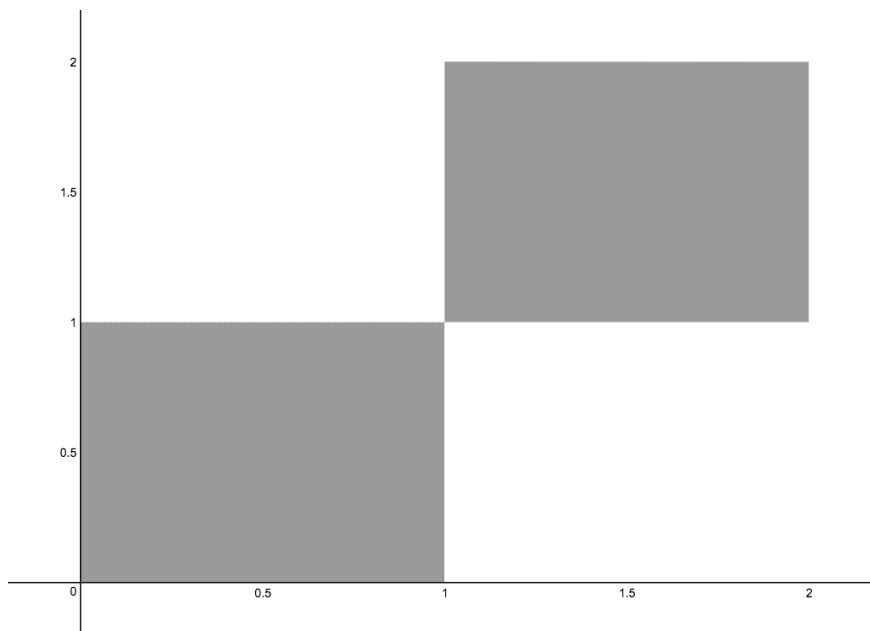


1. **Continuous LLSE**

Suppose that X and Y are uniformly distributed on the following figure:



That is, X and Y have the joint distribution

$$f_{X,Y}(x,y) = \begin{cases} 1/2, & 0 \leq x \leq 1, 0 \leq y \leq 1 \\ 1/2, & 1 \leq x \leq 2, 1 \leq y \leq 2 \end{cases}$$

- (a) Do you expect X and Y to be positively correlated, negatively correlated, or neither?
- (b) Compute the marginal distribution of X .
- (c) Compute $L[Y | X]$.
- (d) What is $E[Y | X]$?

Solution:

- (a) Positively correlated, because high values of Y correspond to high values of X .
- (b) Intuitively, if we slice the joint distribution at any $x \in [0, 2]$, then the probability is the same, so we should expect X to be uniformly distributed on $[0, 2]$. We verify this by

explicit computation:

$$\begin{aligned} f_X(x) &= \int_{-\infty}^{\infty} f_{X,Y}(x,y) dy = 1\{0 \leq x \leq 1\} \int_0^1 \frac{1}{2} dy + 1\{1 \leq x \leq 2\} \int_1^2 \frac{1}{2} dy \\ &= \frac{1}{2} 1\{0 \leq x \leq 2\} \end{aligned}$$

- (c) $E[X] = E[Y] = 1$ by symmetry. Since X is uniform on $[0, 2]$, $\text{var}(X) = 4 \cdot 1/12 = 1/3$ (since the variance of a $\text{Unif}[0, 1]$ random variable is $1/12$). We compute the covariance:

$$\begin{aligned} E[XY] &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} xy f_{X,Y}(x,y) dx dy = \int_0^1 \int_0^1 xy \cdot \frac{1}{2} dx dy + \int_1^2 \int_1^2 xy \cdot \frac{1}{2} dx dy \\ &= \frac{1}{2} \left(\int_0^1 x dx \int_0^1 y dy + \int_1^2 x dx \int_1^2 y dy \right) = \frac{1}{2} \left(\frac{1}{4} + \frac{9}{4} \right) = \frac{5}{4} \end{aligned}$$

So $\text{cov}(X, Y) = 5/4 - 1 \cdot 1 = 1/4$. The LLSE is

$$\begin{aligned} L[Y | X] - 1 &= \frac{1/4}{1/3} (X - 1) \\ L[Y | X] &= \frac{3}{4} X + \frac{1}{4} \end{aligned}$$

