

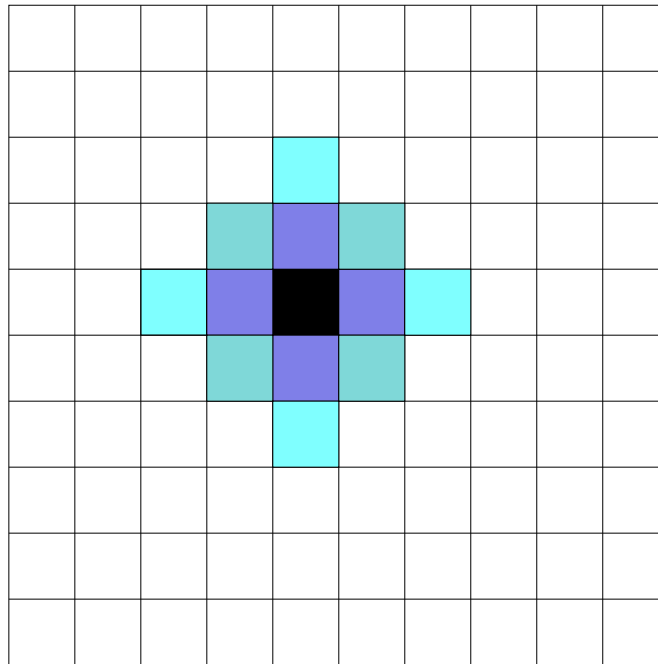
Objective Bayesian analysis for Proper Gaussian Markov Random Fields

Marco A. R. Ferreira (marco@im.ufrj.br)
Universidade Federal do Rio de Janeiro
www.dme.im.ufrj.br/~marco

Outline

- Proper Gaussian Markov random fields
- Confounding of parameters
- Reference prior
- Results for some simulated fields
- Frequentist properties
- Discussion

Proper MRF



Proper MRF

x = observed proper Gaussian Markov random field

$$p(x|\mu, \alpha, \beta) = N(x|\mu 1_n, \Sigma), \quad \Sigma^{-1} = \beta[Z + \alpha I_n]$$

$$\{Z\}_{kl} = \begin{cases} -1, & k \sim l, \\ h_k, & k = l, \\ 0, & \text{otherwise;} \end{cases}$$

h_k = number of neighbors of the block k .

Another look at the likelihood

$$p(x|\mu, \alpha, \beta) = (2\pi)^{-0.5n} \beta^{0.5n} \prod_{k=1}^n (\lambda_k + \alpha)^{0.5} \\ \exp \left\{ -0.5\beta \left[\sum_{k=1}^n (\alpha + h_k)(x_k - \mu)^2 \right. \right. \\ \left. \left. - \sum_{k=1}^n \sum_{l \in \partial k} (x_k - \mu)(x_l - \mu) \right] \right\},$$

$\lambda_1 \geq \lambda_2 \geq \dots, \geq \lambda_{n-1} > \lambda_n = 0$ are the eigenvalues of Z .

Confounding of α and β

When α is big enough:

$$p(x|\mu, \alpha, \beta) \approx (2\pi)^{-0.5n} (\alpha\beta)^{0.5n} \exp \left\{ -0.5\alpha\beta \left[\sum_{k=1}^n (x_k - \mu)^2 \right] \right\}$$

Blocks of x become approximately i.i.d. $N[\mu, (\alpha\beta)^{-1}]$.

Consequence: If we assign independent marginal improper priors for α and β , the marginal posteriors will be improper.

