

① Discrete distribution: Poisson *discrete random variable*

② Continuous random variables, PDF and CDF ✓

③ The Exponential and Gamma distribution  
*continuous.*

# The Poisson distribution

# The Poisson distribution

Some experiments result in counting the **number of times** particular events occur/arrive during a given **time interval**.

## Examples:

- Number of phone calls between 9AM and 10AM; the number of customers that arrive at a ticket window between 12noon and 2pm.
- Number of typos on a 10-page report. (here: 10-page is like the “time interval”)

Usually, need a parameter  $\lambda$  that measures the **average** particular events occur per **unit** time.

$$X = \overset{\text{(generic)}}{\# \text{ arrivals}} \text{ per } \underbrace{\overset{\text{(generic)}}{\text{unit interval}}}$$

$$\begin{aligned} \lambda &= \underline{\text{average}} \# \text{ arrivals per } \underbrace{\text{unit interval}} \\ &= E[\underline{X}]. \end{aligned}$$

A discrete random variable  $X$  has a **Poisson distribution**, denoted by  $X \sim \text{Poisson}(\lambda)$ , if its PMF is of the form

$$p(x) = e^{-\lambda} \frac{\lambda^x}{x!}, \quad x = 0, 1, 2, \dots,$$

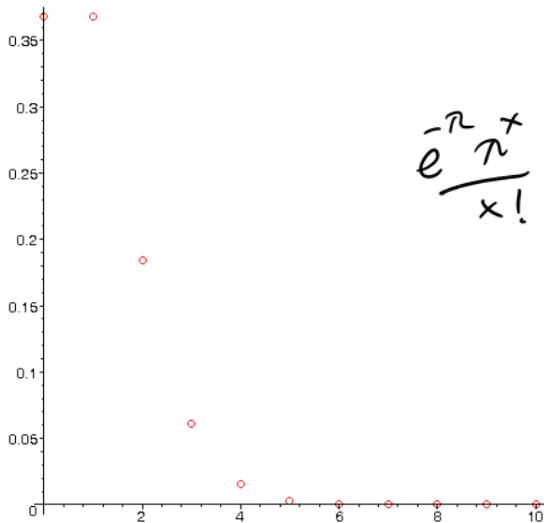
where  $\lambda > 0$ .

### Theorem 1 (Poisson)

If  $X \sim \text{Poisson}(\lambda)$ , then

$$\mathbb{E}[X] = \lambda, \quad \text{Var}[X] = \lambda. \quad \checkmark$$

↑  
verify in Tut 6 later.



PMF of Poisson(1)  $\lambda=1$

## Example 2

On average, there are 2 supernovae in the milky way per century.  
Assuming Poisson distribution, what is the probability that there are 2  
supernovae in the milky way within one decade?

$X = \# \text{ supernovae per } \underline{\text{decade}} . \checkmark$

Want calculate  $P(X=2)$

$$X \sim \text{Poisson}(\underline{\underline{\lambda}})$$

On average,  $\textcircled{2}$  supernovae per century.  $\left. \begin{array}{l} \text{(100 years)} \\ \downarrow \end{array} \right\}$

$\frac{11}{11}$        $\frac{2}{10}$        $\frac{1}{1}$       " decade  
 (10 years) ✓

$$X \sim \text{poisson}(\pi = 0.2)$$

$$P(X=x) = \frac{e^{-\lambda} \lambda^x}{x!}, \quad P(X=2) = \frac{e^{-0.2} 0.2^2}{2!} \approx 0.016 \%$$



### Example 3

In a city, telephone calls to 911 come on the average of two every 3 minutes. If one assumes a Poisson distribution, what is the probability of five or more calls arriving in a 9-minute period?

