

Sec 3.3- The Γ , χ^2 , and β Distributions

① Exponential and Gamma Dist.

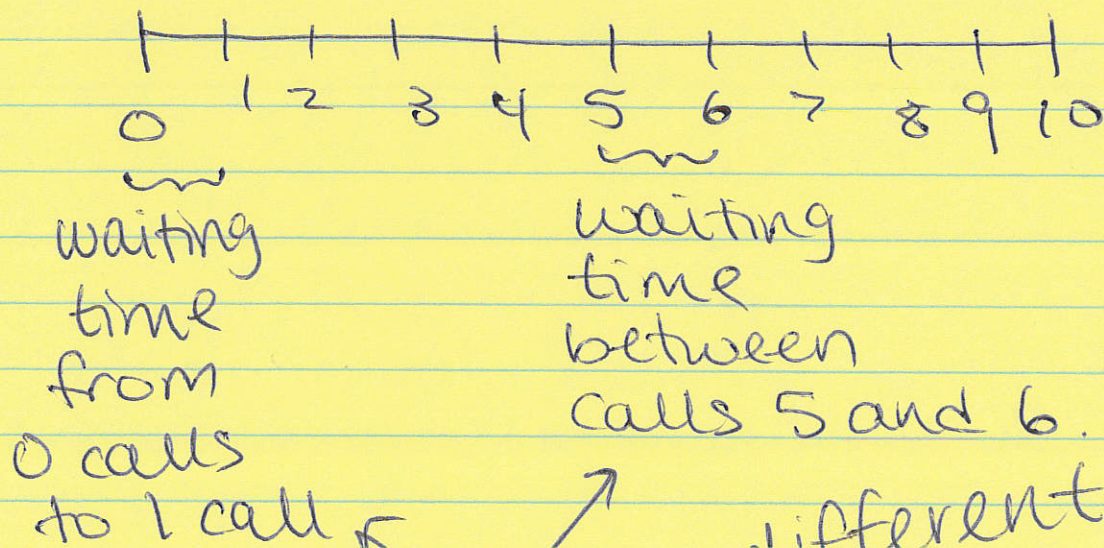
Poisson Dist: # outcomes in a given time interval.

We can look at waiting time between outcomes/events

ex) 10 calls/hr at a helpdesk.
We are interested in a 1 hour time period = w

$\lambda = 10$

$\lambda w = 10 = \mu \leftarrow$ for Poisson.



can be different.

this is a RV. (continuous).

(2)

def) Exponential Dist.

Define RV X to be

X = waiting time for a Poisson process until the 1st change occurs.

X is continuous

$x > 0$

waiting time is positive

The CDF of X =

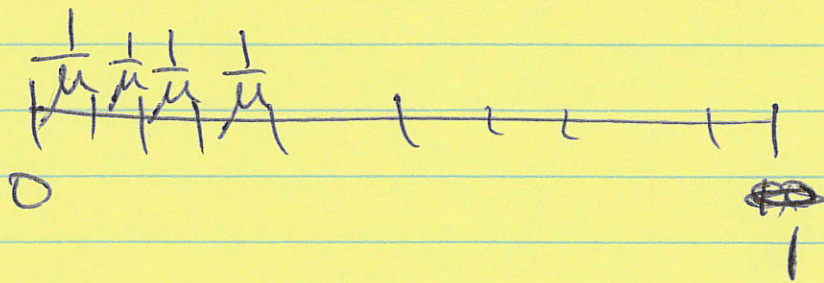
$$F_X(x) = P[X \leq x] = \begin{cases} 1 - e^{-\mu x} & , x > 0 \\ 0 & , \text{o.w.} \\ & (x \leq 0) \end{cases}$$

The PDF of X =

$$f_X(x) = \begin{cases} \mu e^{-\mu x} & , x > 0 \\ 0 & , \text{o.w.} \end{cases}$$