Reference prior

μ is a location parameter, so $\pi^r(\mu|\alpha, \beta) \propto 1$

Integrated likelihood with respect to μ :

$$\begin{aligned} p(x|\alpha, \beta) &= \int_{-\infty}^{\infty} p(x|\mu, \alpha, \beta) \pi^r(\mu|\alpha, \beta) d\mu \\ &= (2\pi)^{-0.5(n-1)} \beta^{0.5(n-1)} n^{-0.5} \prod_{k=1}^{n-1} (\lambda_k + \alpha)^{0.5} \\ &\quad \exp \left\{ -0.5\beta \left[\alpha \sum_{k=1}^n (x_k - \bar{x})^2 + \sum_{k=1}^n h_k x_k^2 - \sum_{k=1}^n \sum_{l \in \partial k} x_k x_l \right] \right\} \end{aligned}$$

Reference prior

The reference prior for (α, β) is:

$$\begin{aligned}\pi^r(\alpha, \beta) &\propto \sqrt{J(\alpha, \beta)} \\ &\propto \beta^{-1} \sqrt{(n-1) \sum_{i=1}^{n-1} (\lambda_i + \alpha)^{-2} - \left[\sum_{i=1}^{n-1} (\lambda_i + \alpha)^{-1} \right]^2},\end{aligned}$$

where $J(\alpha, \beta)$ is the determinant of the Fisher information matrix.

α and β are independent a priori.

Propriety of prior and posterior

When α goes to zero the prior converges to a constant.

Using first order Taylor expansions around one of x^{-1} and x^{-2} :

$$\begin{aligned}\pi^r(\alpha) &\propto \alpha^{-1} \sqrt{(n-1) \sum_{k=1}^{n-1} \left(1 + \frac{\lambda_k}{\alpha}\right)^{-2} - \left[\sum_{k=1}^{n-1} \left(1 + \frac{\lambda_k}{\alpha}\right)^{-1}\right]^2} \\ &\approx \alpha^{-1} \sqrt{(n-1) \sum_{k=1}^{n-1} \left(1 - 2\frac{\lambda_k}{\alpha}\right) - \left[\sum_{k=1}^{n-1} \left(1 - \frac{\lambda_k}{\alpha}\right)\right]^2} \\ &= \alpha^{-2} \left(\sum_{k=1}^{n-1} \lambda_k\right)\end{aligned}$$

Thus, for large α the prior behaves like α^{-2} .

Therefore, the prior of α is integrable and provides a proper marginal posterior distribution for α .

Results for simulated fields

$\alpha^{(true)}$

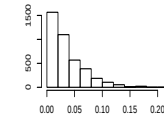
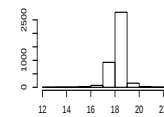
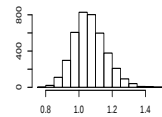
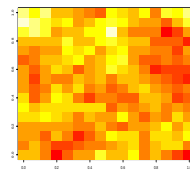
x

β

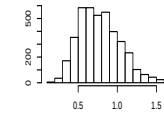
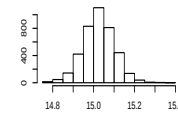
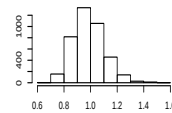
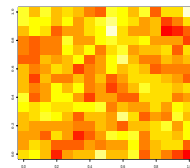
μ

α

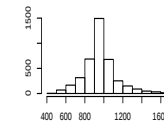
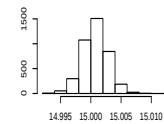
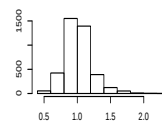
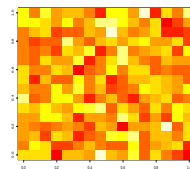
0.001



