

Defn: Random Sample

If  $X_1, X_2, X_3, \dots, X_n$  iid  $f$   $f$  pmf/pdf of same univariate dist.  
 then we call  $\{X_i\}_{i=1}^n$  a random sample from  $f$ .  
 $\uparrow$  of size  $n$

Facts: If  $\{X_i\}$  are a random sample  
 then  $\begin{cases} \underline{X} = (X_1, X_2, \dots, X_n)^T \\ \underline{X} \in \mathbb{R}^n \end{cases}$  notation

then

$$\boxed{f_{\underline{X}}(\underline{x}) = f_{\underline{X}}(x_1, x_2, x_3, \dots, x_n)}$$

$$= f_{X_1}(x_1) f_{X_2}(x_2) f_{X_3}(x_3) \dots f_{X_n}(x_n) \quad (\text{independence})$$

$$= f(x_1) f(x_2) f(x_3) \dots f(x_n)$$

$$\boxed{= \prod_{i=1}^n f(x_i)}$$

(identical distributions)

Defn: Statistic

If  $\{X_i\}$  are a RS and  $T: \mathbb{R}^n \rightarrow \mathbb{R}^d$

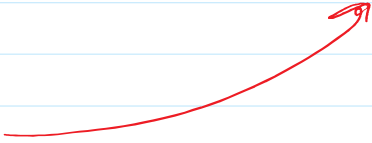
If  $\{X_i\}$  are a RS and  $T: \mathbb{R}^h \rightarrow \mathbb{R}^d$   
then  $T(\underline{X})$  is called a statistic.

Typically  $d < h$  (e.g.  $d=1$ ).

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Ex. (1) Arithmetic mean

$$\bar{X} = \frac{1}{n} (X_1 + X_2 + X_3 + \dots + X_n) = T(\underline{X})$$

here  $T(\underline{X}) = \frac{1}{n} \sum_{i=1}^n X_i$    $d=1$

(2) Sample Variance

$$S^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2$$

(3) Order Statistics

$$X_{(1)} = \text{minimum of } \{X_i\} = \min_{i=1, \dots, n} X_i$$

$$X_{(n)} = \text{maximum of } \{X_i\} = \max_{i=1, \dots, n} X_i$$

$$X_{(r)} = r^{\text{th}} \text{ order statistic} \\ = r^{\text{th}} \text{ smallest value among } \{X_i\}$$

④ range:  $R = X_{(n)} - X_{(1)}$

median: 
$$M = \begin{cases} X_{(\frac{n+1}{2})} & n \text{ odd} \\ \frac{X_{(n/2)} + X_{(n/2+1)}}{2} & n \text{ even} \end{cases}$$

Defn: Sampling Distribution of a Statistic

The sample dist. of a statistic  $T(X)$  is simply the distribution of  $T(X)$

$d=1$   
univariate RV

Ex, Order Statistics

Henceforth:  $\{X_i\}$  are a random sample.

Minimum:  $X_{(1)} = \min_{i=1, \dots, n} X_i$

What is the dist (pmf/pdf) of  $X_{(1)}$ ?

$$\begin{aligned} \underline{P(X_{(1)} > t)} &= P(X_1 > t, X_2 > t, \dots, X_n > t) \\ &= P(X_1 > t) P(X_2 > t) \cdots P(X_n > t) \\ &\quad \text{(independence)} \end{aligned}$$

(independence)

$$\begin{aligned} &= \prod_{i=1}^n P(X_i > t) \\ &= \prod_{i=1}^n (1 - F(t)) \\ &= (1 - F(t))^n \end{aligned}$$

$1 - F(t)$   
 $F = \text{CDF of any } X_i$

$$P(X_{(n)} \leq t) = 1 - (1 - F(t))^n$$

//  
 $F_{X_{(n)}}(t)$

For cts RVs

$$\begin{aligned} f_{X_{(n)}}(t) &= \frac{dF_{X_{(n)}}}{dt} = -n(1 - F(t))^{n-1} \left(-\frac{dF}{dt}\right) \\ &= \boxed{n(1 - F(t))^{n-1} f(t)} \end{aligned}$$

↑ marginal CDF    ↑ marginal pdf

Maximum

$$\begin{aligned} F_{X_{(n)}}(t) &= P(X_{(n)} \leq t) = P(X_1 \leq t, \dots, X_n \leq t) \\ &= P(X_1 \leq t) \dots P(X_n \leq t) \end{aligned}$$

$$= F(t) \cdot \dots \cdot F(t)$$

$$= F(t)^n$$

For cdf:

$$f_{X(n)}(t) = \frac{dF_{X(n)}}{dt} = n F(t)^{n-1} f(t) \quad \leftarrow$$

Ex. Let  $X_i \stackrel{iid}{\sim} \text{Exp}(\lambda)$

$$f(x) = \lambda e^{-\lambda x} \text{ for } x > 0$$

$$F(x) = 1 - e^{-\lambda x} \text{ for } x > 0$$

Minimum?

