non-empty since it contains the field 2^Ω containing all subsets of Ω . Thus by Proposition 1.1(i) $\cap_\lambda \mathcal{A}_\lambda$ is again a field. This is clearly the minimal field containing \mathcal{C} since $\cap_\lambda \mathcal{A}_\lambda \subset \mathcal{A}_{\lambda'}$ for each particular field $\mathcal{A}_{\lambda'}$ containing \mathcal{C} . The proofs for σ -fields and monotone classes is exactly the same.

3. (a) Suppose that $\{\mathcal{A}_n\}$ is an increasing sequence of fields, i.e. $\mathcal{A}_n \subset \mathcal{A}_{n+1}$ for all $n \geq 1$. Show that $\cup_{n=1}^\infty \mathcal{A}_n$ is a field. (b) Suppose that the \mathcal{A}_n of (a) are σ -fields. Show by constructing a counterexample that $\cup_{n=1}^\infty \mathcal{A}_n$ need not be a σ -field.

Solution: (a) Let $\mathcal{A} = \cup_{n=1}^\infty \mathcal{A}_n$.

(i) If $A \in \mathcal{A}$, then $A \in \mathcal{A}_n$ for some n , so $A^c \in \mathcal{A}_n$ (since \mathcal{A}_n is a field), which implies $A^c \in \mathcal{A}$.

(ii) If $A, B \in \mathcal{A}$, then $A \in \mathcal{A}_m, B \in \mathcal{A}_n$ for some m, n . Suppose that $m \leq n$. Then $A \in \mathcal{A}_n$ since $\mathcal{A}_m \subset \mathcal{A}_{m+1} \subset \dots \subset \mathcal{A}_n$. Hence $A \cup B \in \mathcal{A}_n$ (since \mathcal{A}_n is a field). Hence $A \cup B \in \cup_{n=1}^\infty \mathcal{A}_n = \mathcal{A}$.

By (i) and (ii) \mathcal{A} is a field.

(b) Let $\mathcal{A}_1 = \sigma(\{\emptyset, [0, 1]\})$, $\mathcal{A}_2 = \sigma(\mathcal{A}_1 \cup \{[0, 1/2]\})$, $\mathcal{A}_3 = \sigma(\mathcal{A}_2 \cup \{[0, 1 - 1/3]\})$, \dots , $\mathcal{A}_n = \sigma(\mathcal{A}_{n-1} \cup \{[0, 1 - 1/n]\})$, \dots . Then each \mathcal{A}_n is a finite σ -field with $\sharp(\mathcal{A}_n) = 2^n$; clearly $\mathcal{A}_n \subset \mathcal{A}_{n+1}$ by construction. Consider the sets $A_n = [0, 1 - 1/n] \in \mathcal{A}_n$ for $n \geq 2$. Then $A = \cup_{n=2}^\infty A_n = [0, 1) \notin \cup_{n=1}^\infty \mathcal{A}_n$. [If $A = [0, 1) \in \cup_{n=1}^\infty \mathcal{A}_n$, then $A \in \mathcal{A}_{n_0}$ for some n_0 , but this fails. All the sets in $\cup_{n=1}^\infty \mathcal{A}_n$ are closed on the right.]

4. Let μ and ν be Lebesgue-Stieltjes measures on (R, \mathcal{B}) with corresponding generalized d.f.'s F and G . Show that:

(a) $\int_{(a,b]} (F(y) - F(a)) dG(y) = (\mu \times \nu)(\{(x, y) : a < x \leq y \leq b\})$.

(b) $\int_{(a,b]} F(y)dG(y) + \int_{(a,b]} G(y)dF(y) = F(b)G(b) - F(a)G(a) + \sum_{x \in (a,b]} \mu(\{x\})\nu(\{x\})$.

To see that the second term is needed, let $F(x) = G(x) = 1_{[0,\infty)}(x)$ and $a < 0 < b$.

