## 2023 STAT 554/CSSS 554/SOC 534 Statistical Methods for Spatial Data Assignment 2

To be submitted to the canvas site by the start of class on Monday 6th February, 2022.

Hand in your R code as an appendix.

In this question we will carry out disease mapping for lung cancer mortality data for men in the Valencia region of Spain from 1991-2000. For more details on the data, see Martinez-Beneito et al. (2019, Disease Mapping). On the class website is a data file that contains, for each subarea: observed deaths and expected deaths (adjusted for reference rate only), polygon files, and a graph file for INLA.

Use the following code to read in the data:

- # Set your working directory to where HW2data.Rdata and VR.graph are located load(HW2data.Rdata)
- # Expected counts are stored in Exp.mv3[,"Lung"]
- # Observed counts are stored in Obs.mv3[,"Lung"]
- # The shapefile for plotting is VR.cart

Let  $Y_i$  and  $E_i$ ,  $i=1,\ldots,n$ , denote the observed and expected counts in region  $i, i=1,\ldots,n$ . Then consider the model

$$Y_i|\theta_i \sim \mathsf{Poisson}(E_i\theta_i).$$
 (1)

- 1. (a) Provide a map of the observed counts  $Y_i$ .
  - (b) Provide a map of the expected counts  $E_i$ .
  - (c) Provide a map of the SMRs, defined as

$$\mathsf{SMR}_i = \widehat{\theta}_i = \frac{Y_i}{E_i},$$

for i = 1, ..., n. Comment on the variability of the SMRs.

- (d) Plot the SMRs versus the estimated standard errors, which are given by  $\sqrt{\widehat{\theta}_i/E_i}$ .
- 2. In this question we will smooth the SMRs using the disease mapping Poisson-Lognormal model:

$$\begin{split} Y_i|\beta_0, \epsilon_i &\sim_{ind} & \mathsf{Poisson}(E_i \mathsf{e}^{\beta_0} \mathsf{e}^{e_i}) \\ e_i|\sigma_e^2 &\sim_{iid} & \mathsf{N}(0, \sigma_e^2) \end{split}$$

for i, i = 1, ..., n.

- (a) Using the inla function in R fit this model using the default priors for  $\beta_0$  and  $\sigma_e$ . Report the posterior medians and 95% intervals for  $\beta_0$  and for  $\sigma_e$ .
- (b) Extract the posterior medians of the relative risk (RR) estimates and provide a map of these.
- (c) Plot these posterior RR estimates against the SMRs, and comment.
- (d) Plot the posterior standard deviations of the RRs against the standard errors of the SMRs and comment.
- 3. In this question we will smooth the SMRs for the Valencian lung cancer data using the disease mapping Poisson-Lognormal-Spatial model:

$$\begin{array}{ccc} Y_i | \beta_0, S_i, \epsilon_i & \sim_{iid} & \mathsf{Poisson}(E_i \mathsf{e}^{\beta_0} \mathsf{e}^{S_i + \epsilon_i}) \\ & \epsilon_i | \sigma^2_{\epsilon} & \sim_{iid} & \mathsf{N}(0, \sigma^2_{\epsilon}) \\ S_1, \dots, S_n | \sigma^2_s & \sim & \mathsf{ICAR}(\sigma^2_s) \end{array}$$

for i, i = 1, ..., n.

(a) Using the inla function in R fit this model using the bym2 model, with the default prior for  $\beta_0$  and the following prior specification for the spatial and non-spatial random effects (note that you must be in the directory that contains the VR.graph file):

```
f(Region, model="bym2", graph="VR.graph", scale.model=T, constr=T,
hyper=list(phi=list(prior="pc", param=c(0.5, 0.5), initial=1),
prec=list(prior="pc.prec", param=c(0.3,0.01), initial=5)))
```

These choices correspond to the prior belief that there is a 1% chance that the total residual standard deviation is greater than 0.3, and a 50% chance that the proportion of the variance that is spatial is bigger than 0.5. Report both the posterior medians and 95% intervals for  $\beta_0$ , the total variance of the random effects, and the proportion of the total variance attributed to the spatial random effect.

- (b) Extract the relative risk estimates and provide a map of these. **Compare** these estimates with the SMRs and with those obtained from the Poisson-Lognormal model (i.e., the model with **IID random** effects only) that you fit in **Question 2**.
- 4. **Bonus Question:** Suppose that instead of having available the counts and expected numbers we have access to the relative risks  $\widehat{\theta}_i = Y_i/E_i$  and their standard errors  $\sigma_{\epsilon i} = \sqrt{\widehat{\theta}_i/E_i}$ .

Take the data as  $Z_i = \log \widehat{\theta}_i$  and fit the model

$$Z_i = \log \hat{\theta}_i \sim \mathsf{N}(\beta_0 + e_i, \sigma_{\epsilon i}^{\star 2}),$$

where  $\sigma_{\epsilon i}^{\star 2} = \sigma_{\epsilon i}^2/\widehat{\theta}_i^2 = 1/(E_i\widehat{\theta}_i)$ . We then place smoothing priors on via  $e_i \sim_{iid} N(0, \sigma_e^2)$ .

Fit this model using inla and plot the estimated relative risks from this model against the estimates from the Poisson-Lognormal model, and comment.