

LatticeKrige

Group member: Fan, Bai, Pammer

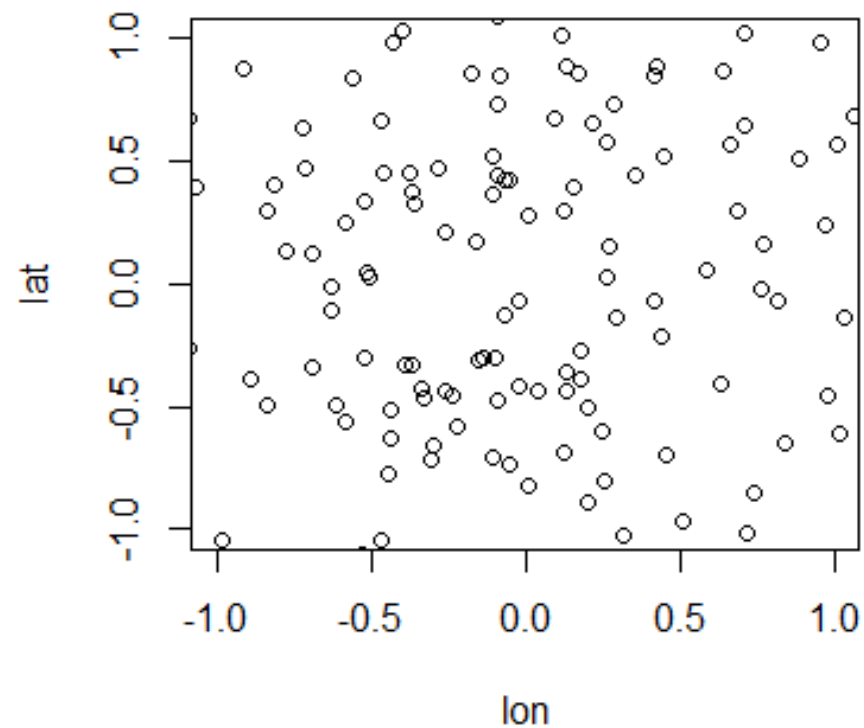
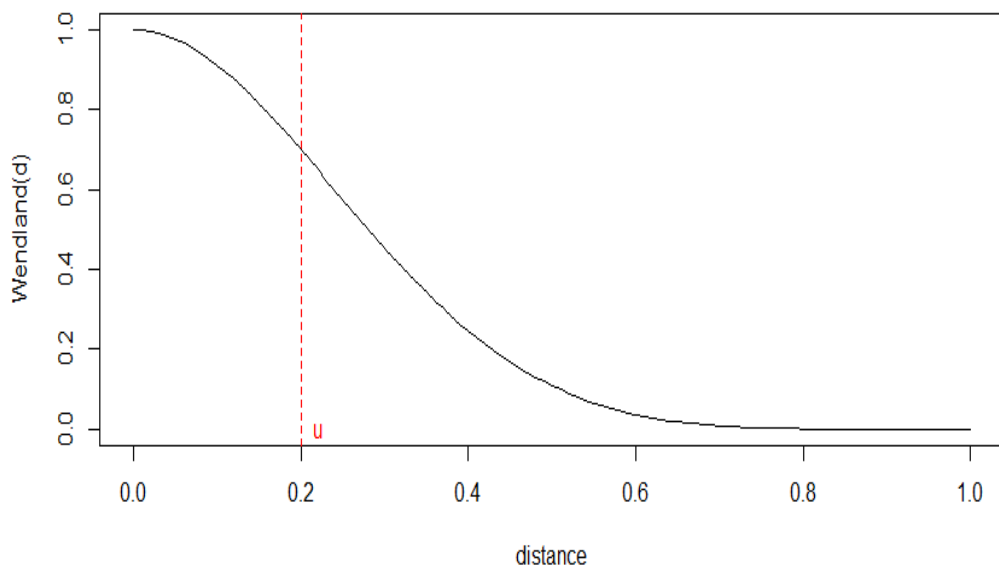
Introduction

- Recall the original model: $Y = X\beta + Z + E$
- If the data is large enough, the calculation of the inverse of $Cov(Z) = \Sigma(\theta)$ is time-consuming when using mle.
- Low rank method: $Z = poly(lon, lat)$. $\Sigma(\theta)$ is hard to approximate.