1.0

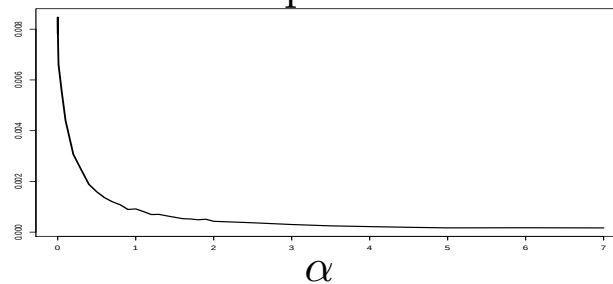


1000.0

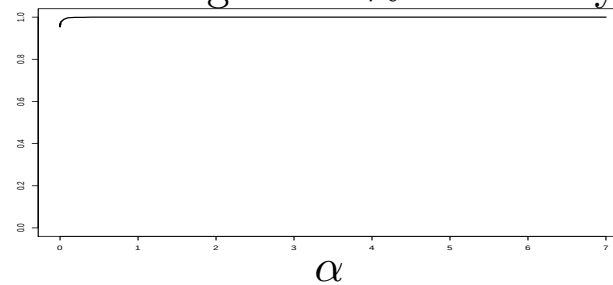


Frequentist properties - Estimation of β

Mean square error

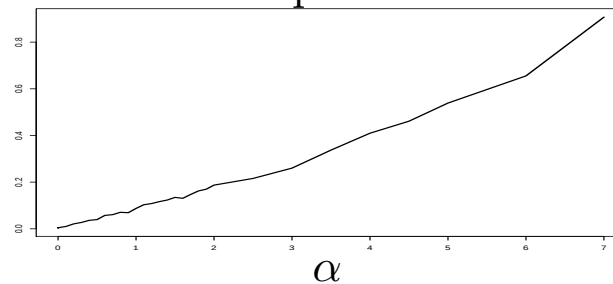


Frequentist coverage of 95% credibility intervals

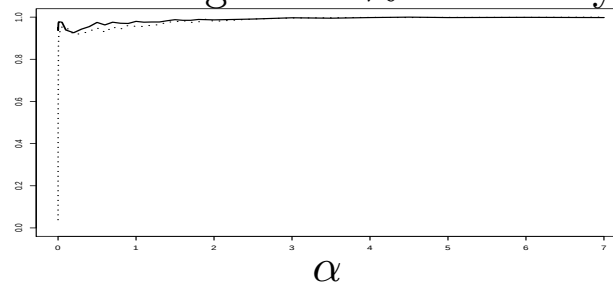


Frequentist properties - Estimation of α

Mean square error

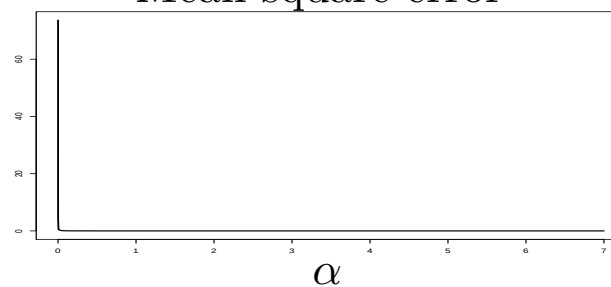


Frequentist coverage of 95% credibility intervals

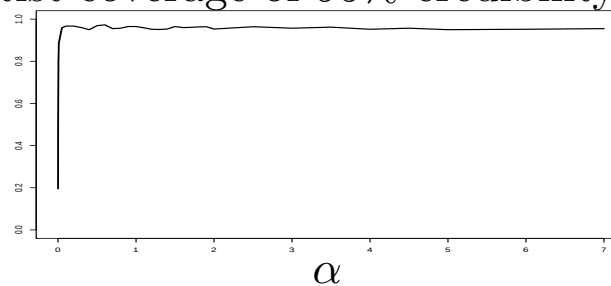


Frequentist properties - Estimation of μ

Mean square error



Frequentist coverage of 95% credibility intervals



Discussion

- Inference for β is robust with respect to $\alpha^{(true)}$
- HPD credibility interval for α works well for all values of $\alpha^{(true)}$
- Proposed objective Bayesian analysis works very well for $\alpha^{(true)} \geq 0.05$:
 - MSE of the estimators of β , μ and α is small
 - Frequentist coverage of central credibility intervals for β , μ and α is close to nominal value
- When $\alpha^{(true)}$ approaches zero field gets closer to improper:
 - Estimation of μ is more problematic, marginal posterior has heavy tails
 - Frequentist coverage of central credibility intervals for μ and α becomes too small