$$f_{X(n)}(t) = n (1 - F(t))^{n-1} f(t) \quad \text{for } t > 0$$

$$= n (e^{-\lambda t})^{n-1} \lambda e^{-\lambda t}$$

$$= (n\lambda) e^{-\lambda t(n-1)} e^{-\lambda t}$$

$$= \underline{(n\lambda)} e^{-\underline{(n\lambda)}t}$$

this is a pdf of  
an exponential dist  
w/ rate  $n\lambda$

So:

$$X_{(1)} \sim \text{Exp}(n\lambda)$$

Ex:  $X_{(1)} \sim \text{Exp}(n\lambda)$

e.g.  $E[X_{(1)}] = n\lambda$

Maximum:

$$f_{X_{(n)}}(t) = n F(t)^{n-1} f(t)$$

$$= n(1 - e^{-\lambda t})^{n-1} \lambda e^{-\lambda t}$$

General Order Statistics:

$X_{(r)} = r^{\text{th}}$  order stat.  
 $= r^{\text{th}}$  smallest among  $X_1, \dots, X_n$

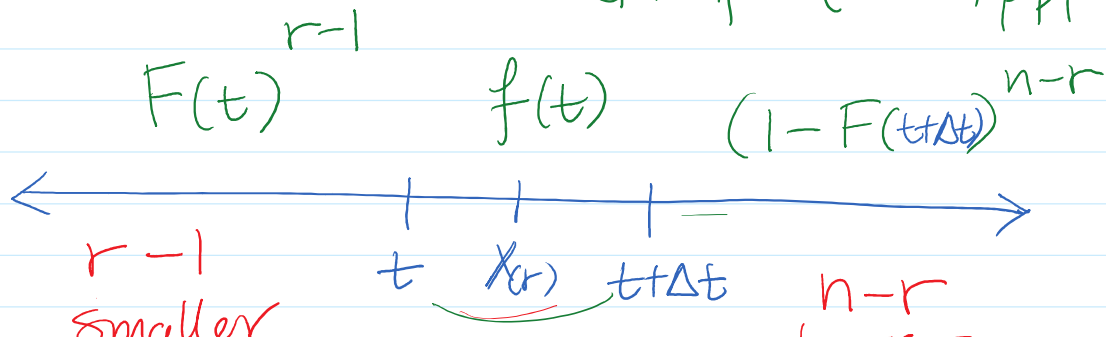
$X_{(n)} \neq X_1$

If  $X_i$  are continuous

$$f_{X_{(r)}}(t) = \frac{n!}{(r-1)!(n-r)!} F(t)^{r-1} (1-F(t))^{n-r} f(t)$$

$\swarrow$  CDF of  $X_i$        $\swarrow$  PDF of  $X_i$

Pf.



$r-1$   
smaller

$t \quad X_{(r)} \quad t+\Delta t$

$n-r$   
larger

$$f_{X_{(r)}}(t) = \lim_{\Delta t \rightarrow 0} \frac{P(t \leq X_{(r)} \leq t + \Delta t)}{\Delta t} \leftarrow$$

$$\lim_{\Delta t \rightarrow 0} \frac{n}{n} \binom{n-1}{r-1} F(t)^{r-1} \left[ \frac{dF/dt}{\Delta t} (F(t+\Delta t) - F(t)) \right] (1 - F(t+\Delta t))^{n-r}$$

$$= \frac{n(n-1)!}{(r-1)!(n-1-(r-1))!} F(t)^{r-1} f(t) (1-F(t))^{n-r}$$

$$= \frac{n!}{(r-1)!(n-r)!} F(t)^{r-1} (1-F(t))^{n-r} f(t)$$

Ex,  $X_i \stackrel{iid}{\sim} U(0,1)$

$$f(t) = 1 \quad \text{for } 0 < t < 1$$

$$F(t) = t \quad \text{for } 0 < t < 1$$

So

$$f_{X_{(r)}}(t) = \frac{n!}{(r-1)!(n-r)!} F(t)^{r-1} (1-F(t))^{n-r} f(t)$$

$$= \frac{n!}{(r-1)!(n-r)!} t^{r-1} (1-t)^{n-r} \cdot 1$$

$$= \frac{t^{r-1} (1-t)^{n-r}}{(r-1)! (n-r)!} \quad \text{for } 0 < t < 1$$

Notice:  $\nearrow$  a Beta

$$X_{(r)} \sim \text{Beta}(r, n-r+1)$$

Theorem: Joint Distribution of Order Stats.

