

Sampling, modelling and uncertainty, statistical approaches to some environmental questions.

Murray Lark

Some generic issues in the life of an environmental statistician at BGS

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Uncertainty in expert interpretation of a geological cross-section

Spatially nested sampling and its optimization

Continental-scale groundwater recharge: a meta-analysis

- Quantifying uncertainty in interpretations
- Efficient sampling of variables at different spatial scales
- How best to use ad hoc data sets

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Confidence interval for an interpretation of the base of the London Clay.

Useful measures of uncertainty: decision analysis

Useful measures of uncertainty: decision analysis

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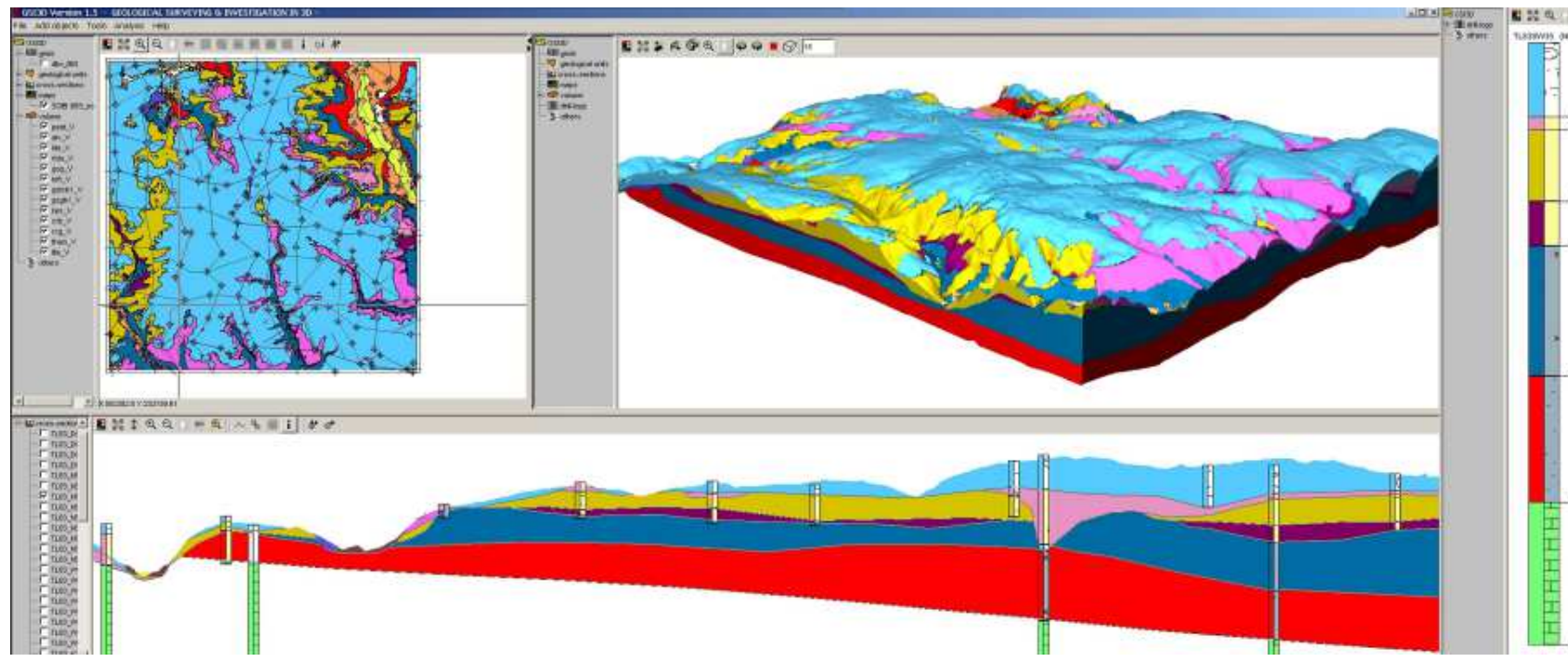
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- An 8-km section in London.
- Four bedrock units the London Clay, the Lambeth Group, the Thanet Sand (all Palaeogene) and the Chalk (Cretaceous), the latter appearing in about 10% of boreholes.
- Target feature here is the base of the London Clay (Eocene) proven by 51 boreholes.
- 28 geologists participated.
- Ten batches each with a unique set of 5 validation boreholes withheld was prepared by independent random sampling.
- Each geologist was allocated to one batch at random.
- A common interpretation of superfcials and information on outcrop (as provide by a map) given to each participant.

