

On October 29, 2010, we discussed the following exercise in class.

Exercise. Based on studies from previous Hallowe'ens, I have determined that trick-or-treating ghouls arrive at my door as follows. The ghouls arrive independently and the time in minutes between the arrival of successive ghouls is exponentially distributed with parameter $\lambda = 2$. If I leave my jack-o-lantern lit for two hours (signifying that the ghouls are welcome to ring my bell) and my pumpkin contains 100 candy bars, do I expect to have enough candy bars to give one to every ghoul who trick-or-treats?

Our intuition told us that since the times in minutes between the arrivals of successive ghouls were independent and exponentially distributed with mean $1/\lambda = 1/2$, we should expect a ghoul every 30 seconds. Hence, in two hours we should expect 240 ghouls meaning that our box of 100 candy bars is not sufficient.

Let's prove that this intuition is correct. In fact, we can prove things in a bit more generality. Instead of assuming that the time in minutes between the arrival of successive ghouls is exponentially distributed with parameter $\lambda = 2$, let's assume that it is exponentially distributed with parameter $\lambda > 0$.

We now need to introduce some notation. Set $T_0 = 0$, and let T_k , $k = 1, 2, 3, \dots$, denote the time that the k th ghoul arrives. Note that the arrival times are *random variables*. We have assumed that the times in minutes *between* the arrivals are independent and exponentially distributed. That is, if

$$W_k = T_k - T_{k-1}$$

then W_k represents the waiting time between the arrival of ghoul $k - 1$ and ghoul k . Furthermore, W_1, W_2, \dots are independent and identically distributed with $W_k \sim \text{Exp}(\lambda)$. Also note that

$$\begin{aligned} T_k &= (T_0 - T_0) + (T_1 - T_0) + (T_2 - T_1) + \dots + (T_{k-1} - T_{k-2}) + T_k \\ &= (T_1 - T_0) + (T_2 - T_1) + \dots + (T_k - T_{k-1}) \\ &= W_1 + W_2 + \dots + W_k. \end{aligned}$$

In other words, T_k is the sum of k iid random variables each having an $\text{Exp}(\lambda)$ distribution. To find the distribution of T_k , we can use moment generating functions. That is,

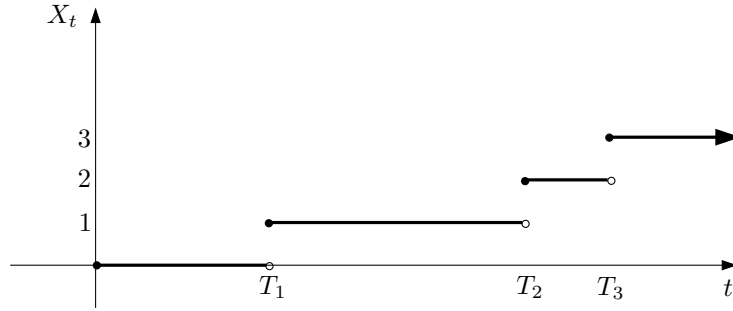
$$\begin{aligned} m_{T_k}(t) &= \mathbb{E}[e^{tT_k}] = \mathbb{E}[e^{t(W_1 + \dots + W_k)}] = \mathbb{E}[e^{tW_1} \dots e^{tW_k}] = \mathbb{E}[e^{tW_1}] \dots \mathbb{E}[e^{tW_k}] = (\mathbb{E}[e^{tW_1}])^k \\ &= \left(\frac{\lambda}{\lambda - t} \right)^k \end{aligned}$$

which just so happens to be the moment generating function of a random variable with a $\text{Gamma}(k, \lambda)$ distribution. Thus,

$$T_k \sim \text{Gamma}(k, \lambda)$$

so that

$$\mathbf{P}\{T_k > t\} = \int_t^\infty \frac{\lambda^k}{\Gamma(k)} x^{k-1} e^{-\lambda x} dx. \quad (*)$$



