## Poisson Random Fields — Exercises

## April 19, 2005

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**1.** Let  $N = (N_1, N_2, ..., N_n)$  be *n* independent Poisson random variables with  $E(N_i) = c_i$ . (Recall that *N* is Poisson with mean *c* if *N* is an integer-valued random variable with  $P(N = n) = e^{-c}c^n/n!$  for n = 0, 1, ...) Prove that

$$E\left(e^{\sum_{i=1}^{n}u_iN_i}\right) = e^{\sum_{i=1}^{n}c_i\left(e^{u_i}-1\right)} \tag{1}$$

for all choices of numbers  $u_i$ . Conversely, if  $N = (N_1, N_2, \ldots, N_n)$  is an arbitrary set of n random variables such that (1) holds for all  $u_i$ , prove that the  $N_i$  are independent Poisson random variables with means  $E(N_i) = c_i$ . (*Hint*: Use the fact that if  $E(e^{\sum_{i=1}^r u_i X_i}) = E(e^{\sum_{i=1}^r u_i Y_i}) < \infty$  for  $-a < u_i < a$  for some a > 0 for two random vectors  $X = (X_1, \ldots, X_r)$  and  $Y = (Y_1, \ldots, Y_r)$ , then X and Y have the same probability distribution.)

**2.** Let  $N = (N_1, \ldots, N_n)$  be as in problem 1. Define a measure  $\mu(A)$  on the set  $X = \{1, 2, \ldots, n\}$  by  $\mu(\{i\}) = c_i$ , so that  $\mu(A) = \sum_{i \in A} c_i$ . Use the vector N to define a random measure N(A) on X by  $N(\{i\}) = N_i$ , so that  $N(A) = \sum_{i \in A} N_i$  for  $A \subseteq X$ . Thus the random measure N(A) is purely atomic with the (random) integer-valued atom  $N_i$  at  $i \in X$ . Prove that

$$E\left(e^{\int_X f(y)N(dy)}\right) = e^{\int_X \left(e^{f(y)}-1\right)\mu(dy)}$$
(2)

for all bounded functions f(y) on X. Conversely, prove that if (2) holds for all bounded functions f(y) on X, then  $N_i$  are independent Poisson random variables with means  $E(N_i) = c_i$ .

**Definition.** A Poisson random field is a random measure  $(X, \mathcal{F}, N(A))$  on a measure space  $(X, \mathcal{F})$  (that is, N(A) are random variables such that, with probability one,  $(X, \mathcal{F}, N(A))$  is a measure) with mean measure  $(X, \mathcal{F}, \mu)$  if (2) holds whenever f(y) is a bounded  $\mathcal{F}$ -measureable function on X with  $\int_X |f(y)| \mu(dy) < \infty$ .

**3.** Assuming that we have been able to construct a Poisson random field  $(X, \mathcal{F}, N(A))$  for a mean measure  $(X, \mathcal{F}, \mu)$ , use (2) to prove

(i) If  $\mu(A) < \infty$  for some  $A \in \mathcal{F}$ , then N(A) is Poisson with mean  $\mu(A)$ . (*Hint*: Try  $f(y) = uI_A(y)$  in (2).)

(ii) If  $\mu(A) < \infty$  and  $\mu(B) < \infty$  for two disjoint sets  $A, B \in \mathcal{F}$ , then N(A) and N(B) are independent. (Note that  $N(A \cup B) = N(A) + N(B)$  since N is a random measure by definition.)

(iii) If  $\{A_n\}$  are disjoint with  $A_n \in \mathcal{F}$  and  $\mu(A_n) < \infty$ , then  $\{N(A_n)\}$  are independent. (Recall that an infinite set of random variables is independent if and only if every finite subset is independent.)

(iv) If  $\mu(A) = \infty$  for some  $A \in \mathcal{F}$  and A is sigma-finite (that is,  $A = \bigcup_{i=1}^{\infty} A_n$ where  $A_n \in \mathcal{F}$ ,  $\{A_n\}$  are disjoint, and  $\mu(A_n) < \infty$ ), then  $N(A) = \infty$  almost surely. (*Hint*: By assumption,  $N(A) = \sum_{i=1}^{\infty} N(A_n)$  almost surely. Consider the random events  $C_n = \{N(A_n) \ge 1\}$  and see if you can use the Borel-Cantelli lemma. Are the  $\{C_n\}$  independent?)

**4.** Let  $N = (N_1, N_2, \ldots, N_n)$  be as in Problem 1. Suppose that there are  $N_i$  objects of some kind at  $i \ (1 \le i \le n)$ . At a particular time, all of the objects move independently of one another to points in a finite set Y. Assume that objects at  $i \ (1 \le i \le n)$  move to  $y \in Y$  with probability  $\pi(i, y)$ , where  $\pi(i, y) \ge 0$  and  $\sum_{y \in Y} \pi(i, y) = 1$  for each i.