(c) If $F = G$ is continuous, then $\int_{(a,b]} F(y)dF(y) = (F^2(b) - F^2(a))/2$.

[Hint: use Fubini's theorem.]

Solution: (a) Now $F(y) - F(a) = \mu((a, y]) = \int_{(a,y]} d\mu(x)$ so, by using the Tonelli part of the Fubini-Tonelli theorem,

$$\begin{aligned} \int_{(a,b]} (F(y) - F(a))dG(y) &= \int_{(a,b]} \left(\int_{(a,y]} d\mu(x) \right) d\nu(y) \\ &= \int_{(a,b]} \int_{(a,b]} 1_{[x \leq y]} d\mu(x) d\nu(y) \\ &= (\mu \times \nu)(\{(x, y) : a < x \leq y \leq b\}) \\ &\equiv (\mu \times \nu)(B) \end{aligned}$$

where the set $B \equiv \{(x, y) : a < x \leq y \leq b\}$.

(b) By using (a) we compute

$$\begin{aligned} \int_{(a,b]} F(y)dG(y) &= \int_{(a,b]} (F(y) - F(a) + F(a))dG(y) \\ &= (\mu \times \nu)(B) + F(a)(G(b) - G(a)). \end{aligned}$$

Similarly,

$$\begin{aligned} \int_{(a,b]} G(y)dF(y) &= \int_{(a,b]} (G(y) - G(a) + G(a))dF(y) \\ &= (\mu \times \nu)(A) + G(a)(F(b) - F(a)) \end{aligned}$$

where $A \equiv \{(x, y) : a < y \leq x \leq b\}$. Note that

$$A \cap B = \{(x, x) : a < x \leq b\} \equiv D.$$

Thus by adding these two expressions we find that

$$\begin{aligned} \int_{(a,b]} F(y)dG(y) + \int_{(a,b]} G(y)dF(y) \\ &= (\mu \times \nu)(B) + F(a)(G(b) - G(a)) \\ &\quad + (\mu \times \nu)(A) + G(a)(F(b) - F(a)) \end{aligned}$$

$$\begin{aligned}
&= (\mu \times \nu)((a, b] \times (a, b]) + (\mu \times \nu)(D) \\
&\quad + F(a)(G(b) - G(a)) + G(a)(F(b) - F(a)) \\
&= (F(b) - F(a))(G(b) - G(a)) + F(a)(G(b) - G(a)) + G(a)(F(b) - F(a)) \\
&\quad + (\mu \times \nu)(D) \\
&= F(b)G(b) - F(a)G(a) + (\mu \times \nu)(D) \\
&= F(b)G(b) - F(a)G(a) + \sum_{x \in (a, b]} \mu(\{x\})\nu(\{x\})
\end{aligned}$$

where the last line follows since the Lebesgue-Stieltjes measures μ and ν can have at most a countable number of discontinuity points in the corresponding generalized d.f.'s F and G in any bounded set.

Note that for the example given with $F(x) = G(x) = 1_{[0, \infty)}(x)$ and $a < 0 < b$, the left side of the formula is $\int_{(a, b]} F(y)dG(y) + \int_{(a, b]} G(y)dF(y) = 1 + 1 = 2$, while the right side is

$$F(b)G(b) - F(a)G(a) + \sum_{x \in (a, b]} \mu(\{x\})\nu(\{x\}) = 1 - 0 + \mu(0)\nu(\{0\}) = 1 + 1 = 2.$$

(c) If $F = G$ is continuous, then from (b) it follows that

$$2 \int_{(a, b]} F dF = \int_{(a, b]} F dG + \int_{(a, b]} G dF = F(b)^2 - F(a)^2 + 0.$$

5. Let $\mathcal{X} = (0, 1)$, $\mathcal{Y} = (0, 1)$, both equipped with the Borel sets and Lebesgue measure. Let

$$g(x, y) = \frac{x^2 - y^2}{(x^2 + y^2)^2} \quad \text{for } (x, y) \in (0, 1) \times (0, 1).$$

Show that:

- (a) $\int_0^1 (\int_0^1 g(x, y) dy) dx = \pi/4$.
- (b) $\int_0^1 (\int_0^1 g(x, y) dx) dy = -\pi/4$.
- (c) Why does Fubini's theorem fail here?

