

# STAT 315: Exponential and Gamma Random Variables

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# Exponential Random Variables

## Exponential Random Variables

A random variable  $Y$  is an **exponential random variable** with parameters  $\beta$  and we write  $Y \sim \text{Exp}(\beta)$  if the **PDF** is

$$f(y) = \frac{1}{\beta} e^{-\frac{y}{\beta}} \quad \beta > 0 \text{ scale parameter}$$

The **CDF** is given by

$$F(y) = \int_0^y \frac{1}{\beta} e^{-\frac{t}{\beta}} dt = 1 - e^{-\frac{y}{\beta}}$$

**Warning:** Very often  $\lambda = 1/\beta$  is used a parameter with

$$f(y) = \lambda e^{-\lambda y} \quad \lambda > 0 \text{ rate parameter}$$

# PDF and CDF

$$\text{PDF: } f(t) = \frac{1}{\beta} e^{-\frac{y}{\beta}}$$

$$\text{CDF: } F(t) = 1 - e^{-\frac{y}{\beta}}$$

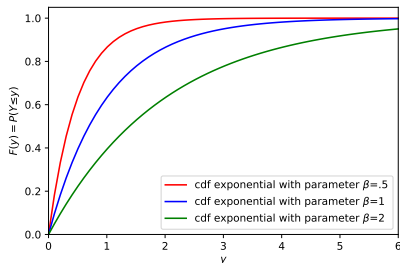
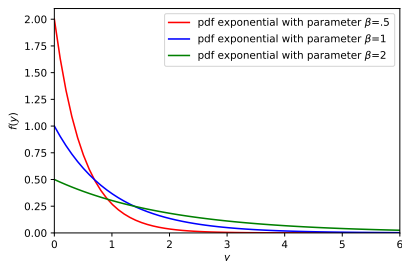


Figure: The Exponential RV with Left:PDF and Right:CDF

# Waiting times and the memoryless property

Often  $Y$  has the interpretation of a **waiting time**

- $Y$  is the time between two consecutive earthquake in California
- $Y$  is the time between the emission of two radioactive particles.
- The time it takes to be served at a cash register at a supermarket.
- ...

## Memoryless property

For an **exponential random variable**  $Y$

$$P(Y \geq t + s | Y \geq t) = P(Y \geq s)$$

"If you have waited for at least 1 hours, the probability you have to wait another 30 minutes is the same as if you just had arrived...."

# Mean and Variance of Exponential RV

## Moments of exponential RV

For a exponential random variable  $Y$  with parameter and  $\beta$  we have

$$E[Y] = \beta \quad V(Y) = \beta^2$$

**Proof:** Integration by parts. See the more general computation later.

# Gamma Random Variable

## Gamma random variables

A gamma random variables  $Y$  with parameters  $\alpha$  (shape parameter) and  $\beta$  (scale parameter) (we write  $Y \sim \Gamma(\alpha, \beta)$ ) has the pdf

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

## Gamma function

The gamma function  $\Gamma(\alpha)$  is given by

$$\Gamma(\alpha) = \int_0^{\infty} y^{\alpha-1} e^{-y} dy$$

It satisfies  $\Gamma(\alpha + 1) = \alpha\Gamma(\alpha)$  and  $\Gamma(n) = (n - 1)!$  and  $\Gamma(1/2) = \sqrt{\pi}$

- Integration by parts with  $u = y^\alpha$  and  $v' = e^{-y}$

$$\Gamma(\alpha + 1) = \int_0^\infty y^\alpha e^{-y} dy = -y^\alpha e^{-y} \Big|_{-\infty}^\infty + \alpha \int_0^\infty y^{\alpha-1} e^{-y} dy = \alpha \Gamma(\alpha)$$

- We have  $\Gamma(1) = \int_0^\infty e^{-y} dy = 1$  and so

$$\Gamma(n) = (n-1)\Gamma(n-2) = (n-1)(n-2)\Gamma(n-3) = \dots = (n-1)!$$

- Using the change of variable  $y = \frac{x^2}{2}$ ,  $dy = x dx$

$$\begin{aligned} \Gamma\left(\frac{1}{2}\right) &= \int_0^\infty y^{-\frac{1}{2}} e^{-y} dy = \int_0^\infty \frac{\sqrt{2}}{x} e^{-\frac{x^2}{2}} x dx = \sqrt{2} \int_0^\infty e^{-\frac{x^2}{2}} dx \\ &= \sqrt{2} \frac{1}{2} \sqrt{2\pi} \underbrace{\int_{-\infty}^\infty \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx}_{=1} = \sqrt{\pi} \end{aligned}$$

Typical examples where Gamma random variables are used to model non-negative quantities such as for example **time until death** or **time between successive insurance claims**).

We can use the two parameters  $\alpha$  and  $\beta$  to adjust the mean and variance

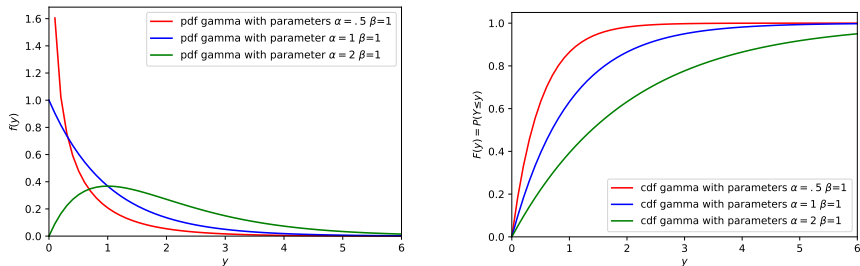


Figure: The gamma RV with Left:PDF and Right:CDF



# Mean and Variance of Gamma RV

## Moments of Gamma RV

For a **gamma random variable**  $Y$  with parameters  $\alpha$  and  $\beta$  we have

$$E[Y] = \alpha\beta \quad V(Y) = \alpha\beta^2$$

### Proof

$$E[Y] = \int_0^{\infty} y \frac{y^{\alpha-1} e^{-\frac{y}{\beta}}}{\beta^{\alpha} \Gamma(\alpha)} dy \underbrace{=}_{t=y/\beta} \frac{1}{\beta^{\alpha} \Gamma(\alpha)} \int_0^{\infty} t^{\alpha} \beta^{\alpha} e^{-t} \beta dt$$

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

$$E[Y^2] = \int_0^{\infty} y^2 \frac{y^{\alpha-1} e^{-\frac{y}{\beta}}}{\beta^{\alpha} \Gamma(\alpha)} dy \underbrace{=}_{t=y/\beta} \frac{\beta^{\alpha+2} \Gamma(\alpha + 2)}{\beta^{\alpha} \Gamma(\alpha)} = \beta^2 \alpha(\alpha + 1)$$

# $\chi^2$ -random variable

## $\chi^2$ -random variable

A random variable  $Y$  is called a  $\chi^2$  random variable with  $k$  degrees of freedom and we write  $Y \sim \chi^2(k)$  if it has the pdf

$$f(y) = \frac{y^{\frac{k}{2}-1} e^{-y/2}}{2^{k/2} \Gamma(k/2)}.$$

- This is a special case of gamma RV with  $\beta = 2$  and  $\alpha = k/2$  (half-integers).
- There is natural relation between  $\chi^2$ -random variable and normal random variable (see later). For example if  $Z \sim N(0, 1)$  then  $Z^2 \sim \chi^2(1)$