

# 1 Problem Statement

We will consider a spatial Poisson process on  $\mathbb{R}^2$  with rate  $\lambda = 1$ , i.e. the number of points in region with area  $A$  follows a Poisson distribution with rate  $A$ . To model increasing larger samples sizes, we look at  $[-\frac{t}{2}, \frac{t}{2}]^2$  for increasing larger  $t$ .

**Problem.** If we construct the nearest-neighbor graph on the points contained in  $[-\frac{t}{2}, \frac{t}{2}]^2$ , what is the expected number of edges that cross the  $y$ -axis?

For  $N \sim \text{Po}(t^2)$ , let  $x_1, \dots, x_N$  be the points in  $[-\frac{t}{2}, \frac{t}{2}]^2$  and let  $x_{n(i)}$  be  $x_i$ 's nearest neighbor. Define

$$X_i = \begin{cases} 1 & x_i \text{ and } x_{n(i)} \text{ belong to opposite halves, } x_i \neq x_{n(i)} \\ \frac{1}{2} & x_i \text{ and } x_{n(i)} \text{ belong to opposite halves, } x_i = x_{n(i)} \\ 0 & \text{otherwise} \end{cases}$$

Let  $T$  be the total number of crossings. Then

$$\mathbb{E}T = \mathbb{E}[\mathbb{E}[T|N]] = \mathbb{E}_N \left[ \sum_{i=1}^N \mathbb{E}X_i \right] = \sum_{n=0}^{\infty} n \mathbb{E}X_1 \frac{t^{2n} e^{-t^2}}{n!} = \mathbb{E}X_1 \sum_{n=0}^{\infty} n \frac{t^{2n} e^{-t^2}}{n!} = t^2 \mathbb{E}X_1$$

It remains to find  $\mathbb{E}X_1 = P(X_1 = 1) + \frac{1}{2}P(X_1 = \frac{1}{2})$ .

## 2 Finding $\mathbb{E}X_1$

### 2.1 Finding $P(X_1 = \frac{1}{2})$

For brevity, define

$$O = \mathbb{1}(x_1 \text{ and } x_{n(1)} \text{ belong to opposite halves})$$

$$D = \mathbb{1}(x_1 = x_{n(1)})$$

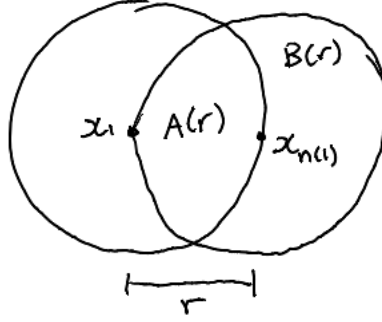
Then  $P(X_1 = 1) = P(O = 1 \cap D = 1)$ . We condition on the distance  $d$  between  $x_1$  and the  $y$ -axis. Note  $d \sim \text{Unif}([0, \frac{t}{2}])$ .

$$\begin{aligned} P(X_1 = 1|d) &= P(O = 1 \cap D = 1|d) \\ &= P(D = 1|O = 1, d)P(O = 1|d) \\ &= \int_d^{\infty} P(D = 1|O = 1, d, r)f(r|O = 1, d) dr * P(O = 1|d), \end{aligned} \tag{1}$$

where  $r = \|x_1 - x_{n(1)}\|$  is the distance between  $x_1$  and its nearest neighbor.

#### 2.1.1 $P(D = 1|O = 1, d, r)$

Given  $O = 1$ ,  $d$ , and  $r$ , we want to find the probability  $x_1 = x_{n(1)}$ . We condition on  $r = \|x_1 - x_{n(1)}\|$ . Consider the following figure depicting the situation.



In order for  $x_1 = x_{n(1)}$ , i.e.  $x_1$  is  $x_{n(1)}$ 's nearest neighbor, there cannot be a point contained in a circle of radius  $r$  round  $x_{n(1)}$ . Since we on condition  $r$ , we know there does not exist a point in the intersection of the two circles. Hence,  $D = 1$  if and only no point lies in the region with area  $B(r)$ .

