

TMA4250 Spatial Statistics

Exam 21.05.2010

Suggested problem solutions  
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Problem 1. Continuous RF

$$\{R(x); x \in D \subset \mathbb{R}^2\} \rightarrow \begin{cases} E\{R(x)\} = \mu_R(x) \\ \text{Var}\{R(x)\} = \sigma_R^2(x) \\ \text{Cov}\{R(x'), R(x'')\} = \rho_R(x', x'') \end{cases}$$

a)

$[R(x_1), \dots, R(x_n)] \rightsquigarrow$  Gaussian pdf

all  $x_1, \dots, x_n$

all  $n > 0$

stationary, isotropic  $\Rightarrow \begin{cases} \mu_R(x) = \mu_R \\ \sigma_R^2(x) = \sigma_R^2 \\ \rho_R(x', x'') = \rho_R(\|x' - x''\|) \end{cases}$

$S(x) = B^T g(x) + R(x)$       ↗ stat, isotr       $B \rightsquigarrow \text{Gauss}(b; 0, \sigma_B^{-2} I)$   
 $B, R(x)$  - indep.       $g(x)$  - known

b)  $S(x) \rightsquigarrow$  Gauss RF since linear comb of Gauss RV and Gauss RF.

$$E\{S(x)\} = E\{B^T\} g(x) + E\{R(x)\} = \mu_R$$

$$\begin{aligned} \sigma_S^2 &= \text{Var}\{S(x)\} = g^T(x) \sigma_B^{-2} I g(x) + \sigma_R^2 \\ &= \sigma_B^{-2} g^T(x) g(x) + \sigma_R^2 \end{aligned}$$

$$C_S \quad C_S(x', x'') = \text{Corr}\{S(x'), S(x'')\} = \frac{\sigma_s^2 g(x')^T g(x'')}{\sigma_s^2 + \rho_R (||x' - x''||)}$$

$$\text{Corr}\{S(x'), S(x'')\} = \frac{C_S(x', x'')}{\sigma_S(x') \sigma_S(x'')}$$

$$S(x^d) = (S(x_1), \dots, S(x_n))^T$$

c)

$$\begin{bmatrix} S(x_0) \\ S(x_1) \\ \vdots \\ S(x_n) \end{bmatrix} \xrightarrow{n+1} \text{Gauss} \begin{pmatrix} \mu_R \\ \vdots \\ \mu_R \end{pmatrix}, \quad \begin{bmatrix} \sigma_s^2(x_0) & C_S(x_1, x_0) \\ C_S(x_0, x_1) & \ddots \\ \vdots & C_S(x_n, x_0) & C_S(x_1, x_n) \\ C_S(x_0, x_n) & C_S(x_1, x_n) & \ddots & \sigma_s^2(x_n) \end{bmatrix}$$

$$\xrightarrow{n+1} \text{Gauss} \begin{pmatrix} \mu_R \\ \vdots \\ \mu_R \end{pmatrix} \begin{bmatrix} \sigma_s^2(x_0) & C_{dd} \\ C_{dd} & C_{dd} \end{bmatrix}$$

$$[S(x_0) | S(x^a)] \xrightarrow{} \text{Gauss}_{\tau}(\mu_{old}, \sigma_{old}^2)$$

$$\mu_{old} = \mu_R + C_{old} C_{dd}^{-1} (S(x^a) - \mu_R)$$

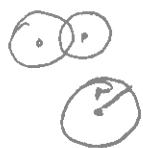
$$\sigma_{old}^2 = \sigma_s^2(x_0) - C_{old} C_{dd}^{-1} C_{old}^T$$

$$\text{Prob}\{S(x_0) > s_0 | S(x^a) = s(x^a)\} = 1 - \bar{\Phi}(s_0 / \mu_{old}, \sigma_{old}^2)$$

↑  
Stand. Gauss cdf

Problem 2. Event  $\mathcal{B}^F$  NB!