$X = \# \text{ calls per } \underline{9\text{-minute period}}.$

Want  $P(X \geq 5)$ .

$$X \sim \text{Poisson}(\underline{\underline{\lambda}})$$

On average, 2 calls per 3-minute. ✓  
 $\times 3$   $\times 3$  ↓

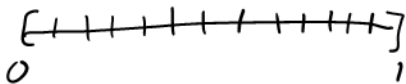
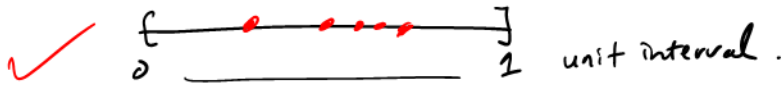
On average, (6) calls per 9-minute

So  $\lambda = 6$ .

$$\begin{aligned} P(X \geq 5) &= 1 - P(X \leq 4) \\ &= 1 - P(0 \leq X \leq 4) \\ &= 1 - \left( \sum_{x=0}^4 \frac{e^{-6} 6^x}{x!} \right) \approx 1 - 0.285 \\ &\approx 0.715 \end{aligned}$$

Not tested:

$\lambda = \text{average \# arrivals}$



$n$  subintervals.

$\frac{\lambda}{n} = \text{average per subinterval}$

$p = \frac{\lambda}{n}$  = "prob" of having  
an arrival per subinterval.

$\lambda = np$ . (success probability  $p$ ).

$$P(X=x) = \binom{n}{x} \left(\frac{\lambda}{n}\right)^x \left(1 - \frac{\lambda}{n}\right)^{n-x}$$
$$= \frac{n!}{x! (n-x)!} \frac{\lambda^x}{n^x} \left(1 - \frac{\lambda}{n}\right)^n \left(1 - \frac{\lambda}{n}\right)^{-x}$$

$n \rightarrow \infty$

Diagram illustrating the limit process:

- $x!$  and  $(n-x)!$  are circled, with an arrow pointing from  $(n-x)!$  to 1.
- $\frac{\lambda^x}{n^x}$  is circled, with an arrow pointing to  $\frac{\lambda^x}{x!} e^{-\lambda}$  (part of a boxed expression).
- $\left(1 - \frac{\lambda}{n}\right)^n$  is circled, with an arrow pointing to  $e^{-\lambda}$  (part of a boxed expression).
- $\left(1 - \frac{\lambda}{n}\right)^{-x}$  is circled, with an arrow pointing to 1.

The PMF of a binomial distribution  $\text{Binomial}(n, p)$  can be approximated by that of  $\text{Poisson}(\lambda)$  with  $\lambda = np$ .

This works well if  $np < 10$  and  $n > 50$ .

### Example 4

Let  $X \sim \text{Binomial}(100, 0.02)$ . ✓

- $\mathbb{P}(X = 2) = \binom{100}{2} 0.02^2 0.98^{98} \approx 0.273$ .
- Approximation by  $\text{Poisson}(\underline{100} \times \underline{0.02})$ , ✓

$$\mathbb{P}(X = 2) = e^{-2} \frac{2^2}{2!} = 0.271.$$

$$n = 100 \quad p = 0.02$$

$$\begin{aligned} \lambda &= 100 \times 0.02 \\ &= np \\ &= 2 \quad \checkmark \end{aligned}$$

# Continuous random variables, PDF and CDF

A **continuous random variable** typically is a random variable whose set of possible values is an interval of real numbers or a union of such intervals.  
 $(a, b), [a, b]$   $(0, 1) \cup [3, 6]$ .

Example: An air sample is analyzed and the fraction  $X$  of oxygen in the sample is determined (e.g.  $X = 0.15$  means that 15% of the volume is taken up by oxygen).

Set of possible values of  $X \in [0, 1]$  (interval of real numbers  $x$  with  $0 \leq x \leq 1$ ).

