#### Problem 1 – Review

Let X and Y be two independent discrete random variables with joint probability mass function  $p_{X,Y}$ . For all  $x \in \mathbb{R}$  and  $y \in \mathbb{R}$ ,  $p_{X,Y}(x,y) = p_X(x)p_Y(y)$ .

We are interested in the probability mass function  $p_Z$  of the discrete random variable Z = X + Y. This is readily constructed as follows. For all  $z \in \mathbb{R}$ ,

$$p_Z(z) = P(Z = z) = \sum_x P(X = x, Y = z - x) = \sum_x p_{X,Y}(x, z - x) = \sum_x p_X(x)p_Y(z - x)$$

We thus found the following result:

$$p_Z(z) = \sum_x p_X(x)p_Y(z-x)$$

The probability distribution of the sum of two independent random variables is the convolution of their individual distributions.

#### Problem 2 – Extension of Problem 1 to continuous random variables

Let X and Y be two independent continuous random variables, with joint probability density function  $f_{X,Y}$ . Following our usual intuition from class, in which we convert formulas for discrete random variables into formulas for continuous random variables by replacing probability mass functions with probability density functions, and sums with integrals, we may guess that the probability density function of Z = X + Y is given by:

$$f_Z(z) = \int_{-\infty}^{+\infty} f_X(x) f_Y(z - x) dx$$

Let us now prove this formula rigorously.

The most natural method to prove this result is to use an object we have not seen in the lectures but which is quite intuitive: the *conditional probability density* function. Specifically, we will consider the conditional probability density function of Z given that X takes the value x:

$$f_{Z|X}(z|x) = f_Y(z-x)$$

Now, recalling the definition of a conditional probability, we may write, for all  $x \in \mathbb{R}$  and  $z \in \mathbb{R}$ :

$$f_{X,Z}(x,z) = f_{Z|X}(z|x)f_X(x) = f_X(x)f_Y(z-x)$$

Now, we are interested in the probability density function of Z, which is just the marginal probability density function for Z obtained from  $f_{X,Z}$ . From Lecture 7, we thus know that we can write:

$$f_Z(z) = \int_{-\infty}^{+\infty} f_X(x) f_Y(z - x) dx$$

This is indeed the result we had guessed. The probability density function of the sum of two independent random variables is the convolution of their individual probability density function.

Let  $T_1$  and  $T_2$  be two independent and identically distributed (i.i.d.) random variables, which have an exponential distribution with parameter  $\lambda$ .

- 1. Characterize the distribution of the random variable  $X = T_1 + T_2$ .
- 2. Let  $T_3$  be a third random variable with the same distribution as  $T_1$  and  $T_2$ , and independent from  $T_1$  and  $T_2$ . Let  $Y = X + T_3 = T_1 + T_2 + T_3$ . Characterize the distribution of Y.

Let  $X_1$  and  $X_2$  be two independent random variables that are uniformly distributed on [0, l]. Let  $M = \min(X_1, X_2)$  and  $T = \max(X_1, X_2)$ . What is the joint cumulative distribution function of M and T?

We consider a Poisson process, and the interval of time [0,t] for that Poisson process. We know that the Poisson process will have two arrivals in that interval  $(N_t = 2 \text{ in the notation of the notes for Lecture 9})$ , and we would like to know the distribution of  $X_1$  and  $X_2$ , corresponding to the time of arrival of the first and second arrival, respectively, in [0,t].

Cars cross a certain line on the highway in accordance with a Poisson process with rate  $\lambda=20$  per minute. If a boar attempts to cross the highway, what is the probability that it will survive if the amount of time that it takes it to cross the road is s seconds? (Assume that if it is on the highway when a car passes by, then it will die.)

Suppose that for Problem 6, we do not have a boar, but instead a fox, which is agile enough to escape from a single car. However, if it encounters two or more cars while attempting to cross the road, then it dies. What is the probability that it survives if it takes it s seconds to cross?