

Lecture 5

Apples into barrels



M barrels

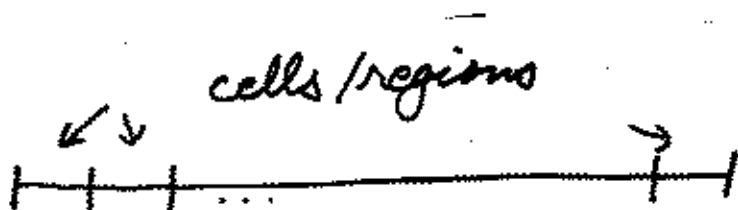
N apples

2 cases $\left\{ \begin{array}{l} \text{apples "identical"} \\ \text{all different} \end{array} \right.$

$\rightarrow M^N$

fewer





N "points" into M cells

If done uniformly & independently then

$$P(Y_1 = k_1, \dots, Y_M = k_M) = \binom{N}{k_1, \dots, k_M} M^{-N}$$

which is the multinomial $(N; \frac{1}{M}, \dots, \frac{1}{M})$ or Maxwell-Boltzmann dist'n ("occupancy statistics")

Remark

(i) If the points are distinguishable then the # of arrangements of the pts in the cells is

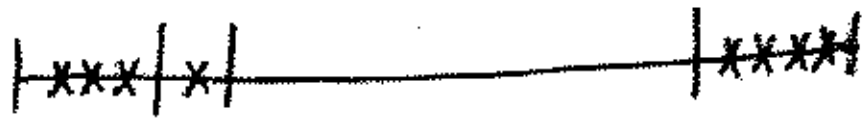
$$\sum_{k_1, \dots, k_M} \binom{N}{k_1, \dots, k_M} = (1 + \dots + 1)^N = M^N$$

Of course different arrangements may lead to the same counts.

(ii) $Y_1 \sim \text{binomial}(N, \frac{1}{M}) \approx \text{Poisson}(\mu)$

if $\frac{N}{M} \approx \mu$ & N is large.

If the points are indistinguishable then the # of arrangements will be fewer. We can count them as follows:



$N - x$'s
 $M - 1$ — | ← longer than the end lines

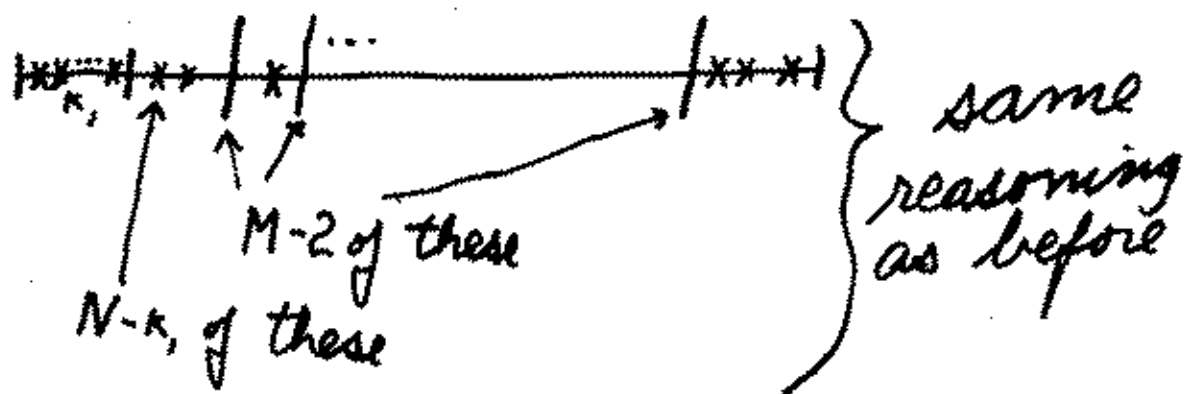
As we move the |'s + x's around we obtain the different ways of putting the N points into the M cells (we are only keeping track of the counts in each cell). This is identical to moving $M - 1$ 1's + N 0's around in $N + M - 1$ cells + asking for the # of arrangements! So we obtain

$$\binom{N + M - 1}{N} \text{ or } \binom{N + M - 1}{M - 1}$$

Notice that each arrangement corresponds to counts in each of the M cells! If equal probability is assigned to each then we have the "Bose-Einstein statistics"

In particular

$$P(Y_1 = k_1) = \frac{\binom{N-k_1+M-2}{M-2}}{\binom{N+M-1}{M-1}}$$



Of course

$$P(Y_1 = k_1) = \frac{\binom{N-k_1+M-2}{N-k_1}}{\binom{N+M-1}{N}}$$

if $\frac{N}{M} \approx \lambda$ & $N \rightarrow \infty$ then

$$P(Y_1 = k_1) \approx \left(\frac{\lambda}{1+\lambda}\right)^{k_1} \left(\frac{1}{1+\lambda}\right), \quad k_1 = 0, 1, \dots$$

which is the geometric (starting at 0) distribution.