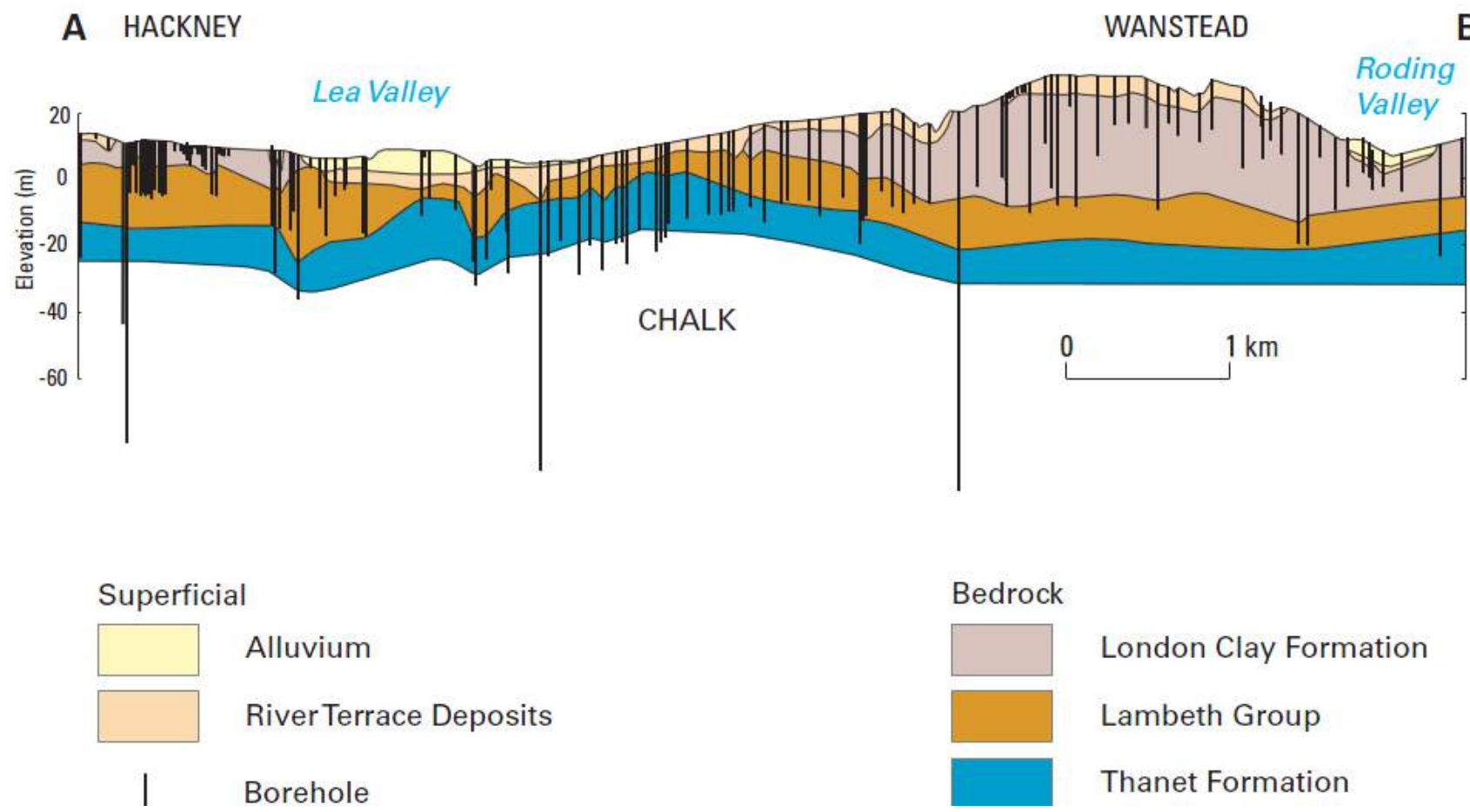
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Questionnaire on modelling experience and responses received

Question: 'Please indicate with a tick which of the 4 descriptions below best reflects your experience of 3D modelling.'

