

2. The Differential Argument and the Gamma Distribution (Blitzstein 5.5, 5.6, 13)

The differential approach to finding a pdf $f_x(x)$ is summarized as follows:

✓
for a continuous variable

This problem continues from the first problem as we move from discrete to continuous variables. For any given discrete variable, the pmf provides the exact probability of X being equal to some number. However, for a continuous variable, the pmf only provides the probability of X being less than or greater than a certain value. However, as we will see through this problem, if we now add a width to a probabilities, of width dx , we can find the exact probability.

Function

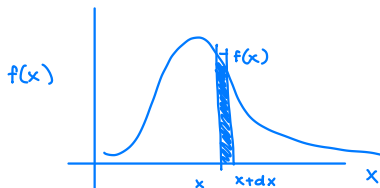
We will see throughout this problem how a discrete variable, like a poisson distribution, which is a process in time with discrete counts, becomes a continuous variable, such as Exponential or Gamma, when we examine the waiting period for the next event.

a) For a continuous random variable X with pdf f_x , explain why, for a small deviation dx ,

$$P(X \in [x, x+dx]) \approx f_x(x)dx$$

Show how this becomes an exact equivalence when you divide both sides by dx and let dx approach 0 (or if the density $f_x(x)$ is a constant function). As an example, derive the Exponential density for X by finding the probability the first Poisson event occurs in the interval $[x, x+dx]$. Then divide by dx and take the limit $dx \rightarrow 0$. Show this agrees with the result you get by finding the Exponential cdf and differentiating.

Let's start with a general function $f(x)$. If we want to find the pdf at x , we instead want to find the pdf at x to the very small increment of dx . So, we find the area of the rectangle of the pdf evaluated at x to the pdf evaluated at $(x+dx)$.



We can see from the graph that $f(x)$ represents the height and dx represents the width of this rectangle.

Therefore, $f(x)dx \approx P(X \in [x, x+dx])$

In order to find this area, we can evaluate the difference in the integrals.

$$f(x)dx \approx F_x(x+dx) - F_x(x)$$

Next, we divide both sides by dx and take the limit as dx approaches 0.

$$f(x) = \lim_{dx \rightarrow 0} \frac{F_x(x+dx) - F_x(x)}{dx} = \frac{d}{dx} F_x(x)$$

We see that this is equivalent to the set up of a derivative and thus the differential approach to find the pmf.

Now, as an example, derive the Exponential density by finding the probability the first poisson event occurs in the interval $[x, x+dx]$.

Count of events in time $x \sim \text{Poisson}(\lambda x)$ $\lambda = \text{events per unit rate}$
 $x = \text{time units}$

$$P(X \in [x, x+dx]) = P(\text{first event} \text{ at } x) = P(0 \text{ events by time } x) \cdot P(1 \text{ event in time } [x, x+dx])$$

These are poisson and independent so we can multiply the individual pmf's.

$$\begin{aligned} P(X \in [x, x+dx]) &= \frac{(\lambda x)^0 e^{-\lambda x}}{0!} \cdot \frac{(\lambda dx)^1 e^{-\lambda dx}}{1!} \\ &= \lambda e^{-\lambda(x+dx)} \cdot dx \quad \text{*note: } e^{-\lambda dx} \rightarrow 1 \text{ as } dx \rightarrow 0 \text{ so we can remove it from the equation in the next step} \\ f(x)dx &\approx \lambda e^{-\lambda x} dx \end{aligned}$$

divide both sides by dx

$$\rightarrow f(x) = \lambda e^{-\lambda x}, \quad x \geq 0 \quad (\text{This evaluates to the exponential pmf})$$

show this agrees with the result you get by finding the exponential cdf and differentiating.