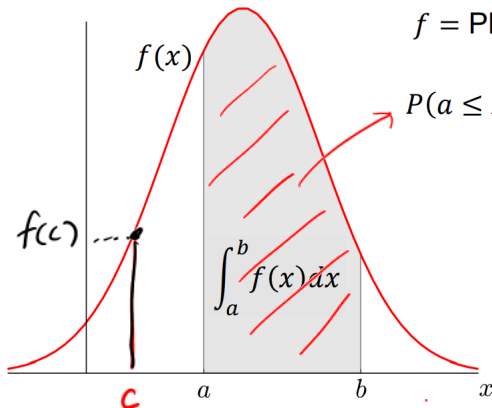
A function  $f$  that assigns a nonnegative real number  $f(x)$  to each real number  $x$  is a **probability density function (PDF)** for a continuous random variable  $X$  if

$$\mathbb{P}(\underline{a \leq X \leq b}) = \int_a^b f(x) dx \quad \checkmark$$

for all real numbers  $a, b, a \leq b$ .

- Note  $\int_a^b f(x) dx$  is the area between the graph of  $f$  and the segment of the  $x$ -axis between  $a$  and  $b$ .
- If necessary, we write  $f_X$  instead of  $f$  to indicate that  $f$  belongs to  $X$ .





$f = \text{PDF of } X$

$$P(a \leq X \leq b) = \int_a^b f(x) dx$$

$$\int_{-\infty}^{\infty} f(x) dx = 1.$$

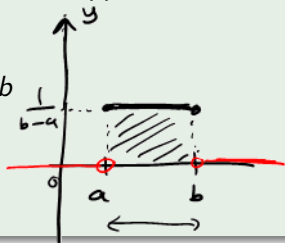
$$P(X=c) \neq f(c).$$

$$\underline{P(X=c)} = 0$$

## Example 5 (Uniform distribution)

The random variable  $X$  has a **uniform distribution** if its PDF  $f(x)$  is equal to a constant on its support. In particular, if the support is the interval  $[a, b]$ , then

$$f(x) = \begin{cases} \frac{1}{b-a} & \text{if } a \leq x \leq b \\ 0 & \text{otherwise.} \end{cases}$$

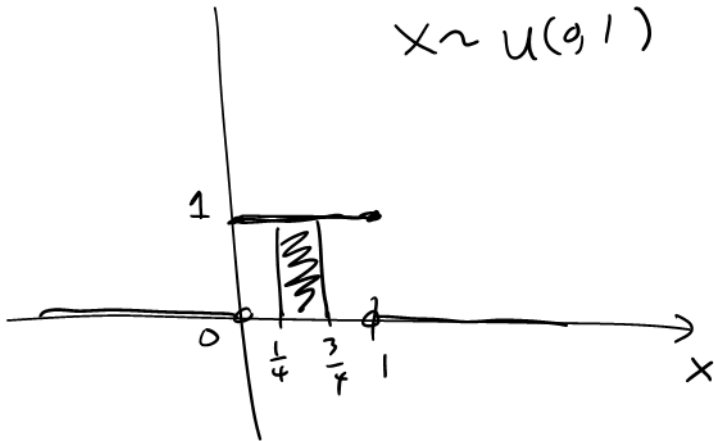


We shall denote it by  $X \sim U(a, b)$ .

Example: If  $X \sim U(0, 1)$ , then

$$\mathbb{P}\left(\frac{1}{4} \leq X \leq \frac{3}{4}\right) = \int_{1/4}^{3/4} f(x) dx = \int_{1/4}^{3/4} 1 dx = \frac{3}{4} - \frac{1}{4} = \frac{2}{4} = \frac{1}{2}.$$

$$X \sim U(0, 1)$$





## Example 6

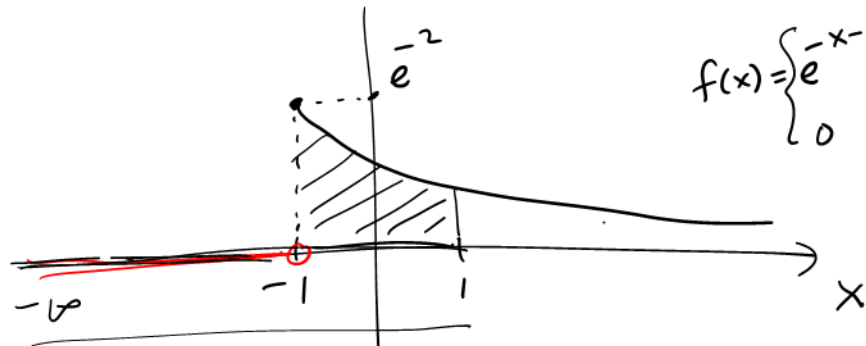
Suppose  $X$  has PDF

$$f(x) = e^{-x-1}, \quad -1 \leq x < \infty.$$

Compute  $\mathbb{P}(X \leq 1)$  and  $\mathbb{P}(X \geq 1)$ .

