

$X \sim \text{Pois}(\lambda)$
 $Y \sim \text{Exp}(\theta)$ we need to invoke
 limiting argument
 on $P(X|Y)$

§ 5.1 Ex

$X \sim \text{Pois}(\lambda)$

$\lambda \sim \text{Exp}(\theta)$

$\theta > 0$

$$P(X=k) = \int_{\alpha=0}^{\alpha=\infty} P(X=k | \lambda=\alpha) P(\lambda=\alpha) d\alpha$$

via the law of total probability.

$$P(X=k) = \int_{\alpha=0}^{\alpha=\infty} \frac{(\alpha)^k e^{-\alpha}}{k!} \theta e^{-\theta} d\alpha$$

We recognize what may be a gamma kernel

$$\Gamma(a, b)(\alpha) = \frac{b}{\Gamma(a)} (b\alpha)^{a-1} e^{-b\alpha}$$

$$b = \theta \quad a = k+1 \Rightarrow \Gamma(k+1, \theta)(\alpha) = \frac{\theta}{k!} \theta^k \alpha^k e^{-\theta\alpha}$$

$$\int_{\alpha=0}^{\alpha=\infty} \frac{\theta}{k!} (\alpha)^k e^{-\alpha(1+\theta)} d\alpha . \quad \text{We recognize what}$$

is a Γ kernel. Let's see if we can evaluate integral using that. Let $Y \sim \Gamma(k+1, \theta+1)$, then

$$f_Y(\alpha) = \frac{(1+\theta)}{k!} (\alpha(1+\theta))^k e^{-(1+\theta)\alpha} \quad \text{using}$$

$$\int_{\alpha=0}^{\alpha=\infty} f_Y(\alpha) d\alpha = 1 \quad \text{we see} \quad P(X=k) = \frac{\theta}{(1+\theta)^{k+1}}$$

§ 5.1 86

$X(t)$ P.P. $\lambda = 3$

a) $P(X(4) - X(0) = 0)$ is what were

looking for. $X(4) - X(0) \sim \text{Pois}(4\lambda) \therefore$

$$P(X(4) - X(0) = 0) = e^{-4\lambda}$$

b) The distribution of the first afternoon message
is the waiting time to first event

$$P(X(k)=1) = \lambda e^{-\lambda} \sim \exp(\lambda)$$

$$\text{S 5.1} \quad P1 \quad X_0^{\xi} \quad \Omega = \mathbb{N} \\ \xi \quad \Omega = [\xi, \infty)$$

$$\xi_i + \text{r.v.} \quad \xi_i \sim \text{Exp}(\lambda) \quad \sum_{j=1}^k \xi_j \sim F(k, \lambda)$$

Let

$$X = \begin{cases} 0 & 1 < \xi_1 \\ 1 & \xi_1 < 1 < \xi_1 + \xi_2 \\ 2 & \xi_1 + \xi_2 < 1 < \xi_1 + \xi_2 + \xi_3 \\ \vdots & \end{cases}$$

$$k \quad \sum_{j=1}^{k-1} \xi_j < 1 \leq \sum_{j=1}^k \xi_j$$

Now let's look at $P(X=k)$

$$P(X=0) = P(\xi_1 > 1) = 1 - P(\xi_1 \leq 1) = 1 - F_{\xi_1}(1)$$

recall $F_{\xi_1}(x) = 1 - e^{-\lambda x}$

$$P(X=0) = e^{-\lambda} \quad \text{which agrees with Pois}(\lambda) \quad k=0$$

Turned the page & saw the rest of the hint!

