- 1. The body temperatures of a healthy person and an infected person are Normal(36.8, 0.5) and Normal(37.8, 1.0) random variables, respectively. About 1% of the population is infected.
 - (a) What is the conditional probability that I am infected given that my temperature is t?
 - (b) For which values of t am I more likely to be infected than not?

Solution:

(a) Let A be the event that I am infected, and T be my body temperature. By the total probability theorem,

$$f_T(t) = \mathcal{P}(A)f_{T|A}(x) + \mathcal{P}(A^c)f_{T|A^c}(t)$$

where T|A is a Normal(37.8, 1.0) random variable and $T|A^c$ is a Normal(36.8, 0.5) random variable. The (unconditional) PDF of X is

$$f_T(t) = \frac{0.01}{\sqrt{2\pi}} e^{-\frac{(t-37.8)^2}{2}} + \frac{0.99}{\sqrt{2\pi}(0.5)} e^{-\frac{(t-36.8)^2}{2(0.5)^2}} = \frac{0.01}{\sqrt{2\pi}} e^{-(t-37.8)^2/2} + \frac{1.98}{\sqrt{2\pi}} e^{-2(t-36.8)^2}$$

By Bayes' rule, the conditional probability of A given T is

$$P(A|T=t) = \frac{P(A)f_{T|A}(t)}{f_T(t)} = \frac{0.01e^{-(t-37.8)^2/2}}{0.01e^{-(t-37.8)^2/2} + 1.98e^{-2(t-36.8)^2}}$$

(b) I am more likely to be infected than not when $P(A) > P(A^c)$, namely when

$$0.01e^{-(t-37.8)^2/2} > 1.98e^{-2(t-36.8)^2}.$$

Taking logarithms of both sides this is equivalent to a quadratic inequality in t. Solving this inequality, we obtain that $P(A) > P(A^c)$ holds when $t < t_-$ or $t > t_+$, where $t_- \approx 34.4742$ and $t_+ \approx 38.4591$.

- 2. A coin has probability P of being heads, where P itself is a Uniform(0, 1) random variable. The coin is flipped twice. Given that it comes out heads both times, what is the (posterior)
 - (a) PDF of P?
 - (b) expected value of P?
 - (c) probability that the next two flips are both heads?

Solution:

(a) Let X be the number of heads. By Bayes' rule,

$$f_{P|X}(p|x) = \frac{f_{X|P}(x|p)f_P(p)}{f_X(x)},$$

where $f_{P|X}(\cdot|x)$ is the conditional PDF of P given x heads were observed, $f_{X|P}(x|p) = \binom{2}{x}p^x(1-p)^{2-x}$ is the conditional p.m.f. of X given the coin has bias p, $f_P(p)$ is the

(prior) PDF of the Uniform(0, 1) random variable P, and f_X is the p.m.f. of X. Plugging in these formulas, we get that

$$f_{P|X}(p|2) = \frac{p^2 \cdot 1}{f_X(x)} = \frac{p^2}{f_X(x)}$$

when $0 \le p \le 1$. Since $f_{P|X}(p|2)$ is a PDF it must integrate to one, so

$$f_X(x) = \int_0^1 p^2 dp = \frac{1}{3}$$

and so $f_{P|X}(p|2) = 3p^2$.

(b) The expected value of P given X = 2 is

$$\mathbf{E}[P|X=2] = \int_0^1 p \cdot f_{P|X}(p|2)dp = \int_0^1 3p^3 dp = \frac{3}{4}.$$

(c) Let Y be the number of heads in the next two flips, By Bayes' rule,

$$f_{Y|X}(y|x) = \frac{f(x,y)}{f_X(x)} = \frac{\int_0^1 f_{XY|P}(x,y|p)f_P(p)dp}{f_X(x)}$$

where $f_{XY|P}(x, y|p) = {\binom{2}{x}} p^x (1-p)^{2-x} \cdot {\binom{2}{y}} p^y (1-p)^{2-y}$ is the conditional joint p.m.f. of X, Y given the coin has bias $p, f_P(p)$ is the (prior) PDF of the Uniform(0,1) random variable P, and $f_X(x) = 1/3$. Plugging in these formulas, we get that

$$f_{Y|X}(2|2) = \frac{\int_0^1 f_{XY|P}(2,2|p)f_P(p)dp}{1/3} = 3\int_0^1 p^4 dp = \frac{3}{5}$$

3. Raindrops hit your head at a rate of 1 per second. What is the PDF of the time at which the second raindrop hits you? How about the third one? (Hint: convolution)

Solution: The time before the second raindrop is $Y = X_1 + X_2$, where X_1 and X_2 are independent Exponential(1) random variables. We calculate the PDF of Y using the convolution formula:

$$f_Y(y) = \int_{-\infty}^{\infty} f_{X_1}(x_1) f_{X_2}(y - x_1) dx_1 = \int_0^y e^{-x_1} e^{-y + x_1} dx_1 = y e^{-y}.$$

The third raindrop hits at time $Z = Y + X_3$, where X_3 is another independent Exponential(1) random variable. By the convolution formula again,

$$f_Z(z) = \int_{-\infty}^{\infty} f_Y(y) f_{X_3}(z-y) dy = \int_0^z y e^{-y} e^{-z+y} dy = \frac{z^2}{2} e^{-z}.$$

- 4. In this question you will calculate the PDF of a *product* XY of two independent Uniform(0, 1) random variables X and Y.
 - (a) What is the PDF of $X' = \ln X$?
 - (b) What is the PDF of $Z = \ln X + \ln Y$?
 - (c) What is the PDF of $e^Z = XY$?

Solution:

(a) The PDF of X' is

$$f_{X'}(x') = \frac{d}{dx'} P(X' \le x') = \frac{d}{dx} P(X \le e^{x'}) = \frac{d}{dx'} e^{x'} = e^{x'}$$

when $x' \leq 0$ (that is, $0 < e^{x'} \leq 1$) and zero otherwise.

(b) Let $Y' = \ln Y$. By the convolution formula,

$$f_Z(z) = \int_{-\infty}^{\infty} f_{X'}(x') f_{Y'}(z - x') dx = \int_{z}^{0} e^{x'} e^{z - x'} dx' = -ze^{z}$$

when $z \leq 0$ and zero otherwise.

(c) The PDF of $Z' = e^Z$ is

$$f_{Z'}(z') = \frac{d}{dz'} P(Z' \le z') = \frac{d}{dz'} P(Z \le \ln z') = \frac{1}{z'} f_Z(\ln z') = -\ln z'$$

when $0 < z' \leq 1$ (that is, $\ln z' \leq 0$) and zero otherwise.