Introduction

Approximate the covariance function

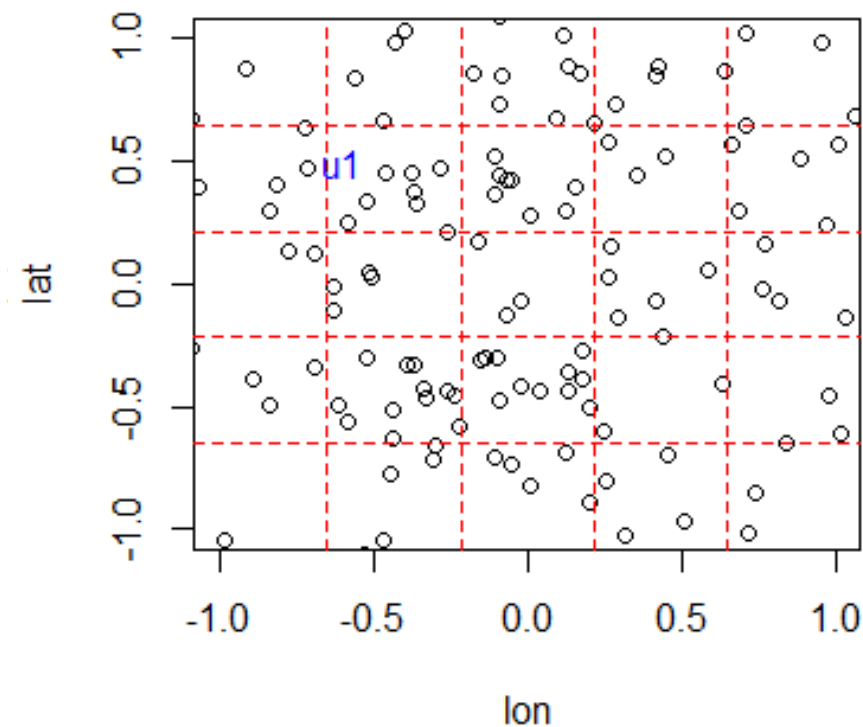
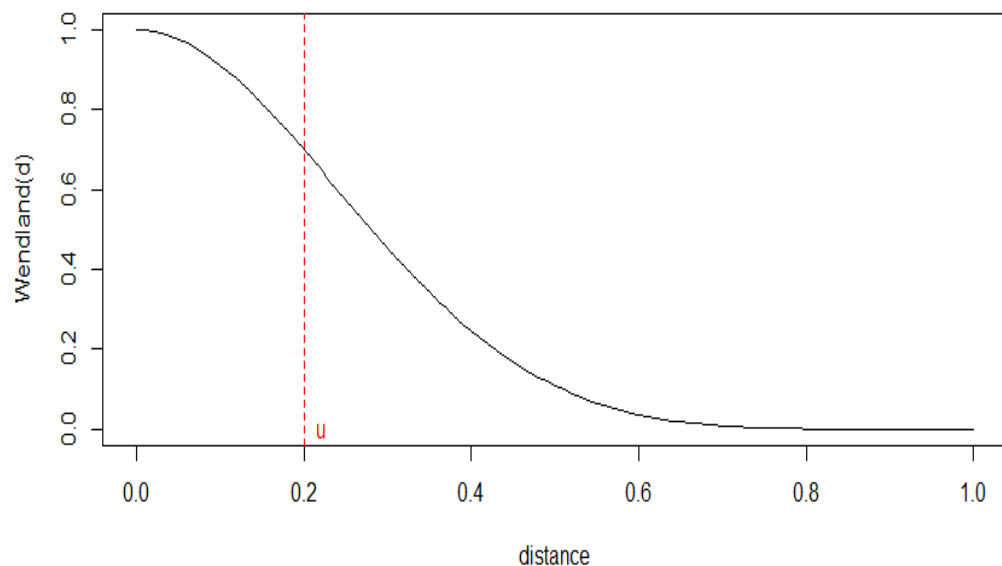
- $Z_i = \sum_{l=1}^L g_l(s_i)$
- $g(s_i) = \sum_{j=1}^m c_j \phi_j(s_i)$
- $\phi_j(s_i) = \phi(\|u_j - s_i\|/\theta)$