Description	Number of participants selecting this description
I have no experience of geological modelling in 3D	2
I have some experience of geological modelling in 3D (perhaps through a training course) but little (up to 6 months) or no experience of modelling independently	8
I have moderate experience of geological modelling in 3D (six months to 2 years of modelling independently)	8
I have substantial experience of geological modelling in 3D (more than 2 years of modelling independently)	10

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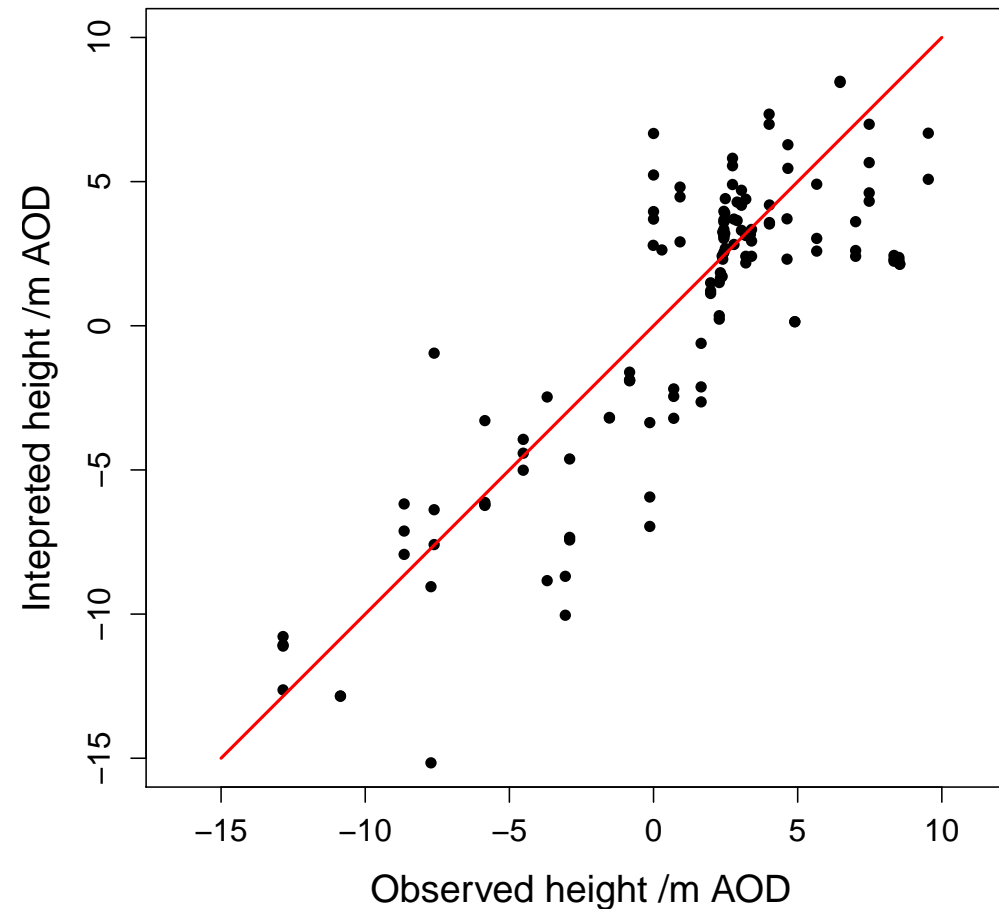
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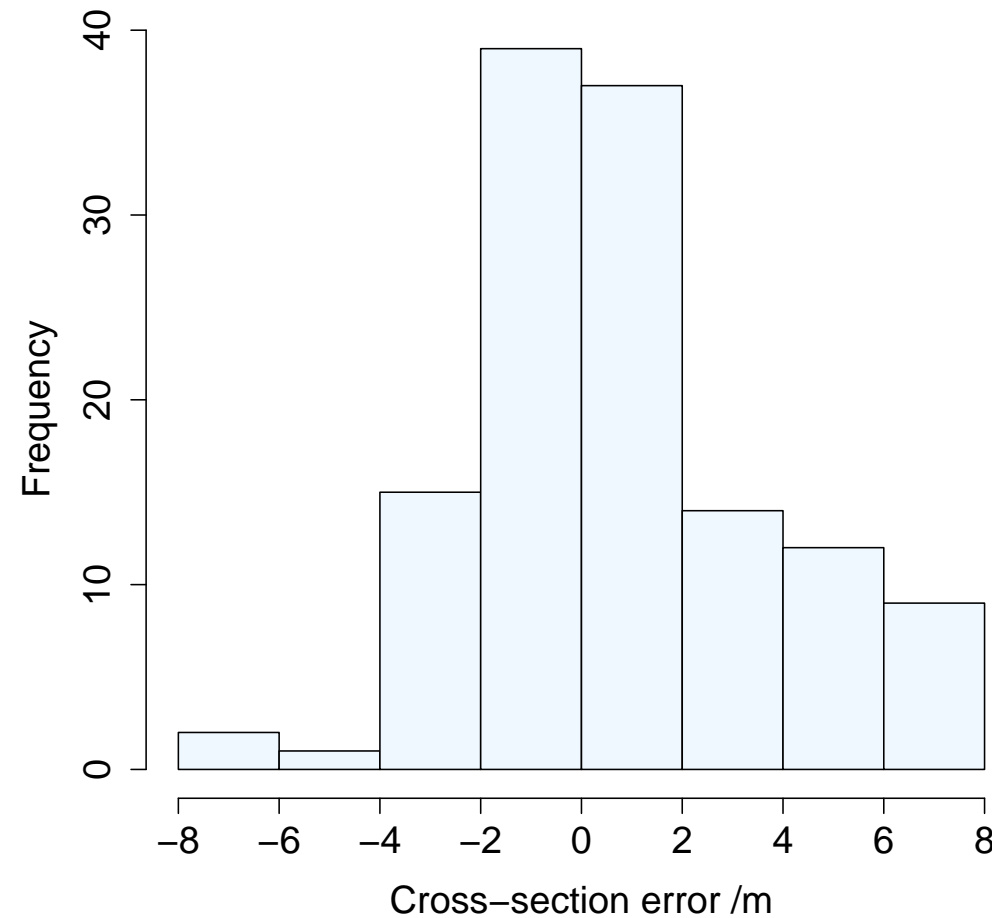
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All validation observations of the interpreted and observed height of the base of the London Clay AOD. The red line is the bisector.



