

# POISSON PROCESSES WITH RANDOM ARRIVAL RATE<sup>1</sup>

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**1. Introduction.** Let  $F$  be a distribution function on  $(0, \infty)$ . A probability  $P_F$  on the integers defined by

$$P_F(n) = (n!)^{-1} \int_0^\infty e^{-\lambda} \lambda^n dF, \quad n \geq 0,$$

will be called a mixture of Poisson probabilities. Since [1] there is a 1-1 correspondence between  $P_F$  and  $F$ , any statistical question about  $F$  can, in principle, be answered by random sampling on  $P_F$ . However,  $F$  can be estimated more easily by random sampling on mixtures of laws of Poisson processes (to be defined below). Even then no unbiased estimate for  $F$  exists; but the Glivenko-Cantelli Lemma [2], p. 20 does hold for the natural estimate of a continuous  $F$ . These two results are proved in Section 3; Section 2 contains some preliminary material.

**2. Independent realizations of mixtures.** Let  $\gamma$  be a nonempty set, and  $B(\gamma)$  a  $\sigma$ -algebra of subsets of  $\gamma$ . Let  $\{P_{\lambda'} : \lambda' \in \Lambda\}$  be a family of probabilities defined on  $B(\gamma)$ . Take  $B(\Lambda)$  to be the smallest  $\sigma$ -algebra of subsets of  $\Lambda$  over which all the functions  $\{P_{\lambda'}(E) : E \in B(\gamma)\}$  are measurable. If  $\mu$  is any probability on  $B(\Lambda)$ , define

$$P_\mu(E) = \int_\Lambda P_{\lambda'}(E) d\mu; E \in B(\gamma).$$

The set function  $P_\mu$  is again a probability on  $B(\gamma)$ , and is called a mixture of the probabilities  $P_{\lambda'}$ . If  $X$  is any  $B(\gamma)$ -measurable function, and  $P$  any probability on  $B(\gamma)$ , define  $E(X | P) = \int_\gamma X dP$ . Then

LEMMA 1.  $E(X | P_\mu) = \int_\Lambda E(X | P_{\lambda'}) d\mu$ , in the sense that if either side exists, both do and they are equal.

PROOF. When  $X$  is the characteristic function of a measurable set, the lemma is a restatement of the definition. Hence the lemma holds for all simple functions by linearity, for nonnegative functions by a monotone passage to the limit, and finally for general functions by linearity.

The purpose of the next lemma is to describe mixtures on product spaces. Define ([2], pp. 90-91)

$$\begin{aligned} (\gamma^J, B(\gamma^J), P^J) &= \prod_{j=1}^J (\gamma, B(\gamma), P) \\ (\Lambda^J, B(\Lambda^J), \mu^J) &= \prod_{j=1}^J (\Lambda, B(\Lambda), \mu), \end{aligned}$$

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where  $J$  is any natural number or  $\infty$ . If  $s \in \gamma^J (\lambda \in \Lambda^J)$ , its  $j$ th coordinate will be written  $s_j(\lambda_j)$ . The convention adopted here (for typographical reasons) is the following:  $\lambda$  is a point of a product space whose factors are  $\Lambda$ ;  $\lambda'$  and  $\lambda_j$  are points of  $\Lambda$ . Let  $P_\lambda = \prod_{j=1}^J P_{\lambda_j}$  for  $\lambda \in \Lambda^J$ .

LEMMA 2. For any  $E \in B(\gamma^J)$ ,  $P_\lambda(E)$  is  $B(\gamma^J)$ -measurable and

$$P^J(E) = \int_{\Lambda^J} P_\lambda(E) d\mu^J.$$

PROOF. Both assertions hold for (finite dimensional) measurable rectangles by Fubini's theorem, and are preserved under complementation and countable unions of the sets  $E$  for which they are true.

LEMMA 3. For each  $j \in Z$  (the set of nonnegative integers) let  $\{Y_{j,n} : n \in Z\}$  be a stochastic process on  $(\gamma^\infty, B(\gamma^\infty))$ . Then  $\lim_{j,n \rightarrow \infty} Y_{j,n} = 0$  a.e.  $[P_\mu^\infty]$ , if and only if, for almost all sequences  $\lambda \in \Lambda^\infty[\mu^\infty]$ ,  $\lim_{j,n \rightarrow \infty} Y_{j,n} = 0$  a.e.  $[P_\lambda]$ .

