

Chapter 9 – Waiting Time Random Variables

Here, we look at two discrete distributions (Geometric and Poisson) and one continuous distribution (Exponential) in greater depth.

Geometric RV's – In Chapter 4 (p.126 ff) we looked at one variant – here, we look at another one. Here, T = *number of trials until the first success*; previously, X = number of *failures* until the first success. The pmf for T is

$$f_T(t) = \pi(1 - \pi)^{t-1}, t = 1, 2, 3 \dots,$$

where π is the ‘success’ probability (for a single trial); $f_T(t) = 0$ for all other values of t . It follows that the cdf of T is $F(t) = \Pr\{T \leq t\} = 1 - (1 - \pi)^t$, and the *survival function* of T is $S(t) = \Pr\{T > t\} = (1 - \pi)^t$. We use this latter function to show that any Geometric RV exhibits the *memory-less property*: $\Pr\{T > s + t \mid T > s\} = \Pr\{T > t\}$. The result is actually ‘IFF’ in the sense that in addition if a discrete RV has the memory-less property, then it must be a Geometric RV.

For T a Geometric RV, it is easy to show (pp. 244-5) that $E(T) = 1/\pi$ and $\text{Var}(T) = \frac{1-\pi}{\pi^2}$. The proofs use the infinite sum (and other derived infinite sums) we gave at the end of the Chapter 4 Notes.

Exponential RV's – In Chapter 7, we defined T as having the Exponential (θ) distribution (written $T \sim \mathcal{E}(\theta)$) provided its pdf is

$$f(t) = \frac{1}{\theta} e^{-t/\theta} \text{ for } t > 0$$

The unknown parameter is $\theta > 0$. We also showed there that the cdf is $F(t) = \Pr\{T \leq t\} = 1 - e^{-t/\theta}$, for $t > 0$. Hence, the survival function is $S(t) = \Pr\{T > t\} = e^{-t/\theta}$ for $t > 0$. It is simple then to show that the Exponential distribution also follows the memory-less property – it is the only continuous RV with this property (proof on p.248). We also showed previously that $E(T) = \theta$ and $\text{var}(T) = \theta^2$.

Poisson RV's – The discrete RV Y has the $\text{Poisson}(\lambda)$ distribution if it has the pmf

$$f_Y(y) = \frac{\lambda^y}{y!} e^{-\lambda}, y = 0, 1, 2, 3 \dots,$$

and we write $Y \sim P(\lambda)$. In order to prove that this pmf sums to one, we need to remind ourselves of the result from Calculus: $e^z = \sum_{k=0}^{\infty} \frac{z^k}{k!}$. We also use this result to show that $E(Y) = \lambda$ and $E(Y^2) = \lambda^2 + \lambda$; hence, $\text{var}(Y) = \lambda$. It's kind of interesting that this distribution has just one parameter (λ), and it is both the mean and the variance.

In the last chapter, we used the Normal distribution to approximate the Binomial distribution, but another approximation of the Binomial distribution can be obtained from the Poisson distribution – provided again that n is large and π is small. Equating means, we set $\lambda = n\pi$. A general rule (p.255) says that the approximation is *excellent* for $n \geq 100$ and $\lambda = n\pi \leq 5$, and is *very good* for $n \geq 100$ and $\lambda = n\pi < 10$.

Example 9.5 (p.256) – First Aid. There are $n = 2000$ spectators and the chance that any one individual needs first aid is $\pi = 0.001$ (one in a thousand). Let's use the Poisson distribution to approximate the probability that at *least 3* spectators need first aid treatment. Here, $\lambda = n\pi = 2$, so the Poisson approximation should be excellent. We want $\Pr\{K \geq 3\} = 1 - \Pr\{K \leq 2\} \approx 1 - \Pr\{Y \leq 2\}$ (with $Y \sim P(\lambda = 2)$) = $1 - \{e^{-2} + 2e^{-2} + 2e^{-2}\} = 1 - 5e^{-2} = 0.3233$ or 32.33%. See p.256 to see how close this approximation is to the true value.

On p.257, two recursion relations are given as is a theorem justifying the Poisson approximation to the Binomial distribution. The proof of this result is based on the important limit from basic Calculus:

$$\lim_{n \rightarrow \infty} \left(1 + \frac{f(x)}{n} \right)^n = e^{f(x)}$$

Also, the Exponential and Poisson distributions are connected via the following (p.260):

Theorem 9.8 – Let T be an exponential random variable of waiting times for an event with mean waiting time of θ units. Let Y be the count of events that occur in a unit of time, $[0, 1]$. Then, Y has a Poisson distribution with mean $\lambda = 1/\theta$.

A nice application is given in Example 9.6.