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Histogram of cross-section errors.



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The basic model

$$\varepsilon = \mu + \mathbf{X}_b\boldsymbol{\beta}_b + \mathbf{X}_s\mathbf{Z}_s + \eta_g,$$

- Between-batch effect (negligible and dropped)
- Between-site within batch effect (spatially dependent, Matérn correlation).
- Between-geologist within-site effect (correlated within geologists at different sites).

Refinements

1. We tested the hypothesis that the between-geologist variance depends on distance to the nearest borehole.
2. We also tested the hypothesis that the between-geologist variance depends on the geologists experience (below).
3. Both hypotheses were supported, but (1) most strongly, and a model combining the effects did not improve on it.

	Between-geologist variance			
Experience	None	Some, <6 months	6 months – 2 years	>2 years
	4.44	2.25	1.32	0.46

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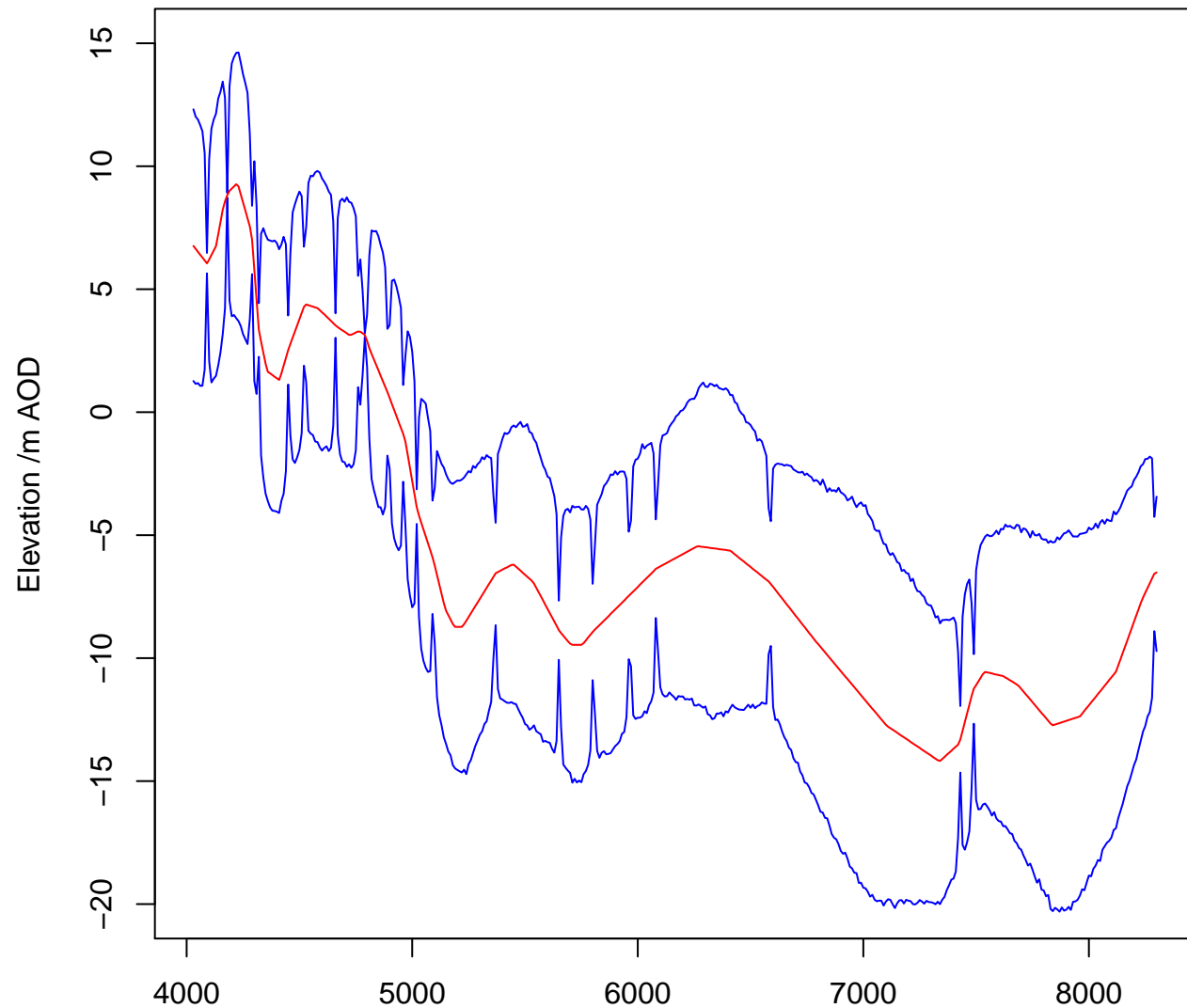
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- Point-wise measures of uncertainty (e.g. the confidence intervals) are of limited value
- Users may often be interested in what the model predicts at more than one location (e.g. a route, piers).
- We are therefore interested in the joint uncertainty at these locations.