*Give it a try!*

$$f(x) = \begin{cases} e^{-x-1} & x \geq -1 \\ 0 & x < -1 \end{cases}$$



$$\begin{aligned} P(X \leq 1) &= \int_{-\infty}^1 f(x) dx = \int_{-\infty}^{-1} \textcircled{0} dx + \int_{-1}^1 e^{-x-1} dx \\ &= \int_{-1}^1 e^{-x-1} dx \end{aligned}$$

$$\begin{aligned}
 &= \bar{e}' \left[ \int_{-1}^1 \bar{e}^x dx \right] \\
 &= \bar{e}' \left[ \left[ -\bar{e}^x \right]_{-1}^1 \right] \\
 &= \bar{e}' \left[ -\bar{e}^1 - (-\bar{e}^{-1}) \right] \\
 &= -\bar{e}^{-2} + 1
 \end{aligned}$$

$$P(X \geq 1) = 1 - P(X < 1) \quad \checkmark$$

$$= 1 - P(X \leq 1)$$

$$= 1 - (1 - e^{-2})$$

$$= e^{-2}$$

#.



If  $X$  is a continuous random variable with PDF  $f(x)$ , then the **Cumulative Density Function (CDF)** of  $X$  is defined by

$$\underline{\underline{F(x) = \mathbb{P}(X \leq x) = \int_{-\infty}^x f(t) dt.}}$$

Note:

- $F(x)$  is nondecreasing
- $0 \leq F(x) \leq 1$ . ✓

~~•~~  $F'(x) = \frac{dF}{dx} = f(x)$  (PDF) ✓



$$\frac{dF}{dx} = \frac{d}{dx} \int_{-\infty}^x f(t) dt = f(x)$$

Fundamental  
Theorem of  
Calculus.

Discrete

PMF  $P(X=x) = p(x)$

CDF  $P(X \leq x) = \sum_{t \leq x} p(t)$

Continuous

PDF  $P(a \leq X \leq b)$   
 $= \int_a^b f(x) dx$

CDF  $P(X \leq x) = \int_{-\infty}^x f(t) dt$

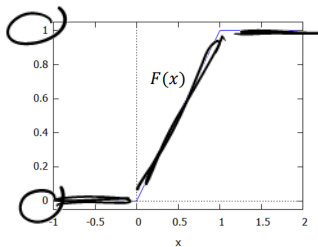
## Example 7

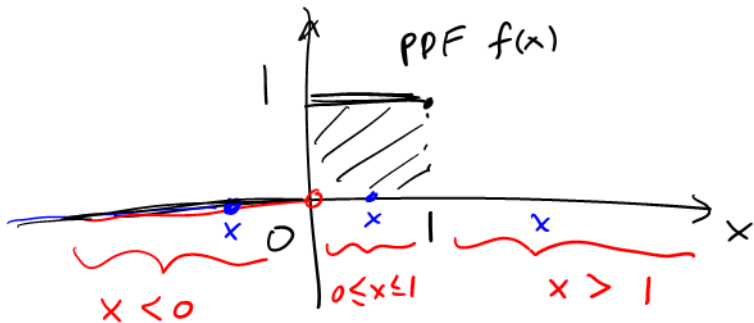
Let  $X \sim U(0, 1)$  be the uniform distribution on  $[0, 1]$ . Its CDF is given by

$$F(x) = \begin{cases} 0 & \text{if } x \leq 0 \\ x & \text{if } 0 \leq x \leq 1 \\ 1 & \text{if } x \geq 1. \end{cases}$$