$$\{\mathbb{X}_i; i=1, \dots, N; \mathbb{D} \subset \mathbb{R}^3\}$$



constant circular radius  $r$

a)  $1 \text{ d... } \mathcal{B} \subset \mathcal{D} \quad |\mathcal{B}| = 1 \text{ dm}^3$

$$N(\mathcal{B}) \xrightarrow{\text{Poisson}} \text{Poisson}(n; \lambda |\mathcal{B}|)$$

$$\text{Prob}\{N(\mathcal{B}) = k\} = \frac{[\lambda |\mathcal{B}|]^k}{k!} e^{-\lambda |\mathcal{B}|} = \frac{\lambda^k}{k!} e^{-\lambda}$$

b)

$$E\{\text{weight of } dx \mid \text{cheese in } dx\} = \rho dx \quad dx \subset \mathcal{D}$$

$$E\{\text{weight of } dx \mid \text{hole in } dx\} = 0$$

hence

$$\begin{aligned} E\{\text{weight of } dx\} &= E\{w \text{ of } dx \mid \text{ch in } dx\} \cdot \text{Prob}\{\text{ch in } dx\} \\ &\quad + E\{w \text{ of } dx \mid \text{ho in } dx\} \cdot \text{Prob}\{\text{ho in } dx\} \\ &= 0 \end{aligned}$$

$$= \rho \cdot dx \cdot \text{Prob}\{\text{ch in } dx\}$$

$$= \rho \cdot dx \cdot \text{Prob}\{\text{dx dist closest ho to centre} > r\}$$

$$= \rho \cdot dx \cdot \frac{[\lambda \frac{4}{3} \pi r^3]^0}{0!} e^{-\lambda \frac{4}{3} \pi r^3}$$

$$= \rho e^{-\lambda \frac{4}{3} \pi r^3} dx$$

$$E\{\text{weight of } \mathcal{B}\} = \int \rho e^{-\lambda \frac{4}{3} \pi r^3} dx = \rho e^{-\lambda \frac{4}{3} \pi r^3} |\mathcal{B}|$$

$$= \underline{\rho e^{-\lambda \frac{4}{3} \pi r^3}}$$

c)

$D_1$  = dist. from arb. loc  $x \in D$  to closest ho-centre

= dist from ho-centre to closest ho-centre

Property of Poisson RT

$$\text{Prob}\{D_1 > d\} = \text{Prob}\left\{\text{--- empty}\right\} = \frac{\left[\lambda \frac{4}{3}\pi d^3\right]^0}{0!} e^{-\lambda \frac{4}{3}\pi d^3}$$

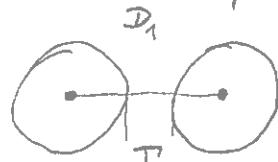
$$= e^{-\lambda \frac{4}{3}\pi d^3} \quad d > 0$$

$$\bar{F}_{D_1}(d) = \text{Prob}\{D_1 \leq d\} = 1 - e^{-\lambda \frac{4}{3}\pi d^3} \quad d > 0$$

$$f_{D_1}(d) = \frac{d\bar{F}_{D_1}}{dd} = 3\lambda \frac{4}{3}\pi d^2 e^{-\lambda \frac{4}{3}\pi d^3} \quad d > 0$$

$T$  - minimum cheese-thickness/circular hole

$$= [D_1 - 2r \mid D_1 > 2r]$$



Change of variable:

$$t = d - 2r \Rightarrow d = t + 2r$$

$$t > 0 \Rightarrow d > 2r$$

$$f_T(t) = f_{D_1}(t+2r \mid d > 2r) \quad t > 0$$

$$= \frac{f(t+2r)}{1 - \bar{F}_{D_1}(2r)}$$

$$= e^{+\lambda \frac{4}{3}\pi (2r)^3} \cdot 3\lambda \frac{4}{3}\pi (t+2r)^2 e^{-\lambda \frac{4}{3}\pi (t+2r)^3}; t > 0$$

$$= 3\lambda \frac{4}{3}\pi (t+2r)^3 e^{-\lambda \frac{4}{3}\pi [(t+2r)^3 - (2r)^3]} \quad t > 0$$