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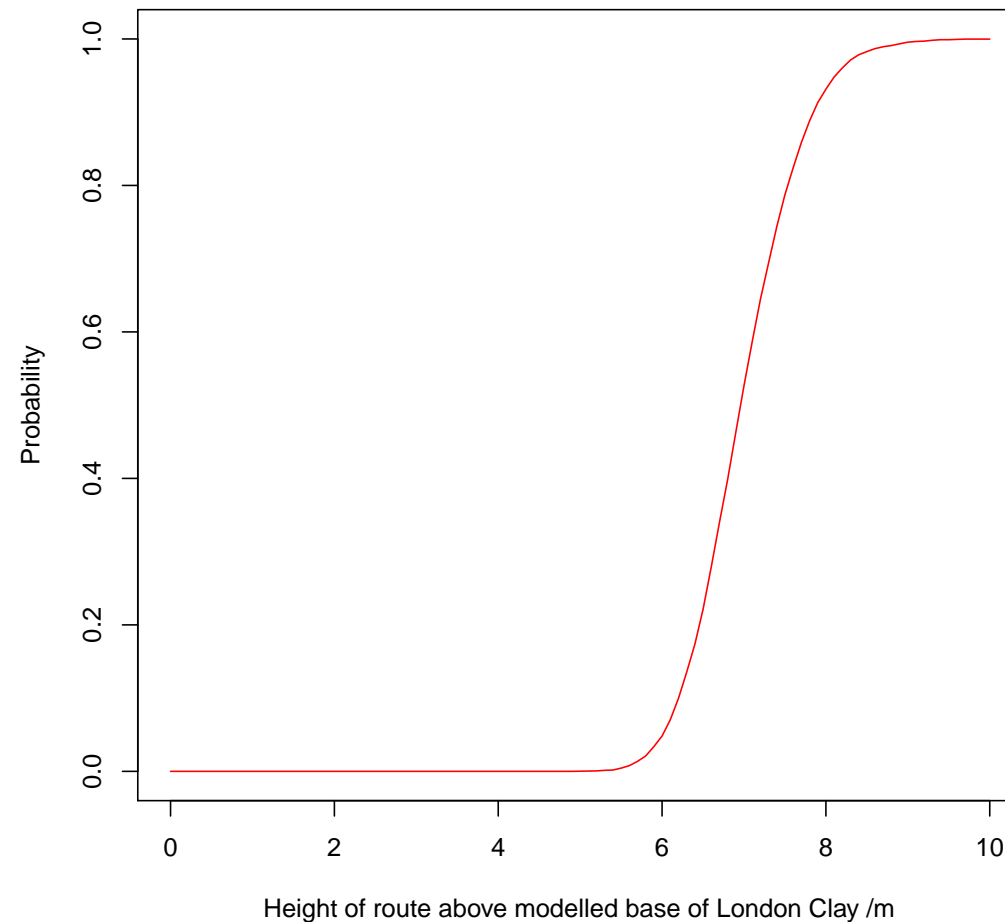
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Probability that the route is in Lambeth Group for no more than 1% of its length