Note: For  $0 \leq x \leq 1$ , we have

$$F(x) = \int_{-\infty}^x f(t) dt = \int_0^x 1 dt = x.$$





$$\boxed{x < 0} : F(x) = \int_{-\infty}^x f(t) dt = \int_{-\infty}^x 0 dt = 0.$$

$$\boxed{0 \leq x \leq 1} : F(x) = \int_{-\infty}^x f(t) dt = \int_{-\infty}^0 0 dt + \int_0^x 1 dt$$

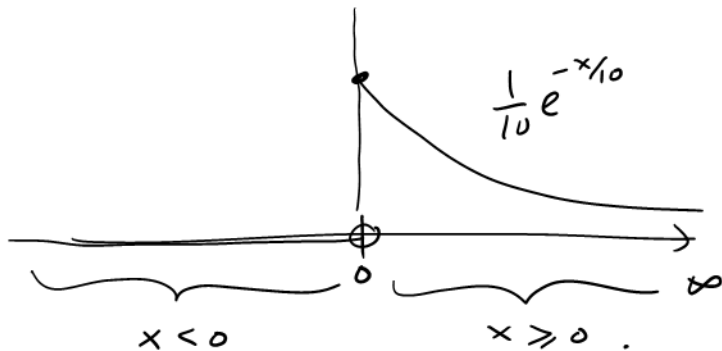
$$\boxed{x > 1} : F(x) = 1 = x.$$

## Example 8

Suppose  $X$  has PDF

$$f(x) = \begin{cases} \frac{1}{10} e^{-x/10} & \text{if } x \geq 0, \\ 0 & \text{otherwise.} \end{cases}$$

Compute the CDF of  $X$ .



$$\boxed{x < 0} : F(x) = \int_{-\infty}^x 0 \, dt = 0$$

$$\boxed{x \geq 0} : F(x) = \int_{-\infty}^x f(t) \, dt = \underbrace{\int_{-\infty}^0 0 \, dt}_0 + \int_0^x \frac{1}{10} e^{-t/10} \, dt$$

$$= \frac{1}{10} \left[ \frac{e^{-\frac{t}{10}}}{-\frac{1}{10}} \right]_0^x$$

$$= \left( -e^{-\frac{x}{10}} \right) - (-1)$$

$$= 1 - e^{-\frac{x}{10}}$$

$$F(x) = \begin{cases} 0 & \text{if } x < 0 \\ 1 - e^{-\frac{x}{10}} & \text{if } x \geq 0 \end{cases} \quad \checkmark$$

Let  $X$  be a continuous random variable with PDF  $f(x)$ .

- Its **expected value** or **mean** is

$$\mathbb{E}[X] = \int_{-\infty}^{\infty} x \cdot f(x) dx.$$

Discrete

$$\mathbb{E}[x] = \sum_x x \cdot p(x)$$

- If  $g(X)$  is a function of  $X$ , then

$$\mathbb{E}[g(X)] = \int_{-\infty}^{\infty} g(x) f(x) dx.$$

$$\mathbb{E}[g(x)] = \sum_x g(x) p(x)$$

- Similar to discrete random variables, the **variance** and **standard deviation**  $\sigma$  of  $X$  can be calculated as follows (where  $\mu = \mathbb{E}[X]$ ):

$$\text{Var}[X] = \mathbb{E}[(X - \mu)^2] = \underline{\underline{\mathbb{E}[X^2] - \mu^2}}, \quad \sigma = \sqrt{\text{Var}[X]}.$$



## Example 9

The total amount of medical claims (in millions) of the employees of a company has the PDF given by


$$f(x) = \begin{cases} 30x(1-x)^4, & 0 < x < 1. \\ 0 & \text{otherwise.} \end{cases} \quad \checkmark$$


Find


- (i) The mean and the standard deviation of the total in dollars.
- (ii) The probability that the total exceeds \$0.2 millions.

