

# Inverse Problems in Statistics

## Chapter 2: Examples

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### 1 Wicksell's corpuscle problem

Our goal is to estimate the distribution function of the radii  $R_i$  of some spheres, only knowing the distribution of the radii  $Z_i$  of Spheresections.

#### 1.1 Model Assumption

- The centers of the spheres,  $m_1, m_2, \dots$ , are Homogeneous Poisson distributed
- The radii,  $R_1, R_2, \dots$ , are F-distributed and pairwise independent
- The probability to cut a sphere is proportional to its radius.
- Once the sphere  $i$  has been cut, the distances from the center of the sphere to the cutting plane is uniformly  $U(0, R_i)$  distributed

#### 1.2 Model Derivation

$$X := R^2$$

According to the assumption that the probability that a sphere is cut is proportional to its radius:  $\rightarrow$

$$F \longrightarrow F^b = \frac{1}{m_F} \int_0^x \sqrt{Y} dF(y)$$

According to the last model-assumption:  $\rightarrow$

$$Z = R^2 - U^2 R^2 = X(1 - U^2) =: XY$$

Now we can calculate the density of Y:  $\rightarrow$

$$k_Y(y) = 1/2(1 - y)^{1/2} 1_{(0,1)}(y)$$

We find the density of the observed radii:  $\rightarrow$

$$g(z) = \frac{1}{2m_F} \int_{(z,\infty)} \frac{dF(x)}{\sqrt{x - z}}$$

## 2 Linear probe problem

It's useless here doing the derivations because are similar of those in section 1

### 2.1 Model Assumption

- The centers of the spheres,  $m_1, m_2, \dots$ , are Homogeneous Poisson distributed
- The radii,  $R_1, R_2, \dots$ , are F-distributed and pairwise independent
- The probability for a line to go through a sphere is proportional to its squared radius.
- Once the sphere  $i$  has been hit, be  $V\sqrt{X}$  the distances from the center of the sphere to the line. Then  $V$  is  $H(v) = v^2 \cdot 1_{(0,1)}$  distributed

## 3 Censoring Problems

In many statistical problems, the quantities we would like to observe, are subjected to some censoring mechanism.

### 3.1 Right censoring

- sample  $X_1, X_2, \dots, X_n$  with unknown distribution  $F$  on  $[0, \infty)$
- sample  $R_1, R_2, \dots, R_n$  with distribution  $H$  on  $[0, \infty)$ , independent of the  $X$ 's
- observable data  $Z_i = (T_i, \Delta_i) = (X_i \wedge R_i, 1_{\{X_i \leq R_i\}})$
- distribution of  $Z$ :  $G(y) = P(Y \leq y) = F(y) + H(y)(1 - F(y))$
- density of  $Z$ :  $g(y, \delta) = (f(y)(1 - H(y))^\delta)(h(y)(1 - F(y)))^{1-\delta}$
- possible inverse relation:  $F(y) = \frac{G(y) - H(y)}{1 - H(y)}$

### 3.2 Double censoring

- sample  $X_1, X_2, \dots, X_n$  with unknown distribution  $F$  on  $[0, \infty)$
- sample  $(L_1, R_1), (L_2, R_2), \dots, (L_n, R_n)$  with distribution  $H$  on  $[0, \infty)^2$ , independent of the  $X$ 's and with  $P[L < R] = 1$
- observable data  $Z_i = (Y_i, \Delta_i, \Gamma_i) = ((L_i \vee X_i) \wedge R_i, 1_{\{X_i \leq L_i\}}, 1_{\{L_i < X_i \leq R_i\}})$
- density of  $Z$ :  $g(y, \delta, \gamma) = (F(y)h_L(y))^\delta((H_L(y) - H_R(y))f(y))^\gamma((1 - F(y))h_R(y))^{1-\gamma-\delta}$
- possible inverse relation:  $F(x) = \int_0^x \frac{dG(y, 0, 1)}{H_L(y) - H_R(y)}$

### 3.3 Interval censoring

interval censoring case I

- sample  $X_1, X_2, \dots, X_n$  with unknown distribution  $F$  on  $[0, \infty)$
- sample  $T_1, T_2, \dots, T_n$  with distribution  $H$  on  $[0, \infty)$ , independent of the  $X$ 's
- observable data  $Z_i = (T_i, \Delta_i) = (T_i, 1_{\{X_i \leq T_i\}})$
- density of  $Z$ :  $g(t, \delta) = F(t)^\delta (1 - F(t))^{1-\delta} h(t)$
- distribution of  $X$ :  $F(t) = \frac{g(t,1)}{h(t)} = 1 - \frac{g(t,0)}{h(t)}$

interval censoring case II

- sample  $X_1, X_2, \dots, X_n$  with unknown distribution  $F$  on  $[0, \infty)$
- sample  $T_1, T_2, \dots, T_n$  and  $U_1, U_2, \dots, U_n$  with joint distribution  $H$  on  $[0, \infty)^2$ , independent of the  $X$ 's and with  $P[T < U] = 1$
- observable data  $Z_i = (T_i, U_i, \Delta_i, \Gamma_i) = (T_i, U_i, 1_{\{X \leq T\}}, 1_{\{T < X \leq U\}})$
- density of  $Z$ :  $g(t, u, \delta, \gamma) = F(t)^\delta (F(u) - F(t))^\gamma (1 - F(u))^{1-\gamma-\delta} h(t, u)$
- possible inverse relation:  $F(t) = \frac{g(t,u,1,0)}{h(t,u)}$

## 4 Migrating birds problem

Problem: we want to estimate the time spent by a generic bird at an oasis

- $X$  positive random variable with unknown distribution  $F$ , denotes the time spent by the bird at an oasis
- $N$  homogeneous Poisson Process with intensity  $\lambda$ , describes the times the birds is caught
- $N(X)$  number of times the birds is caught

We will derive the distribution of time between two catches in terms of  $F$

- distribution of the sojourn time of a bird given it is caught exactly twice:
- $F'(x) = P(X \leq x | N(X) = 2) \approx \frac{\int_0^x y^2 dF(y)}{\int_0^\infty y^2 dF(y)}$
- distribution of  $Z$  (time between two catches) given that  $X = x$  and  $N(X) = 2$
- $P(Z \leq z) = \frac{\int_z^\infty (x-z)^2 dF(x)}{\int_0^\infty x^2 dF(x)}$

We now can express  $F$  in terms of the derivative of the density of  $Z$

$$F(x) = 1 - \frac{g'(x)}{g'(0)}$$

## 5 Deconvolution problem

Our goal is to get the distribution function  $F$  out of the known density of  $Z$  :

$$g(z) = \frac{d}{dz} \int_{-\infty}^{\infty} k(z-x) \cdot dF(x)$$

### 5.1 Model Assumption

- Let  $X_1, X_2, \dots, X_n$  be  $F$  distributed,  $F$  unknown
- Let  $Y_1, Y_2, \dots, Y_n$  have a known density function  $k(x)$
- We can observe  $Z_1, Z_2, \dots, Z_n$ , where  $Z_i := X_i + Y_i$