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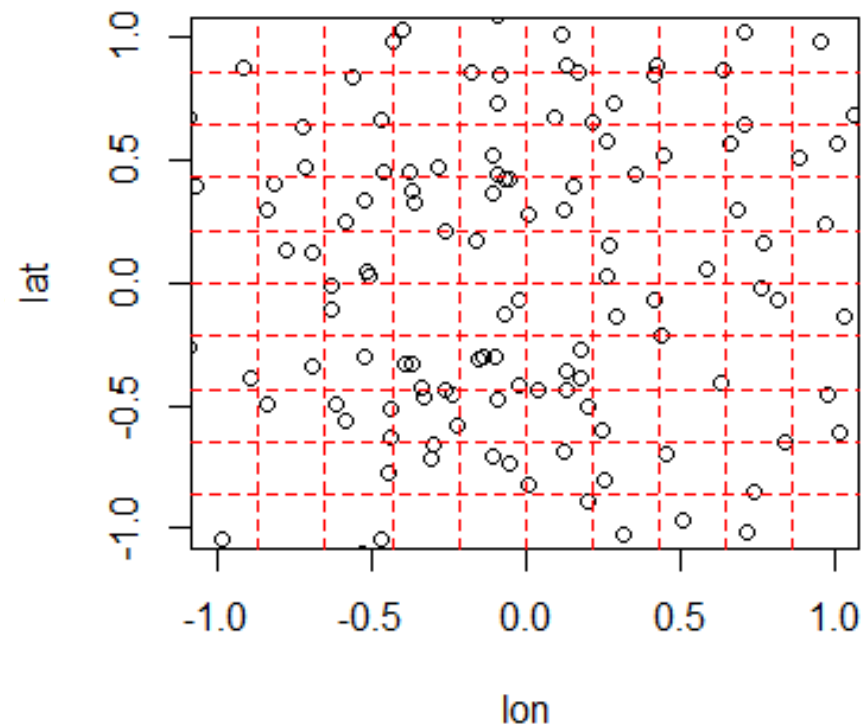
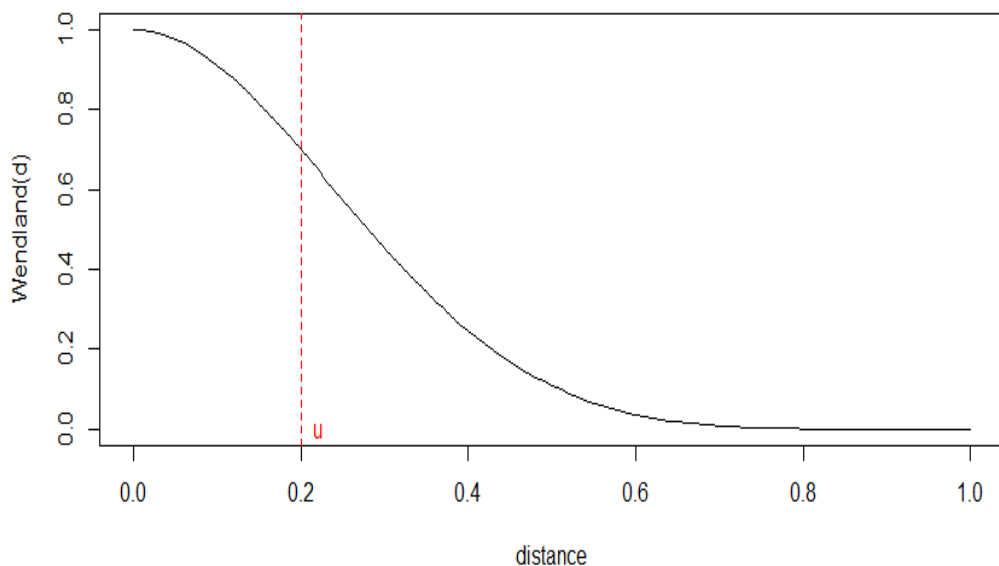
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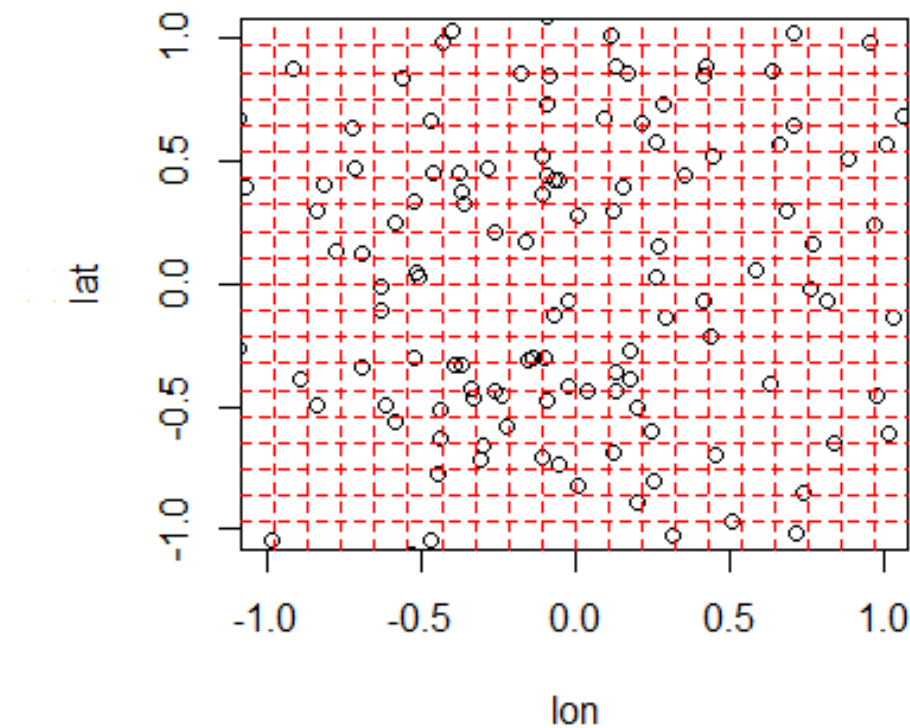
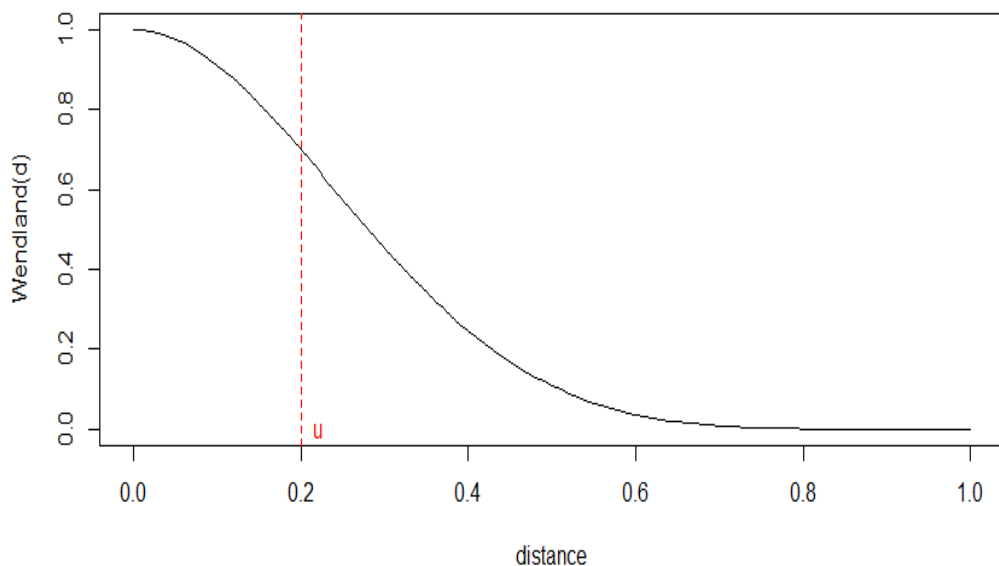
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Introduction

