

## Edge Corrections for Spatial Point Processes

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**Some Background: The Kaplan Meier Estimator.** Let

$$\begin{aligned} X &= \text{lifetime,} \\ Y &= \text{censoring time,} \\ Z &= X \wedge Y = \min[X, Y] \end{aligned}$$

and

$$\delta = \mathbf{1}_{X \leq Y}.$$

Suppose that  $X$  and  $Y$  are independent, that  $X \sim F$ , and that  $Y \sim G$ , so that

$$\begin{aligned} F(t) &= P[X \leq t], \\ G(t) &= P[Y \leq t]. \end{aligned}$$

Suppose also that  $F(0) = 0 = G(0)$  and that  $0 < F(t), G(t) < 1$  for all  $0 < t < \infty$ .

*The Problem:* Given an sample  $(\delta_1, Z_1), \dots, (\delta_n, Z_n)$ , estimate  $F$ . Ignoring the censoring leads to a biased estimator here, as does only considering uncensored cases. Let

$$0 = t_0 < t_1 < \dots < t_m < \infty$$

and

$$\Lambda_i = P[X > t_i | Z > t_{i-1}].$$

Then

$$\Lambda_i = P[X > t_i | X > t_{i-1}] = \frac{1 - F(t_i)}{1 - F(t_{i-1})}$$

and

$$\hat{\Lambda}_i = \frac{\#\{j : X_j > t_i\}}{\#\{j : Z_j > t_{i-1}\}}$$

is an unbiased estimator of  $\Lambda_i$ . Here

$$1 - F(t_k) = \Lambda_1 \times \dots \times \Lambda_k.$$

So,

$$\hat{\Lambda}_1 \times \dots \times \hat{\Lambda}_k$$

provides an estimator of  $1 - F(t_k)$ .

Letting  $0 < z_1 < \dots < z_n$  denote the ordered values of  $Z_1, \dots, Z_n$  and taking a limit, the Kaplan-Meier estimator is

$$1 - F(\hat{z}_k) = \prod_{i=1}^k \frac{\#\{j : X_j \geq z_k\}}{\#\mathcal{R}_i},$$

where

$$\mathcal{R}_i = \{j : Z_j \geq z_j\}$$

is called the *risk set*.

**Euclidean Space:** Let  $R^p$  be the set of lists  $x = (x_1, \dots, x_p)$  of real numbers,  $\|x\|^2 = x_1^2 + \dots + x_p^2$ , and  $d(x, y) = \|x - y\|$ .

**Point Processes:** Informally, a point process in  $\mathbb{R}^p$  is a collection of points, placed at random. It can also be described in terms of counts. Let  $N_B = N\{B\}$ ,  $B \subseteq \mathbb{R}^p$ , be jointly distributed, non-negative integer or  $\infty$  valued random variables for which

$$N_B < \infty \text{ w.p.1}$$

for bounded  $B$ .

$$N_{A \cup B} = N_A + N_B$$

whenever  $A \cap B = \emptyset$ , and some other more technical conditions are satisfied.

**Interpretation:** Here  $N_B$  denotes the number of points in the set  $B$ . Let  $\{x\}$  be the set consisting of  $x$ . Then, there is a point at  $x$  iff  $N_{\{x\}} > 0$ . Suppose that there are no multiple points; that is,  $N_{\{x\}} \leq 1$  w.p.1 for all  $x$ . Let

$$\mathcal{N} = \{x : N_{\{x\}} > 0\}.$$

This is the set of points.

**Example: Poisson Processes.** Recall that a random variable  $X$ , say, has the Poisson distribution with parameter  $\mu$  if

$$P[X = k] = \frac{1}{k!} \mu^k e^{-\mu}$$

for  $k = 0, 1, 2, \dots$  in which case  $E(X) = \mu$ . Thus

$$P[X = 0] = e^{-\mu}.$$

A point process  $N$  is called a *Poisson process*, if  $N_B$  has a Poisson distribution for every  $B$  and  $N_{A_1}, \dots, N_{A_m}$  are independent whenever  $A_1, \dots, A_m$  are mutually exclusive.

