Poisson Distribution

discrete RV

- Support:
$$N_0 = \{0, 1, 2, 3, ... \}$$

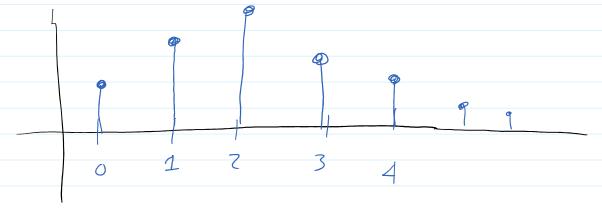
Canonical Experient:

cant the number of events in some time period

of occurences

PMF:

$$f(x) = \frac{e^{-\lambda} x}{x!} \quad \text{for } x = 0, 1, 2, 3, \dots$$



Expected Valve

Expected Valve

$$\mathbb{E}[X] = \sum_{\chi=0}^{\infty} \chi f(\chi) = \sum_{\chi=\phi_1}^{\infty} \chi e^{-\lambda} \chi \frac{\chi}{\chi!} = \frac{1}{(\chi-1)!}$$

$$e = \frac{\chi}{\sum_{i=0}^{\infty} \frac{\chi^{i}}{i!}}$$

$$\chi = \frac{\chi}{2} \frac{\chi}{\chi} = \frac{\chi}{(\chi - 1)}$$

$$\chi = \chi = 1 \frac{\chi}{(\chi - 1)}$$

$$=e^{-\lambda} \frac{\infty}{\chi_{+}} = \lambda e^{-\lambda} \frac{\infty}{\chi_{-}} \frac{\lambda}{\chi_{-}}$$

$$= \lambda e^{-\lambda} \frac{\infty}{\chi_{-}} \frac{\lambda}{\chi_{-}} = \lambda e^{-\lambda} \frac{\infty}{\chi_{-}} \frac{\lambda}{\chi_{-}}$$

$$= \lambda e^{-\lambda} \lambda = \mathbb{E}[X]$$

onsider:
$$\frac{\chi(\chi-1)}{\mathbb{E}\left[\chi(\chi-1)\right]} = \frac{\chi(\chi-1)}{\chi(\chi-1)} = \frac{\chi(\chi-1)}{\chi(\chi-1)(\chi-2)!} = \frac{\chi(\chi-1)}{\chi(\chi-1)(\chi-2)!} = \frac{\chi(\chi-1)}{\chi(\chi-2)!}$$

$$= \sum_{\chi=2}^{\infty} \frac{e^{-\lambda} \chi}{(\chi-2)!}$$

$$= e^{-\lambda} \sum_{\chi=0}^{\infty} \frac{\chi+2}{\chi!}$$

$$= \lambda e^{2} \frac{\lambda}{\chi = 0} \frac{\lambda}{\chi!} = \lambda e^{2} = \lambda^{2}$$

$$= \lambda e^{2} \frac{\lambda}{\chi!} = \lambda e^{2} = \lambda^{2}$$

$$E[X(X-I)] = E[X^2-X] = E[X^2] - E[X] = \lambda^2$$
$$E[X^2] = \lambda^2 + E[X] = \lambda^2 + \lambda$$

$$Var(X) = E[X^2] - E[X]^2 = \lambda^2 + \lambda - \lambda^2 = \lambda^2$$

$$M(t) = E[e^{tx}] = \sum_{x=0}^{\infty} e^{tx} - xx$$

$$\frac{\partial}{\partial y} = e^{y} = e^{y}$$

$$= e^{-\lambda \omega} \underbrace{(xe^{t})}_{\chi=0}^{\chi} = e^{-\lambda} \underbrace{(xe^{t})}_{\chi!}^{\chi} = e^{-\lambda} \underbrace{(xe^{t})}_{\chi!}^{\chi}$$

$$= \exp(\chi(e^{t}-1)) = M(t)$$

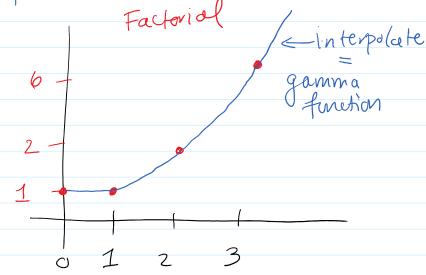
Gamma Distribution:

let's talk about gamma function:

Extends factorial to pos. numbers.

$$\Gamma: \mathbb{R}^+ \to \mathbb{R}^+$$

$$\Gamma(k) = \int_{0}^{\infty} \chi e^{-1-x} dx$$



Properties of P:

1) If k is an integer then

$$\Gamma(k) = (k-1)!$$

$$\int (k+1) = k!$$

$$\frac{\xi_{X}}{\Gamma(1)} = 0! = 1$$

$$\frac{\Gamma(2) = 1! = 1}{\Gamma(3) = 2! = 2}$$

Notice! $\chi! = \chi(\chi-1)!$

So fer an integer k,

$$\Gamma(k) = (k-1)! = (k-1)(k-2)! = (k-1)\Gamma(k-1)$$

This is also true for all k > 0.

Important Facts about ?

