

# Week #4

## binomial

icd Bernoulli trials yielding

$$\begin{matrix} X_1 & X_2 & \dots & X_N \\ \sim & \sim & & \sim \\ 2 \times 1 & 2 \times 1 & & 2 \times 1 \end{matrix}$$

$\rightarrow$  icd  $\tilde{X}$

Note possible values of  $\tilde{X}$  are  $\begin{pmatrix} p_2 \\ 0 \\ 1 \end{pmatrix}$  or  $\begin{pmatrix} p_1 \\ 1 \\ 0 \end{pmatrix}$   $p_1 + p_2 = 1$

jt pgf of the components of  $\tilde{X}$  or just the

the pgf of  $\tilde{X}$

$$G_{\tilde{X}}(\underline{z}) = E(\underline{z}^{\tilde{X}}) = E\left(\begin{matrix} \text{1st comp of } \tilde{X} \\ z_1 \\ \text{2nd comp of } \tilde{X} \\ z_2 \end{matrix}\right)$$

$$= z_2 p_2 + z_1 p_1$$

$$= p_1 z_1 + p_2 z_2$$

= pgf of a "vector Bernoulli"

Now set

$$\underline{Y} = \underline{X}_1 + \dots + \underline{X}_N$$

$$\Rightarrow G_{\underline{Y}}(\underline{z}) = (p_1 z_1 + p_2 z_2)^N = \sum_{y_1 + y_2 = N} \binom{N}{y_1, y_2} p_1^{y_1} p_2^{y_2} z_1^{y_1} z_2^{y_2}$$

$$\Rightarrow P(\underline{Y} = \underline{y}) = \binom{N}{y_1, y_2} p_1^{y_1} p_2^{y_2}$$

Note  $z_2 = 1 \Rightarrow (p_1 z_1 + p_2)^N$  is the pgf of the 1st component of  $\underline{X}$  (this is the binomial  $(N, p_1)$  as you know it).

Extend to the multinomial

$$\underbrace{X_1}_{M \times 1}, \underbrace{X_2}_{M \times 1}, \dots, \underbrace{X_M}_{M \times 1} \quad \text{i.i.d. } \underline{X}$$

$\underline{X}$  has one component = 1 + the rest 0. The probability that the 1 is in the  $i$ th place we call  $p_i$  ( $i=1, \dots, M$ ). The pgf of  $\underline{X}$  is

$$(p_1 z_1 + p_2 z_2 + \dots + p_M z_M)$$

& letting

$$\underline{Y} = \underline{X}_1 + \dots + \underline{X}_M$$

we get

$$G_{\underline{Y}}(\underline{z}) = (p_1 z_1 + \dots + p_M z_M)^N$$

$$\Rightarrow P(\underline{Y} = \underline{y}) = \underbrace{\binom{N}{y_1}}_{\binom{N}{y_1, \dots, y_M}} p_1^{y_1} \dots p_M^{y_M}$$

Note  $\frac{\partial^2}{\partial z_1 \partial z_2} G(\underline{z}) = E(Y_1 z_1^{Y_1-1} Y_2 z_2^{Y_2-1} z_3^{Y_3} \dots)$  } works for all counting vectors

$\stackrel{z=1}{=} E(Y_1, Y_2)$

etc...  
 Can use to get the cov( $Y_i, Y_j$ ) for counting vec's  $\downarrow$ .

Properties of covariances ( $\text{cov}(X, Y) = E(XY) - E(X)E(Y)$ )

vein

$$\left\{ \begin{array}{l} \text{cov}(X, X) = \text{Var}(X) \\ \text{cov}(X+c, Y+d) = \text{cov}(X, Y) \\ \text{cov}(aX, bY) = ab \text{cov}(X, Y) \\ \text{cov}\left(\sum_i X_i, \sum_j Y_j\right) = \sum_{i,j} \text{cov}(X_i, Y_j) \end{array} \right.$$

eg  $X_1 \sim \text{Poisson}(\lambda_1), X_2 \sim \text{Poisson}(\lambda_2), X_3 \sim \text{Poisson}(\lambda_3)$