Set

$$P(X=k) = P\left(\sum_{j=1}^k \xi_j < 1 < \sum_{j=1}^{k+1} \xi_j\right) \quad (\text{let } A = \sum_{j=1}^k \xi_j)$$

$$= P(1 < \xi_{k+1} + \alpha \mid \sum_{j=1}^k \xi_j < 1) \\ = \int P(X=k \mid A=\alpha) P(A=\alpha) d\alpha \quad \text{what limits}$$

make sense for integral? $[0, 1]$ since

$\text{if } A > 1 \quad P(X=k) = 0 \quad \text{and} \quad (\Omega_{\xi_i} = [\xi, \infty)) \text{ gives}$

lower limit

$$P(X=k) = \int_0^\infty P(X=k \mid A=\alpha) P(A=\alpha) d\alpha$$

§ 5.4 P1

remember $\alpha = \sum_{j=1}^k \xi_j$

(P2)

Now $P(X=k | A=\alpha, \alpha \in [0,1]) =$

$$P(\alpha < 1 < \alpha + \xi_{k+1}, \alpha \in [0,1])$$

$$P(\xi_{k+1} > 1 - \alpha) = 1 - F_\xi[1 - \alpha]$$

Using $\alpha = \sum_{j=1}^k \xi_j \sim \text{Gamma}(k, \lambda)$ we have

$$P(X=k) = \int_0^1 P(X=k | A=\alpha) P(A=\alpha) d\alpha$$

$$= \int_0^1 P(X=k | A=\alpha, \alpha \in [0,1]) P(A=\alpha) d\alpha$$

$$= \int_0^1 (1 - F_\xi[1 - \alpha]) \Gamma(k, \lambda)(\alpha) d\alpha$$

Where $\Gamma(k, \lambda)(\alpha) = \frac{(\lambda x)^{\alpha-1}}{\Gamma(\alpha)} e^{-\lambda x}$ is pdf of

Gamma(k, λ), and

$F_\xi[x] = 1 - e^{-\lambda x}$ is CDF of $\text{Exp}(\lambda)$ r.v.

§ 5.1 PS

$$X(h) \sim \text{Pois}(\lambda h) \quad h > 0$$

$$P_k(h) = P(X(h)=k) = \frac{(\lambda h)^k e^{-\lambda h}}{k!}$$

a) $P_0(h) = e^{-\lambda h}$ expanding the Taylor series

of $e^{-\lambda h}$ about 0 ($\lim_{h \rightarrow 0}$) we have

$$e^{-\lambda h} = \sum_{n=0}^{\infty} \frac{(-\lambda h)^n}{n!} = 1 - \lambda h + O(h)$$

$$\lim_{h \rightarrow 0} \frac{1 - P_0(h)}{h} = \lim_{h \rightarrow 0} \frac{1 - [1 - \lambda h + O(h)]}{h} = \lim_{h \rightarrow 0} \frac{\lambda h + O(h)}{h} = \lambda$$

b) $P_1(h) = \lambda h e^{-\lambda h} = \lambda h [1 - \lambda h + O(h)] \Rightarrow$

$$\lim_{h \rightarrow 0} \frac{P_1(h)}{h} = \lim_{h \rightarrow 0} \frac{\lambda h [1 - \lambda h + O(h)]}{h}$$

$$= \lim_{h \rightarrow 0} \lambda - \lambda^2 h + O(h) = \lambda$$

c) $P_2(h) = \frac{(\lambda h)^2}{2} e^{-\lambda h}$

$$\lim_{h \rightarrow 0} \frac{P_2(h)}{h} = \frac{(\lambda h)^2}{2} \underbrace{[1 - \lambda h + O(h)]}_{h} =$$

$$\lim_{h \rightarrow 0} \frac{\lambda^2 h^2}{2} - \lambda^3 h^2 + (\lambda h)^2 O(h) = 0$$

1 5.2 82

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3 Let $P = \frac{240}{600}$ prob of error

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5 Consider P.P. ($N \cdot p$) approx to

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7 $S_n = \varepsilon_1 + \dots + \varepsilon_n$ $\varepsilon_i \sim \text{Bernoulli}(p)$

8

9 $Y \sim \text{Binom}(3, p)$ $P(Y=0)$ is probability

10

11 of 3 successive error free pages.

