Spacetime models in R-INLA

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Outline

Separable space-time models

PM-10 concentration in Piemonte, Italy

Kronecker product models

•
$$\mathbf{x} = \{x_{11}, ..., x_{n1}, x_{12}, ..., x_{nT}\}$$

assume

$$\pi(\mathbf{x}) \propto (|\mathbf{Q}\mathbf{1}\otimes\mathbf{Q}\mathbf{2}|^*)^{1/2} \exp\left(-rac{1}{2}\mathbf{x}^{\mathcal{T}}\{\mathbf{Q}\mathbf{1}\otimes\mathbf{Q}\mathbf{2}\}\mathbf{x}
ight)$$

where $|.|^*$ is the generalized determinant

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kronecker product model example in R-INLA

f(spatial, model='besagproper2',
 group=time, control.group=list(model='ar1'))

Spacetime interactions

- kronecker product models follows Clayton's rule
- combine Q1 and Q2 available
- warning care when main effects are in the model
- WARNING super care when Q1 and/or Q2 have rank deficiency
- ▶ the described dynamic model is type IV and uses Q2 as AR(1)

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Space-time dynamic intercept

► The (linear) measurement equation

$$\mathbf{y}_{it} = F'_{it}\beta + \mathbf{A}_{i(t)}\mathbf{x}_t + \epsilon_{it}$$

- **F**_t is a matrix of covariates
- $\blacktriangleright~\beta$ are the fixed effects
- $\mathbf{A}_{(t)}$ picks out the appropriate values of \mathbf{x}_t

•
$$\epsilon_t \stackrel{\text{i.i.d.}}{\sim} N(0, \sigma^2 \mathbf{I})$$

Space-time dynamic intercept

The (linear) measurement equation

$$\mathbf{y}_{it} = \mathbf{F}_{it}^{\prime} \boldsymbol{\beta} + \mathbf{A}_{i(t)} \mathbf{x}_t + \epsilon_{it}$$

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vector AR(1) process for x

$$\mathbf{x}_t = \rho \mathbf{x}_{t-1} + \boldsymbol{\omega}_t$$

• ω_t : spatial SPDE model

$$\omega_t \overset{\text{i.i.d.}}{\sim} N(\mathbf{0}, \mathbf{Q}^{-1}),$$

 $\blacktriangleright \ \rho$ is the time correlation

PM-10 concentration in Piemonte, Italy

Cameletti et al. (2011), on r-inla.org

- 24 monitoring stations
- Daily data from 10/05 to 03/06

Space model part

```
    Make the latent model
```

```
spde = inla.create.spde(mesh,model="matern")
```

Using the group feature

Construct a kronecker product model using the group feature formula = y ~ -1 + intercept + WS + HMIX + ... + f(field, model=spde, group =time, control.group=list(model="ar1"))

- This tells INLA that the observations are grouped in a certain way.
- control.group contains the grouping model (ar1, exchangable, rw1, and others) as well as their prior specifications.

Make an A matrix

```
Use the group argument
LocationMatrix = inla.spde.make.A(mesh = mesh,
loc =dataLoc, group=time, n.group=nT)
```

- data locations in all group=time level
- builds an A matrix in an appropriate way

Organising the data

Covariates at the data points, but the latent field only defined their through the A matrix

We need to make sure that **A** only applies to the random effect.