In one dimension, it is conventional to let

$$M_t = N_{(0,t]},$$

the number of points in the (time) interval  $(0, t]$ . Then the points occur at the discontinuities of  $M_t$ .

**Stationarity:** Let  $B + x = \{y : y - x \in B\} = \{x + z : z \in B\}$  and suppose that

$$[N_{B_1+x}, \dots, N_{B_m+x}] \stackrel{\mathcal{D}}{=} [N_{B_1}, \dots, N_{B_m}]$$

for all  $B_1, \dots, B_m$  and  $x$ . Then

$$E[N_B] = \lambda \nu_p(B),$$

where  $\nu_p$  is  $p$ -dimensional volume and  $\lambda$  is a constant, called the *intensity*

**Some Quantities of Interest:** Let

$$\begin{aligned} B(x, t) &= \{y : d(x, y) \leq t\}, \\ B_-(x, t) &= B(x, t) - \{x\}. \end{aligned}$$

*The Occupied Space Function:*

$$p(t) = P[N_{B_-(x, t)} > 0].$$

In view of the assumed stationarity, this does not depend on  $x$

*The Interpoint Distribution Function:* Let

$$G(t) = P[N_{B_-(0, t)} > 0 | 0 \in \mathcal{N}].$$

*The Interpoint Distance Function:* Let

$$K(t) = \frac{1}{\lambda} E[N_{B_-(0, t)} | 0 \in \mathcal{N}].$$

**Note:** Defining the conditional probabilities, in any generality, requires some subtlety in the form Palm distribution.

**Examples.** If  $N$  is a Poisson process in  $\mathbb{R}^2$ , then

$$E[N_{B(0, t)}] = \lambda \pi t^2$$

and

$$p(t) = 1 - P[N_{B(0, t)} = 0] = 1 - e^{-\lambda \pi t^2}.$$

Also, using the independence,

$$G(t) = p(t)$$

and

$$K(t) = \lambda \pi t^2.$$

**Estimating  $p$ ,  $G$ , and  $K$  From Data.** Let  $E \subseteq \mathbb{R}^p$  be a bounded set with a non-empty interior—for example

$$E = \{x : |x_i| \leq 1, i = 1, 2\}$$

in  $\mathbb{R}^2$ . Suppose that we observe (only) the points that fall in  $E$ . Let

$$n = N_E$$

and

$$\mathcal{N}_E = \mathcal{N} \cap E.$$

*Estimating  $p$ : The Problem.* Let

$$W = \{w^1, \dots, w^m\} \subseteq E$$

be a finite subset of  $E$ ;

$$\tilde{d}_i = \min_{x \in \mathcal{N}} d(w^i, x)$$

and

$$\tilde{p}(t) = \frac{1}{m} \#\{i : d_i \leq t\}$$

Then

$$P[\tilde{d}_i \leq t] = P[B(w^i, t) > 0] = p(t)$$

and

$$E[\tilde{p}(t)] = \frac{1}{m} \sum_{i=1}^m P[\tilde{d}_i \leq t] = p(t).$$

The problem is that we cannot compute  $\tilde{d}_i$ , because it may require points that are outside of  $E$ .

*Estimating  $p$ : The Solution.* Let

$$d_i = \min_{x \in \mathcal{N}_E} d(w^i, x)$$

and

$$r_i = d(w^i, E^c),$$

the distance from  $w^i$  to the complement of  $E$ . Observe that: If  $r_i \geq t$ , then  $d_i \leq t$  iff  $\tilde{d}_i \leq t$ , so that

$$P[d_i \leq t] = p(t)$$

for  $t \leq d_i$ . Let

$$\hat{p}(t) = \frac{\#\{i : d_i \leq t \ \& \ r_i \geq t\}}{\#\{i : r_i \geq t\}}.$$

Then  $\hat{p}(t)$  is unbiased. For, letting

$$m_i = \#\{i : r_i \geq t\},$$

$$E[\hat{p}(t)] = \frac{1}{m_i} \sum_{i:r_i \geq t} P[d_i \leq t] = p(t).$$

*Some Problems.*  $p(t)$  is an increasing function of  $f$ ,  $\hat{p}(t)$  need not be.

Also,  $m_i$  may be very small for large values of  $t$ .