(3)

$$\text{Let } \beta = \frac{1}{\mu}$$



PDF of X :

$$f_X(x) = \begin{cases} \frac{1}{\beta} e^{-\frac{x}{\beta}}, & x > 0 \\ 0, & \text{o.w.} \end{cases}$$

$$\mu = E[X] = \beta$$

$$\sigma^2 = \text{Var}[X] = \beta^2$$

We say $X \sim \text{Exp}(\beta)$

$$\text{MGF } M_X(t) = \frac{1}{1 - \beta t} \quad \text{for } t < \frac{1}{\beta}$$

(4)

def) Gamma Dist.

Define RV X to be

X = waiting time ~~for~~ until the α^{th} change occurs.

Let $\beta = \frac{1}{\mu}$ \nwarrow rate from Poisson.

The PDF of X is

$$f_X(x) = \begin{cases} \frac{1}{\beta^\alpha \Gamma(\alpha)} x^{\alpha-1} e^{-\frac{x}{\beta}} & , x > 0 \\ 0 & , \text{o.w.} \end{cases}$$

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

We say $X \sim \Gamma(\alpha, \beta)$

$\Gamma(\alpha)$ is a Gamma Function

$$\Gamma(\alpha) = \int_0^{\infty} x^{\alpha-1} e^{-x} dx \text{ for } \alpha > 0$$

5

$$\Gamma(1) = 1$$

$$\Gamma(\alpha) = (\alpha-1)! \quad \text{if } \alpha \in \mathbb{Z}^+ \text{ and } \alpha > 1$$

Find using Int. by parts.

MGF for X

$$M_X(t) = \mathbb{E}[e^{tx}]$$

$$= \int_0^{\infty} e^{tx} \frac{1}{\beta^\alpha \Gamma(\alpha)} x^{\alpha-1} e^{-\frac{x}{\beta}} dx$$

$$= \int_0^{\infty} \frac{1}{\beta^\alpha \Gamma(\alpha)} x^{\alpha-1} e^{-x\left(\frac{1}{\beta} - t\right)} dx$$

$\frac{1-\beta t}{\beta}$

Let $y = \frac{1-\beta t}{\beta} x$

$$\Rightarrow x = \frac{y\beta}{1-\beta t} \Rightarrow \frac{\beta}{1-\beta t} dy = dx$$

If $x=0, y=0$

If $x=\infty, y=\infty$

restriction on t!

$$1-\beta t > 0$$
$$t < \frac{1}{\beta}$$

$$= \int_0^{\infty} \frac{1}{\beta^\alpha \Gamma(\alpha)} \left(\frac{\beta}{1-\beta t} \right)^{\alpha-1} y^{\alpha-1} e^{-y \frac{\beta}{1-\beta t}} \frac{\beta}{1-\beta t} dy \quad (6)$$

$$= \int_0^{\infty} \frac{1}{\beta^\alpha \Gamma(\alpha)} \frac{\beta^{\alpha-1}}{(1-\beta t)^{\alpha-1}} y^{\alpha-1} e^{-\frac{\beta y}{1-\beta t}} dy$$

$$= \frac{1}{(1-\beta t)^\alpha} \int_0^{\infty} \frac{1}{\beta^\alpha \Gamma(\alpha)} y^{\alpha-1} e^{-\frac{y}{t}} dy.$$

if $\beta=1$

= 1 by PDF properties

$$= \frac{1}{(1-\beta t)^\alpha} \text{ for } t < \frac{1}{\beta}$$

$$E[X] = \alpha \beta$$

$$\text{Var}[X] = \alpha \beta^2$$

(7)

Thm Let X_1, \dots, X_n be indep RVs

Let $X_i \sim \Gamma(\alpha_i, \beta)$

Let $Y = \sum_{i=1}^n X_i$

then $Y \sim \Gamma\left(\sum_{i=1}^n \alpha_i, \beta\right)$

Proof by MGF

(2) Chi-Square Dist.

