

Geometric Random Variable:

First success -

It is the number X of trials to get the first success. Its PMF is given by $X = 1, 2, 3, \dots$

Suppose probability of success is p

$$P(X=1) = p$$

$$P(X=2) = (1-p)p$$

$$P(X=k) = P_x(k) = (1-p)^{k-1} \cdot p, \quad k=1, 2, 3, \dots$$

Prove that:

$$\sum_{k=1}^{\infty} P_x(k) = 1$$

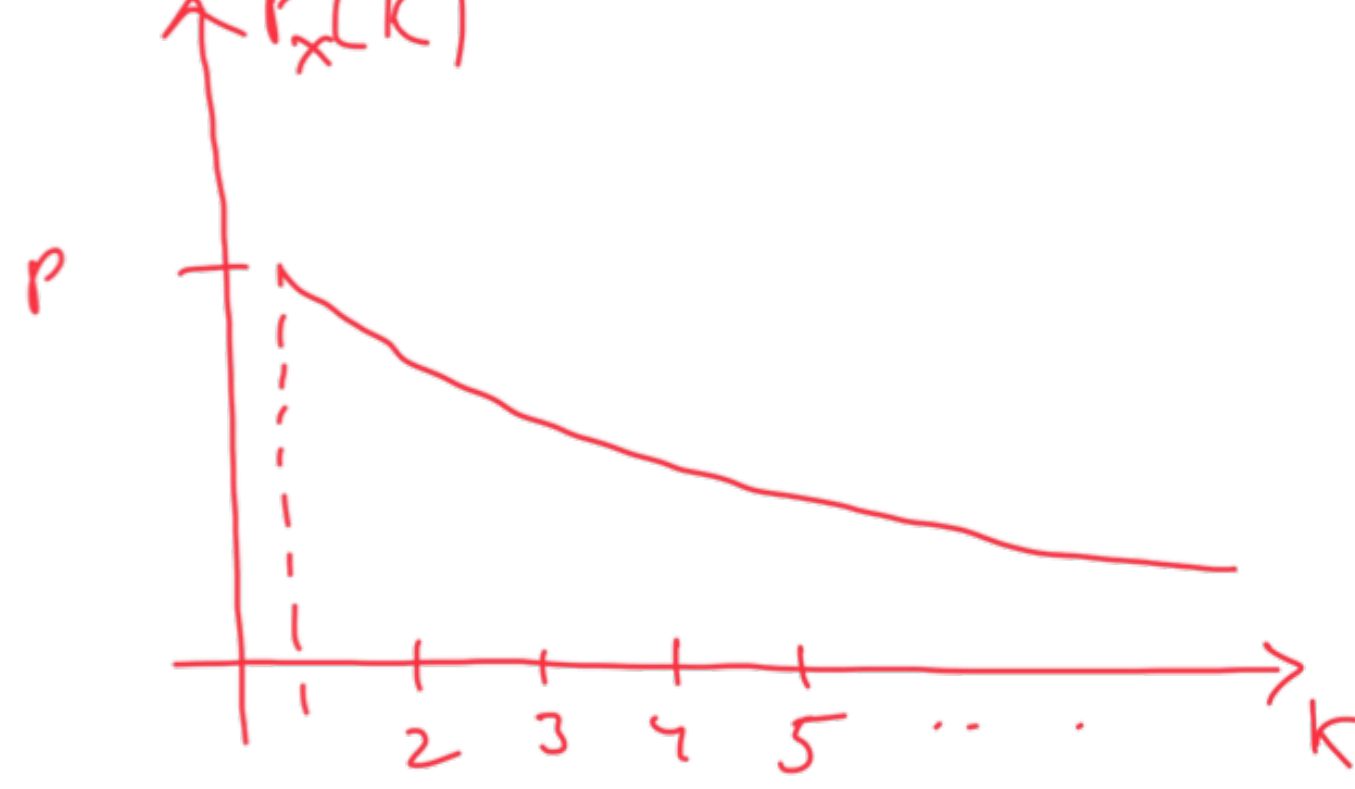
$$\text{Proof} = \sum_{k=1}^{\infty} P_x(k) = \sum_{k=1}^{\infty} (1-p)^{k-1} \cdot p$$

$$a = p \quad \text{So, Sum of infinite G.P.} = \frac{a}{1-r}$$

$$r = 1-p$$

$$\text{So, } \sum_{k=1}^{\infty} P_x(k) = \frac{p}{1-(1-p)} = \frac{p}{1-1+p} = \frac{p}{p} = 1$$

$$\text{Thus } \sum_{k=1}^{\infty} P_x(k) = 1 \quad \text{Hence proved}$$



The Poisson Random Variable:

A poisson Random variable has the following

PMF:

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

$$\& \lambda > 0$$

This is used when we want to know that in a fixed time interval how many times an event happens.

$$\text{Prove that } \sum_{k=0}^{\infty} \frac{e^{-\lambda} \lambda^k}{k!} = 1$$

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

↓ Taylor expansion of e^{λ}

$$= e^{-\lambda} \cdot e^{\lambda} = 1$$

Take a binomial random variable with very small 'p' & very large n.