$$\text{CDF method: } 1 - F_x(x) = P(X > x) = P(0 \text{ events by time } x) = e^{-\lambda x}$$

$$F_x(x) = 1 - e^{-\lambda x}$$

$$f(x) = \lambda e^{-\lambda x} \quad (\text{same result as above}).$$

- b) Use a differential argument to derive the $\text{Gamma}(k, \lambda)$ density. Let Y be the time of the k th event for a Poisson process with rate λ . Find the probability of $k-1$ events occurring before time y and a k th event occurring in the interval $[y, y+dy)$. Divide by dy and take the limit as $dy \rightarrow 0$. Give intuition for why, if $X_1 \sim \text{Gamma}(k_1, \lambda)$ is independent of $X_2 \sim \text{Gamma}(k_2, \lambda)$, then $X_1 + X_2 \sim \text{Gamma}(k_1 + k_2, \lambda)$, and for why the sum is not Gamma if the λ 's differ.

We want to find the probability of $k-1$ events occurring before time y and a k th event occurring in the interval $[y, y+dy)$

These individual and independent events are once again poisson.

$$P(k-1 \text{ events by time } y) = \frac{e^{-\lambda y} (\lambda y)^{k-1}}{(k-1)!} \sim \text{Poisson}(\lambda y)$$

$$P(k\text{th event in } [y, y+dy)) = \lambda dy \sim \text{Poisson}(\lambda dy) \quad (\text{given that there are } k-1 \text{ events by time } y)$$

The total probability of having $k-1$ events by time y and the k th event in $[y, y+dy)$ is

$$P(X \in [y, y+dy)) = \frac{(\lambda y)^{k-1} e^{-\lambda y}}{(k-1)!} \cdot \lambda dy = \frac{\lambda^k y^{k-1} e^{-\lambda y}}{(k-1)!} dy \quad (\text{combine like terms})$$

divide by dy to get the pdf

$$\frac{P(X \in [y, y+dy))}{dy} = \frac{(\lambda y)^{k-1} e^{-\lambda y}}{(k-1)!} \cdot \lambda$$

Take the limit as $dy \rightarrow 0$. This evaluates to the pdf of the Gamma distribution.

$$f_y(y) = \frac{\lambda^k y^{k-1} e^{-\lambda y}}{(k-1)!}, \quad y \geq 0$$

Lastly, give the intuition for why, if $X_1 \sim \text{Gamma}(k_1, \lambda)$ is independent of $X_2 \sim \text{Gamma}(k_2, \lambda)$, then $X_1 + X_2 \sim \text{Gamma}(k_1 + k_2, \lambda)$ and for why the sum is not Gamma if the λ 's differ.

$X_1 + X_2 \rightarrow k_1 + k_2 = \text{total number of events}$ (wait for k_1 events and then wait for k_2)

$X_1 + X_2 \rightarrow \lambda$ remains the rate of events

if λ_1 and λ_2 were to differ, then they would not be in the same poisson process.

c) For a constant $c > 0$, show that if $X \sim \text{Gamma}(\alpha, \lambda)$, then $Y = cX \sim \text{Gamma}(\alpha, \lambda/c)$. Give intuition based on units of measure. For example, consider X to be a waiting time in hours and $Y = 60X$ to be the time in minutes.

For a constant $c > 0$, show that if $X \sim \text{Gamma}(\alpha, \lambda)$, then $Y = cX \sim \text{Gamma}(\alpha, \lambda/c)$

$$X \sim \text{Gamma}(\alpha, \lambda) \quad Y = cX \quad c > 0$$

$$X = Y/c$$

$$F_Y(y) = P(Y \leq y) = P(cX \leq y)$$

$$= P(X \leq y/c) = F_X(y/c)$$

$$f_Y(y) = \underbrace{\frac{1}{c} f_X(y/c)}_{\text{chain rule}} = \frac{\lambda^\alpha}{\Gamma(\alpha)} (y/c)^{\alpha-1} e^{-\lambda y/c} \cdot \frac{1}{c} \quad \text{* plug into gamma pdf}$$

$$= \frac{(\lambda/c)^\alpha}{\Gamma(\alpha)} \cdot y^{\alpha-1} e^{-(\lambda/c)y}, \quad y > 0$$

This confirms $Y = cX \sim \text{Gamma}(\alpha, \lambda/c)$

We can now look at an example of this.