Solution: (a) It is easily seen that

$$\int_0^1 g(x, y) dy = \frac{y}{x^2 + y^2} \Big|_{y=0}^1 = \left(\frac{1}{1 + x^2} - 0 \right) = \frac{1}{1 + x^2},$$

and hence

$$\int_0^1 \left(\int_0^1 g(x, y) dy \right) dx = \int_0^1 \frac{1}{1+x^2} dx = \arctan(x) \Big|_0^1 = \pi/4.$$

(b) Similarly,

$$\int_0^1 g(x, y) dx = -\frac{x}{x^2+y^2} \Big|_{x=0}^1 = -\left(\frac{1}{1+y^2} - 0 \right) = \frac{-1}{1+y^2},$$

and hence

$$\int_0^1 \left(\int_0^1 g(x, y) dx \right) dy = -\int_0^1 \frac{1}{1+y^2} dy = -\arctan(y) \Big|_0^1 = -\pi/4.$$

(c) It is clear that the hypothesis $g \in L_1(\lambda \times \lambda)$ fails in this example; if it held, the two iterated integrals in (a) and (b) must be the same by the Fubini part of the Fubini-Tonelli theorem. To see that $g \notin L_1(\mu \times \nu)$ directly, note that the function g is non-negative on the set $0 \leq y \leq x \leq 1$, and non-positive on the set $0 \leq x \leq y \leq 1$. Thus

$$\int_0^1 \int_0^1 |g(x, y)| dx dy = \int \int_{\{0 \leq y \leq x \leq 1\}} g(x, y) dx dy + \int \int_{\{0 \leq x \leq y \leq 1\}} -g(x, y) dx dy,$$

where, by the Tonelli part of the Fubini-Tonelli theorem

$$\begin{aligned} \int \int_{\{0 \leq y \leq x \leq 1\}} g(x, y) dx dy &= \int_0^1 \left(\int_y^1 g(x, y) dx \right) dy \\ &= \int_0^1 \left(\frac{-x}{x^2+y^2} \Big|_{x=y}^1 \right) dy \\ &= \int_0^1 \left(\frac{-1}{1+y^2} - \frac{-y}{2y^2} \right) dy \\ &= \int_0^1 \left(\frac{1}{2y} - \frac{1}{1+y^2} \right) dy \\ &= \infty - \pi/4 = \infty. \end{aligned}$$

By symmetry it follows that

$$\int \int_{\{0 \leq x \leq y \leq 1\}} -g(x, y) dx dy = \infty,$$

and hence we have $\int \int |g(x, y)| dx dy = \infty$. Thus $g \notin L_1(\lambda \times \lambda)$ on $[0, 1] \times [0, 1]$.

6. Lehmann and Casella, TPE, problem 1.4, page 62: Let X and Y have common expectation θ , variances σ^2 and τ^2 , and correlation coefficient ρ . Determine the conditions on σ , τ , and ρ under which:
- (a) $Var(X) < Var[(X + Y)/2]$;
- (b) The value of α that minimizes $Var(\alpha X + (1 - \alpha)Y)$ is negative.