If  $r < s$  and consider  $X_{(r)}$  and  $X_{(s)}$

note:  $X_{(r)} < X_{(s)}$ .

$$f_{X_{(r)}, X_{(s)}}(u, v) = \frac{n!}{(r-1)! (s-r-1)! (n-s)!} \cdot \underbrace{F(u)^{r-1}}_{r-1} \underbrace{(F(v)-F(u))^{s-r-1}}_{s-r-1} \underbrace{(1-F(v))^{n-s}}_{n-s} f(u) f(v)$$

pf.

$$\xleftarrow{\quad \underbrace{F(u)^{r-1}}_{r-1} \quad \underbrace{f(u)}_u \quad \underbrace{(F(v)-F(u))^{s-r-1}}_{s-r-1} \quad \underbrace{f(v)}_v \quad \underbrace{(1-F(v))^{n-s}}_{n-s} \quad \rightarrow$$



smaller  $X_{(r)}$  between  $X_{(s)}$   
1 1

Ex.  $X_i \stackrel{iid}{\sim} U(0,1)$   $\left\{ F(t)=t, f(t)=1 \quad 0 < t < 1 \right\}$

$f_{X_{(r)}, X_{(s)}}(u,v) = \frac{n!}{(r-1)!(n-s)!(s-r-1)!} u^{r-1} (v-u)^{s-r-1} (1-v)^{n-s}$

Ex. U  $\left[ R = X_{(n)} - X_{(1)} \right]$   $X_i \stackrel{iid}{\sim} U(0,1)$   
 What is the dist of  $R$ ?

$V = X_{(1)}$

$(X_{(1)}, X_{(n)}) \xrightarrow{g} (R, V)$

$\xleftarrow{g^{-1}}$

Inverse transf.

$X_{(1)} = g_1^{-1}(R, V) = \underline{V} \quad \parallel \quad X_{(n)} = g_2^{-1}(R, V) = \underline{R+V}$

$\frac{\partial g^{-1}}{\partial (R, V)} = J = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix}$

$|\det J| = 1$

$$f_{R,V}(r,v) = f_{X_{(1)}, X_{(n)}}(v, r+v) \cdot 1$$

Aside :

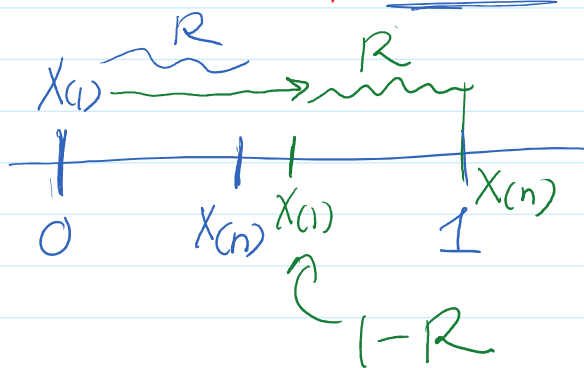
$$f_{X_{(1)}, X_{(n)}}(a,b) = n(n-1)(F(b)-F(a)) \cancel{f(a)} \cancel{f(b)}$$

$(b-a)^{n-2}$   
 $\uparrow \quad \uparrow$   
 $1 \quad 1$

$$f(r,v) = n(n-1)(r+v-v)^{n-2}$$

joint of  $R = X_{(n)} - X_{(1)}$ ,  $V = X_{(1)}$ .

$$= n(n-1)r^{n-2}$$



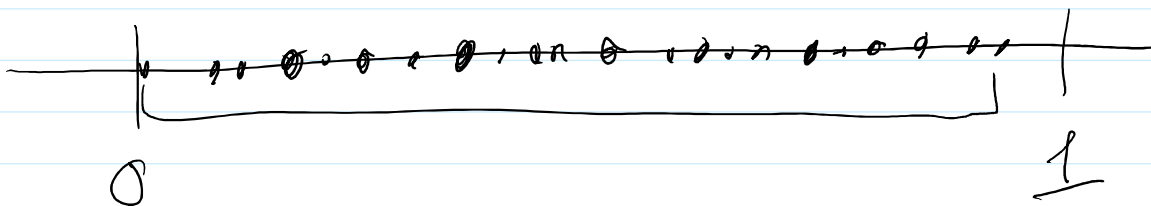
$$f_R(r) = \int_0^{1-r} n(n-1)r^{n-2} dv$$

$$= n(n-1)r^{n-2} \int_0^{1-r} dv$$

$$= n(n-1)r^{n-2}(1-r)$$

$$R \sim \text{Beta}(n, 2)$$

$$E[R] = \frac{n-1}{n+1} \xrightarrow{n \rightarrow \infty} 1$$



Theorem:

Joint dist of all order stats

$$f_{X_{(1)}, X_{(2)}, X_{(3)}, \dots, X_{(n)}}(u_1, u_2, u_3, \dots, u_n) = n! \prod_{i=1}^n f(u_i)$$