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Optimization: Objective
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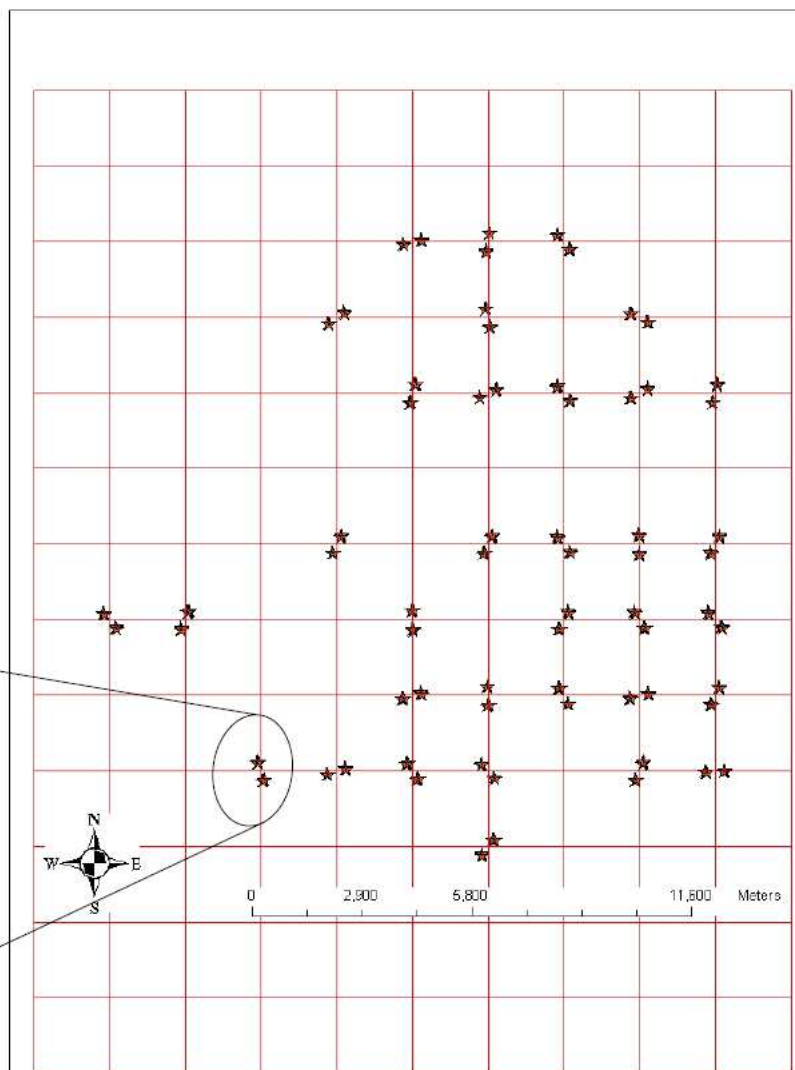
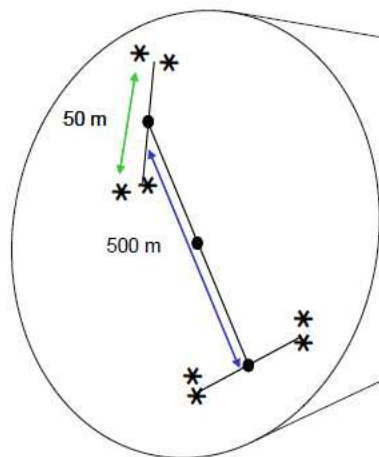