PROOF. For  $m \in Z$  let  $E_m = \{s : s \in \gamma^\infty \text{ and for any } j_0 \text{ and } n_0 \text{ there exist } j > j_0 \text{ and } n > n_0 \text{ with } |Y_{j,n}(s)| > 1/m\}$ . Then each statement below is equivalent to the one following it.

- (i)  $Y_{j,n} \rightarrow 0$  a.e.  $[P_\mu^\infty]$ .
- (ii)  $P_\mu^\infty(E_m) = 0$ , all  $m$ .
- (iii)  $\int_{\Lambda^\infty} P_\lambda(E_m) d\mu^\infty = 0$ , all  $m$ .
- (iv)  $P_\lambda(E_m) = 0$ ,  $[\mu^\infty]$ , all  $m$ .
- (v)  $P_\lambda(E_m) = 0$ , all  $m$ ,  $[\mu^\infty]$ .
- (vi)  $Y_{j,n} \rightarrow 0$  a.e.  $[P_\lambda][\mu^\infty]$ .

**3. Poisson processes.** Let  $B(Z)$  denote the family of all subsets of  $Z$ , and define  $(\gamma, B(\gamma)) = \prod_1^\infty (Z, B(Z))$ . If  $s \in \gamma$ , its  $n$ th coordinate will be written  $s_n$ . Define (using [2], p. 93, Theorem A)  $P_{\lambda'}$  as the probability on  $B(\gamma)$  making  $X_n(s) = s_n$ ,  $n \geq 1$ ,  $X_0(s) = 0$  a Poisson process with parameter  $\lambda'$  for  $\lambda' \in \Lambda = (0, \infty)$ ;  $B(\Lambda)$  is easily seen to consist of the Borel subsets of  $\Lambda$ . If  $\mu$  is a probability on  $B(\Lambda)$ , its distribution function will be denoted by  $F$ .

When  $P_\mu$  is constructed on  $B(\gamma)$ , the law of the process  $\{X_n\}$  is called a mixture of laws of Poisson processes, with mixing distribution  $\mu$ .

Recall that  $(\gamma^\infty, B(\gamma^\infty)) = \prod_1^\infty (\gamma, B(\gamma)) = \prod_1^\infty \prod_1^\infty (Z, B(Z))$  so that  $\gamma^\infty$  is the space of infinite matrices of nonnegative integers.

For  $s \in \gamma^\infty$ ,  $s_j$  is the  $j$ th coordinate of  $s$  and is a point of  $\gamma$ . Hence  $s_{j,n}$  is the  $n$ th coordinate of the  $j$ th of  $s$ , and is an integer (namely, the entry in the  $j$ th row and  $n$ th column of  $s$ ). Define  $X_{j,n}(s) = s_{j,n}$ ,  $n \geq 1$ ;  $X_{j,0}(s) = 0 : j \geq 1$ . In the balance of the paper, the probability on  $(\gamma^\infty, B(\gamma^\infty))$  will be  $P_\mu^\infty$ .

Less formally, there is an unknown random mechanism which selects a parameter  $\lambda_j \in \Lambda = (0, \infty)$  according to the prior probability  $\mu$ . This is done repeatedly and independently, which corresponds to selecting a point  $\lambda = \{\lambda_1, \lambda_2, \dots\}$

from  $\Lambda^\infty$  according to probability  $\mu^\infty$ . Once a  $\lambda_j$  has been selected, the statistician observes the evolution of a discrete-time Poisson process  $\{X_{j,n} : n = 1, 2, \dots\}$  with arrival rate  $\lambda_j$ . The processes are independent in  $j$ . These processes are defined on the common probability space  $(\gamma^\infty, B(\gamma^\infty), P_\mu^\infty)$ , and are a sequence of independent realizations of a process whose law is a mixture of Poisson laws (with mixing distribution  $\mu$ ). Having observed  $\{X_{j,n} : 1 \leq n \leq N\}$  for  $1 \leq j \leq J$ , the statistician wishes to make inferences about  $\mu$ . The first theorem gives a limitation on the type of estimates available.