- (d) The easiest way to solve this is to look at the picture of the joint density, and draw horizontal lines through middles of each of the two squares. Intuitively, $E[Y | X]$ means “for each slice of $X = x$, what is the best guess of Y ”? Slightly more formally, one can argue that conditioned on $X = x$ for $0 < x < 1$, $Y \sim U[0, 1]$, so $E[Y | X = x] = 1/2$ in this region. Conditioned on $X = x$ for $1 < x < 2$, $Y \sim U[1, 2]$, so $E[Y | X = x] = 3/2$ in this region.

$$E[Y | X = x] = \begin{cases} 1/2, & 0 \leq x \leq 1 \\ 3/2, & 1 \leq x \leq 2 \end{cases}$$

2. Conditioning on Exponentials

Let X_i be i.i.d. $\text{Expo}(\lambda)$ random variables.

- Compute $E[Y | Z]$, where $Y = \max\{X_1, X_2\}$ and $Z = \min\{X_1, X_2\}$.
- Compute $E[X_1 + X_2 | Z]$. (*Hint*: Use part (a).)
- Use part (b) to compute $E[Z]$.
- Compute $E[X_1 + X_2 | X_1 + X_2 + X_3]$.

Solution:

- View the exponential distributions as light bulbs burning out. Conditioned on Z , we are looking for the expected time until the next light bulb burns out, which by the memory-less property is $1/\lambda$. Therefore, $E[Y | Z] = Z + 1/\lambda$.

(b) Observe that $X_1 + X_2 = \min\{X_1, X_2\} + \max\{X_1, X_2\}$.

$$E[X_1 + X_2 | Z] = E[\min\{X_1, X_2\} | Z] + E[\max\{X_1, X_2\} | Z] = Z + Z + 1/\lambda = 2Z + 1/\lambda$$

(c) Take expectations of both sides and apply iterated expectation.

$$E[X_1 + X_2] = E[E[X_1 + X_2 | Z]] = 2E[Z] + \frac{1}{\lambda}$$

Note that $E[X_1 + X_2] = 2/\lambda$ by linearity, so we find that $E[Z] = 1/(2\lambda)$. (In fact, we can prove a stronger result: the minimum of two independent exponentials with parameter λ is exponential with parameter 2λ .)

(d) Given $X_1 + X_2 + X_3$, the i.i.d. nature of the exponentials tells us that we should expect $E[X_1 + X_2 | X_1 + X_2 + X_3] = (2/3)(X_1 + X_2 + X_3)$.

3. Erlang Distribution

In lecture, we proved the following **Fact**: if the lifetimes of light bulbs are i.i.d. $\text{Expo}(1)$ and we replace the light bulb as soon as one dies out, then the number of light bulbs we replace by time t follows the Poisson distribution with mean t . Using this fact, find the density of the sum of two i.i.d. $\text{Expo}(1)$ random variables.

Solution:

First, we compute the probability that a light bulb will die in the first dt seconds, where dt is an infinitesimal amount of time. Let $T \sim \text{Expo}(1)$.

$$\Pr[T \in (0, dt)] \approx f_T(0) dt = \lambda dt$$

By the memoryless property, this is the same as the probability that the light bulb will die in *any* interval of dt seconds, conditioned on the fact that the light bulb has survived until that point in time. We compute now $\Pr[T \in (t, t + dt)]$, the probability that the lifetime of two light bulbs is within $(t, t + dt)$. In order for this to happen, we must have the second light bulb die in the interval $(t, t + dt)$, and the probability of this, we have just argued, is $1 dt$. What about the first light bulb? We must have replaced exactly one light bulb in the first t seconds, which by the **Fact**, is the probability that a Poisson distribution with mean t takes on the value 1. Putting this together,

$$\Pr[T \in (t, t + dt)] = 1 dt \cdot \frac{t \cdot e^{-t}}{1!} = te^{-t} dt$$

Therefore, we identify the density as

$$f_{T_1+T_2}(t) = te^{-t}, \quad t > 0$$

This is a special case of the **Erlang** distribution, which is the density of the sum of k i.i.d. $\text{Expo}(\lambda)$ distributions. The intimate connection between the exponential distribution and the Poisson distribution is part of the study of stochastic processes.