Extra material (basic conditioning, a review)

Def'm If $P(A) > 0$ define $E(Y|A) = \frac{E[Y I(A)]}{E[I(A)]}$.

It is easily seen that, for fixed A , $E(\cdot|A)$ satisfies the Axioms for E (verify this).

Def'm $P(B|A) = E(I(B)|A)$

Of course $P(B|A) = P(BA)/P(A)$ as before

If \tilde{X} is discrete then we set

$$r(x) = E(Y | \tilde{X} = x)$$

and

$$E(Y | X) = r(\tilde{X})$$

If Y is also discrete then it is easily verified

$$E(Y) = E[E(Y | \tilde{X})]$$

and

$$E(Y | \tilde{X}_1) = E[E(Y | \tilde{X}_1, \tilde{X}_2) | \tilde{X}_1]$$

A further property is

$$E[Y g(\tilde{X})] = E[E(Y | \tilde{X}) g(\tilde{X})], \quad \forall \text{ real } g$$

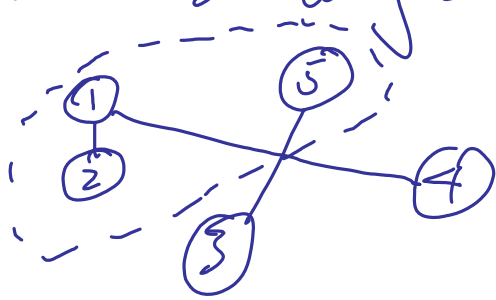
These last three properties are true in general and are easily verified when $\begin{pmatrix} Y \\ \tilde{X} \end{pmatrix}$ is its with $\exists d \forall (y, x)$.

Finally, it can be shown that $E(Y | \tilde{X})$ satisfies (see Th 5.3.1)

$$E(Y - E(Y | \tilde{X}))^2 = \min_{\text{real } h} E(Y - h(\tilde{X}))^2$$

Application of the uniform + Bernoulli / Indicator

N - towns } at most 1 road between
 m - roads } any 2 towns



$\frac{2}{3}$ of roads
go into to
dotted collection

In fact there is always a collection
of towns \rightarrow at least half of roads
lead into it.

Sol'n Go to each town + toss a
fair coin. If H put town into
a collection S . S is a random
collection of towns. Note that there
are 2^N possible "values" that S can be.
Let $X = \#$ of roads leading into S
from outside. We need to show there is

a possible value of $X \geq m/2$. Since X is a (positive) counting rv this will hold if $E(X) \geq m/2$. Now, let

$$I_j = I(\{\text{road } j \text{ leads into } S\})$$

Then

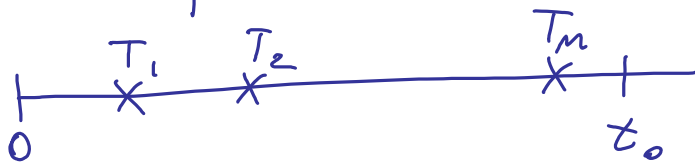
$$X = \sum_{j=1}^m I_j$$

$$\begin{aligned} * E(X) &= m E(I_1) = m P(\text{road 1 leads into } S) \\ &= m P(\text{one H + one T for the 2 tosses}) \\ &= m/2 \end{aligned}$$

Let $\{N(t) : t \geq 0\}$ be a Poisson process of rate λ on $t \geq 0$.

Suppose you know $N(t_0) = m$. Call the

times of the points $T_1 < T_2 < \dots < T_m$



We want the pdf of T_m .

Approach

Use def F or $\bar{F} = 1 - F$

↖ Aside

$$F(x) = P(X \leq x), \quad \forall x \in \mathbb{R}$$

For a cts r.v. X

$$F(x) = \int_{-\infty}^x \underset{\substack{\uparrow \\ \text{pdf}}}{f(t)} dt$$

\Rightarrow at continuity pts

$$F'(x) = f(x)$$

In fact we could take F' to be an equivalent pdf.

$$F(b) - F(a) = P(a < X \leq b)$$

$$\left. \begin{array}{l} \{0, P \\ \emptyset \end{array} \right\} \left\{ \overbrace{P(X \leq a)}^{F(a)} + P(a < X \leq b) = \overbrace{P(X \leq b)}^{F(b)} \right\}$$
$$= \int_a^b F'(z) dz \quad \boxed{\text{FLAC}}$$

Remark If F is cto it is possible Feller
 for $F'(x) = 0$ for almost every x
 (places where not true have length 0)
 + so in that case $\int_{-\infty}^{\infty} F'(x) dx = 0$.