$$B(r) = \pi r^2 - r^2 \left( \frac{2\pi}{3} - \frac{\sqrt{3}}{2} \right) = r^2 \left( \frac{\pi}{3} + \frac{\sqrt{3}}{2} \right),$$

and thus,

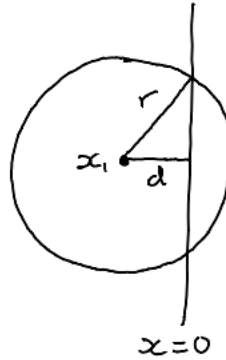
$$P(D = 1|O = 1, d, r) = e^{-r^2 \left( \frac{\pi}{3} + \frac{\sqrt{3}}{2} \right)}.$$

### 2.1.2 $f(r|O = 1, d)$

According to Bayes' rule,

$$f(r|O = 1, d) = \frac{P(O = 1|r, d)f(r|d)}{P(O = 1|d)}.$$

For  $f(O = 1|r, d)f(r|d)$  consider the following situation.



$x_{n(1)}$  lies somewhere along the circle of radius  $r$  centered at  $x_1$ , which is a distance  $d$  away from the  $y$ -axis.  $O = 1$  if and only if  $x_{n(1)}$  lies on the arc on the other side of the  $y$ -axis.

Since  $x_{n(1)}$  is equally likely to lie anywhere along the circle, this probability is equal to the ratio of the arc length to the circle's circumference,

$$P(O = 1|r, d) = \frac{2 \cos^{-1} \left( \frac{d}{r} \right) r}{2\pi r} = \frac{\cos^{-1} \left( \frac{d}{r} \right)}{\pi}.$$

$r$  is independent of  $d$ . For all  $r_0 \geq 0$ ,

$$P(r \leq r_0) = 1 - e^{-\pi r_0^2} \Rightarrow f(r) = 2\pi r e^{-\pi r^2}.$$

For  $P(O = 1|d)$ , we condition on  $r$  again.

$$\begin{aligned} P(O = 1|d) &= \int_d^\infty P(O = 1|r, d) f(r|d) dr \\ &= \int_d^\infty \frac{\cos^{-1} \left( \frac{d}{r} \right)}{\pi} 2\pi r e^{-\pi r^2} dr \\ &= 2 \int_d^\infty \cos^{-1} \left( \frac{d}{r} \right) r e^{-\pi r^2} dr \end{aligned} \tag{2}$$

Putting it all together,

$$\begin{aligned} f(r|O = 1, d) &= \frac{P(O = 1|r, d) f(r|d)}{P(O = 1|d)} \\ &= \frac{\frac{\cos^{-1} \left( \frac{d}{r} \right)}{\pi} * 2\pi r e^{-\pi r^2}}{P(O = 1|d)} \\ &= \frac{\cos^{-1} \left( \frac{d}{r} \right) * 2r e^{-\pi r^2}}{P(O = 1|d)} \end{aligned} \tag{3}$$

## 2.2 Plugging everything in

$$\begin{aligned} P(X_1 = 1|d) &= \int_d^\infty P(D = 1|O = 1, d, r) f(r|O = 1, d) dr * P(O = 1|d) \\ &= \int_d^\infty P(D = 1|O = 1, d, r) \frac{P(O = 1|r, d) f(r|d)}{P(O = 1|d)} dr * P(O = 1|d) \\ &= \int_d^\infty P(D = 1|O = 1, d, r) P(O = 1|r, d) f(r|d) dr \\ &= \int_d^\infty e^{-r^2 \left( \frac{\pi}{3} + \frac{\sqrt{3}}{2} \right)} \cos^{-1} \left( \frac{d}{r} \right) 2r e^{-\pi r^2} dr \end{aligned} \tag{4}$$

Since  $d \sim \text{Unif}([0, \frac{t}{2}])$ , it follows that

$$\begin{aligned} P(X_1 = 1) &= \int_0^{t/2} P(X_1 = 1|d) \frac{2}{t} dd \\ &= \frac{4}{t} \int_0^{t/2} \int_d^\infty e^{-r^2 \left( \frac{\pi}{3} + \frac{\sqrt{3}}{2} \right)} \cos^{-1} \left( \frac{d}{r} \right) r e^{-\pi r^2} dr dd. \end{aligned} \tag{5}$$