*Solution.*

(i)

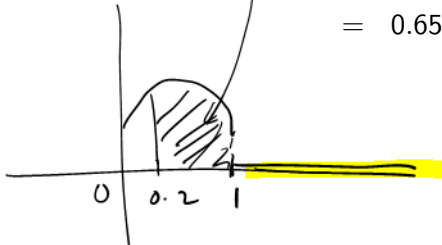
$$\begin{aligned}\text{mean } \mu = \mathbb{E}[X] &= \int_{-\infty}^{\infty} x f(x) dx = \int_0^1 x(30x)(1-x)^4 dx \\ &= 0.286 \text{ millions.}\end{aligned}$$


$$\begin{aligned}\text{variance} &= \underline{\underline{\mathbb{E}[X^2]}} - \mu^2 = \int_{-\infty}^{\infty} x^2 f(x) dx - (0.286)^2 \\ &= \int_0^1 x^2(30x)(1-x)^4 dx - (0.286)^2 \\ &= 0.107 - (0.286)^2 = 0.025204 \text{ millions.}\end{aligned}$$


$$\sigma = \sqrt{\text{Var}[X]} = \sqrt{0.025204} = 0.159 \text{ millions.}$$


(ii)

$$\begin{aligned}\mathbb{P}(X > 0.2) &= \int_{0.2}^{\infty} 30x(1-x)^4 dx \\ &= \int_{0.2}^1 30x(1-x)^4 dx \\ &= 0.6554.\end{aligned}$$



# The Exponential and Gamma distribution

# The Exponential distribution

We now turn to a continuous random variable that is related to Poisson distribution.

- Let  $\lambda$  be the mean/average number of occurrences per unit interval.  $[0, 1]$
- Let  $X \sim \text{Poisson}(\lambda w)$  be the random variable that counts the number of occurrences in an interval of size  $w$ . ✓
- Then  $\mathbb{P}(\text{no occurrences in } \underline{[0, w]}) = \mathbb{P}(X = \underline{0}) = \underline{e^{-\lambda w}}$ . ✓

$$PMF = \frac{e^{-\lambda} \lambda^x}{x!} = P(X=x) \quad \text{if } X \sim \text{Poisson}(\lambda)$$

$$\downarrow \quad \frac{e^{-\lambda w} (\lambda w)^0}{0!} = e^{-\lambda w}$$

$\lambda$  = # arrivals per unit interval.  
 $\downarrow$   $\times w$   $\downarrow$   
 $\lambda w$  = # arrivals per interval of site w.

$$X \sim \text{Poisson}(\lambda w)$$

$X$  = # arrivals. in  $[0, w]$

$$\begin{aligned} &P[\text{no occurrence in } [0, \omega]] \\ &= P(X = 0) \end{aligned}$$

Let  $W$  = **waiting time** until the **first** occurrence. Then its CDF  $F(w)$  is given by

$$\begin{aligned} F(w) &= \mathbb{P}(W \leq w) = 1 - \mathbb{P}(W > w) \\ &= 1 - \mathbb{P}(\text{no occurrences in } [0, w]) = 1 - e^{-\lambda w} \end{aligned} \quad \checkmark$$

Note that  $W$  is nonnegative. For  $w \geq 0$ , the PDF of  $W$  is

$$\frac{dF}{dw} = f(w) = \lambda e^{-\lambda w}. \quad \checkmark$$

$W$       PDF       $f(w) = \lambda e^{-\lambda w}$        $w \geq 0$

$$F(w) = 1 - e^{-\lambda w}$$



We often let  $\lambda = \frac{1}{\theta}$ , and say that the random variable  $X$  has an **exponential distribution**, denoted by  $X \sim \text{Exp}(\underline{\theta})$ , if its PDF is defined by

$$f(x) = \frac{1}{\theta} e^{-x/\theta}, \quad 0 \leq x < \infty. \quad \checkmark$$
$$= \lambda e^{-\lambda x} \quad \checkmark$$

### Theorem 10 (Exponential distribution)

If  $W \sim \text{Exp}(\theta)$ , then

$$\mathbb{E}[W] = \theta, \quad \text{Var}[W] = \theta^2. \quad \checkmark$$

$$\mathbb{E}[W] = \frac{1}{\lambda}, \quad \text{Var}[W] = \frac{1}{\lambda^2}. \quad \checkmark$$

## Example 11

Customers arrive in a certain shop according to a Poisson process at mean rate of 20 per hour. What is the probability that the shopkeeper will have to wait more than 5 minutes for the arrival of the first customer?