12

13 $P(Y=0) \approx P(X \text{ (P.P.) } 3 \cdot p \text{ has no}$

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15 arrivals in first period) by the Poisson

16

17 approximation. Let X be P.P. $3 \cdot p$

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19 $P(X(1)=0) = e^{-3p} = e^{-1.2}$

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21 To me $240/600$ is not small. Also

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23 how are errors distributed? What if they
24

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are all on the same page.

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$$\S 5.2 \quad X_{(n,p)} \sim \text{Binom}(n,p)$$

$$\begin{array}{l} n \rightarrow \infty \\ p \rightarrow 0 \end{array} \Rightarrow np = \lambda$$

$$a) \text{ Show } \lim_{n \rightarrow \infty} P(X_{(n,p)} = 0) = e^{-\lambda}$$

$$P(X_{(n,p)} = k) = \binom{n}{k} p^k (1-p)^{n-k}$$

$$= \frac{n!}{k!(n-k)!} p^k (1-p)^{n-k}$$

\Rightarrow

~~$$\frac{n!}{k!(n-k)!} p^k (1-p)^{n-k}$$~~

$$\frac{n!}{(n-k)!} \xrightarrow{n \rightarrow \infty} 1 \quad (\checkmark)$$

$$p = \frac{\lambda}{n} \Rightarrow (1-p)^{n-k} = \left(1 - \frac{\lambda}{n}\right)^n \xrightarrow{n \rightarrow \infty} e^{-\lambda}$$

$$\lim_{n \rightarrow \infty} \frac{n!}{k!(n-k)!} \frac{\lambda^k}{n^k} \frac{\left(1 - \frac{\lambda}{n}\right)^n}{\left(1 - \frac{\lambda}{n}\right)^k}$$

$$= \frac{\lambda^k}{k!} \lim_{n \rightarrow \infty} \frac{n!}{(n-k)! n^k} \frac{\left(1 - \frac{\lambda}{n}\right)^n}{\left(1 - \frac{\lambda}{n}\right)^k} \leftarrow$$

$$= \frac{\lambda^k}{k!} e^{-\lambda}$$

$$\lim P(X_{(n,p)} = 0) = e^{-\lambda}$$

§ 5.2 p1

$X(n, p) \sim \text{Binom}(n, p)$

$$P(X(n, p) = k+1) = \frac{n!}{(k+1)! (n-(k+1))!} p^{(k+1)} (1-p)^{n-(k+1)}$$

$$P(X(n, p) = k) = \frac{n!}{k! (n-k)!} p^k (1-p)^{n-k}$$

$$\frac{P(X(n, p) = k+1)}{P(X(n, p) = k)} = \frac{\frac{k! (n-k)!}{(k+1)! (n-(k+1))!} \frac{p^{k+1}}{p^k} \frac{(1-p)^{n-(k+1)}}{(1-p)^{n-k}}}{\frac{(n-k)}{(k+1)} \frac{p}{1-p}} = \frac{n p - k p}{(k+1) (1-p)}$$

$$\lim_{n \rightarrow \infty} \frac{P(X(n, p) = k+1)}{P(X(n, p) = k)} = \lim_{n \rightarrow \infty} \frac{n p - k p}{(k+1) (1-p)} \quad \cancel{(k+1) (1-p)}$$

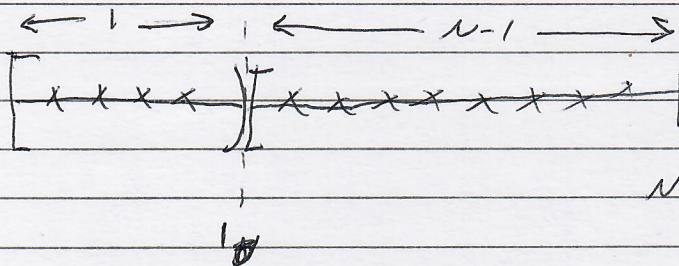
~~$\frac{k}{n} = p$~~

$$= \lim_{n \rightarrow \infty} \left[\frac{\cancel{k}}{(k+1)(1-\frac{\cancel{k}}{n})} + \frac{k \frac{\cancel{k}}{n}}{(k+1)(1-\frac{\cancel{k}}{n})} \right]$$

