

# Probability and Statistics

## MA20205

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Lecture 8  
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# Special discrete distributions

From Poisson process to Poisson distribution

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## From Poisson process to Poisson distribution

### Poisson process (PP)

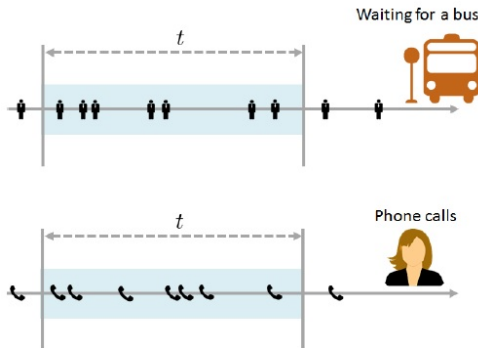
A Poisson Process is a model for a series of discrete events where the average time between events is known, but the exact timing of events is random. The arrival of an event is independent of the event before (waiting time between events is memoryless).

# Special discrete distributions

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## Criteria of PP

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- 3 Two events cannot occur at the same time.

**Note** Events are not simultaneous — means we can think of each sub-interval of a Poisson process as a Bernoulli Trial, that is, either a success or a failure. (Bus arrivals need not perfectly follow all the criteria)



## Special discrete distributions

Poisson Distribution probability mass function gives the probability of observing  $k$  events in a time period given the length of the period and the average events per time:

$$P(k \text{ events in time period}) = e^{-\frac{\text{events}}{\text{time}} \times \text{time period}} \frac{\left(\frac{\text{events}}{\text{time}} \times \text{time period}\right)^k}{k!}$$

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Setting

$$\begin{aligned}\lambda &= \frac{\text{events}}{\text{time}} \times \text{time period} \\ &= \text{expected number of events in the time interval (time period)} \\ &= \text{rate parameter}\end{aligned}$$

we obtain the desired pmf.

Click the link: [The Waiting Time Paradox, or, Why Is My Bus Always Late?](#)

# Special discrete distributions

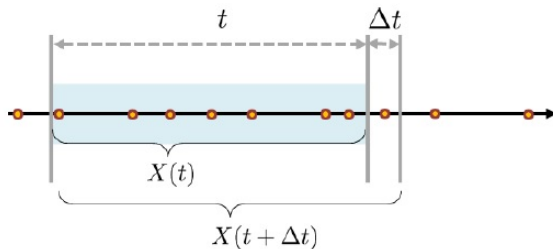
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For small  $\Delta t$ ,

$$P(X(t + \Delta t) - X(t) = 1) = \lambda \cdot \Delta t$$

where  $\lambda$  is the rate of occurrence of events. Here  $X(t + \Delta t) - X(t)$  can be thought as a Bernoulli trial.

## Special discrete distributions

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Then

$$\begin{aligned} & P(X(t + \Delta t) = k) \\ = & P[X(t) = k] \cdot (1 - \lambda \Delta t) + P[X(t) = k - 1] \cdot (\lambda \Delta t) \\ = & P[X(t) = k] - P[X(t) = k] \lambda \Delta t + P[X(t) = k - 1] \lambda \Delta t. \end{aligned}$$

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Setting  $\Delta t \rightarrow 0$  we obtain the ODE

$$\frac{d}{dt} P[X(t) = k] = \lambda (P[X(t) = k - 1] - P[X(t) = k])$$

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Setting  $t = 1$  we obtain the Poisson distribution

## Special discrete distributions

Binomial approximation to Poisson If  $n$  is large in  $\text{Bin}(n, p)$  then calculation the probabilities is a prohibited amount of work. For instance, let  $n = 3000$ ,  $p = 0.005$  and try to calculate  $P(X = 29)$  when  $X \sim \text{Bin}(n, p)$ .

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Setting  $n \rightarrow \infty$  and  $p \rightarrow 0$  with  $\lambda = np$ , a constant, the binomial probability is given by

$$\begin{aligned}\binom{n}{k} p^k (1-p)^{n-k} &= \frac{n!}{k!(n-k)!} \left(\frac{\lambda}{n}\right)^k \left(1 - \frac{\lambda}{n}\right)^{n-k} \\ &= \frac{\lambda^k}{k!} \left(1 - \frac{\lambda}{n}\right)^n \\ &\rightarrow \frac{\lambda^k}{k!} e^{-\lambda} \text{ as } n \rightarrow \infty\end{aligned}$$

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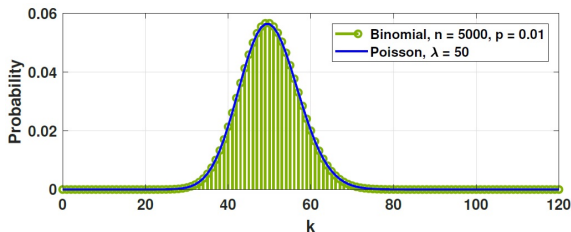
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**Question** How does this derivation relate to Poisson Process?

