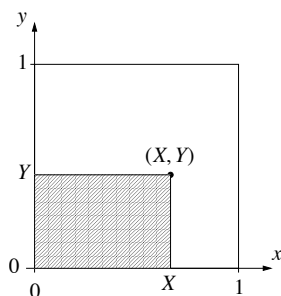


Problem 1. Let X and Y be independent random variables, each with distribution $\text{Uniform}(0, 1)$. Then the point with coordinates (X, Y) is a random vector that is uniformly distributed in the unit square, i.e., its probability density function is

$$f_{X,Y}(x, y) = \begin{cases} 1 & \text{if } 0 \leq x \leq 1, \ 0 \leq y \leq 1, \\ 0 & \text{otherwise.} \end{cases}$$

Let A be the random variable equal to the area of the rectangle with vertices at the points $(0, 0)$, $(0, Y)$, $(X, 0)$, and (X, Y) – see the figure. Clearly, A is a continuous random variable taking values in the interval $[0, 1]$.



- (a) In the (x, y) plane, sketch the domain determined by the inequalities $0 \leq x \leq 1$, $0 \leq y \leq 1$, $xy \leq a$ (where a is a value between 0 and 1).
- (b) Show that the cumulative distribution function of A ,

$$F_A(a) = \mathbb{P}(A \leq a) = \mathbb{P}(XY \leq a) = \iint_{xy \leq a} f_{X,Y}(x, y) \, dx \, dy,$$

is given by

$$F_A(a) = \begin{cases} 0 & \text{if } a \in (-\infty, 0], \\ a(1 - \ln a) & \text{if } a \in (0, 1], \\ 1 & \text{if } a \in [1, \infty). \end{cases}$$

Hint: What you drew in part (a) will be useful.

- (c) Find the probability density function, $f_A(a)$, of A . Be sure to specify $f_A(a)$ for all values of a .
- (d) Determine the expected value $\mathbb{E}[A]$ of the area A of the random square with sides of length X and Y .

Problem 2. The joint p.m.f. $p_{X,Y}(x_k, y_m) = \mathbb{P}(X = x_k, Y = y_m)$ of the discrete RVs X and Y has values given in the table.

	$Y = 1$	$Y = 3$	$Y = 4$
$X = 5$	0	$\frac{1}{15}$	$\frac{2}{15}$
$X = 7$	$\frac{3}{15}$	$\frac{4}{15}$	$\frac{5}{15}$

- Find the marginal p.m.f.s $p_X(x_k)$ and $p_Y(y_m)$.
- Find the expected value $\mathbb{E}[X]$ of X .
- Compute the conditional p.m.f.s $p_{X|Y}(x_k|y_m) = \mathbb{P}(X = x_k|Y = y_m)$ for $y_m = 1, 3, 4$.
- Find the conditional expectations $\mathbb{E}[X|Y = y_m]$ for $y_m = 1, 3, 4$.
- The quantity $\mathbb{E}[X|Y]$ can be considered as a random variable which is a function of Y , and written as a linear combination of the indicator functions of the sets $\{Y = y_m\} = Y^{-1}(y_m)$:

$$\mathbb{E}[X|Y] = \sum_m \mathbb{E}[X|Y = y_m] I_{Y^{-1}(y_m)} .$$

Use your result from part (d) to write $\mathbb{E}[X|Y]$ in this form (using concrete numbers).

- Use the representation of $\mathbb{E}[X|Y]$ from part (e) to find its expectation, $\mathbb{E}[\mathbb{E}[X|Y]]$.

Hint: Recall the linearity property of the expectation, and use that the expectation of the indication function of any $A \in \mathcal{F}$ is $\mathbb{E}(I_A) = 0 \cdot \mathbb{P}(A^c) + 1 \cdot \mathbb{P}(A) = \mathbb{P}(A)$. According to the so-called *tower rule*, $\mathbb{E}[\mathbb{E}[X|Y]] = \mathbb{E}[X]$, you should obtain in a different way the same value as in part (b).