Simplify the calculation

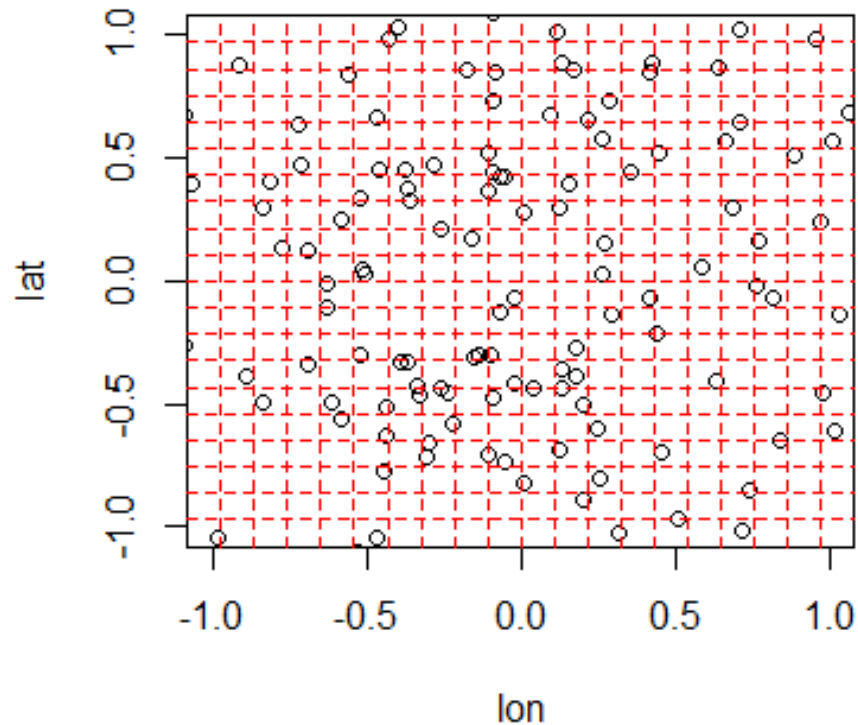
$$g_l(s_i) = \sum_{j=1}^{m(l)} c_j^l \phi_{j,l}(s_i) \quad \text{cov}(g_l(s), g_l(s')) = \sum_{j,k=1}^{m(l)} \rho Q_{j,k}^{-1} \phi_{j,l}(s), \phi_{k,l}(s')$$

For each level l:

$$C = (c_1, \dots, c_{m(l)}) \sim MN(0, Q^{-1}) \quad \hat{C} = Q^{-1} \phi^T M_\lambda^{-1} (Y - X\beta) \quad M_\lambda^{-1} = \phi Q^{-1} \phi^T + \lambda W^{-1}$$

The calculation of \hat{C} is fast

Introduction



Turning parameters

NC: grid points

nlevel: level of grid

nu: control of variance for each grid level

$$\alpha_j = \text{var}(g_l(s_i)), \alpha_j = e^{-2j\nu}$$

Implementation details

- R package: LatticeKrig
- Functions used:
 - LKrigSetup(): create a list describing the LatticeKrig spatial model.
`LKinfo <- LKrigSetup(x=, NC = , nlevel = , a.wght = , nu =)`
 - LatticeKrig(): fit the LatticeKrig spatial model to a dataset.
`fitFinal <- LatticeKrig(x=, y=, LKinfo= LKinfo)`
 - Predict(): spatial prediction.
`Y_hat <- predict(fitFinal, xnew =)`

Implementation details

- Tuning parameters

- **NC**

The maximum number of lattice grid points at the coarsest level of resolution. For a example, for a square region, $NC=5$ results in a $5 \times 5 = 25$ lattice

- **nlevel**

Number of levels in multi-resolution. Note that each subsequent level increases the number of basis functions within the spatial domain size by a factor of roughly 4.