$\bar{F}'(x) = -f(x)$. So knowing F or \bar{F}
 \Rightarrow know the dist'n.

By the way,

$$x_1 < x_2 \Rightarrow P(X \leq x_1) \leq P(X \leq x_2)$$

$$\circ \circ \{X \leq x_1\} \Rightarrow \{X \leq x_2\}$$

$\circ \circ F$ is increasing.

$$\text{If } x_n \downarrow a \Rightarrow \{X \leq x_n\} \downarrow \{X \leq a\}$$

$$\Rightarrow P(X \leq x_n) \rightarrow P(X \leq a)$$

(continuity property of P)

$\circ \circ F$ is right cto

$$\begin{aligned} \downarrow x_m \uparrow \infty &\Rightarrow \{X \leq x_m\} \uparrow \{X < \infty\} \\ &\Rightarrow F(x_m) \rightarrow P(X < \infty) = 1 \end{aligned}$$

$$\therefore \underbrace{x \rightarrow \infty \Rightarrow F(x) \rightarrow 1}_{F(\infty) = 1}$$

$$\begin{aligned} \downarrow x_m \downarrow -\infty &\Rightarrow \{X \leq x_m\} \downarrow \emptyset \\ &\Rightarrow F(x_m) \rightarrow 0 \end{aligned}$$

$$\therefore \underbrace{x \rightarrow -\infty \Rightarrow F(x) \rightarrow 0}_{F(-\infty) = 0}$$

Note ① $x_m \uparrow a \Rightarrow \{X \leq x_m\} \uparrow \{X < a\}$

$$\therefore \lim_{x \uparrow a} F(x) = P(X < a) \neq F(a)$$

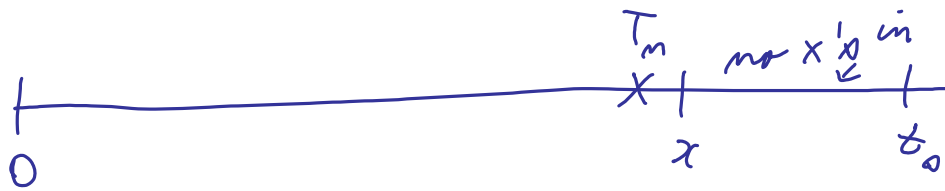
$$P(X = a) = F(a) - \lim_{x \uparrow a} F(x)$$

② Knowing $F \Rightarrow$ know $P(a \leq X \leq b)$
not obvious
 \Rightarrow know the dist ln \checkmark

Back to our Poisson process.



$$F_{(m)}(x) = P(T_m \leq x) \left\{ = P(T_m \leq x \mid N(t_0) = m) \right\}$$



$$= P(N((x, t_0]) = 0 \mid N(t_0) = m)$$

$$= \frac{P(N((x, t_0]) = 0, N(t_0) = m)}{P(N(t_0) = m)}$$

$$= \frac{P(N(x) = m, N((x, t_0]) = 0)}{P(N(t_0) = m)}$$

$$= \frac{P(N(x) = m) P(N((x, t_0]) = 0)}{P(N(t_0) = m)}$$

$$\begin{aligned}
&= \frac{P(N(x) = m) P(N((x, t_0]) = 0)}{P(N(t_0) = m)} \\
&= \frac{e^{-\lambda x} \frac{(\lambda x)^m}{m!} e^{-\lambda(t_0 - x)}}{e^{-\lambda t_0} \frac{(\lambda t_0)^m}{m!}} \\
&= \left(\frac{x}{t_0}\right)^m
\end{aligned}$$

