TMA4300 Project 2

Magnus Grytten & Petter J. Gudbrandsen

```r
# Import classes
library("INLA")
```

```
## Loading required package: Matrix

## Loading required package: foreach

## Warning: package 'foreach' was built under R version 4.1.3

## Loading required package: parallel

## Loading required package: sp

## Warning: package 'sp' was built under R version 4.1.3

## This is INLA_21.11.22 built 2021-11-21 16:10:15 UTC.
##  - See www.r-inla.org/contact-us for how to get help.
##  - Save 80Mb of storage running 'inla.prune()'
```
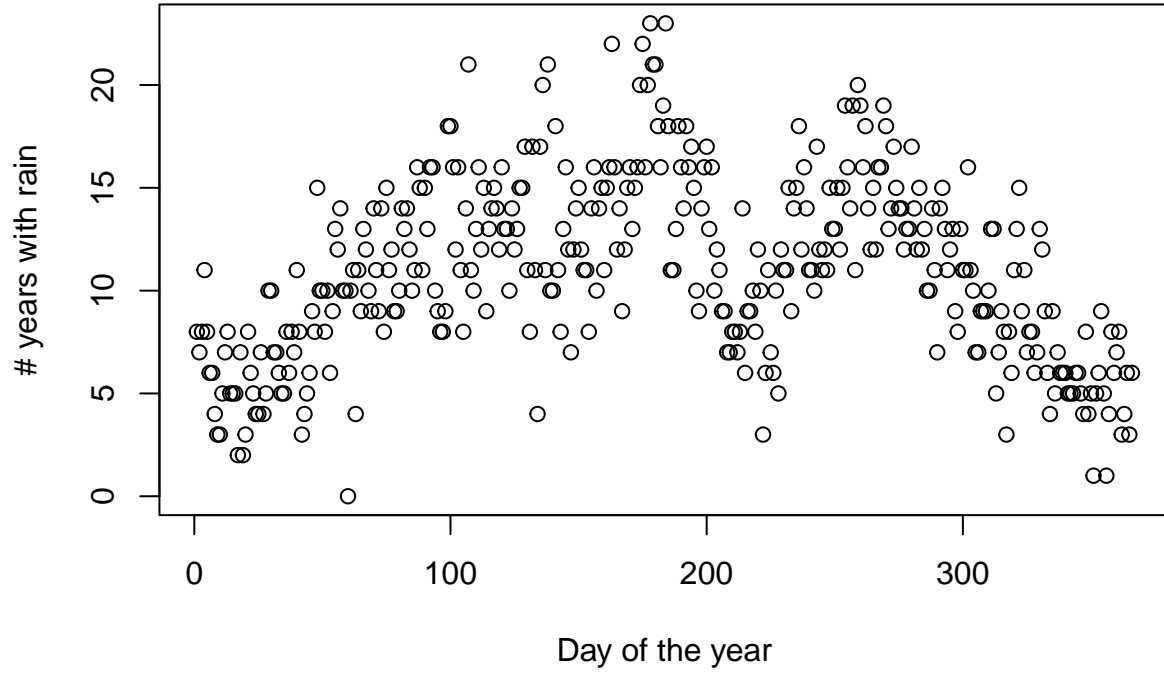
```r
# Load data
load("rain.rda") # set working directory correctly
```

Problem 1

a)

Explore the Tokyo rainfall dataset, plot the response as a function of t, and describe any patterns that you see.

```r
plot(rain$day, rain$n.rain, xlab = 'Day of the year', ylab = '# years with rain')
```

Day of the year

C: The plot shows the number of days with rain in each day of the year over a period of 39 years.

During the winter period there is little rain. It increases during the spring. After that it seems to have two 'down' bumps during the summer. During the autumn, from about day 275 and until new year, it decreases.