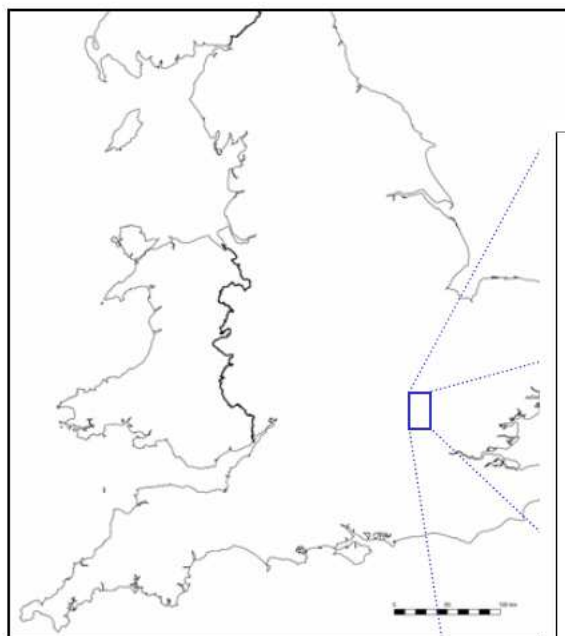
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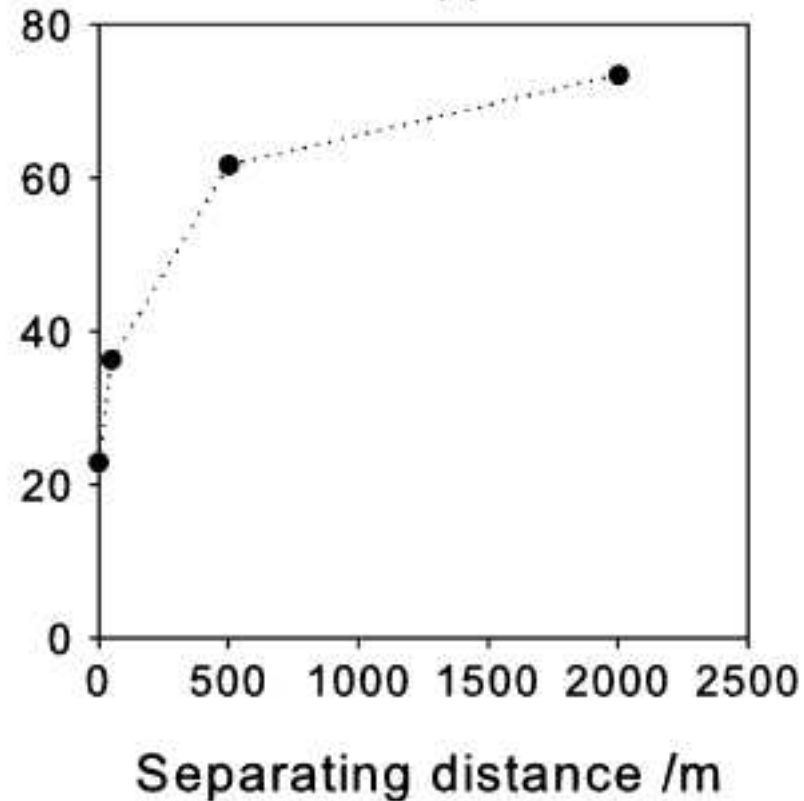
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**Ammonia-N volatilized,
% of applied**