∴ conditional on $N(t_0) = m$,

$$F_{(m)}(x) = \left(\frac{x}{t_0}\right)^m, \quad 0 \leq x \leq t_0$$

$$= 1, \quad x > t_0$$

$$= 0, \quad \text{otherwise}$$

QED

$$f_{(m)}(x) = \frac{m}{t_0} \left(\frac{x}{t_0}\right)^{m-1}, \quad 0 \leq x \leq t_0$$

$$= 0, \quad \text{otherwise}$$

if $t_0 = 1$ then

$$f_{(m)}(x) = m x^{m-1}, \quad 0 \leq x \leq 1$$

$$= 0, \quad \text{otherwise}$$

Remark

$$r(x) = E(Y | X=x) = \sum_y y \underbrace{f(y|x)}_{\substack{\uparrow \\ \text{discret}}} = \sum_y y \underbrace{f(x,y)}_{\substack{\uparrow \\ f(x)}} \underbrace{P(X=x)}_{\substack{\uparrow \\ P(X=x)}}$$

$$E(Y|X) = r(X)$$

$$E[E(Y|X)] = E(Y)$$

$$E[E(z^Y|X)] = E(z^Y)$$

$$r(x) = E(Y|X=x) = \int_{-\infty}^{\infty} y \underbrace{f(y|x)}_{\text{cond pdf}} dy$$

can be extended to the cts case + then

stochastic process = collection of random elements

Gaussian process = collection of rv's \Rightarrow dist's are normal

Markov property

$\{X_t : t \in T\}$
↑ time ↑ index set

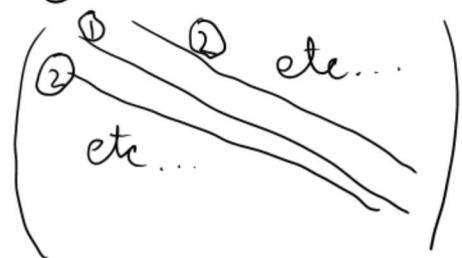
For each $t_0 \in T$,

$$g(\{X_t : t > t_0\}) | \{X_t : t \leq t_0\}$$

" = " $g(\{X_t : t > t_0\}) | X_{t_0}$ "A" g

stationary process: $\begin{pmatrix} X_{t_1} \\ \vdots \\ X_{t_k} \end{pmatrix} \stackrel{d}{=} \begin{pmatrix} X_{t_1+\Delta} \\ \vdots \\ X_{t_k+\Delta} \end{pmatrix}$ "A" t, Δ

Toeplitz matrix



elements on ① same
" " ② same

square

Poisson processes & order stats

sample X_1, \dots, X_m
order stats $X_{(1)} < \dots < X_{(m)}$
(assume cts)

Poisson process

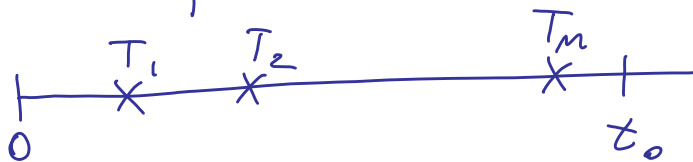
- throw points \rightarrow Poisson process
- have process via some other way & condition \rightarrow throw jets!

So

Let $\{N(t) : t \geq 0\}$ be a Poisson process of rate λ on $t \geq 0$.

Suppose you know $N(t_0) = m$. Call the

times of the points $T_1 < T_2 < \dots < T_m$

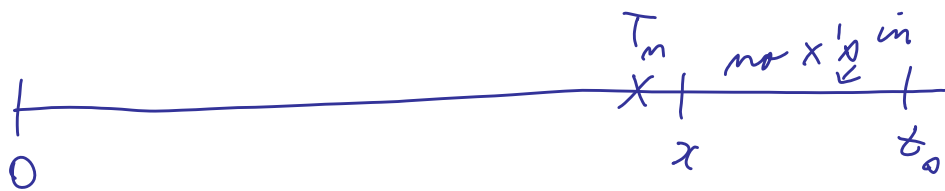


We want the pdf of T_m .

Back to our Poisson process.



$$F_{(m)}(x) = P(T_m \leq x) \left\{ = P(T_m \leq x \mid N(t_0) = m) \right\}$$



≡ } conditional calculations

$$F_{(m)}(x) = \left(\frac{x}{t_0}\right)^m, \quad 0 \leq x \leq t_0$$

$$= 1, \quad x > t_0$$

$$= 0, \quad \text{otherwise}$$

q so

$$f_{(m)}(x) = \frac{m}{t_0} \left(\frac{x}{t_0}\right)^{m-1}, \quad 0 \leq x \leq t_0$$

$$= 0, \quad \text{otherwise}$$

if $t_0 = 1$ then

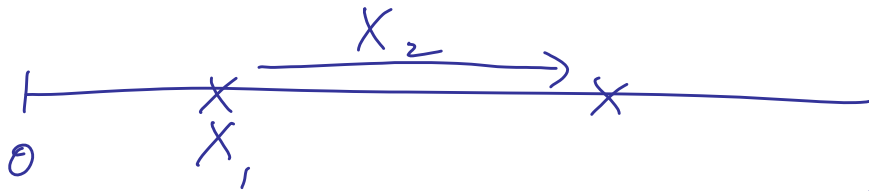
$$f_{(m)}(x) = m x^{m-1}, \quad 0 \leq x \leq 1$$

$$= 0, \quad \text{otherwise}$$

As we will see this is what we get when looking at the order statistics from a uniform.