# Special discrete distributions

## Comparing Poisson with Binomial





# Special continuous distributions

**Continuous uniform random variable** The pdf of the uniform random variable  $X$  is

$$f(x) = \begin{cases} \frac{1}{b-a} & \text{if } x \in [a, b] \\ 0, & \text{otherwise} \end{cases}$$

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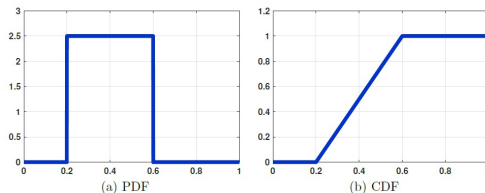
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 $\text{Unif}(0.2, 0.6)$  looks like



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(b)  $\text{Var}(X) = \frac{(b-a)^2}{12}$

(c)

$$M(t) = \begin{cases} 1, & \text{if } t = 0 \\ \frac{e^{tb} - e^{ta}}{t(b-a)} & \text{if } t \neq 0 \end{cases}$$

# Special continuous distributions

Recall: Gamma function

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

where  $z > 0$



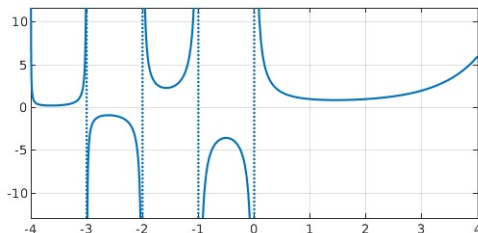
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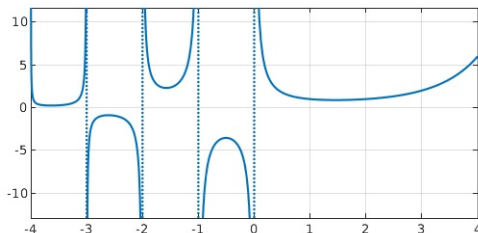
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**Note** There is an alternative way to define Gamma function when  $z \leq 0$ .  
The above expression on the rhs is also known as Euler's second integral

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(e)  $\Gamma(n+1) = n!$  for any positive integer  $n$

# Special continuous distributions

**Gamma distribution** A random variable  $X$  has Gamma distribution if the pdf is given by

$$f(x) = \begin{cases} \frac{\beta^\alpha x^{\alpha-1} e^{-\beta x}}{\Gamma(\alpha)} & \text{if } 0 < x < \infty \\ 0, & \text{otherwise} \end{cases}$$

where  $\alpha > 0, \beta > 0$



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Then we write  $X \sim \text{Gamma}(\alpha, \beta)$

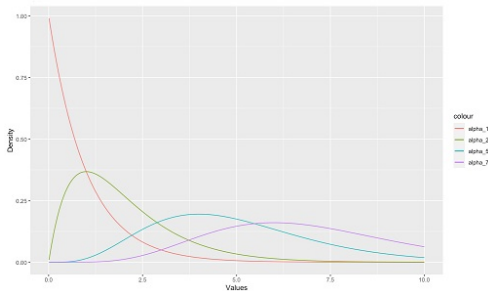
# Special continuous distributions

Shape parameter:  $\alpha$

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The pdf for  $\text{Gamma}(\alpha, 1)$  looks like



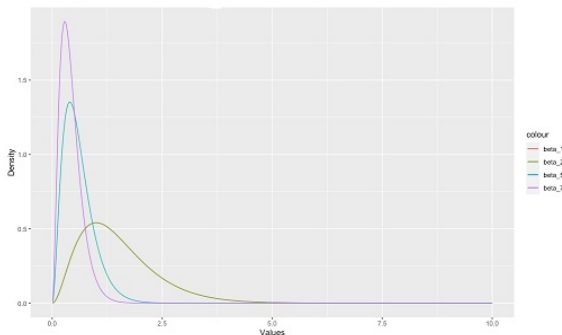
# Special continuous distributions

Scale parameter:  $\beta$

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Scale parameter:  $\beta$

The pdf for  $\text{Gamma}(3, \beta)$  looks like

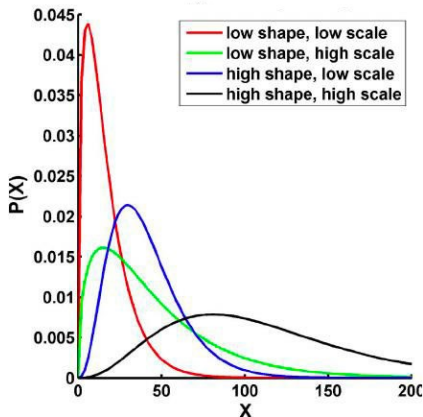


# Special continuous distributions

pdf of  $\text{Gamma}(\alpha, \beta)$

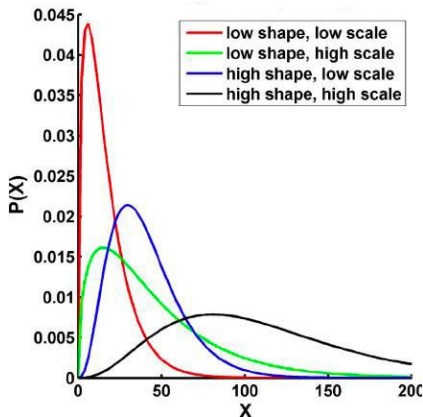
# Special continuous distributions

pdf of  $\text{Gamma}(\alpha, \beta)$



# Special continuous distributions

pdf of  $\text{Gamma}(\alpha, \beta)$



Properties:

$$E(X) = \frac{\alpha}{\beta}, \quad \text{Var}(X) = \frac{\alpha}{\beta^2}$$