### 2.3 Finding $P(X_1 = 1)$

Note,

$$P(X_1 = 1) = P(O = 1) - P(X_1 = 1).$$

It remains to solve for  $P(O = 1)$ . We condition on  $r$  and  $d$ :

$$\begin{aligned} P(O = 1) &= \int_0^{t/2} P(O = 1|d) \frac{2}{t} dd \\ &= \int_0^{t/2} \int_d^\infty P(O = 1|d, r) f(r|d) dr \frac{2}{t} dd \\ &= \int_0^{t/2} \int_d^\infty \frac{\cos^{-1}\left(\frac{d}{r}\right)}{\pi} 2\pi r e^{-\pi r^2} dr \frac{2}{t} dd \\ &= \frac{4}{t} \int_0^{t/2} \int_d^\infty \cos^{-1}\left(\frac{d}{r}\right) r e^{-\pi r^2} dr dd \end{aligned} \tag{6}$$

## 3 Solution

If we define

$$\begin{aligned} \varphi(t) &= \frac{4}{t} \int_0^{t/2} \int_d^\infty \cos^{-1}\left(\frac{d}{r}\right) r e^{-\pi r^2} dr dd \\ \theta(t) &= \frac{4}{t} \int_0^{t/2} \int_d^\infty e^{-r^2\left(\frac{\pi}{3} + \frac{\sqrt{3}}{2}\right)} \cos^{-1}\left(\frac{d}{r}\right) r e^{-\pi r^2} dr dd \end{aligned}$$

then

$$\mathbb{E}X_1 = (\varphi(t) - \theta(t)) - \frac{1}{2}\theta(t) = \varphi(t) - \frac{1}{2}\theta(t).$$

It follows that

$$\mathbb{E}T = t^2 \mathbb{E}X_1 = t^2 \left[ \varphi(t) - \frac{1}{2}\theta(t) \right].$$

## 4 Generalizing to Higher Dimension

If we generalize to higher dimension, the following functions are required:

- $s_n(d, r)$  = the ratio of surface areas in an  $n$ -dimension hypersphere
- $f_n(r)$  = the density of distance to the nearest neighbor
- $c_n(r)$  = the volume of the crescent created by two overlapping  $n$ -dimensional hyperspheres

In two dimensions, these functions are

- $s_2(d, r) = \frac{2r \cos^{-1}\left(\frac{d}{r}\right)}{2\pi r}$ ,
- $f_2(r) = 2\pi r e^{-\pi r^2}$ , and
- $c_2(r) = -r^2 \left( \frac{\pi}{3} + \frac{\sqrt{3}}{2} \right)$ .

Using these functions,

- $\varphi_n(t) = \frac{2}{t} \int_0^{t/2} \int_0^\infty s_n(d, r) f_n(r) dr dd$  and
- $\theta_n(t) = \frac{2}{t} \int_0^{t/2} \int_d^\infty e^{-c_n(r)} s_n(d, r) f_n(r) dr dd$ .

#### 4.1 $s_n$ , $f_n$ , and $c_n$ in arbitrary dimension

For  $s_n$ , see *Concise Formulas for the Area and Volume of a Hyperspherical Cap*.

$$s_n(d, r) = \frac{1}{2} I_{1 - (\frac{d}{r})^2} \left( \frac{n-1}{2}, \frac{1}{2} \right),$$

where  $I_z(z, b) = \frac{B_z(a, b)}{B(z, b)}$  is the regularized Beta function.

For  $f_n(r)$ ,

$$F_n(r) = 1 - e^{-V_n(r)}$$

$$f_n(r) = \frac{\partial V_n(r)}{\partial r} e^{-V_n(r)},$$

where  $V_n(r) = \frac{\pi^{n/2}}{\Gamma(\frac{n}{2}+1)} r^n$  is the volume of a  $n$ -dimensional hypersphere with radius  $r$ .

For  $c_n(r)$ , we first calculate the volume of the intersection of two overlapping  $n$ -dimensional hyperspheres. This regions consists of two hyperspherical caps with  $d = \frac{r}{2}$ . According to *Concise Formulas for the Area and Volume of a Hyperspherical Cap*, each of these hyperspherical caps has volume

$$\frac{1}{2} V_n(r) I_{\frac{3}{4}} \left( \frac{n+1}{2}, \frac{1}{2} \right),$$

so the overlapping volume is

$$V_n(r) I_{\frac{3}{4}} \left( \frac{n+1}{2}, \frac{1}{2} \right).$$

Therefore, the crescent volume is

$$c_n(r) = V_n(r) - V_n(r) I_{\frac{3}{4}} \left( \frac{n+1}{2}, \frac{1}{2} \right) = V_n(r) \left[ 1 - I_{\frac{3}{4}} \left( \frac{n+1}{2}, \frac{1}{2} \right) \right].$$