**THEOREM 1.** *For fixed  $x > 0$ ,  $F(x)$  has no unbiased estimate measurable on a finite number of the  $X_{j,n}$ .*

**PROOF.** Let  $J$  be the largest  $j$ -subscript available, and for  $1 \leq j \leq J$  let  $n_j$  be the largest  $n$ -subscript for each  $j$ . Thus  $1 \leq J < \infty, 0 \leq n_j < \infty$ .

It suffices to consider functions of the sufficient statistic  $\{X_{j,n_j} : 1 \leq j \leq J\}$ . Let  $T$  be a function on  $Z^J$ , and suppose that, contrary to the theorem,

$$E[T(X_{j,n_j} : 1 \leq j \leq J) | P_\mu^J] = F(x),$$

for all  $\mu$  (or even all  $\mu$  with carriers of  $J$  points).

By Lemmas 1 and 2, the finiteness of  $E(T | P_\mu^J)$  implies the finiteness of  $E(T | P_\lambda)$  for almost all vectors  $\lambda \in \Lambda^J[\mu^J]$ . Since this holds for all  $\mu$ ,  $|E(T | P_\lambda)| < \infty$  for all  $\lambda \in \Lambda^J$ . Then  $E(T | P_\lambda)$  is  $\exp(-n_1\lambda_1 - \dots - n_J\lambda_J)$  times a multiple power series, absolutely and uniformly convergent on any bounded set in  $\Lambda^J$ ; and is a continuous function of  $\lambda$ .

Let  $\mu_y$  assign mass 1 to  $y > 0 : F_y$  being the corresponding distribution function. Then  $\int_{\Lambda^J} E(T | P_\lambda) d\mu_y^J = F_y(x)$ , and the left-hand side is continuous, as a function of  $y$ , while the right-hand side, as a function of  $y$ , is discontinuous at  $y = x$ ; a contradiction which completes the proof.

On the other hand, for large  $n$ ,  $\{n^{-1}X_{j,n} : 1 \leq j \leq J\}$  is approximately a random sample from  $F$ , and the sample cumulative distribution function provides a natural estimate for  $F$ . Let  $f(y) = f(y, x) = 1(y \leq x)$  and  $O(y > x)$  and put  $F_{J,n}(x) = J^{-1} \sum_{j=1}^J f(n^{-1}X_{j,n})$ .

**THEOREM 2.** *If  $F$  is continuous at  $x$ ,  $\lim_{J,n \rightarrow \infty} F_{J,n} = F(x)$  a.e.  $[P_\mu^\infty]$ .*

**PROOF.** By Lemma 3, this is equivalent to showing that  $F_{J,n}(x) \rightarrow F(x)$  a.e.  $[P_\lambda]$  for almost all  $[\mu^\infty]$  sequences  $\lambda \in \Lambda^\infty$ .

Choose  $\epsilon > 0$  and  $\delta > 0$  so that  $F(x + \delta) - F(x - \delta) < \epsilon$ . The idea of the proof is to discard  $\lambda_j \in (x - \delta, x + \delta)$ , committing only a small error. Outside this region the Markov inequality gives sharp enough estimates to secure the theorem.

The construction is in terms of the following functions (whose dependence on  $\epsilon, \delta$ , and  $x$  is understood):

$$\begin{aligned} f_1(\lambda) &= 1, \lambda \leq x - \delta; = 0, & \lambda > x - \delta \\ f_2(\lambda) &= 0, \lambda < x + \delta; = 1, & \lambda \geq x + \delta \\ f_3(\lambda) &= 1, \lambda \in (x - \delta, x + \delta); = 0, & \lambda \notin (x - \delta, x + \delta) \end{aligned}$$

$$\begin{aligned}
 u_i(\lambda) &= \lambda(\lambda - x)^{-4}f_i(\lambda), & i = 1, 2 \\
 v_i(\lambda) &= \lambda^2(\lambda - x)^{-4}f_i(\lambda), & i = 1, 2
 \end{aligned}$$