(1)
$$\Gamma(k+1) = k\Gamma(k)$$
 or $\Gamma(k) = (k-1)\Gamma(k-1)$

2) If k is an integer,

$$\Gamma(k) = (k-1)! \text{ or } \Gamma(k+1) = k!$$

Gamma Dist:

Shape

$$f(x) = \frac{\lambda e^{-\lambda x} (\lambda x)^{k-1}}{\lambda (k)} \quad \forall x > 0$$

Note: le = 1 then we have EXP(X) f(x)

change k

Expectation:

$$E[X] = \int \frac{x}{x} \frac{\lambda e^{-\lambda x}(\lambda x)^{k-1}}{\Gamma(k)} dx$$

$$= \frac{1}{\lambda} \int \frac{\lambda e^{-\lambda x}}{\lambda e^{-\lambda x}} dx \qquad \frac{\log k}{\log k} \int \frac{\log k}{k} dx$$

$$= \frac{1}{\lambda} \int \frac{\lambda e^{-\lambda x}}{\Lambda(k)} dx \qquad \frac{\log k}{\log k} \int \frac{\log k}{(\lambda x)^{k-1}} dx$$

$$= \frac{1}{\lambda \Gamma(k+1)} \frac{1}{\lambda e} \frac{1}{\lambda e} \frac{1}{\lambda x} \frac{1}{\lambda$$

$$= \frac{1}{\lambda} \frac{\lceil (k+1) \rceil}{\lceil (k) \rceil}$$

$$=\frac{1}{\lambda}\frac{\ell \ell(\ell e)}{\ell'(\ell e)}=\frac{\ell e}{\lambda}=\ell(\chi)$$

What about E(X')

$$\mathbb{E}(X^r) = \int_{C}^{\infty} \frac{\chi^r \lambda e^{-\lambda \chi} e^{-\lambda \chi}}{\Gamma(E)} d\chi$$

$$=\frac{\Gamma(h+r)_{1}}{\Gamma(h)}\int_{0}^{\infty}\frac{\lambda e^{-\lambda x} e^{-\lambda r} x^{+}r^{-1}}{\lambda e^{-\lambda x}} dx$$

Gamma(
$$k+r$$
, λ)
$$\frac{\lambda e^{-\lambda x}}{\Gamma(k+r)}$$

$$=\frac{\Gamma(k+r)}{\Gamma(k)} = E[x^r]$$

$$\frac{\mathcal{E}_{K, r=2}}{\mathbb{E}[\chi^2]} = \frac{\binom{(k+2)}{l}}{\binom{(k)}{k}} = \frac{(k+1)\binom{(k+1)}{l}}{\binom{(k)}{k}}$$

$$= \frac{(k+1)kR(k)}{\Gamma(k)}$$

$$= \frac{(k+1)k}{\lambda^2}$$

$$Var(X) = E[X^{2}] - E[X]^{2}$$

$$= \frac{k(k+1)}{\lambda^{2}} - \left(\frac{k}{\lambda}\right)^{2}$$

$$= \frac{k}{\lambda^{2}} + \frac{k}{\lambda^{2}}$$

Geometric Distribution (discrete)

Canonical Experiment

If I flip coins (independently), each w/ a prob. p of a H, and I do this until I get my first H.

outcome	*
+	1
TH	2
TTH	3
{	

X ~ Geometric (p)

PMF:
$$f(x) = (1-p)^{x-1}$$
 p for $x=0,1,2,3,...$

$$\frac{\text{CDF:}}{\text{F}(x)} = \begin{cases} 1 - (1-p) & , x \ge 1 \\ 0 & \text{else} \end{cases}$$

Pecall: geometric series

$$\sum_{i=0}^{\infty} r^i = \frac{1}{1-r} \quad \text{for} \quad |r| < 1$$

Expected Valve

$$\mathbb{E}[X] = \sum_{x=1}^{\infty} \chi(1-p)^{x-1} p$$

$$= -p \sum_{x=1}^{\infty} \frac{d}{dp}(1-p)^{x}$$

$$= -P \left(\frac{dp}{dp} (1-p) \right) = (1-p) \stackrel{\approx}{\geq} (1-p) = (1-p) = p$$

$$= -P \left(\frac{dp}{dp} (1-p) \stackrel{\perp}{p} \right) = -P \left(\frac{dp}{p^2} \right) = -P \left(\frac{$$

MGF:

$$M(t) = \mathbb{E}\left[e^{tX}\right] = \sum_{x=1}^{\infty} e^{tx} \xrightarrow{x-1}$$

$$= \frac{P}{1-P} \sum_{x=0}^{\infty} (1-p)e^{t} \xrightarrow{x+1}$$

$$= \frac{pe}{(-(1-p)e^{t})} = M(t)$$

So.
$$-\frac{d^2M}{dt}\Big|_{t=0} = \frac{2-p}{p^2} = \mathbb{E}[\chi^2]$$

hence
$$Var(X) = E[X^2] - E(X)^2 = \frac{2-p}{p^2} - (\frac{1}{p})^2$$

$$= \frac{1-p}{p^2} = Var(x)$$