SPDE one dimensional example

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In this example we show row to analyse a time series of daily temperature using a one dimension SPDE model. More details about it are on the paber at https://www.jstatsoft.org/article/view/v063i19

1 The data

We consider the daily weather data available at http://www.yr.no/. We have the following set the URL for the daily data for Trondheim considerint in the last 13 months

One can read and extract the desired data table (the second one at the URL) using the **XML** package with the readHTMLTable() function. However, it still need some pre-processing.

We do not consider it and just read the web page as a text and play with the text and its structure directly.

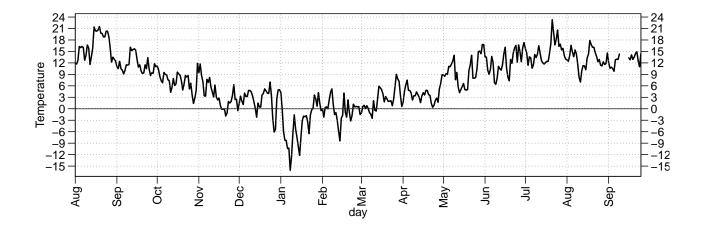
```
d0 <- readLines(u0) ### read it as text (Done at 26 September 2016)
```

First, we have to find the index for each table line and consider only those for the main table:

```
i <- grep('<tr>', d0) ### index for each table line
i <- i[i>grep('', d0)[2]] ### select those for the second table
```

The desired data we would like to analyse is the minimum and maximum temperature. Commands to extract and pre-process these data

Visualize it with the following commands



2 Model fitting

Mesh in 1d it is a matter of chosing a set of knots, the order of the basis functions and the boundary. Choosing first order basis function and Neumann boundary.

Projector matrix is the matrix built to project the process at the mesh nodes to the locations

```
A <- inla.spde.make.A( ## projector creator
    mesh=mesh, ## provide the mesh
    loc=coo) ### locations where to project the field
dim(A) ## an 'n' by 'm' projector matrix
## [1] 422 60
summary(rowSums(A)) ### each line sums up to one
##
      Min. 1st Qu.
                    Median
                              Mean 3rd Qu.
                                               Max.
##
         1
                 1
                         1
                                 1
summary(colSums(A)) ### 'how many' observations per knot
##
      Min. 1st Qu.
                    Median
                              Mean 3rd Qu.
                                               Max.
     7.000 7.000 7.000
                             7.033 7.000
                                              8.500
```

Build the SPDE model on the mesh, choosing the smoothness (α) , to define the precision structure, and priors for the hyperparameters. We just set $\alpha = 2$ and keep the default priors.

```
spde <- inla.spde2.pcmatern( ## model for the precision
   mesh=mesh, ## mesh supplied
   alpha=2, ## smoothness parameter
   prior.range = c(1, 0.01), ## P(range < 1) = 0.01
   prior.sigma = c(1, 0.5)) ## P(sigma > 1) = 0.5
```

Create a data stack in order to organize the data. This is a way to allow models with complex linear predictors. In our case, we have a SPDE model defined on m nodes. It must be combined with the covariate (and the intercept) effect at n locations. We do it using different projector matrices.

```
stk.e <- inla.stack( ## stack creator
data=list(y=tmed), ## response
effects=list(## two elements:
    data.frame(b0=rep(1, n)), ## regressor part
    i=1:spde$n.spde), ## RF index
A=list(## projector list of each effect
    1, ## for the covariates
    A), ## for the RF
tag='est') ## tag</pre>
```

Fit the posterior marginal distributions for all model parameters, syplying the model formula, data and some additional controls to the main function in the **INLA** package

```
formula <- y ~ 0 + b0 + ## fixed part
  f(i, model=spde) ## RF term
res <- inla( ## main function in INLA package
  formula, ## model formula
  data=inla.stack.data(stk.e), ## dataset
  control.predictor=list( ## inform projector needed in SPDE models
    A = inla.stack.A(stk.e), compute=TRUE)) ## projector from the stack data</pre>
```

3 Posterior marginal distributions - PMDs

Summary of the regression coefficients PMDs

```
round(res$summary.fixed, 4)

## mean sd 0.025quant 0.5quant 0.975quant mode kld
## b0 7.438 1.7138    4.0181    7.4395    10.8444 7.4421    0
```

The PMDs summary for the Gaussian likelihood precision and the two RF parameters

```
round(res$summary.hyperpar, 4)
## data frame with 0 columns and 0 rows
```