$$Q(n, \lambda) = P_\lambda(X_{j,n} \leq nx).$$

By the strong law of large numbers [2], p. 239,  $J^{-1} \sum_{j=1}^J \xi(\lambda_j) \rightarrow \int_\Lambda \xi d\mu$  a.e.  $[\mu^\infty]$  simultaneously for  $\xi = f_i, i = 1, 2, 3, = u_i, = v_i, i = 1, 2$ , since these functions are bounded and therefore summable. Let  $N_{\epsilon, \delta}$  be the exceptional  $\mu^\infty$ -null set. Select a sequence  $\epsilon_k \rightarrow 0$  with corresponding  $\delta_k \rightarrow 0$ . Put  $N = \bigcup_{k=1}^\infty N_{\epsilon_k, \delta_k}$ , so that  $\mu^\infty(N) = 0$ . In what follows,  $\lambda \in \Lambda^\infty - N$ , while  $(\epsilon, \delta)$  takes values in the sequence  $(\epsilon_k, \delta_k)$ . Two estimates of  $P(n, \lambda)$  are required. By the Markov inequality [2], p. 158, with  $r = 4$  and  $(n^{-1}X_{j,n} - \lambda)$  for  $X$ , if  $\lambda \leq x - \delta$

$$(1) \quad 1 - Q(n, \lambda) \leq n^{-3}u_1(\lambda) + 3n^{-2}v_1(\lambda),$$

while if  $\lambda \geq x + \delta$ ,

$$(2) \quad Q(n, \lambda) \leq n^{-3}u_2(\lambda) + 3n^{-2}v_2(\lambda).$$

These estimates can be used to prove

$$(3) \quad \lim_{J, n \rightarrow \infty} J^{-1} \sum_{j=1}^J Q(n, \lambda_j) = F(x) \text{ a.e. } [\mu^\infty].$$

Indeed, let  $A_i = J^{-1} \sum_{j=1}^J Q(n, \lambda_j)f_i(\lambda_j), i = 1, 2, 3$ . By inequality (1), with  $0 \leq \theta \leq 1$ ,

$$A_1 = J^{-1} \sum_{j=1}^J f_1(\lambda_j) - \theta n^{-3} J^{-1} \sum_{j=1}^J u_1(\lambda_j) - 3\theta n^{-2} J^{-1} \sum_{j=1}^J v_1(\lambda_j),$$

the first term converging to  $F(x - \delta)$ , the second and third to 0. Similarly, (2) implies that  $A_2 \rightarrow 0$ , and clearly  $\limsup_{J, n \rightarrow \infty} A_3 < \epsilon$ . In summary

$$\liminf_{J, n \rightarrow \infty} J^{-1} \sum_{j=1}^J Q(n, \lambda_j) \geq F(x - \delta) > F(x) - \epsilon,$$

$$\limsup_{J, n \rightarrow \infty} J^{-1} \sum_{j=1}^J Q(n, \lambda_j) < F(x - \delta) + \epsilon < F(x) + \epsilon.$$

Allowing  $k \rightarrow \infty$ , so that  $\epsilon \rightarrow 0$ , completes the proof of (3).

The next step is to prove

$$(4) \quad \lim_{J, n \rightarrow \infty} J^{-1} \sum_{j=1}^J [f(n^{-1}X_{j,n}) - Q(n, \lambda_j)] = 0 \text{ a.e. } [P_\lambda].$$

As before, put  $B_i = J^{-1} \sum_{j=1}^J [f(n^{-1}X_{j,n}) - Q(n, \lambda_j)]f_i(\lambda_j), i = 1, 2, 3$ . Then

$$E\{[f(n^{-1}X_{j,n}) - Q(n, \lambda_j)]f_2(\lambda_j) \mid P_{\lambda_j}\} = 0$$

and

$$\begin{aligned}
 E\{[f(n^{-1}X_{j,n}) - Q(n, \lambda_j)]^2 f_2(\lambda_j) \mid P_{\lambda_j}\} &= Q(n, \lambda_j)(1 - Q(n, \lambda_j))f_2(\lambda_j) \\
 (5) \qquad \qquad \qquad &\leq Q(n, \lambda_j)f_2(\lambda_j) \\
 &\leq n^{-3}u_2(\lambda_j) + 3n^{-2}v_2(\lambda_j)
 \end{aligned}$$

by (2).

