


## Poisson Distribution

$$P(x; \lambda) = \frac{\lambda^x e^{-\lambda}}{x!}$$

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## 1 Poisson Distribution

- Poisson Distribution from Binomial Distribution
- Expectation Value of Poisson Distribution  $E(X)$ :
- Varaince  $\sigma^2(X)$
- Poisson Distribution (Alternative Method)
- Problems on Poisson Distribution (Home Work)

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- The Probability here is proportional to size of sample say region,time volume or number

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- with mean  $np = 1 \cdot \frac{1}{2}$  and variance  $npq = 1 \cdot \frac{1}{2} \frac{1}{2} = \frac{1}{4}$

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- For two coins having four possible events, each with Probability as  $\frac{1}{2^2}$
- And for  $n$  it is  $\frac{1}{2^n} \rightarrow 0$  for large  $n$ .

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- The information with us is just average happenings defined above
- we can develop this Poisson Distribution independently as well as from Binomial as  $n \rightarrow \infty$



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- put  $np = \lambda$  (mean of PD)

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$$P(X = x) = f(x) = \frac{\lambda^x e^{-\lambda}}{x!}$$

Or we write a random variable follows a Poisson with mean  $\lambda$  as

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- $$E(X) = e^{-\lambda} \left[ \lambda + \frac{\lambda^2}{1!} + \frac{\lambda^3}{2!} + \frac{\lambda^4}{3!} + \dots \right]$$

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- so the expectation value of Poisson Distribution is  $\lambda$

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Need to know

$$E(X(X - 1)) = \sum_{x=0}^n (x(x - 1)) \frac{\lambda^x e^{-\lambda}}{x!}$$

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- $$E(X(X - 1)) = \lambda^2 \sum_{y=0}^m \frac{e^{-\lambda} \lambda^y}{y!} = \lambda^2 \cdot 1 = \lambda^2$$

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- Therefore

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- Hence we see expectation as well as the variance for Poisson Distribution is same as  $\lambda$

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$t + \delta t$	$x$ events	$P(x - 1; t).P(1; = \delta t) + P(x; t).P(0; = \delta t)$



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- Total Probability in time  $\delta t$

$$P(1, \delta t) + P(0, \delta t) = 1$$

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- $$\ln P(0, t) = -\lambda t + c$$

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$$P(0, t) = e^{-\lambda t}$$

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- $$P(x; t + \delta t) = P(x; t)P(0, \delta t) + P(x - 1; t).P(1, \delta t)$$

- $$P(x; t + \delta t) = P(x; t)(1 - \lambda\delta t) + P(x - 1; t).\lambda\delta t$$



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$$P(1; t) = \lambda t e^{-\lambda t} = \left( \frac{\lambda t}{1!} \right)^1 e^{-\lambda t}$$

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- This is the desired Poisson Distribution with mean  $\mu$  and variance  $\mu$

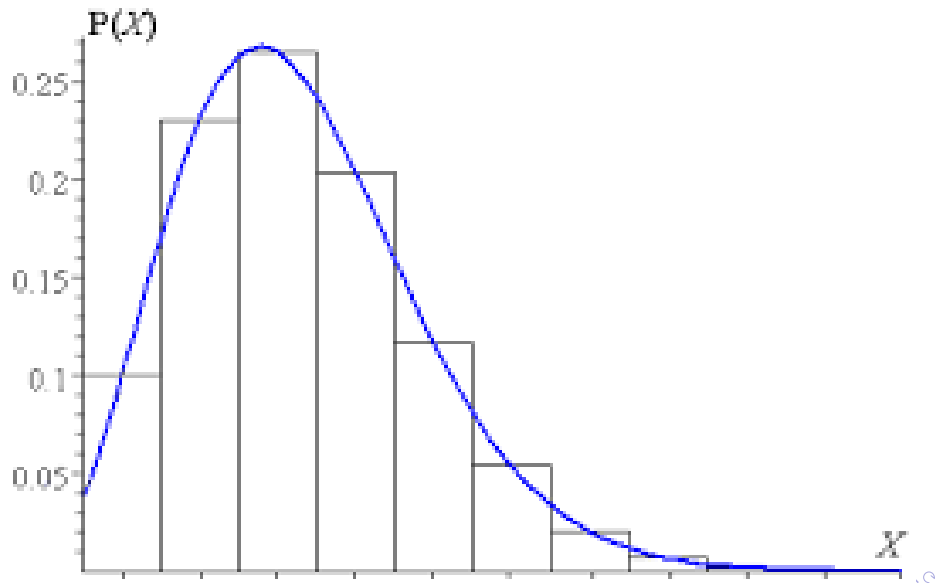
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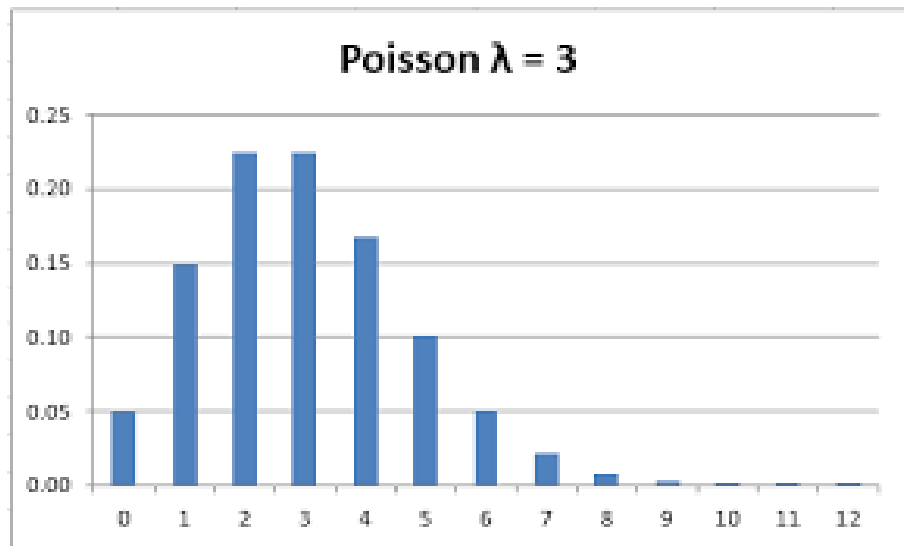
- we see the total Probability is one as

