

# Home Assignment - 1

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## 1 Introduction

## 2 Theory

Let  $\mathbf{Y}_k$  represent the  $n$  known observations of spatial data and  $\mathbf{Y}_u$  represent the  $m$  data points that needs to be estimated. For ordinary Kriging ( $\mu = \mathbf{1}\beta$ ), the optimal predictions are given by

$$\begin{aligned}\hat{\mathbf{Y}}_u &= \mathbf{1}_u \hat{\beta} + \Sigma_{uk} \Sigma_{kk}^{-1} (\mathbf{Y}_k - \mathbf{1}_k \hat{\beta}) \\ \hat{\beta} &= (\mathbf{1}_k^T \Sigma_{kk}^{-1} \mathbf{1}_k)^{-1} \mathbf{1}_k^T \Sigma_{kk}^{-1} \mathbf{Y}_k,\end{aligned}$$

where

$$\begin{bmatrix} \mathbf{Y}_k \\ \mathbf{Y}_u \end{bmatrix} \in \mathbf{N} \left( \begin{bmatrix} \mathbf{1}_k \beta \\ \mathbf{1}_u \beta \end{bmatrix}, \begin{bmatrix} \Sigma_{kk} & \Sigma_{ku} \\ \Sigma_{uk} & \Sigma_{uu} \end{bmatrix} \right).$$

Firstly, we show that the predictions are linear in the observations, i.e.,  $\hat{\mathbf{Y}}_u = \boldsymbol{\lambda}^T \mathbf{Y}_k$  for some  $\boldsymbol{\lambda}$ . We may re-write  $\hat{\beta}$  as

$$\begin{aligned}\hat{\beta} &= \boldsymbol{\gamma}^T \mathbf{Y}_k \\ \boldsymbol{\gamma} &= \left( (\mathbf{1}_k^T \Sigma_{kk}^{-1} \mathbf{1}_k)^{-1} \mathbf{1}_k^T \Sigma_{kk}^{-1} \right)^T.\end{aligned}$$

We note that the vector  $\boldsymbol{\gamma}^T$  is of dimensions  $1 \times n$ , and therefore,  $\hat{\beta}$  is a scalar. Substituting the above in the expression for  $\hat{\mathbf{Y}}_u$ , we obtain

$$\begin{aligned}\hat{\mathbf{Y}}_u &= \mathbf{1}_u \boldsymbol{\gamma}^T \mathbf{Y}_k + \Sigma_{uk} \Sigma_{kk}^{-1} \mathbf{Y}_k - \Sigma_{uk} \Sigma_{kk}^{-1} \mathbf{1}_k \boldsymbol{\gamma}^T \mathbf{Y}_k \\ &= (\mathbf{1}_u \boldsymbol{\gamma}^T + \Sigma_{uk} \Sigma_{kk}^{-1} - \Sigma_{uk} \Sigma_{kk}^{-1} \mathbf{1}_k \boldsymbol{\gamma}^T) \mathbf{Y}_k \\ &= \boldsymbol{\lambda}^T \mathbf{Y}_k,\end{aligned}$$

where  $\boldsymbol{\lambda} = (\mathbf{1}_u \boldsymbol{\gamma}^T + \Sigma_{uk} \Sigma_{kk}^{-1} - \Sigma_{uk} \Sigma_{kk}^{-1} \mathbf{1}_k \boldsymbol{\gamma}^T)^T$ .

Secondly, we show that the predictions are unbiased.

## 3 Conclusions