Let  $B(J, n) = JB_2$ . To show  $\lim_{J, n \rightarrow \infty} B_2 = 0$ , a.e.  $[P_\lambda]$  select  $\eta > 0$  and define the sets  $C_{r,n} = \{s : s \varepsilon \gamma^\infty \text{ and there is a } J \text{ with (i) } 2^{r-1} < J \leq 2^r, \text{ (ii) } B(J, n) \geq J\eta\}$ ,  $D_{r,n} = \{s : s \varepsilon \gamma^\infty \text{ and there is a } J \text{ with (i) } 1 \leq J \leq 2^r, \text{ (ii) } B(J, n) \geq \eta/2 \cdot 2^r\}$ . Then  $C_{r,n} \subset D_{r,n}$  so that  $\sum_{r,n} P_\lambda(C_{r,n}) \leq \sum_{r,n} P_\lambda(D_{r,n})$ . Now by Kolmogorov's inequality [2], p. 235, and the estimate (5)

$$P_\lambda(D_{r,n}) \leq 4\eta^{-2}2^{-2r} \sum_{j=1}^{2^r} [n^{-3}u_2(\lambda_j) + 3n^{-2}v_2(\lambda_j)],$$

so that  $\sum_{r,n} P_\lambda(C_{r,n}) < \infty$ . Hence by the Borel-Cantelli Lemma [2], p. 228,  $B_2 \geq \eta$  only finitely often  $[P_\lambda]$ ; allowing  $\eta \rightarrow 0$  through some sequence of values shows that  $\lim_{J, n \rightarrow \infty} B_2 = 0$  a.e.  $[P_\lambda]$ . Similarly;  $\lim_{J, n \rightarrow \infty} B_1 = 0$  a.e.  $[P_\lambda]$  and clearly  $\limsup_{J, n \rightarrow \infty} |B_3| < \epsilon$  a.e.  $[P_\lambda]$ . These facts show that

$$\limsup_{J, n \rightarrow \infty} J^{-1} \sum_{j=1}^J [f(n^{-1}X_{j,n}) - Q(n, \lambda_j)] < \epsilon \text{ a.e. } [P_\lambda],$$

and allowing  $k \rightarrow \infty$  establishes (4).

Finally, combining (3) and (4) gives

$$\begin{aligned}
 \lim_{J, n \rightarrow \infty} F_{J,n}(x) &= \\
 \lim_{J, n \rightarrow \infty} J^{-1} \sum_{j=1}^J Q(n, \lambda_j) + \lim_{J, n \rightarrow \infty} J^{-1} \sum_{j=1}^J [f(n^{-1}X_{j,n}) - Q(n, \lambda_j)] \\
 &= F(x), \text{ a.e. } [P_\lambda],
 \end{aligned}$$

which is the requisite conclusion.

**COROLLARY.** *If  $F$  is continuous,*

$$\lim_{J, n \rightarrow \infty} (\sup_{-\infty < x < \infty} |F_{J,n}(x) - F(x)|) = 0 \text{ a.e. } [P_\mu^\infty].$$

The condition that  $F$  be continuous is indispensable. Indeed, by the central limit theorem,  $\lim_{n \rightarrow \infty} Q(n, x) = \frac{1}{2}$ ; moreover, for a rapidly increasing sequence  $n_\nu$ , the events  $n_\nu^{-1}X_{j, n_\nu} \leq x$  are almost independent. Put  $f_4(\lambda) = 1$  or  $0$  according as  $\lambda = x$  or not. Then by a slight modification of the Borel-Cantelli Lemma, for any  $J$

$$\sum_{j=1}^J f(n^{-1}X_{j, n_\nu})f_4(\lambda_j) / \sum_{j=1}^J f_4(\lambda_j)$$

is equal to 0 infinitely often and equal to 1 infinitely often as  $\nu \rightarrow \infty$ ,  $[P_\lambda]$ . Hence

$$\begin{aligned}\liminf_{J,n \rightarrow \infty} F_{J,n} &= F(x-), \\ \limsup_{J,n \rightarrow \infty} F_{J,n} &= F(x) \text{ a.e. } [P_\mu^\infty].\end{aligned}$$

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