$$\left. \begin{array}{l} U = X_1 + X_2 \\ V = X_2 + X_3 \end{array} \right\} \text{easy to get the pgf}$$

$$E(z_1^U z_2^V) = E(z_1^{X_1+X_2} z_2^{X_2+X_3})$$

$$\begin{aligned}
&= E(z_1^{X_1} (z_1 z_2)^{X_2} z_2^{X_3}) \\
&= E(z_1^{X_1}) E(z_1 z_2)^{X_2} E(z_2^{X_3}) \\
&= e^{\lambda_1(z_1-1)} e^{\lambda_2(z_1 z_2 - 1)} e^{\lambda_3(z_2-1)}
\end{aligned}$$

- Use this to get  $E(UV)$  - imp } 15 minutes  
 & then get  $\text{cov}(U, V)$

Another way

$$\begin{aligned}
\text{cov}(U, V) &= \text{cov}(X_1 + X_2, X_2 + X_3) \\
&= \text{cov}(X_1, X_2) + \text{cov}(X_1, X_3) \\
&\quad + \text{cov}(X_2, X_2) + \text{cov}(X_2, X_3) \\
&= \text{cov}(X_2, X_2) = \text{Var}(X_2) = \lambda_2
\end{aligned}$$

Note  $X, Y$  ind  $\Rightarrow E(XY) = E(X)E(Y) \Rightarrow \text{cov}(X, Y) = 0$

eg  $Z \sim N(0, 1)$ . Let  $X = Z$ ,  $Y = Z^2$ . Then

$X$  &  $Y$  are dependent but

$$E(XY) = E(Z Z^2) = E(Z^3) = 0$$

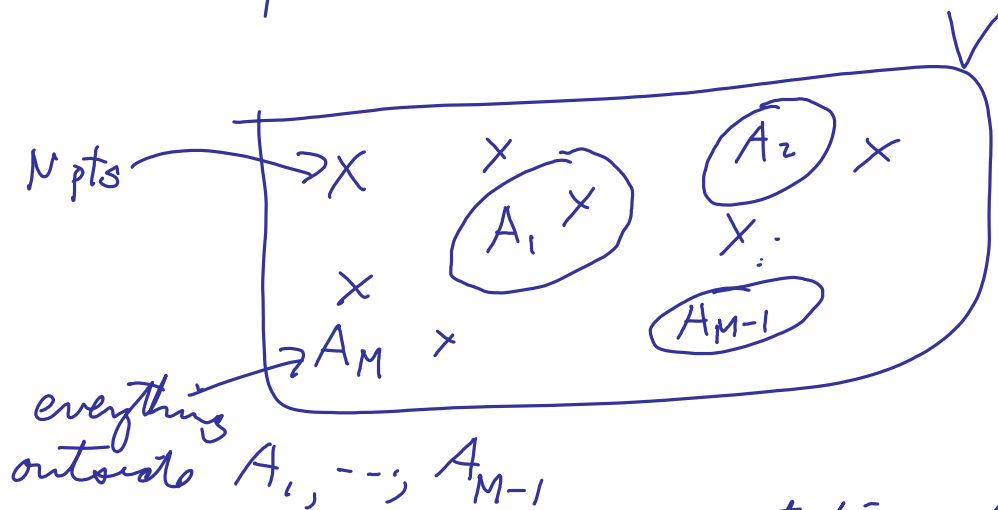
& so  $X$  &  $Y$  are uncorrelated.

$$\begin{array}{cc}
E(X) & E(Y) \\
0 & 1
\end{array}$$

# Spatial Poisson process

Background A rvec is uniform on a set if its Pdf or pdf is constant there.

Throw  $N$  points onto  $V$  in a uniform way



$\{A_1, \dots, A_{M-1}, A_M\}$  is a partition of  $V$ .