Let  $M_y$  be the total number of objects that end up at  $y \in Y$  (that is, from all starting points *i*). Prove that  $\{M_y : y \in Y\}$  are independent Poisson random variables and find  $E(M_y)$ . Does the form of  $E(M_y)$  seem plausible? (*Hint*: Verify (1) by calculating

$$E\left(e^{\sum_{y\in Y}v_{y}M_{y}}\right) = E\left(E\left(e^{\sum_{y\in Y}v_{y}M_{y}}\mid N\right)\right)$$
(3)

for  $N = (N_1, N_2, ..., N_n)$ . Note that, given N, the positions of the objects in Y are independent Y-valued random variables with probability distributions depending on their initial position.)

5. Let  $N = (N_1, N_2, \ldots, N_n)$  be as in Problem 1 and suppose that there are  $N_i$  objects at i, as in Problem 4. Suppose that, independently for all objects, each object at i is colored green with probability  $g_i$  and colored red with probability  $r_i$ , where  $g_i + r_i \leq 1$ . If an object is not colored either red or green, then it remains its original dull gray color. Let  $G_i$  be the number of green objects and  $R_i$  be the number of red objects at i.

(i) Prove that the  $\{G_1, G_2, \ldots, G_n\}$  are independent Poisson random variables. Find  $E(G_i)$ . (This is called a *thinning* of  $\{N_1, N_2, \ldots, N_n\}$ .)

(ii) Prove that  $\{G_1, \ldots, G_n, R_1, \ldots, R_n\}$  together are independent Poisson random variables. (In this case,  $\{G_1, G_2, \ldots, G_n\}$  and  $\{R_1, R_2, \ldots, R_n\}$  are called *orthogonal thinnings* of  $\{N_1, N_2, \ldots, N_n\}$ .)

orthogonal thinnings of  $\{N_1, N_2, \dots, N_n\}$ .) (iii) Let  $G = \sum_{i=1}^n G_i$  and  $R = \sum_{i=1}^n R_i$ . Prove that G and R are independent Poisson random variables. Find E(G) and E(R).

(*Hint*: Apply Problem 4 for an appropriate choice of a set Y. Apply (2) on Y for various functions f(y). What set Y did you choose?)

**6.** Let  $\pi(i, j)$  be a Markov transition function on the set  $X = \{0, 1, 2, ..., K\}$  for some integer K. Assume that  $\{1, 2, ..., K - 1\}$  are transient states and that 0, K are traps.

Suppose that, at each time n = 0, 1, 2, ..., a random number  $V_n$  of objects is placed at state  $1 \in X$ , where  $\{V_n : n \ge 0\}$  are independent Poisson random variables with  $E(V_n) = v > 0$  and there were no objects in X before time 0. Immediately after the  $n^{\text{th}}$  set of immigrants arrive  $(n \ge 0)$ , all objects move one step independently according to  $\pi(i, j)$ . In particular, the probability that one of the objects that originally arrived at 1 at time n is at i at time n + t is given by  $\pi^{(t)}(1, i)$ , where  $\pi^{(t)}$  is the  $t^{\text{th}}$  matrix power of  $\pi(i, j)$ .

Let N(n, i) be the number of objects at i at time  $n \ge 0$ , where the population is counted in the  $n^{\text{th}}$  time step just after the arrival of the new immigrants. (Thus  $N(0, 1) = V_0$  and  $N(n, 1) \ge V_n \ge 0$ .) Then

(i) Prove that, for each n, { N(n, i) } are independent Poisson for 0 ≤ i ≤ K.
(ii) Prove that { N(n, i) : 1 ≤ i ≤ K − 1 } converges in distribution as n → ∞ to independent Poisson random variables N(i) (1 ≤ i ≤ K − 1). Find E(N(i)).

(iii) Find an expression for

$$\nu_K = \lim_{n \to \infty} \frac{1}{n} E \left( N(n, K) \right)$$

and explain what  $\nu_K$  means in terms of the objects. (*Hint*: It is the asymptotic rate at which objects are trapped at K.)

(*Hints*: Note that

$$N(n,i) = \sum_{t=0}^{n} N(t,n,i)$$
(4)

where N(t, n, i) are the objects that began at state 1 at time  $t \leq n$  and then ended up at *i* at time *n*. (In particular,  $\sum_{i=0}^{K} N(t, n, i) = V_t$  for  $n \geq t$ , since the objects don't leave  $X = \{0, 1, 2, \dots, K\}$ .) Argue as in Problem 4 to show that, for fixed *n* and  $t, \{N(t, n, i)\}$  are independent Poisson for  $0 \leq i \leq K$ . Use the independence of the  $V_t$  to conclude that  $\{N(t, n, i)\}$  are also independent for  $0 \leq t \leq n$  and  $0 \leq i \leq K$ . Use (4) to conclude that  $\{N(n, i)\}$  ( $0 \leq i \leq K$ ) are independent Poisson for each *n*. Finally, use either (1) or (2) and take limits as  $n \to \infty$ .)