# Big Geospatial Data - Home assignment 2

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Note

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#### Problem - Simulation of dependent random processes

Simulate spatial lattice data  $\{Z(s): s=(s_1,s_2)'\in Z^2, 1\leq s_1,s_2\leq d\}$  from a gaussian distribution, where the covariance is

$$Cov(Z(s_i), Z(s_i)) = a \exp(-b||h||) \qquad \text{for all} \quad i, j = 1, \dots, n$$

with  $h=s_i-s_j,\, a>0,\, b\geq 0.$  Choose initially  $d=25,\, a=0.2,\, b=0.15,$  and  $||\cdot||$  as Euclidean norm.

- 1. Compare the resulting random fields  $\{z(s)\}$  for different choices of a and b. How a and b can be interpreted?
- 2. Compare the results for this so-called exponential covariance model to
  - a) a spherical model,
  - b) a linear model,
  - c) and a power model.
- 3. Compare the results for different norms  $||\cdot||$ , e.g.  $||\cdot||_p$  for  $p \in \{1, 2, \infty\}$ . Do the norms have an impact on the spatial dependence? In order to compare the results, the same random seed should be specified.

# Problem - Monte Carlo simulation study - computation time

Simulate the above described spatial model for increasing sizes of the spatial random field. Perform a Monte Carlo simulation study with m = 100 replications to evaluate the computation time, if the number of observations n is increasing  $(n \in \{16, 100, 1024, 4900\})$ . Visualize your results graphically and shortly explain them.

# **Problem: Covariance tapering**

For covariance tapering, zeros are introduced into C in order to make it sparse The tapered covariance function is then given by the product

$$C_{tap}(s_i - s_i) = C(s_i - s_i)C_{\theta}(s_i - s_i)$$

Explain why the tapering matrix  $C_{\theta}$  must be chosen as valid covariance function (i.e., positive definite).

### **Problem: Simulation of sparse matrices**

Simulate binary matrices with 60, 80, 95 per cent zero elements. The matrices should be of dimension  $100 \times 100$  and the random number generator should be initialized by a random seed.

- 1. Explain the difference between sparse matrices and band matrices.
- 2. Permute the simulated matrix using the Cuthill-McKee algorithm. Compute the bandwidth of the permuted matrices.
- 3. Why band matrices are (sometimes) preferred in computational statistics?
- 4. Compute the determinant!

#### Problem: Monte Carlo simulation study - minimal bandwidth

Simulate the above described binary matrices and perform a Monte Carlo simulation study with m=1000 replications. Compute the average minimal bandwidth, which can be achieved by the Cuthill-McKee algorithm. Visualize your results graphically and shortly explain them.

# Problem: Monte Carlo simulation study - sparse matrices

Simulate the above described binary matrices and perform a Monte Carlo simulation study with m=1000 replications to evaluate the computation time and required memory (RAM) for computing the determinant. Assess the computational advantages

- 1. if a class for sparse matrices is used (e.g., C++ Eigen::SparseMatrix, Python scipy.sparse, R Matrix, ...),
- 2. if the matrices are permuted by the Cuthill-McKee algorithm.

Furthermore, how the computation time changes if the dimension of the matrices increases  $(20 \times 20, 50, 100 \times 100)$  Visualize all results graphically and shortly explain them.