Let  $N(A) = \#$  of pts in a set  $A \subset V$ .  
 $\sim$  binomial  $(N, |A|/|V|)$

Also  $N(A_1), \dots, N(A_M)$  is our multinomial with  $p_i = |A_i|/|V|$

+ the pdf is

$$(p_1 z_1 + \dots + p_{M-1} z_{M-1} + p_M z_M)^N$$

$\Rightarrow$  p.g.f. of  $N(A_1), \dots, N(A_{M-1})$

is  $(p_1 z_1 + \dots + p_{M-1} z_{M-1} + p_M)^N$   
 $\downarrow$   
 $= 1 - (p_1 + \dots + p_{M-1})$

Let  $V \rightarrow \mathbb{R}^{(d)}$   
and  $N \rightarrow \infty$  } keep  $\frac{N}{|V|} = \rho$  fixed  
 $\downarrow$   
called  $\lambda$  eventually  
 $\rho$  is like a density of pts

In the limit you get a Poisson point process on  $\mathbb{R}^{(d)}$  of "rate"  $\rho$ .

Poisson spatial process, Poisson counting process, Poisson process.

Why is it called Poisson?

The j'th p.g.f. of  $N(A_1), \dots, N(A_{M-1})$  is

$$(*) \left[ p_1(z_1-1) + \dots + p_{M-1}(z_{M-1}-1) + 1 \right]^N$$

$$* p_i = \frac{|A_i|}{|V|} = \frac{|A_i| N}{|V| N} = \frac{\rho |A_i|}{N}$$

so that (\*) is

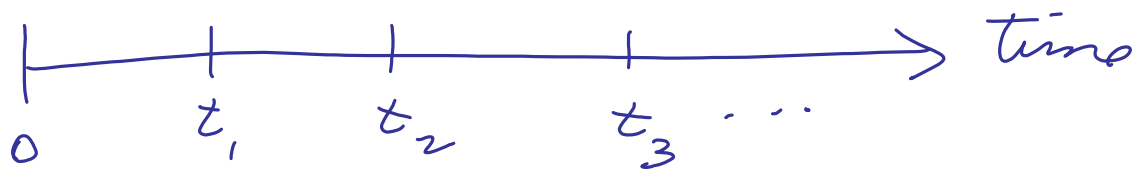
$$\left[ 1 + \frac{\rho |A_1| (z_1-1) + \rho |A_2| (z_2-1) + \dots + \rho |A_{M-1}| (z_{M-1}-1)}{N} \right]^N$$

$$\xrightarrow{N \rightarrow \infty} e^{\rho |A_1| (z_1-1) + \dots + \rho |A_{M-1}| (z_{M-1}-1)}$$

= part of  $M-1$  independent Poisson's  
with means  $\rho |A_1|, \dots, \rho |A_{M-1}|$

Note For a Poisson process the # 's  
in disjoint regions are independent  
Poisson's & the # in a set  $A$  is  
Poisson with mean rate  $\times |A|$   
 $\rho$  or  $\lambda$   
 $\uparrow$   
 usual

Special case Poisson process of rate  $\lambda$  on  $t \geq 0$ .



$\Downarrow$   $t_0 < t_1 < t_2 < \dots$   $\wedge$   $N(t) = \#$  of points in  $[0, t]$

then  $N(t_i) - N(t_{i-1}) = \#$  of pts in  $(t_{i-1}, t_i]$ ,  $i=1, 2, \dots$

$\{N(t); t \geq 0\}$  is a Poisson counting process of rate  $\lambda$  on  $t \geq 0$ .

Note that the  $N(t_i) - N(t_{i-1})$  are independent (independent increments).

Times between pts are rv's } want to study these  
Time to the  $n$ th pt is a rv }



Let  $S_n =$  time to the  $n$ th pt +  
 $X_1, X_2, \dots$  be the times between pts  
 $\uparrow$  time from 1st pt to 2nd  
time from 0 to the 1st pt

dist'n of  $X_1$

Let  $F(x) = P(X_1 \leq x)$  — df of  $X_1$   
 $\bar{F}(x) = P(X_1 > x)$  — Tail probability  
of  $X_1$

Assume  $F$  (or  $\bar{F}$ ) determine the dist'n  
(which it does but hard to prove).