First Interarrival times ind?
 conditional pdf / pdf / mgf / pgf / df = unconditional
 \Rightarrow ind

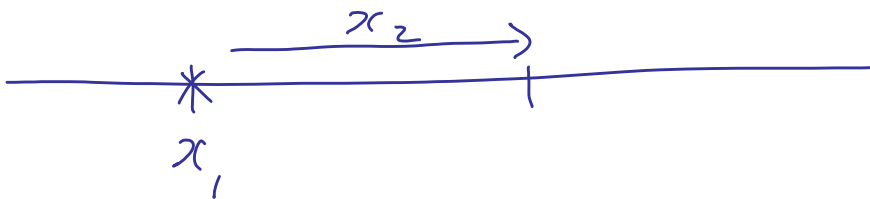
eg Poisson process, rate λ on $t \geq 0$



We know $X_1 \sim \text{exponential}(\lambda)$

Look at

$$P(X_2 > x_2 \mid X_1 = x_1)$$



$$= P\left(\underbrace{N((x_1, x_1 + x_2]) = 0}_{\text{ind}} \mid \underbrace{X_1 = x_1}_{\text{ind}}\right)$$

$$= P(N((x_1, x_1 + x_2]) = 0)$$

$$= e^{-\lambda x_2}$$

which is the tail probability f'm of an

exponential(λ).

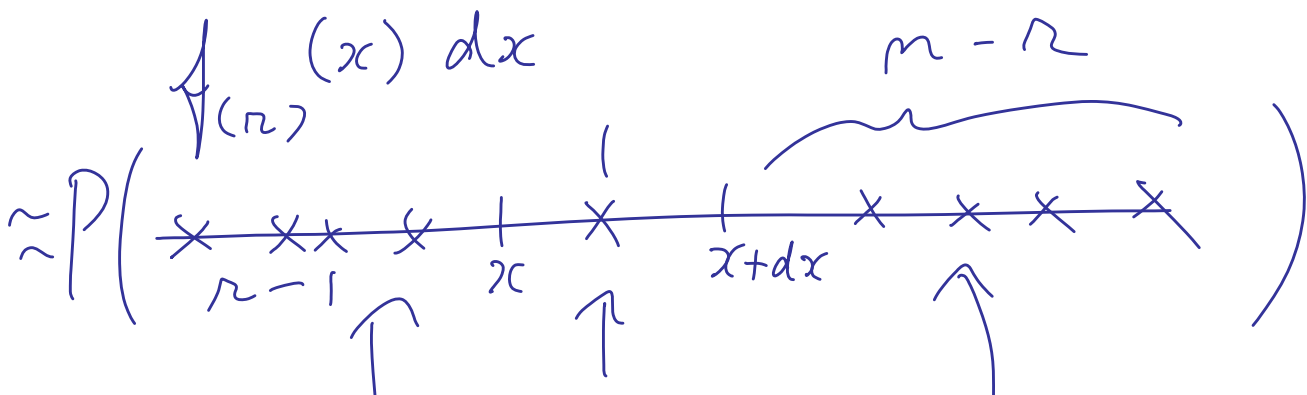
$\therefore X_1, \dots, X_m$ are independent & the unconditional tail probability f'm is as above. That is, X_1, X_2 are iid exponential(λ). Continue to get X_1, X_2, \dots iid exponential

Order Stats

X_1, X_2, \dots, X_m iid $\begin{matrix} \text{pdf} \\ \downarrow \\ dF \end{matrix}$

$X_{(1)} < \dots < X_{(m)}$ ← order stats

pdf of $X_{(r)}$



$$F(x) \approx \int f(x) dx \approx \bar{F}(x)$$

$$\approx \binom{m}{r-1, 1, m-r} F(x)^{r-1} \int f(x) dx \bar{F}(x)^m$$

"

$$f_{(r)}(x) \approx \binom{m}{r-1, 1, m-r} F(x)^{r-1} \int f(x) \bar{F}(x)^m$$

For a uniform $(0, 1)$

$$f_{(r)}(x) = m x^{m-1}, \quad 0 < x < 1$$