- **a.wght**

For a stationary model, at level k the center point has weight 1 with the 4 nearest neighbors given weight $-1/a.wght[k]$. In this case $a.wght$ must be greater than 4.

- **nu**

A smoothness parameter that controls relative sizes of alpha. Currently this parameter only makes sense for the 2D rectangular domain.

Implementation details

- Tuning method

Use observations with $g=1$ to fit the model. Predict response for $g=21$ and calculate prediction error. Compare results based on RMSE and fitting time.

- Tuning grid

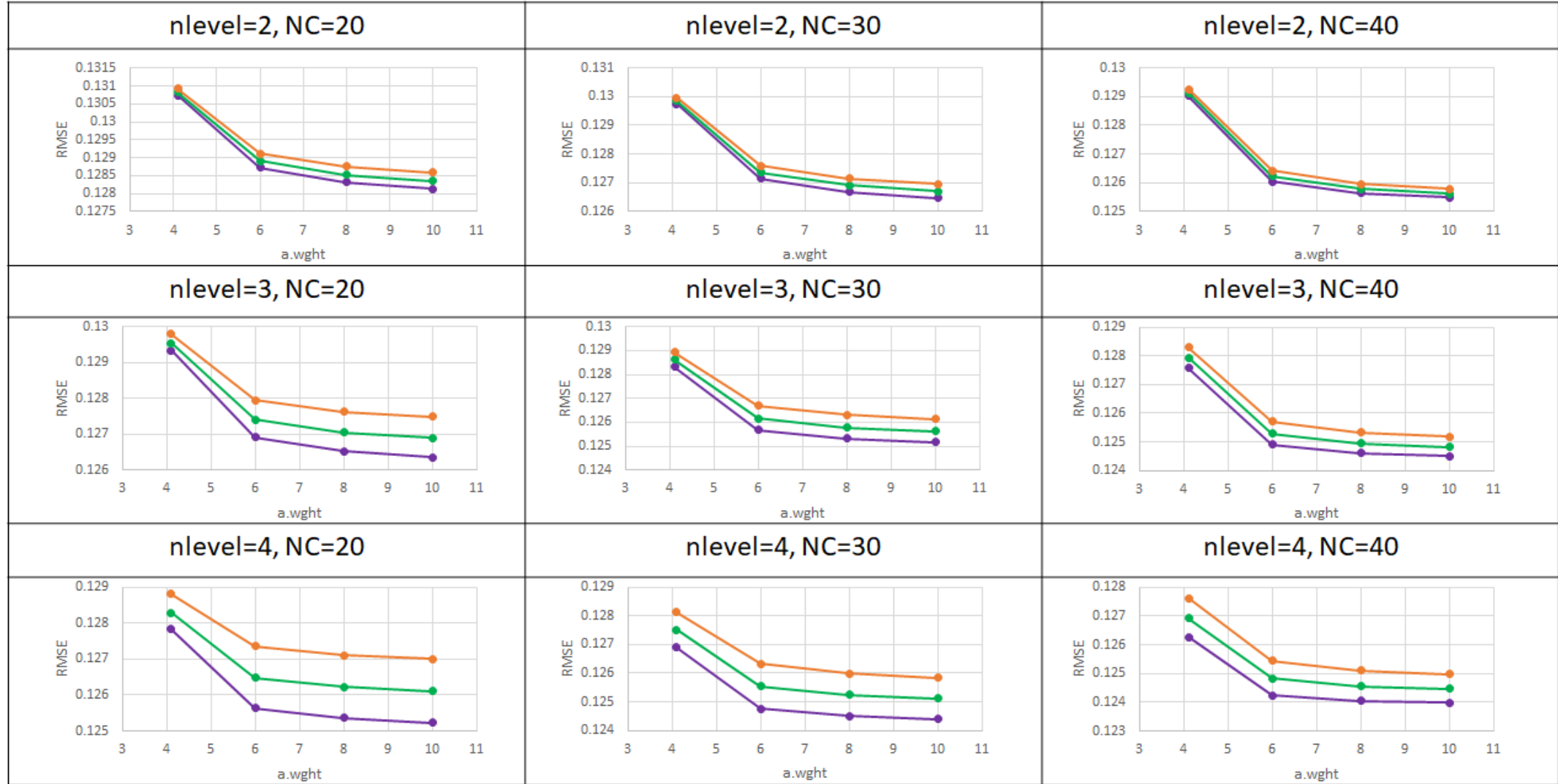
NC = 20, 30, 40

nlevel = 2, 3, 4

a.wght = 4.1, 6, 8, 10

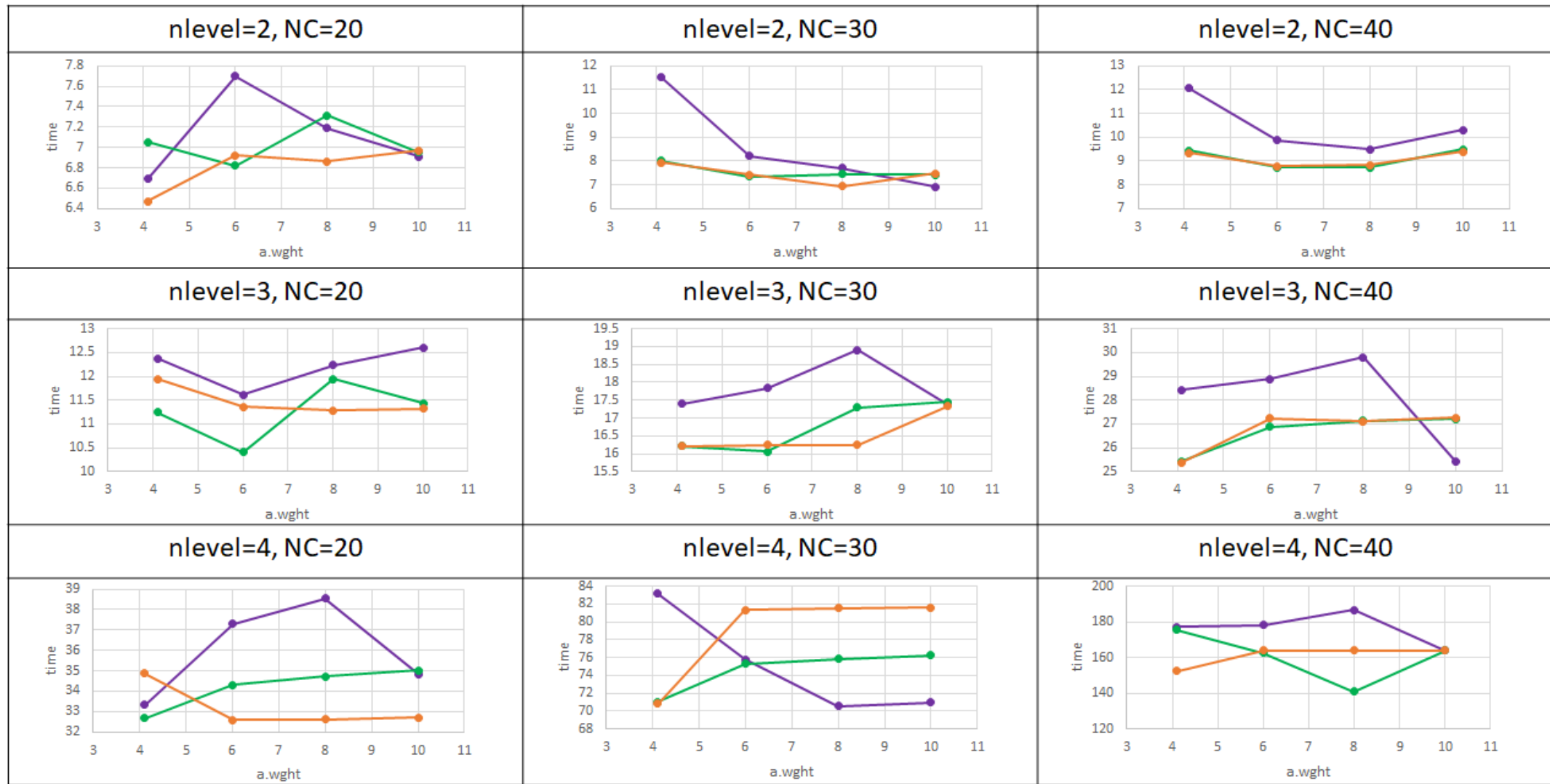
nu = 0.1, 0.3, 0.5

RMSE



—●— $\nu=0.1$ —●— $\nu=0.3$ —●— $\nu=0.5$

Fitting time

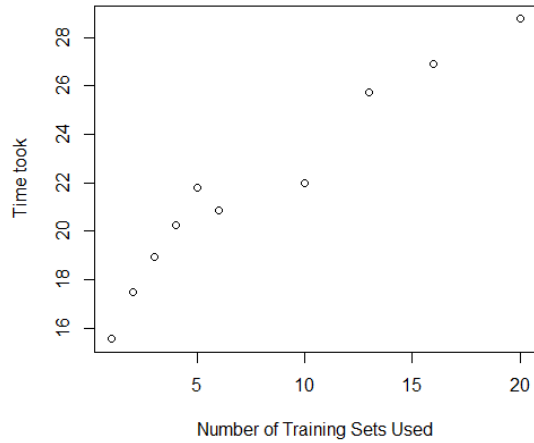


—●— nu=0.1 —●— nu=0.3 —●— nu=0.5

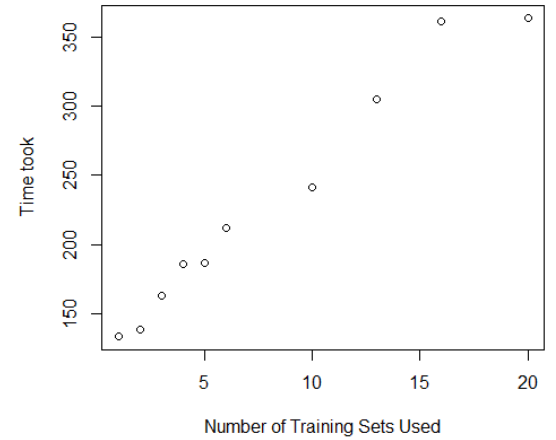