~~\cancel{k}~~

$$= \frac{\cancel{k}}{k+1} \quad \text{Second term} \rightarrow 0$$

§ 5.2 P4



Let $X = \# \text{ points}$
in $(0, 1)$

$P(X = k) = ?$ There are 2 bins

$(0, 1) [1, n]$ If a point falls in $(0, 1)$ call it
a success.

$$P(\text{Point falls in } (0, 1)) = 1/N$$

$$P(\text{Point falls in } [1, N]) = \frac{N-1}{N} = 1 - \frac{1}{N}$$

Let $p = 1/N$ since we have to account
for the ways in which $X=k$, we see

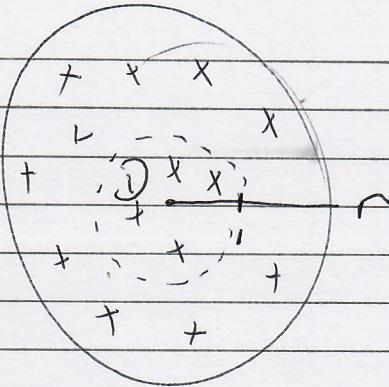
$X \sim \text{Binom}(N, p)$ and now we can
take a limit

$$P(X = k) = \binom{N}{k} p^k (1-p)^{N-k}$$

$$\lim_{N \rightarrow \infty} P(X = k) = \lim_{N \rightarrow \infty} \frac{N!}{k!(N-k)!} \left(\frac{1}{N}\right)^k \left(1 - \frac{1}{N}\right)^{N-k}$$

$$= \lim_{N \rightarrow \infty} \frac{N! N^k}{k!(N-k)!} \left(\frac{1 - \frac{1}{N}}{N}\right)^N = \frac{e^{-1}}{k!} \sim \text{exp}(1) \text{ Pois}(1)$$

S. 2 PS



Let $X = \# \text{ points}$
in D

Same setup as prob 5.2 (let $D = B(1)$)
 $A = B(r) \setminus B(1)$ annulus disk unit radius

$$|D| = \pi \quad |A| = \pi(r^2 - 1)$$

$$P(\text{point falls in } D) = \frac{|D|}{|A|} \quad P(\text{point falls in } A) = \frac{|A|}{\pi r^2}$$

~~πr^2~~

$$= \frac{\pi(r^2 - 1)}{\pi r^2} = \left(1 - \frac{1}{r^2}\right)$$

Taking into account the ways

in which points can be arranged in A, D .

$$P(X=k) = \binom{N}{k} \left(\frac{1}{r^2}\right)^k \left(1 - \frac{1}{r^2}\right)^{N-k}$$

$$\lim_{\substack{N \rightarrow \infty \\ r \rightarrow \infty}} \frac{N!}{\pi r^2} = \lambda \quad \left\{ \frac{N!}{k!(N-k)!} \left(\frac{1}{r^2}\right)^k \left(1 - \frac{1}{r^2}\right)^{N-k} \right.$$

$$\frac{1}{r^2} = \frac{2\pi}{N}$$

$$\lim_{\substack{N \rightarrow \infty \\ r \rightarrow \infty}} \frac{N!}{\pi r^2} = \left(\frac{2\pi}{N}\right)^k \left(1 - \frac{2\pi}{N}\right)^{N-k}$$

$$\frac{1}{k!} \left(\frac{2\pi}{N}\right)^k \lim_{\substack{N \rightarrow \infty \\ r \rightarrow \infty}} \frac{\left(1 - \frac{2\pi}{N}\right)^N}{\left(1 - \frac{2\pi}{N}\right)^k} = \frac{\left(\frac{2\pi}{N}\right)^k}{k!} e^{-2\pi} \sim \text{Pois}\left(\frac{2\pi}{N}\right)$$

