

Andrew Gard - equitable.equations@gmail.com



The Gamma Distribution

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$$\begin{aligned} F(x) &= P(X \leq x) = 1 - P(\text{fewer than } \alpha \text{ occurrences before time } x) \\ &= 1 - \sum_{k=0}^{\alpha-1} \frac{x^k e^{-x/\theta}}{\theta^k k!}, \quad x \geq 0 \end{aligned}$$

The expression inside the summation represents the probability of exactly k occurrences in time x in the Poisson distribution with parameter θ .



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the version usually cited in books.



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Since θ is the average time between occurrences, λ represents the average number of occurrences per unit time. For this reason, it's referred to as the *rate* parameter.



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Each of these facts can be proved using the moment-generating function,

$$M(t) = \frac{1}{(1 - \theta t)^\alpha}.$$



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$$\begin{aligned} P(X \leq 10) &= \int_0^{10} \frac{x^{3-1} e^{-x/4}}{(3-1)!4^3} dx = \int_0^{10} \frac{x^2 e^{-x/4}}{128} dx \\ &\approx 0.456 \end{aligned}$$



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$$\begin{aligned}\mu &= \theta\alpha \\ &= (4)(3) \\ &= 12 \text{ minutes.}\end{aligned}$$



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where λ is the average number of occurrences and x is the observed number of occurrences .



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$$P(X = 15) = \frac{15^{15} e^{-15}}{15!} \approx 0.102$$



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- When $\alpha = 1$, we have the *exponential distribution*, which models the waiting time between occurrences in a Poisson process. The pdf of that distribution is

$$f(x) = \frac{1}{\theta} e^{-x/\theta}, \quad x \geq 0.$$



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- When $\alpha = 1$, we have the *exponential distribution*, which models the waiting time between occurrences in a Poisson process. The pdf of that distribution is

$$f(x) = \frac{1}{\theta} e^{-x/\theta}, \quad x \geq 0.$$

- When instead $\theta = 2$ and $\alpha = r/2$, where r is a positive integer, we have the *χ -squared distribution with r degrees of freedom*, which models the distribution of the sum of the squares of r random variables, each with the standard normal distribution. The pdf of that distribution is

$$f(x) = \frac{x^{r/2-1} e^{-x/2}}{\Gamma(r/2) 2^{r/2}}, \quad x \geq 0.$$