Results

- MLE is Cubic
- Lattice Kriging shows relatively consistent linearity
- When there are more grid points and more n levels it grows at a much faster rate. (10X)

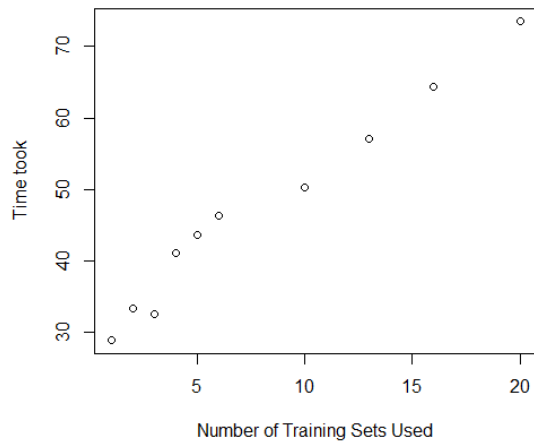
NC = 20, NLevel = 3, aweight = 6, nu = .3



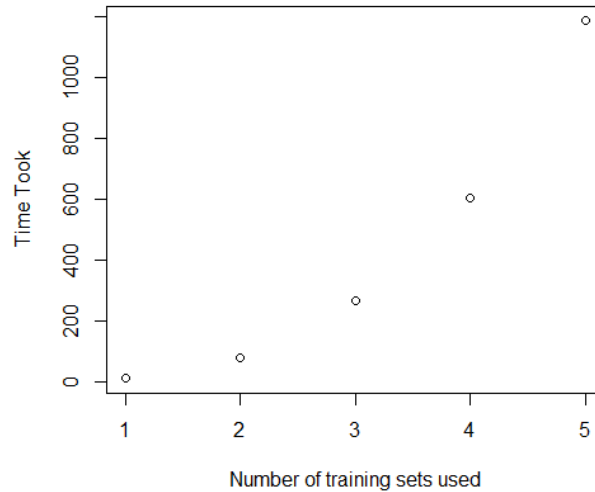
NC = 40, NLevel = 4, aweight = 6, nu = .3



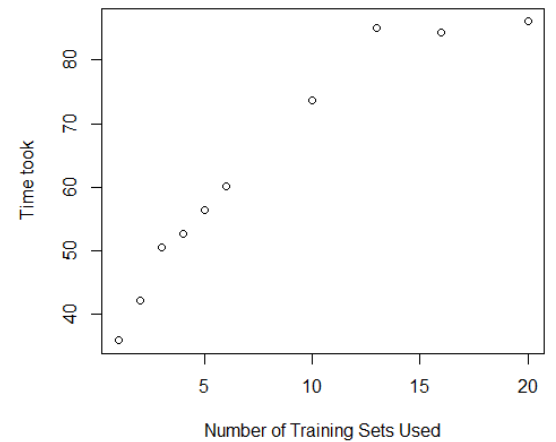
NC = 40, NLevel = 3, aweight = 6, nu = .3