Problem 3. A frog lays Y eggs, where $Y \sim \text{Poi}(\lambda)$ is a Poisson RV with parameter $\lambda > 0$, i.e., the p.m.f. of Y is $p_Y(n) = e^{-\lambda} \frac{\lambda^n}{n!}$, $n = 0, 1, 2, \dots$, where $0! := 1$.

Each egg survives independently of the survival of the other eggs, with probability $p \in (0, 1)$. From this one can derive that the number X of surviving eggs is a binomial RV with parameters Y and p , which can be written symbolically as $X \sim \text{Bin}(Y, p)$. This means that the RV X , conditioned on the event $\{Y = n\}$, is binomial with parameters n and p : $X | \{Y = n\} \sim \text{Bin}(n, p)$, i.e., that the conditional p.m.f. of X conditioned on the value of Y is

$$p_{X|Y}(k|n) = \binom{n}{k} p^k (1-p)^{n-k}, \quad k = 0, 1, \dots, n,$$

where $\binom{n}{k} := \frac{n!}{k!(n-k)!}$ are the binomial coefficients.

Show by a direct calculation of the p.m.f. of X that $X \sim \text{Poi}(\lambda p)$.

Hint: Use equation (3) derived in Food For Thought Problem 1 below. You may also find some of the following facts useful:

- series expansion of the exponential function: $e^t = \sum_{k=0}^{\infty} \frac{t^k}{k!}$, $t \in \mathbb{R}$;

- binomial formula: $(x + y)^n = \sum_{k=0}^n \binom{n}{k} x^k y^{n-k}$ for $x, y \in \mathbb{R}$, $n \in \{0, 1, 2, 3, \dots\}$;
- geometric series: $\frac{1}{1-q} = \sum_{k=0}^{\infty} q^k$ whenever $|q| < 1$.

Problem 4. Let X be a stationary discrete-time discrete-state space Markov chain with state space S consisting of two states, 0 and 1, and let the 1-step transition probability matrix of the stochastic process be

$$\mathbf{P} = \begin{bmatrix} p_{00} & p_{01} \\ p_{10} & p_{11} \end{bmatrix} = \begin{bmatrix} \frac{1}{3} & \frac{2}{3} \\ 1 & 0 \end{bmatrix}, \quad (1)$$

where $p_{ij} = \mathbb{P}(X_{n+1} = j | X_n = i)$. Let

$$\rho_{ij}^{(n)} := \mathbb{P}(X_n = j, X_{n-1} \neq j, \dots, X_1 \neq j | X_0 = i), \quad n \geq 1, \quad i, j \in \{0, 1\}$$

be the probability of moving to state j , from the initial state i , *for the first time* at the n th transition. Recall the relations

$$p_{ij}^{(n)} = \sum_{k=1}^n \rho_{ij}^{(k)} p_{jj}^{(n-k)}, \quad (2)$$

where $p_{ij}^{(n)}$ are the entries of the n -step transition probability matrix $\mathbf{P}^{(n)}$. Assume (quite reasonably) that $\mathbf{P}^{(0)} = \mathbf{I}$ (the identity matrix).

- Compute explicitly $\mathbf{P}^{(0)}$, $\mathbf{P}^{(1)}$, $\mathbf{P}^{(2)}$, and $\mathbf{P}^{(3)}$.
- Use (2) to compute the value of $\rho_{00}^{(1)}$.
- Use (2) and the value of $\rho_{00}^{(1)}$ (found in (b)) to compute the value of $\rho_{00}^{(2)}$.
- Use (2) and the values of $\rho_{00}^{(1)}$ and $\rho_{00}^{(2)}$ (found in (b) and (c)) to compute the value of $\rho_{00}^{(3)}$.

Food for Thought Problem 1. Let X and Y be discrete RVs taking values x_k and y_m , where the indices k and m run over discrete sets (finite or infinite).