5.3. 89

1 W_n - waiting time to n^{th} event

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$$3 \quad W_n = t \Leftrightarrow X(t) = N(0, t) = n$$

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5 $W_r \leq t$ means r events happened before t

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7 This $\Rightarrow X(t) \geq r$. Similarly if we've

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9 observed more than r events in $[0, t)$

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11 we know the waiting time time is less than

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13 t . We can determine cdf of $P(n, \lambda)$ as

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15 sum of Poisson

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$$17 \quad P(W_n \leq t) = P(X(t) \geq n) = 1 - P(X(t) < n)$$

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$$19 \quad = 1 - \sum_{i=0}^{n-1} \frac{\lambda^i e^{-\lambda}}{i!}$$

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1 5.3 P1

2
3 $X(t) \text{ P.P. } \Rightarrow \text{ Show } \{W_1 > w_1, W_2 > w_2\}$ 4
5 iff $\{X(w_1) = 0, X(w_2) - X(w_1) \in \{0, 1\}\}$ 6
7 $W_1 > w_1$ means no arrivals by w_1 , ie $X(w_1) = 0$ 8
9 $W_2 > w_2$ means we have not observed 210
11 arrivals by time $w_2 \Leftrightarrow 0$ arrivals12
13 or 1 arrival between w_1 and w_2 14
15 $P(W_1 > w_1, W_2 > w_2) = P(X(w_1) = 0, X(w_2) - X(w_1) \in \{0, 1\})$ 16
17 Invoking \dagger increments for PP18
19 $P(W_1 > w_1, W_2 > w_2) = P(X(w_1) = 0)P(X(w_2) - X(w_1) \in \{0, 1\})$ 20
21 $X(w_1) \sim \text{Pois}(w_1, \lambda)$ $X(w_2) - X(w_1) \sim \text{Pois}(w_2 - w_1, \lambda)$ 22
23 $P(X(w_1) = 0) = e^{-\lambda w_1}$ 24
25 $P(X(w_2) - X(w_1) \in \{0, 1\}) = e^{-\lambda(w_2 - w_1)} +$ 26
27 $\lambda(w_2 - w_1) e^{-\lambda(w_2 - w_1)}$ union of \neq events

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(P2)

1 5.3 P2

$$2 \frac{\partial^2}{\partial \omega_1 \partial \omega_2} P(X(\omega_1=0)) P(X(\omega_2) - X(\omega_1) \in \xi_{0,13})$$

$$5 \frac{\partial^2}{\partial \omega_1 \partial \omega_2} e^{-\gamma \omega_1} [1 + \gamma(\omega_2 - \omega_1)] e^{-\gamma \omega_2} e^{\gamma \omega_1} =$$

$$8 \frac{\partial^2}{\partial \omega_1 \partial \omega_2} [1 + \gamma(\omega_2 - \omega_1)] e^{-\gamma \omega_2} =$$

$$11 \frac{\partial}{\partial \omega_2} -\gamma e^{-\gamma \omega_2} = \gamma^2 e^{-\gamma \omega_2}$$

5.3 p5

$$X(t) \text{ P.P. } T \sim \text{Exp}(\theta)$$

Find pmf of $X(T)$

$$P(X(T) = k) = P(X(T) = k | T)$$

$$= \int_{t=0}^{\infty} P(X(t) = k | T=t) P(T=t)$$

use result from 5.1, ES

$$P(X(t) = k | T=t) = \frac{(\lambda t)^k e^{-\lambda t}}{k!}$$

$$P(T=t) = \theta e^{-\theta t}$$

$$P(X(T) = k) = \int_{t=0}^{t=\infty} \frac{(\lambda t)^k e^{-\lambda t}}{k!} \theta e^{-\theta t} dt$$