MLE estimator



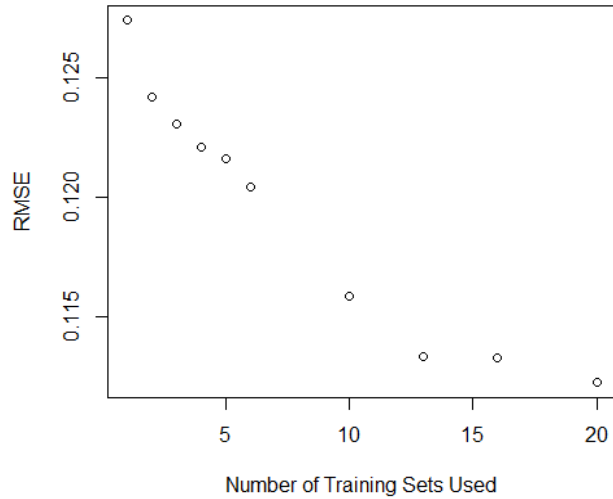
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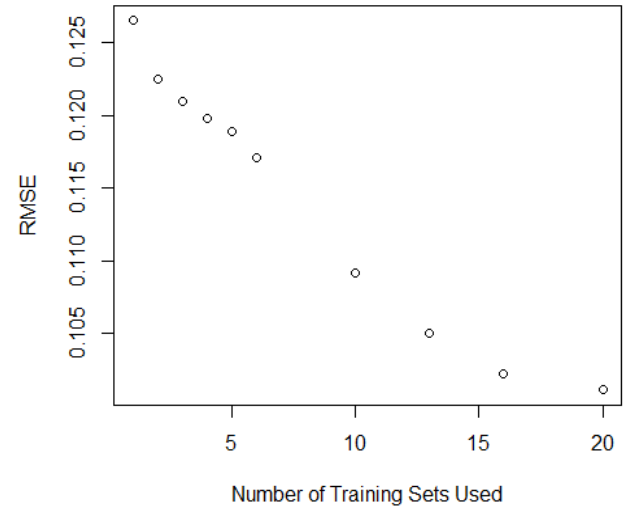
Results

- Lattice has worse RMSE values (not by much)
- The use of 20 training sets in kriging is more accurate than just 5 for MLE

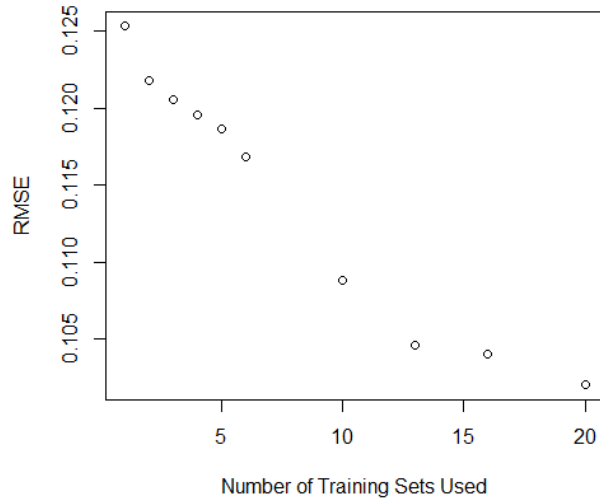
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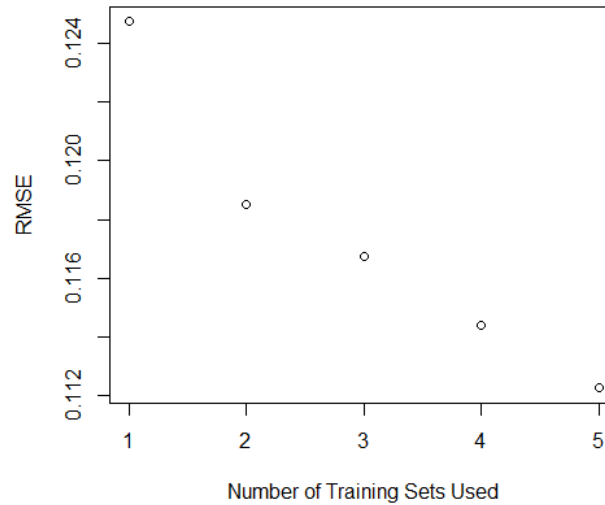
NC = 20, NLevel = 4, aweight = 6, nu = .3



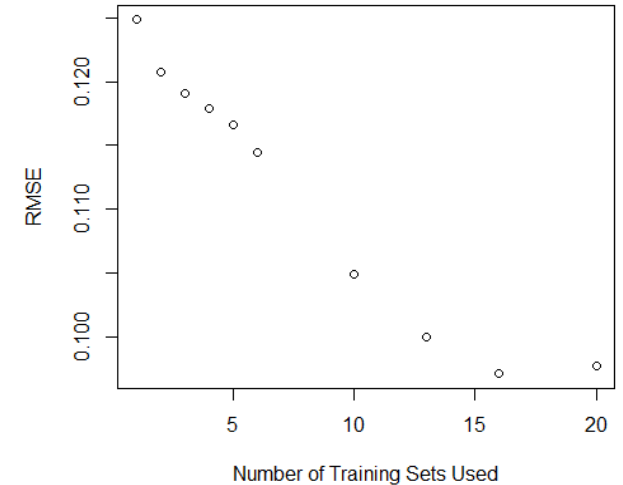
NC = 40, NLevel = 3, aweight = 6, nu = .3



MLE Estimator



NC = 40, NLevel = 4, aweight = 6, nu = .3



Results

- When Lattice Kriging splits up the data into grids, it enables the computer to process far less computations.
- It does not compromise accuracy too much
- You can have many gridlines (Adding time and accuracy)