The points seem to follow a continuous function with a similar error around it throughout the year.

b)

The likelihood of y_t given $\pi(\tau_t)$ is given to be a binomial distribution.

$$p(y_t|\pi(\tau_t)) = \binom{n_t}{y_t}\pi(\tau_t)^{y_t}(1 - \pi(\tau_t))^{n_t - y_t}$$

c)

$$p(\sigma^2|\boldsymbol{y},\boldsymbol{\tau}) \propto p(\boldsymbol{y},\boldsymbol{\tau}|\sigma^2) \cdot p(\sigma^2)$$
$$= p(\boldsymbol{y}|\boldsymbol{\tau},\sigma^2) \cdot p(\boldsymbol{\tau}|\mathbf{y},\sigma^2) \cdot p(\sigma^2)$$
$$= p(\boldsymbol{y}|\boldsymbol{\tau}) \cdot p(\boldsymbol{\tau}|\sigma^2) \cdot p(\sigma^2)$$
$$= \prod_{t=1}^{T} \binom{n_t}{y_t} \pi(\tau_t)^{y_t} (1 - \pi(\tau_t))^{n_t - y_t}$$
$$\cdot \prod_{t=2}^{T} \frac{1}{\sigma_u} e^{-\frac{1}{2\sigma_u^2}(\tau_t - \tau_{t-1})^2}$$
$$\cdot \frac{\beta^\alpha}{\Gamma(\alpha)} \left(\frac{1}{\sigma_u^2}\right)^{\alpha+1} e^{-\frac{\beta}{\sigma_u^2}}$$

We remove factors without σ_u.

$$\propto \prod_{t=2}^{T} \frac{1}{\sigma_u} e^{-\frac{1}{2\sigma_u^2}(\tau_t - \tau_{t-1})^2} \cdot \left(\frac{1}{\sigma_u^2}\right)^{\alpha+1} \cdot e^{-\frac{\beta}{\sigma_u^2}}$$
$$= \frac{1}{\sigma_u^{T-1}} e^{-\frac{1}{2\sigma_u^2}\boldsymbol{\tau}^t \boldsymbol{Q}\boldsymbol{\tau}} \cdot \left(\frac{1}{\sigma_u^2}\right)^{\alpha+1} \cdot e^{-\frac{\beta}{\sigma_u^2}}$$
$$= \left(\frac{1}{\sigma_u^2}\right)^{\alpha+1+\frac{T-1}{2}} e^{-\frac{1}{2\sigma_u^2}(\boldsymbol{\tau}^t \boldsymbol{Q}\boldsymbol{\tau} + \beta)}$$

In the last expression we recognize the core of an inverse gamma distribution:

$$IG \sim \left(\alpha + \frac{T-1}{2}, \frac{\boldsymbol{\tau}^t \boldsymbol{Q}\boldsymbol{\tau} + \beta}{2}\right)$$

for shape α and rate β.

d)

$$p(\boldsymbol{\tau},\sigma^2|\boldsymbol{y}) \propto p(\boldsymbol{y}|\boldsymbol{\tau},\sigma^2) \cdot p(\boldsymbol{\tau},\sigma^2)$$
$$= p(\boldsymbol{y}|\boldsymbol{\tau}) \cdot p(\boldsymbol{\tau}|\sigma^2) \cdot p(\sigma^2)$$
$$= p(\boldsymbol{y}_I|\boldsymbol{\tau}_I) \cdot p(\boldsymbol{y}_{-I}|\boldsymbol{\tau}_{-I}) \cdot p(\boldsymbol{\tau}_I|\sigma^2) \cdot p(\boldsymbol{\tau}_{-I}|\boldsymbol{\tau}_I,\sigma^2) \cdot p(\sigma^2)$$

$$\alpha = min\left(1, \frac{p(\boldsymbol{\tau'}_I,\sigma^2|\boldsymbol{y}) \cdot p(\boldsymbol{\tau}_I|\boldsymbol{\tau'}_{-I},\sigma^2)}{p(\boldsymbol{\tau}_I,\sigma^2|\boldsymbol{y}) \cdot p(\boldsymbol{\tau'}_I|\boldsymbol{\tau}_{-I},\sigma^2)}\right)$$