### Problem 3 Mosaic Random Field

$$\mathcal{L} : \{l_s; s \in \mathbb{L}_D\} \quad l_s \in [-1, 1] \quad \mathbb{L}_D \text{ reg. latt over } D \subset \mathbb{R}^2$$

Ising model - par  $\beta$

2)  $\text{Prob}\{l=l\} = \text{const} \times \exp\left\{-\sum_{\substack{u, v \in \mathbb{L}_D \\ \langle u, v \rangle}} \beta l_u l_v\right\}$

3) Markov formulation

$$\begin{aligned} & \text{Prob}\{l_s = l_s | l_r = l_r; r \in \delta(s)\} \\ &= \frac{\exp\left\{-\sum_{\substack{u \in \mathbb{L}_D \\ u \in \delta(s)}} \beta l_s l_u\right\}}{\sum_{\substack{l'_s \in \{-1, 1\} \\ l'_s \neq l_s}} \exp\left\{-\sum_{\substack{u \in \mathbb{L}_D \\ u \in \delta(s)}} \beta l'_s l_u\right\}} \quad ; \forall s \in \mathbb{L}_D \end{aligned}$$

$$\begin{aligned} \delta(s) &= \begin{array}{|c|c|} \hline \bullet & \cdot \\ \hline \cdot & \bullet \\ \hline \end{array} \quad - \quad s = (i, j) \\ \delta(s) &= [(i-1, j), (i, j-1), (i+1, j), (j, i+1)] \end{aligned}$$

$$b) f(d | L = l) = \prod_{t \in L_D} f(d_t | L_r = l_r, r \in V(t))$$

$$\begin{aligned} \text{Prob}\{L | d\} &= \text{const} \times f(d | L = l) \text{Prob}\{L = l\} \\ &= \text{const} \prod_{t \in L_D} f(d_t | L_r = l_r; r \in V(t)) \exp\left\{-\sum_{\substack{u, v \in L \\ \langle u, v \rangle}} \beta_{uv} l_u l_v\right\} \end{aligned}$$

~~$\sum$~~

Markov formulation:

$$\text{Prob}\{L_s = l_s | L_{-s} = l_{-s}, d\}$$

$$= \frac{\text{Prob}\left\{ \overbrace{L_s = l_s, L_{-s} = l_{-s}}^{L = l} | d \right\}}{\sum_{l_s \in \{-1, 1\}} \text{Prob}\{L_s = l_s, L_{-s} = l_{-s} | d\}}$$

$$= \frac{\left[ \text{const} \prod_{\substack{t \in L_D \\ t \in V(s)}} f(d_t | L_r = l_r; r \in V(t)) \exp\left\{-\sum_{\substack{u, v \in L \\ \langle u, v \rangle \\ u, v \neq s}} \beta_{uv} l_u l_v\right\} \right]}{\left[ \text{const} \prod_{\substack{t \in L_D \\ t \notin V(s)}} f(d_t | L_r = l_r; r \in V(t)) \exp\left\{-\sum_{\substack{u, v \in L \\ \langle u, v \rangle \\ u, v \neq s}} \beta_{uv} l_u l_v\right\} \right]}$$

$$\frac{\prod_{\substack{t \in L_D \\ t \in V(s)}} f(d_t | L_r = l_r; r \in V(t)) \cdot \exp\left\{-\sum_{u \in \delta(s)} \beta_{us} l_u\right\}}{\sum_{l_s \in \{-1, 1\}} \prod_{\substack{t \in L_D \\ t \in V(s)}} f(d_t | L_r = l_r; r \in V(t)) \cdot \exp\left\{-\sum_{u \in \delta(s)} \beta_{us} l_u\right\}}$$

$$= \text{Prob}\{L_s = l_s | d, L_r = l_r; r \in \delta(s) \cup \beta(s)\}$$

with  $\beta(s)$  - neighbor assoc with clique  $V(s)$   $\rightarrow$