$W$  = waiting time (minutes) until first customer.

Want  $P(W > 5)$

$$W \sim \text{Exp}(\theta) \quad \theta = \frac{1}{\lambda}$$

Poisson :

average of 20 per 60 minutes.

" "  $\boxed{\frac{20}{60}}$  per minute.

$$\lambda = \frac{1}{3}.$$

$$W \sim \text{Exp}(\theta = \frac{1}{\lambda} = \frac{1}{1/3} = 3).$$

$$P(W > 5) = \int_5^{\infty} \frac{1}{\underline{\underline{\theta}}} e^{-\frac{x}{\theta}} dx$$

$$= \int_5^{\infty} \frac{1}{3} e^{-x/3} dx$$

$$= \frac{1}{3} \left[ \frac{e^{-x/3}}{-1/3} \right]_5^{\infty} = 0 + e^{-5/3} \approx 0.1889$$

#.

Let  $W$  denote the **waiting time until the  $\alpha$ th occurrence** in a Poisson process with  $\lambda = \frac{1}{\theta}$ . Then  $W$  has a **Gamma distribution** with **shape** parameter  $\alpha$  and **scale** parameter  $\theta$ , denoted by  $W \sim \text{Gamma}(\alpha, \theta)$ , with PDF given by

$$f(w) = \frac{1}{\Gamma(\alpha)\theta^\alpha} w^{\alpha-1} e^{-w/\theta}, \quad 0 \leq w < \infty.$$

✓

$\xrightarrow{\alpha=1} \frac{1}{\theta} e^{-w/\theta}$

Here:

$$\Gamma(\alpha) = \int_0^\infty y^{\alpha-1} e^{-y} dy, \quad \alpha > 0.$$

For our purpose,  $\alpha$  is usually a positive integer, and so

$$\Gamma(\alpha) = (\alpha - 1)!.$$

$$\text{Gamma}(\alpha = \underline{1}, \theta) = \text{Exp}(\theta)$$

## Theorem 12 (Gamma distribution)

If  $W \sim \text{Gamma}(\alpha, \theta)$ , then

$$\mathbb{E}[W] = \alpha\theta, \quad \text{Var}[W] = \alpha\theta^2.$$

average 30 per 60 minutes  
average.  $\frac{30}{60} = \frac{1}{2}$  per minute.

### Example 13

Suppose the number of customers per hour arriving at a shop follows a Poisson distribution with mean 30. What is the probability that the shopkeeper will wait for more than 5 minutes until the second customer arrives?

$$\alpha = 2$$

$$\lambda = \frac{1}{2}$$

$$\Rightarrow \theta = \frac{1}{\lambda} = 2.$$

*Solution.*  $W$  = waiting time (in minutes) until the second customer arrives. Then

$$W \sim \text{Gamma}(\alpha = 2, \theta = 2). \text{ (why?)}$$

Want to compute  $\mathbb{P}(W > 5)$ .



$$\mathbb{P}(W > 5) = \int_5^{\infty} \frac{1}{\Gamma(2)(2)^2} w^{2-1} e^{-w/2} dw$$

$$= \frac{1}{4} \int_5^{\infty} w e^{-w/2} dw$$

$$= \frac{1}{4} \left( \left[ -2w e^{-w/2} \right]_5^{\infty} - \int_5^{\infty} (-2) e^{-w/2} dw \right)$$

(by integration-by-parts)

$$= \frac{1}{4} \left( 10e^{-5/2} + 2 \left[ (-2) e^{-w/2} \right]_5^{\infty} \right)$$

$$= \frac{1}{4} \left( 10e^{-5/2} + 4e^{-5/2} \right)$$

$$= 0.2873.$$

