

# The Exponential and Gamma Distributions

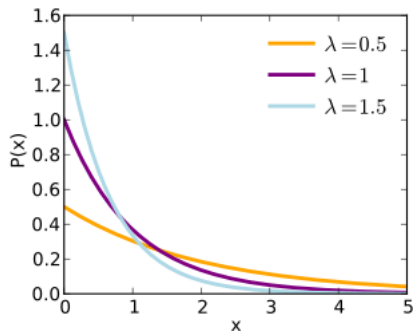
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# The Exponential Distribution

$$f(x|\beta) = \frac{1}{\beta} e^{(-x/\beta)}, 0 \leq x < \infty, \beta > 0$$

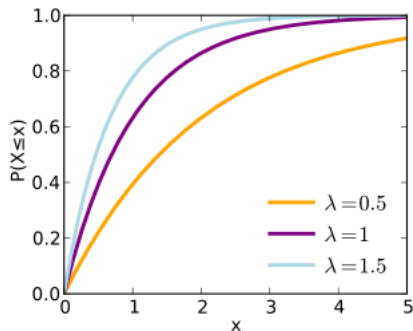
The parameter  $\beta$  is a scale parameter.



Note:  $\lambda = \frac{1}{\beta}$

# CDF

$$F(x) = 1 - e^{(-x/\beta)}, 0 \leq x < \infty$$



Note:  $\lambda = \frac{1}{\beta}$

# Exponential Mean, Variance, and MGF

$$E(X) = \beta$$

$$\text{Var}(X) = \beta^2$$

$$M_x(t) = \frac{1}{1 - \beta t}, t < \frac{1}{\beta}$$

# Relationship to Other Distributions

The Exponential Distribution is related to the Uniform, Gamma and Weibull Distributions.

If  $X \sim \text{Uniform}$  and  $Y = -\log(X)$ , then  $Y \sim \text{Exp}$ .

If  $X \sim \text{Gamma}(\alpha = 1, \beta)$ , then  $X \sim \text{Exp}(\beta)$ .

If  $X \sim \text{Exp}$  and  $Y = X^{(1/\gamma)}$ , then  $Y \sim \text{Weibull}$ .

The Exponential Distribution is the continuous form of the Geometric Distribution.

# Example of Exponential Distribution

If the average waiting time at a doctor's office is  $\beta = 20$  minutes, what is the probability that you will have to wait less than 10 minutes?

$$P(X \leq 10) = \int_0^{10} f(x)dx = 0.3934693$$

# More about the Exponential Distribution

The exponential distribution can also be used to model time between Poisson events and lifetimes.

An interesting characteristic is that the Exponential distribution is memoryless.

$$P(X > s | X > t) = P(X > s - t)$$

For example: Given you have waited 20 minutes, the probability you'll wait 30 minutes is the same as the probability you'll wait 10 minutes.

# Gamma Function:

The Gamma Function is defined as:

$$\Gamma(\alpha) = \int_0^{\infty} t^{\alpha-1} e^{-t} dt,$$

where  $\alpha > 0$ .



# Useful Identities:

Properties of the Gamma Function:

$$\Gamma(\alpha + 1) = \alpha\Gamma(\alpha), \quad \alpha > 0$$

$$\Gamma(n) = (n - 1)!, \quad n \in \mathbb{N}$$

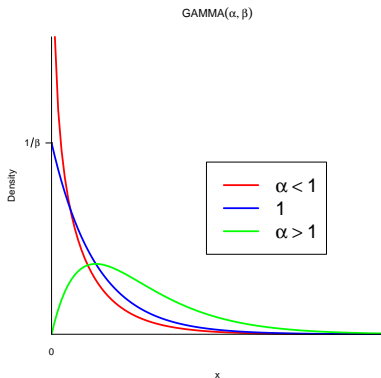
$$\Gamma(1) = 1$$

$$\Gamma\left(\frac{1}{2}\right) = \sqrt{\pi}$$

# Gamma Distribution:

If  $X \sim \text{Gamma}(\alpha, \beta)$ , then the pdf of  $X$  is:

$$f(x) = \frac{1}{\Gamma(\alpha)\beta^\alpha} x^{\alpha-1} e^{-x/\beta}$$



# Gamma Distribution:

And the CDF is:

$$\frac{\gamma(k, x/\theta)}{\Gamma(k)},$$

where  $k > 0$ , and

$$\gamma(k, x/\theta) = \int_0^{x/\theta} t^{k-1} e^{-t} dt$$

# Gamma Mean, Variance, and Moment Generating Function:

Mean:

$$E[X] = \alpha\beta$$

Variance:

$$\text{Var}[X] = \alpha\beta^2$$

MGF:

$$M_X(t) = \left( \frac{1}{1 - \beta t} \right)^\alpha$$

# Relation to Other Distributions:

**Poisson-Gamma.** The Gamma distribution is a conjugate prior for the mean of the Poisson distribution.

**Exponential-Inverse Gamma.** The Inverse Gamma distribution is a conjugate prior for the mean of the exponential distribution.

**Exponential.**  $\text{Gamma}(1, \beta) = \text{EXP}(\beta)$ .

**Chi-Squared.**  $\text{Gamma}(p/2, 2) = \chi^2(p)$ .

**Maxwell.** Let  $X$  be distributed as a Gamma. If  $\alpha = 3/2$ , then  $Y = \sqrt{X/\beta}$  is distributed as a Maxwell.

**Inverse Gamma.** Let  $X \sim \text{Gamma}(\alpha, \beta)$ . Then  $Y = 1/X \sim \text{InverseGamma}(\alpha, \beta)$ .

# Relation to Other Distributions:

**More on Poisson-Gamma.** The Negative Binomial can be derived as a gamma mixture of Poissons.

Specifically, if

$$X \sim NB(r, \beta)$$

and

$$X|\lambda \sim POI(\lambda),$$

then

$$\lambda \sim Gamma(\alpha, \beta).$$

# Applications of the Gamma Distribution:

The Gamma distribution is positive and right-skewed, so it is can be used to model a variety of events, including the following:

- The amount of rainfall accumulated in a reservoir
- The size of loan defaults of aggregate insurance claims
- The flow of items through manufacturing and distribution processes
- The load on web servers
- The many and varied forms of telecom exchange

## Very Hypothetical Situation:

Suppose that for a graduate Statistical Computing course, the average time required for every student to come up with a feasible solution for each exam problem is two hours. Compute the probability that a student, say Mickey, will take 6 or more hours to come up with the solutions to all four problems of the exam.

One problem every 2 hours means we would expect to "solve"  $\beta = 1/2$  exam problems every hour on average. Using  $\beta = 1/2$  and  $\alpha = 16$ , we can compute this as follows:

$$P(X \geq 8) = \int_8^{\infty} \frac{x^{4-1} e^{-x/.5}}{\Gamma(4)2^4} dx = .466745$$

Hypothetical Inference???