Use the notation $\{Y = y_m\} = Y^{-1}(y_m) = \{\omega \in \Omega | Y(\omega) = y_m\}$ and the fact that the events $\{Y = y_m\}$ form a partition of the sample space Ω , to convince yourselves that the following derivation of the p.m.f. of the RV X through conditioning on the RV Y is correct:

$$\begin{aligned} p_X(x_k) &= \mathbb{P}(X = x_k) = \mathbb{P}(\{X = x_k\} \cap \Omega) = \mathbb{P}\left(\{X = x_k\} \cap \bigcup_m \{Y = y_m\}\right) \\ &= \mathbb{P}\left(\bigcup_m \{X = x_k, Y = y_m\}\right) = \sum_m \mathbb{P}(X = x_k, Y = y_m) \\ &= \sum_m \mathbb{P}(X = x_k | Y = y_m) \mathbb{P}(Y = y_m) = \sum_m p_{X|Y}(x_k | y_m) p_Y(y_m). \end{aligned} \quad (3)$$

Food for Thought Problem 2. Prove the recursive relation (2).

Solution: Let $i \in \{0, 1\}$, $j \in \{0, 1\}$, and $n \geq 1$ be fixed. Notice that the events

$$A_k := \{ \text{the MC reaches state } j \text{ for the first time in exactly } k \text{ steps} \} .$$

Clearly, for a given n , the time k when the MC reaches state j for the first time (starting from state i) can take values $1, 2, \dots, n$. It is obvious that the events A_k form a partition of the sample space Ω (why?):

$$\bigcup_{k=1}^n A_k = \Omega , \quad A_k \cap A_{k'} = \emptyset \text{ for } k \neq k' . \quad (4)$$

Note that

$$\rho_{ij}^{(k)} = \mathbb{P}(A_k \mid X_0 = i) . \quad (5)$$

Using (4) and (5), we obtain

$$\begin{aligned} p_{ij}^{(n)} &= \mathbb{P}(X_n = j \mid X_0 = i) \\ &= \mathbb{P}(\{X_n = j\} \cap \Omega \mid X_0 = i) \\ &= \mathbb{P}\left(\{X_n = j\} \cap \bigcup_{k=1}^n A_k \mid X_0 = i\right) \\ &= \mathbb{P}\left(\bigcup_{k=1}^n (\{X_n = j\} \cap A_k) \mid X_0 = i\right) \\ &= \sum_{k=1}^n \mathbb{P}(\{X_n = j\} \cap A_k \mid X_0 = i) \\ &= \sum_{k=1}^n \frac{\mathbb{P}(\{X_n = j\} \cap A_k \cap \{X_0 = i\})}{\mathbb{P}(X_0 = i)} \\ &= \sum_{k=1}^n \frac{\mathbb{P}(\{X_n = j\} \cap A_k \cap \{X_0 = i\})}{\mathbb{P}(A_k \cap \{X_0 = i\})} \frac{\mathbb{P}(A_k \cap \{X_0 = i\})}{\mathbb{P}(X_0 = i)} \\ &= \sum_{k=1}^n \mathbb{P}(X_n = j \mid A_k \cap \{X_0 = i\}) \mathbb{P}(A_k \mid X_0 = i) \\ &= \sum_{k=1}^n \mathbb{P}(X_n = j \mid A_k) \mathbb{P}(A_k \mid X_0 = i) \\ &= \sum_{k=1}^n \mathbb{P}(X_n = j \mid X_k = j) \mathbb{P}(A_k \mid X_0 = i) \\ &= \sum_{k=1}^n \mathbb{P}(X_{n-k} = j \mid X_0 = j) \mathbb{P}(A_k \mid X_0 = i) \\ &= \sum_{k=1}^n p_{jj}^{(n-k)} \rho_{ij}^{(k)} . \end{aligned}$$