The poisson PMF with parameter λ is a good approximation for a binomial PMF with parameter $n \in p$, i.e.

$$\frac{e^{-\lambda} \lambda^k}{k!} \approx \binom{n}{k} p^k (1-p)^{n-k}$$

$$k=0, 1, 2, \dots, n$$

provided $\lambda = np$

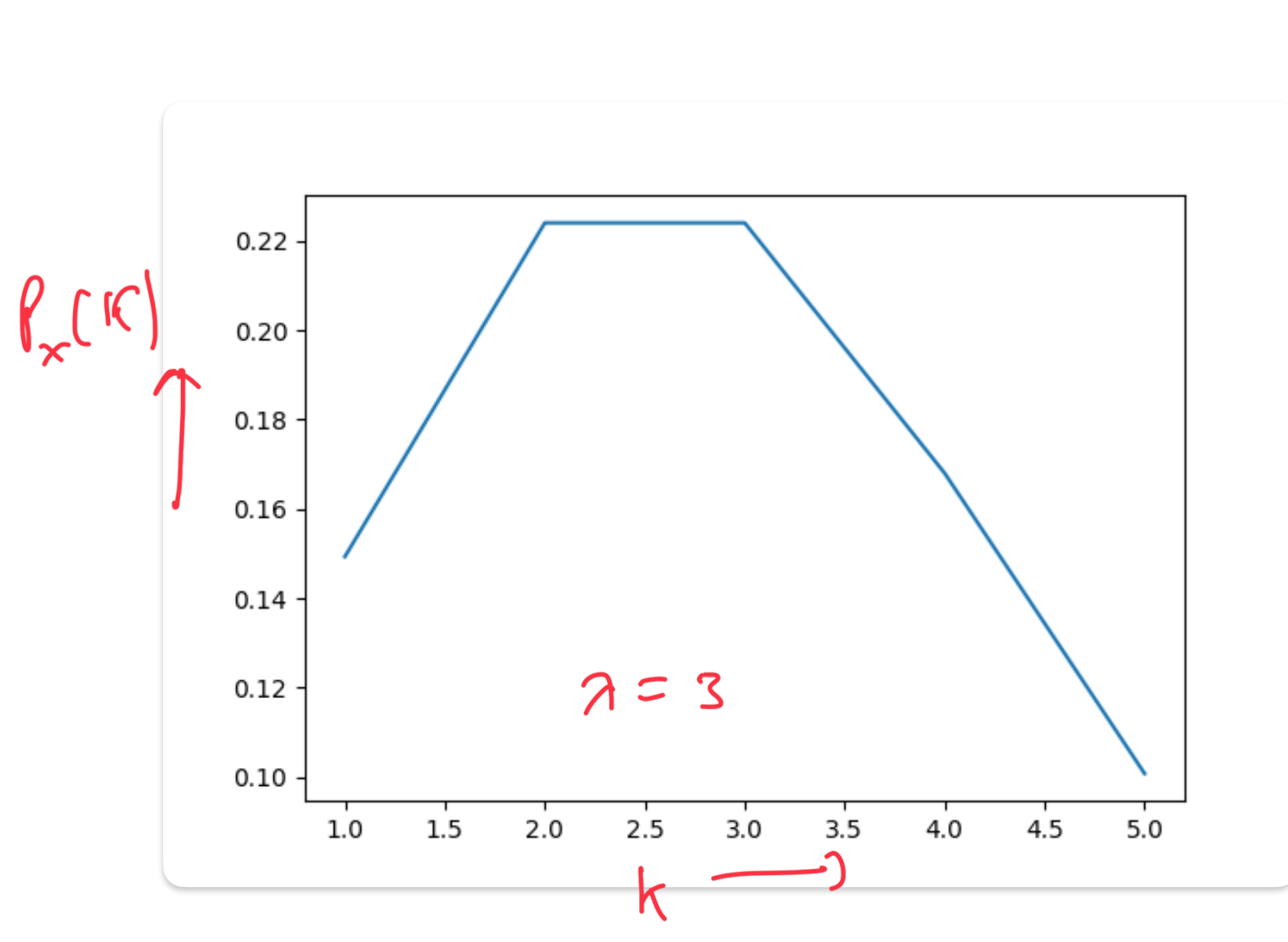
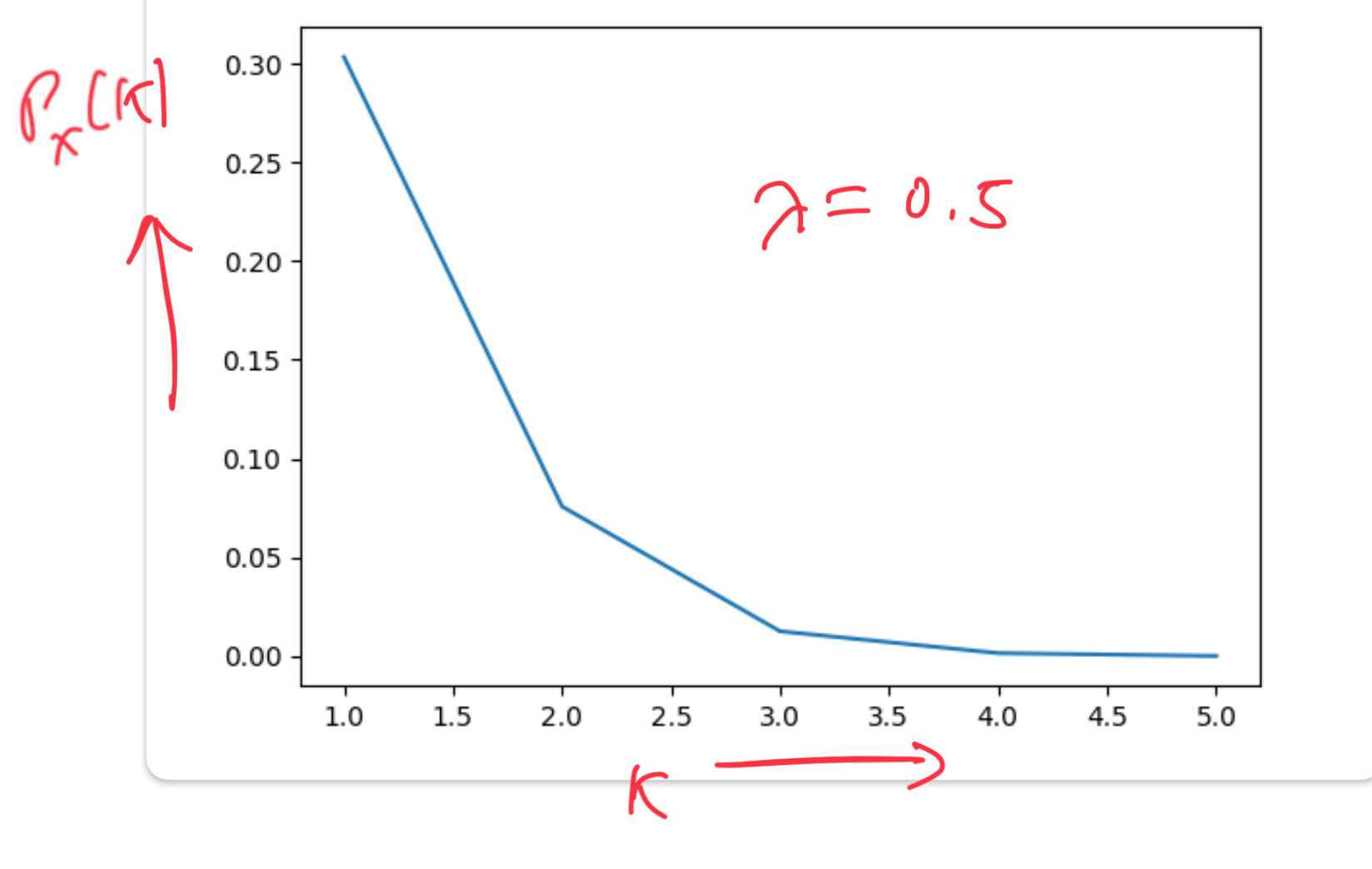
Ex: $n=100, p=0.01$, then probability of 5 successes in 100 trials is