$$\sum_{x=0}^{\infty} P(x; \mu) = \sum_{x=0}^{\infty} \frac{\mu^x e^{-\mu}}{x!} = e^{-\mu} \sum_{x=0}^{\infty} \frac{\mu^x}{x!} = e^{-\mu} e^{\mu} = 1$$

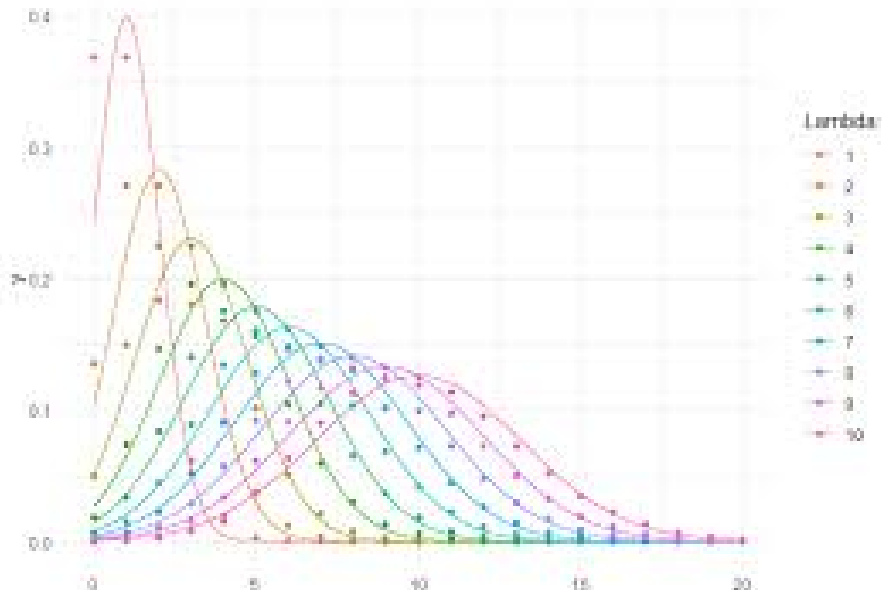
# Poisson Plots



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$$np = \lambda = 2$$

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- Now using complement

$$\begin{aligned} P(x \geq 4; 2) &= 1 - P(x < 4; 2) = 1 - [P(x = 0; 2) + P(x = 1; 2) + P(x = 2; 2)] \\ &= 1 - e^{-2} \left[ \frac{2^0}{0!} + \frac{2^1}{1!} + \frac{2^2}{2!} + \frac{2^3}{3!} \right] = 14.3\% \end{aligned}$$

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- *Solution:* Here,  $\lambda = 5$  and you need to Calculate  $P(x = 3, 5)$  .  
Answer: 14%

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- a) Calculate the probability that it emits 3 particles during a 5-second period.
- b) Calculate the probability that it emits at least one particle during a 5 second period.
- c) During a ten-second period, what is the probability that 6 particles are emitted?

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  - a) the expected number of larvae in the flask? b) the probability that the flask contains at least one mosquito larva?



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- 4. If the light bulbs in a house fail according to a Poisson law, and over the last 15 weeks there have been 5 failures, find the probability that there will not be more than one failure next week.

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- 4. If the light bulbs in a house fail according to a Poisson law, and over the last 15 weeks there have been 5 failures, find the probability that there will not be more than one failure next week.
- 5. What is the probability of making 2 to 4 sales in a week if the average sales rate is 3 per week?