→ special case of Gamma where

$$\alpha = \frac{r}{2} \quad \text{and} \quad \beta = 2$$

$$r \in \mathbb{Z}^+$$

The PDF is :

$$f_X(x) = \begin{cases} \frac{1}{2^{r/2} \Gamma(\frac{r}{2})} x^{\frac{r}{2}-1} e^{-\frac{x}{2}}, & x > 0 \\ 0, & \text{o.w.} \end{cases}$$

8

$$E[X] = r$$

$$\text{Var}[X] = \alpha\beta^2 = \frac{r}{2} \cdot 2^2 = 2r$$

$$\text{MGF: } M_X(t) = (1 - 2t)^{-\frac{r}{2}}, \quad t < \frac{1}{2}$$

r = degrees of freedom

we say $X \sim \chi^2(r)$

ex) ~~1~~ ~~2~~ 3 4

Pick #1 : 2

Pick #2 : 4

Pick #3 : 1

Pick #4 : 3

← chosen for me
no choice.

No freedom to choose.

d.f = degrees of freedom = 3

③ Beta Distribution

Why use?

- defined on interval from 0 to 1 ~~is~~ bounded.
- scale or shift to any finite interval of interest.

Where

- PERT (program evaluation + review technique).
- critical path methods
- Joint cost schedule modeling
- Bayesian Inference
- order statistics.

How to Derive this Dist?

- start w/ 2 indep Gamma RVs, then transform.

$$\left. \begin{aligned} X_1 &\sim \Gamma(\alpha, 1) \\ X_2 &\sim \Gamma(\beta, 1) \end{aligned} \right\} \text{indep.}$$

$$f(x_1, x_2) = \begin{cases} \frac{1}{\Gamma(\alpha)\Gamma(\beta)} x_1^{\alpha-1} e^{-x_1} x_2^{\beta-1} e^{-x_2} & \text{for } x_1 > 0, x_2 > 0 \\ 0 & \text{o.w.} \end{cases}$$

by indep.

$$\text{Let } Y_1 = X_1 + X_2$$

$$Y_2 = \frac{X_1}{X_1 + X_2}$$

Step 1 (Find pdf of (Y_1, Y_2) .)

$$y_2(x_1 + x_2) = x_1 = y_1 y_2$$

$$y_1 = x_1 + x_2 = x_2 = y_1 - y_1 y_2 = y_1(1 - y_2)$$

$$J = -y_1$$

$$Dy = \begin{matrix} 0 < y_1 < \infty \\ 0 < y_2 < 1 \end{matrix}$$

$$0 < y_2 < 1$$

$$|J| = y_1$$

$$= |-y_1|$$

$$f_{Y_1, Y_2}(y_1, y_2) = \begin{cases} \frac{1}{\Gamma(\alpha)\Gamma(\beta)} y_1^{\alpha+\beta-1} y_2^{\alpha-1} (1-y_2)^{\beta-1} e^{-y_1} & \text{for } 0 < y_1 < \infty \\ & 0 < y_2 < 1 \\ 0, & \text{o.w.} \end{cases}$$

Step 2: Find marginal pdf for Y_2 .

$$\frac{y_2^{\alpha-1} (1-y_2)^{\beta-1}}{\Gamma(\alpha)\Gamma(\beta)} \int_0^{\infty} y_1^{\alpha+\beta-1} e^{-y_1} dy_1$$

Gamma Funct

$$= \Gamma(\alpha+\beta)$$

Beta Dist

$$f_X(x) = \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha)\Gamma(\beta)} x^{\alpha-1} (1-x)^{\beta-1}$$

$0 < x < 1$

$0, \text{ o.w.}$

$$E[X] = \frac{\alpha}{\alpha + \beta}$$

$$\text{Var}[X] = \frac{\alpha\beta}{(\alpha + \beta + 1)(\alpha + \beta)^2}$$