$$\text{binomial} \rightarrow \binom{100}{5} (0.01)^5 (1-0.01)^{100-5} = 0.0029$$

Using Poisson PMF with $\lambda = np = 100 \times 0.01 = 1$

$$\frac{e^{-1} (1)^5}{5!} = 0.0030$$

Plot of Poisson PMF:



Function of random variables:-

Let X be a random variable. If $Y = g(X)$ is a function of X , then Y is also a random variable, since it provides a numerical value for each possible outcome.

If X is discrete the Y is also a discrete random variable. The PMF of Y is calculated from PMF of X .

The PMF of Y is

$$P_Y(y) = \sum_{\{x | g(x)=y\}} P_X(x)$$

$$\text{Ex: } P_X(x) = \begin{cases} 1/9 & \text{if } x \text{ is an integer in } [-4, 4] \\ 0 & \text{otherwise} \end{cases}$$

Let $Y = |X| \Rightarrow Y$ is a discrete random variable

Possible values of $Y = 0, 1, 2, 3, 4$.

$$P(Y=0) = P(X=0) = 1/9$$

$$P(Y=1) = P(X=1) + P(X=-1) = \frac{2}{9}$$

$$P(Y=2) = P(X=2) + P(X=-2) = \frac{2}{9}$$

$$P(Y=3) = P(X=3) + P(X=-3) = \frac{2}{9}$$

$$P(Y=4) = P(X=4) + P(X=-4) = \frac{2}{9}$$

The PMF of Y is:

$$P_Y(y) = \begin{cases} 1/9 & y=0 \\ 2/9 & y=1, 2, 3, 4 \\ 0 & \text{otherwise} \end{cases}$$