Consider X to be a waiting time in hours and $Y = 60X$ to be time in minutes.

Let's look at the set up.

$$\begin{aligned} X &= \text{time in hours} & \lambda &= \text{events per hour} \\ Y &= \text{time in minutes} & = 60X &\rightarrow \text{events per minute} = \frac{\lambda}{60} \end{aligned}$$

Thus, through the change of variables, we see how this now becomes λ/c or in this case, $\lambda/60$

d) Define the Gamma function and show how it normalizes the $\text{Gamma}(\alpha, \lambda)$ density for any value $\alpha > 0$. Give the recursive property of the Gamma function and the connection to the factorial function.

We have previously defined the Gamma function for an Integer K .

$Y \sim \text{Gamma}(K, \lambda) \rightarrow Y$ is the time until K^{th} event

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

For an integer K , we believe this pdf is correct because of the story in the previous problems: $K-1$ events before time x , 1 event in dx , $n-K$ events after $x+dx$

But what about a non integer K , which I will call α ?

non-integer α in general: $\frac{\lambda^\alpha}{\Gamma(\alpha)} y^{\alpha-1} e^{-\lambda y}, \quad y > 0 \quad \alpha > 0 \text{ (real number)}$

What we will see is that the Gamma function will normalize the Gamma density.

Example: Let $\lambda = 1$

$$f(x) = \frac{x^{\alpha-1} e^{-x}}{\Gamma(\alpha)}, \quad x > 0$$

for an integer K , we used $\Gamma(K) = (K-1)!$ $K = 1, 2, \dots$

for non integer α , we can use $\Gamma(\alpha) = \int_0^\infty x^{\alpha-1} e^{-x} dx$

$$\text{This normalizes the probability as } \int_0^\infty f(x) dx = \int_0^\infty \frac{x^{\alpha-1} e^{-x} dx}{\Gamma(\alpha)} = \frac{\int_0^\infty x^{\alpha-1} e^{-x} dx}{\int_0^\infty x^{\alpha-1} e^{-x} dx} = 1$$

using a change of variables, we see the normalisation regardless of the λ value

$$y = x/\lambda \quad x = \lambda y \quad Y \sim \text{Gamma}(\alpha, \lambda)$$

$$f_Y(y) = \frac{\lambda^\alpha}{\Gamma(\alpha)} y^{\alpha-1} e^{-\lambda y}, \quad y > 0$$

Give the recursive property of the gamma function.

$$\text{Recursive property: } \Gamma(\alpha+1) = \alpha \Gamma(\alpha)$$

$$\text{Broadly: } \Gamma(\alpha+1) = \int_0^\infty x^\alpha e^{-x} dx = \underbrace{\alpha \int_0^\infty x^{\alpha-1} e^{-x} dx}_{\text{integration by parts}} = \alpha \Gamma(\alpha)$$

connection to the factorial function:

recursive: $\Gamma(\alpha+1) = \alpha \Gamma(\alpha) \rightarrow$ we can think of this like a factorial $\rightarrow (x+1)! = (x+1) \cdot x!$

$$\Gamma(1) = \int_0^\infty x^{1-1} e^{-x} dx = 1 = 0!$$

\rightarrow rewrite using recursive property

$$\Gamma(2) = 1 \cdot \Gamma(1) = 1 = 1!$$

$$\Gamma(3) = 2 \cdot \Gamma(2) = 2 = 2!$$

$$\Gamma(4) = 3 \Gamma(3) = 3 \cdot 2 = 6 = 3!$$

Thus, we can see that:

$$\Gamma(n+1) = n!$$

Therefore, the recursive property is equivalent to the factorial function.