Now, here is the next observation. Let X_t denote the number of ghouls that have arrived by time t . Note that X_t is a discrete random variable since the number of ghouls who have arrived is necessarily a non-negative integer. Observe that at most k ghouls have arrived at time t if and only if ghouls $k + 1$ arrived after time t ; that is, $\{X_t \leq k\} = \{T_{k+1} > t\}$ and so

$$\mathbf{P}\{X_t \leq k\} = \mathbf{P}\{T_{k+1} > t\}. \quad (**)$$

Our goal is to compute

$$\mathbb{E}(X_t) = \sum_{k=0}^{\infty} k \mathbf{P}\{X_t = k\} = \sum_{k=1}^{\infty} k \mathbf{P}\{X_t = k\}$$

but we do not yet have an expression for $\mathbf{P}\{X_t = k\}$. However, we do have an expression for $\mathbf{P}\{X_t \leq k\}$. Observe that

$$\mathbf{P}\{X_t = k\} = \mathbf{P}\{X_t \leq k\} - \mathbf{P}\{X_t \leq k - 1\}$$

and so using (*) and (**) we conclude

$$\begin{aligned} \mathbf{P}\{X_t = k\} &= \mathbf{P}\{T_{k+1} > t\} - \mathbf{P}\{T_k > t\} \\ &= \int_t^{\infty} \frac{\lambda^{k+1}}{\Gamma(k+1)} x^k e^{-\lambda x} dx - \int_t^{\infty} \frac{\lambda^k}{\Gamma(k)} x^{k-1} e^{-\lambda x} dx. \end{aligned} \quad (\dagger)$$

The next step is to analyze the difference of integrals of gamma functions. The “usual” way to analyze integrals of gamma function is to try integration by parts. Hence, consider

$$\int_t^{\infty} \lambda x^k e^{-\lambda x} dx.$$

If we let $u = x^k$ and $dv = \lambda e^{-\lambda x} dx$, then

$$\int_t^{\infty} \lambda x^k e^{-\lambda x} dx = x^k e^{-\lambda x} \Big|_t^{\infty} + \int_t^{\infty} k x^{k-1} e^{-\lambda x} dx = t^k e^{-\lambda t} + k \int_t^{\infty} x^{k-1} e^{-\lambda x} dx.$$

Multiplying both sides by $\lambda^k / \Gamma(k + 1)$ implies

$$\frac{\lambda^k}{\Gamma(k+1)} \int_t^{\infty} \lambda x^k e^{-\lambda x} dx = \frac{\lambda^k}{\Gamma(k+1)} t^k e^{-\lambda t} + \frac{\lambda^k}{\Gamma(k+1)} k \int_t^{\infty} x^{k-1} e^{-\lambda x} dx$$

and so

$$\int_t^\infty \frac{\lambda^{k+1}}{\Gamma(k+1)} x^k e^{-\lambda x} dx = \frac{(\lambda t)^k}{\Gamma(k+1)} e^{-\lambda t} + \int_t^\infty \frac{\lambda^k}{\Gamma(k)} x^{k-1} e^{-\lambda x} dx$$

using the fact that $\Gamma(k+1) = k\Gamma(k)$ which in turn implies that

$$\int_t^\infty \frac{\lambda^{k+1}}{\Gamma(k+1)} x^k e^{-\lambda x} dx - \int_t^\infty \frac{\lambda^k}{\Gamma(k)} x^{k-1} e^{-\lambda x} dx = \frac{(\lambda t)^k}{\Gamma(k+1)} e^{-\lambda t}.$$

In other words, we have found our required expression for the particular difference of integrals of gamma functions in (†). Thus,

$$\mathbf{P}\{X_t = k\} = \frac{(\lambda t)^k}{\Gamma(k+1)} e^{-\lambda t} = \frac{(\lambda t)^k}{k!} e^{-\lambda t}$$

since $\Gamma(k+1) = k!$ when k is a positive integer.

Finally, we conclude

$$\begin{aligned} \mathbb{E}(X_t) &= \sum_{k=1}^{\infty} k \mathbf{P}\{X_t = k\} = \sum_{k=1}^{\infty} k \frac{(\lambda t)^k}{k!} e^{-\lambda t} = e^{-\lambda t} \sum_{k=1}^{\infty} \frac{(\lambda t)^k}{(k-1)!} = e^{-\lambda t} \sum_{k=1}^{\infty} \frac{(\lambda t)(\lambda t)^{k-1}}{(k-1)!} \\ &= \lambda t e^{-\lambda t} \sum_{k=1}^{\infty} \frac{(\lambda t)^{k-1}}{(k-1)!} \\ &= \lambda t e^{-\lambda t} \sum_{k=0}^{\infty} \frac{(\lambda t)^k}{k!} \\ &= \lambda t e^{-\lambda t} e^{\lambda t} \\ &= \lambda t. \end{aligned}$$

Thus, if we return to the ghoulish exercise that started this whole mess and take $t = 120$ minutes and $\lambda = 2$, then

$$\mathbb{E}(X_{120}) = 2 \cdot 120 = 240$$

as suspected!