Solution: (a) First note that

$$Var((X + Y)/2) = \frac{1}{4}(\sigma^2 + \tau^2 + 2\rho\sigma\tau).$$

Thus the inequality in question holds if

$$\sigma^2 + \tau^2 + 2\rho\sigma\tau > 4\sigma^2,$$

or equivalently if

$$3\sigma^2 - 2\rho\tau\sigma - \tau^2 < 0.$$

Since the left side, viewed as a function of σ has zeros at

$$\begin{aligned}\sigma_0 &= \frac{2\rho\tau \pm \sqrt{4\rho^2\tau^2 - 4(3)(-\tau^2)}}{6} \\ &= \frac{\tau}{3} \left(\rho \pm \sqrt{\rho^2 + 3} \right),\end{aligned}$$

it follows that the inequality in (a) holds for given τ and ρ if

$$0 \leq \sigma < \frac{\tau}{3}(\rho + \sqrt{\rho^2 + 3}).$$

When $\rho = -1$, this requires $\sigma < \tau/3$; when $\rho = 0$, we need $\sigma < \tau/\sqrt{3}$; when $\rho = 1$, we only need $\sigma < \tau$.

(b) First note that

$$\begin{aligned}f(\alpha) &\equiv Var(\alpha X + (1 - \alpha)Y) = Var(\alpha(X - Y) + Y) \\ &= \alpha^2 Var(X - Y) + 2\alpha Cov(X - Y, Y) + Var(Y).\end{aligned}$$

Hence

$$f'(\alpha) = 2\alpha Var(X - Y) + 2Cov(X - Y, Y)$$

and $f''(\alpha) = 2Var(X - Y)$. It follows that f is minimized as a function of α by

$$\alpha_{min} = \frac{-Cov(X - Y, Y)}{Var(X - Y)} = \frac{Cov(Y - X, Y)}{Var(X - Y)} = \frac{\tau^2 - \rho\sigma\tau}{\sigma^2 + \tau^2 - 2\rho\sigma\tau},$$

and this is negative if $\tau^2 - \rho\sigma\tau < 0$, or equivalently if $\tau < \rho\sigma$. For example, if $\rho = 1/2$, $\sigma = 1$ and $\tau = 1/4$, then the minimizing value of α is

$$\alpha_{min} = \frac{1/16 - (1/2)(1)(1/4)}{1 + 1/16 - 2(1/2)(1)(1/4)} = \frac{-1/16}{13/16} = \frac{-1}{13}.$$

Note that if $\rho = 0$, then $\alpha_{min} = \tau^2/(\sigma^2 + \tau^2)$; if the variances are equal, this further simplifies to $\alpha_{min} = 1/2$.

7. Lehmann and Casella, TPE, problem 1.10, page 62. Show that for real numbers x_1, \dots, x_n in the domain of a monotone real-valued functions h , the function

$$H(a) \equiv \frac{1}{n} \sum_{i=1}^n (h(x_i) - h(a))^2$$

is minimized by that value of a , say a_{min} , given by

$$a_{min} = h^{-1} \left(\frac{1}{n} \sum_1^n h(x_i) \right).$$

Find particular functions h so that this yields the arithmetic, harmonic, and geometric means of the x_i 's.

Solution: First note that that $H(a)$ is minimized by any value of a , say a_{min} , satisfying

$$h(a_{min}) = \frac{1}{n} \sum_1^n h(x_i).$$

This follows by noting that

$$H(a) = H(a_{min}) + (h(a_{min}) - h(a))^2.$$

Since we have assumed that h is (strictly) monotone, it follows easily that

$$a_{min} = h^{-1} \left(\frac{1}{n} \sum_1^n h(x_i) \right).$$

When $h(x) = x$,

$$a_{min} = \frac{1}{n} \sum_1^n (x_i) = \text{the arithmetic mean}.$$

When $h(x) = 1/x$,

$$a_{min} = \frac{1}{\frac{1}{n} \sum_1^n \frac{1}{x_i}} = \text{the harmonic mean.}$$

When $h(x) = \log(x)$,

$$\begin{aligned} a_{min} &= \exp\left(\frac{1}{n} \sum_1^n \log(x_i)\right) = \exp\left(\frac{1}{n} \log\left(\prod_1^n x_i\right)\right) \\ &= \exp\left(\log\left(\left(\prod_1^n x_i\right)^{1/n}\right)\right) = \left(\prod_1^n x_i\right)^{1/n} \\ &= \text{the geometric mean.} \end{aligned}$$