We have to transform the likelihood precision PMD to have the variance PMD. It can be done by

```
m.prec <- res$marginals.hyperpar$'Precision for the Gaussian observations' ## the marginal
post.s2e <- inla.tmarginal(## function to compute a transformation
  function(x) sqrt(1/x), ## inverse transformation and square root
  m.prec) ## marginal to be applied

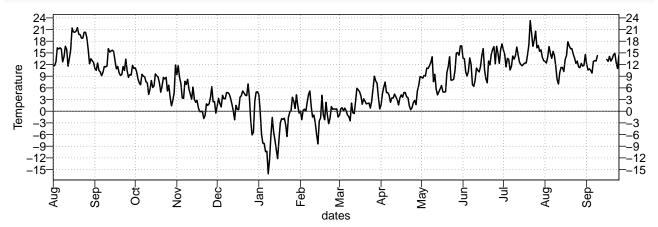
## Error in seq.default(xmin, xmax, len = n): 'from' must be a finite number</pre>
```

The PMDs for likelihood standard deviation and the RF parameters can be visualized by

4 Predicted

Visualize it with the commands bellow

```
par(mfrow=c(1,1), mar=c(3,3,0.3,2), mgp=c(2,0.5,0), las=2, xaxs='i')
id <- inla.stack.index(stk.e, tag='est')$data
plot(dates, tmed, type='l', axes=FALSE, ylab='Temperature', lwd=2)
for (j in 3:5)
   lines(dates, res$summary.fitted.values[id, j], lty=2, lwd=2)
box(); axis(2, 3*(-8:9)); axis(4, 3*(-8:9))
axis(1, pd, months(pd, T))
abline(h=0)
abline(h=3*(-8:9), v=pd, lty=3, col=gray(.5))</pre>
```



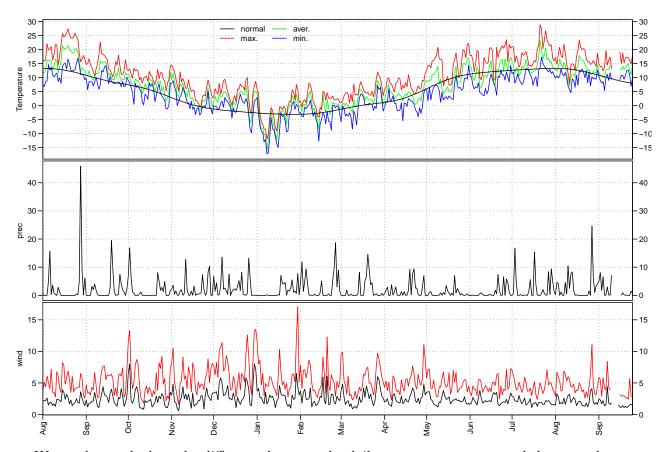
5 Just a look to the rest of the data

Pre-processing the maximum, minimum and normal temperature, the precipitation, and the average and maximum wind:

```
tmax <- as.numeric(gsub('<td>', '', gsub('°', '', d0[i+2])))
tmin <- as.numeric(gsub('<td>', '', gsub('°', '', d0[i+3])))
tnormal <- as.numeric(gsub('<td>', '', gsub('°', '', d0[i+5])))
prec <- as.numeric(gsub('<td>', '', gsub('mm', '', d0[i+6])))
wind <- as.numeric(gsub('<td>', '', gsub('m/s', '', d0[i+10])))
wmax <- as.numeric(gsub('<td>', '', gsub('m/s', '', d0[i+9])))
```

Visualize it

```
par(mfrow=c(3,1), mar=c(0.1,3,0.1,2), mgp=c(2,.7,0), las=2, xaxs='i')
plot(dates, tmed, type='l', ylim=range(tmin, tmax, na.rm=TRUE),
     axes=FALSE, xlab='', ylab='Temperature', col='green')
lines(dates, tmin, col='blue')
lines(dates, tmax, col='red')
lines(dates, tnormal)
legend(dates[which.min(tmin)], par()$usr[4], c('normal', 'max.', 'aver.', 'min.'),
      col=1:4, lty=1, ncol=2, xjust=0.5, bty='n')
abline(h=5*(-5:6), v=pd, lty=3, col=gray(.5))
box(); axis(2, 5*(-5:6)); axis(4, 5*(-5:6))
plot(dates, prec, type='l', axes=FALSE, xlab='')
box(); axis(2); axis(4)
abline(v=pd, h=10*(1:4), lty=3, col=gray(0.5))
par(mar=c(3, 3, 0.1, 2), new=FALSE)
plot(dates, wind, type='l', axes=FALSE, xlab='',
     ylim=range(wind, wmax, na.rm=TRUE))
lines(dates, wmax, col=2)
box(); axis(2); axis(4)
abline(v=pd, h=5*(1:3), lty=3, col=gray(0.5))
axis(1, pd, months(pd, TRUE))
```



We can have a look at the difference between the daily mean temperature and the normal temperature. The current normal is the average over the period from 1961 to 1990.