Here

$$F(x) = 0 \text{ unless } x > 0$$

Let  $x > 0$ . Then

$$\bar{F}(x) = P(X_1 > x) = P(N(x) = 0)$$

But

$$N(x) \sim \text{Poisson}(\lambda x)$$

$$\Rightarrow P(N(x) = 0) = e^{-\lambda x}$$

$$\begin{aligned} \therefore \bar{F}(x) &= e^{-\lambda x}, & x > 0 \\ &= 1, & \text{ow} \end{aligned}$$

Notice  $\bar{F}'(x) = -\lambda e^{-\lambda x}, x > 0$

$$* f(x) = \lambda e^{-\lambda x}, x > 0$$

which is the exponential ( $\lambda$ ) pdf. So  $X_1 \sim \text{exponential}(\lambda)$

Note If a pdf exists then  $\bar{F}'(x) = -f(x)$ .  
 The exponential( $\lambda$ ) dist'n, by def'n, has pdf

$$f(x) = \begin{cases} \lambda e^{-\lambda x}, & x > 0 \\ 0, & \text{ow} \end{cases} \left. \vphantom{f(x)} \right\} \text{mean} = \frac{1}{\lambda}$$

If  $\bar{F}'(x)$  is a pdf then we are dealing with an absolutely cts dist'n.

For  $S_n =$  time to  $n$ th point

$$\bar{F}_n(x) = P(S_n > x)$$

Let  $x > 0$  then

$$\bar{F}_r(x) = P(S_r > x) = P(N(x) \leq r-1)$$

$$= \sum_{k=0}^{r-1} e^{-\lambda x} \frac{(\lambda x)^k}{k!}$$

Now calculate  $\bar{F}_r'(x)$  to get - the pdf of a gamma (check this).

Extra material (basic conditioning, a review)

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

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

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

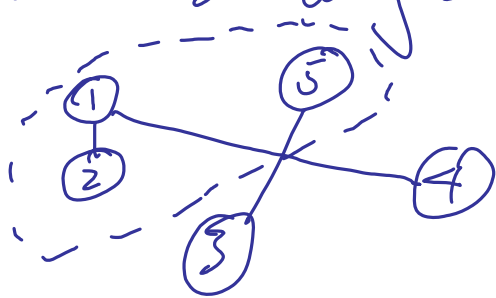
These last three properties are true in general and are easily verified when  $\begin{pmatrix} Y \\ \tilde{X} \end{pmatrix}$  is its with  $\exists d \forall (y, x)$ .

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

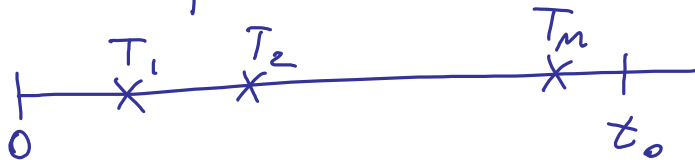
$$X = \sum_{j=1}^m I_j$$

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

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

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

$\Rightarrow$  at continuity pts

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

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

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

$$\left. \begin{array}{l} \{0, P \\ \emptyset \end{array} \right\} \overbrace{P(X \leq a) + P(a < X \leq b)}^{F(a)} = \overbrace{P(X \leq b)}^{F(b)} \left. \right\} \\ = \int_a^b F'(t) dt \quad \boxed{\text{FLAC}}$$

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

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

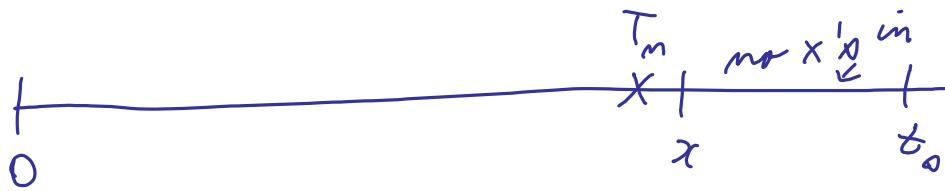
$$P(X = a) = F(a) - \lim_{x \uparrow a} F(x)$$

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

$\nearrow f(x,y)$

$\downarrow f(x)$

$\uparrow P(X=x)$

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

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

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

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

can be extended to the cts case + then