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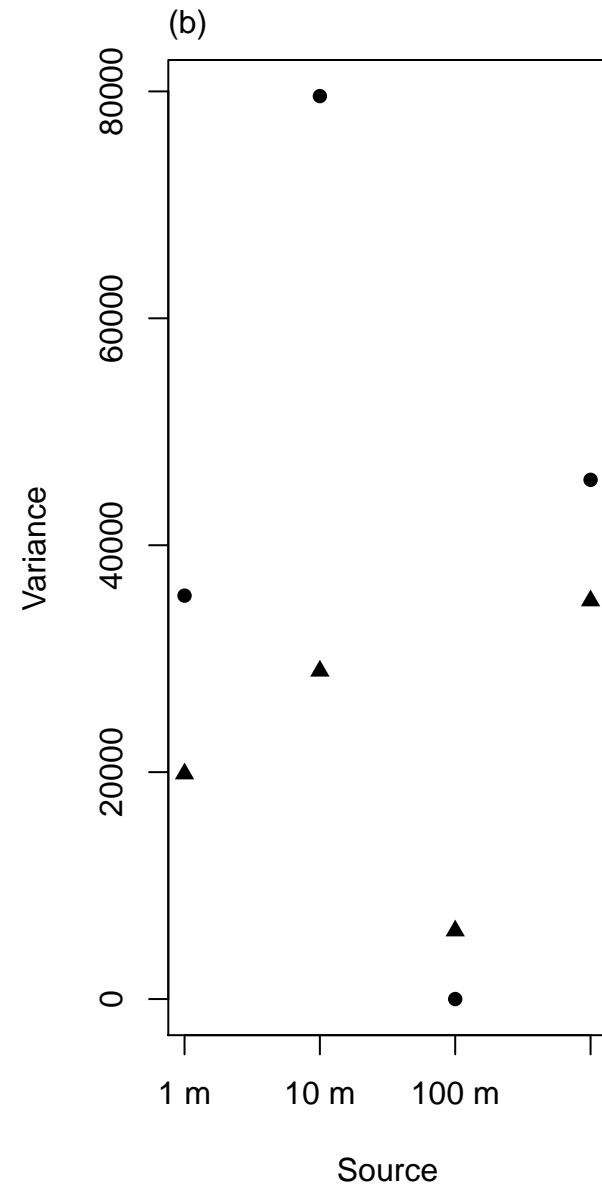
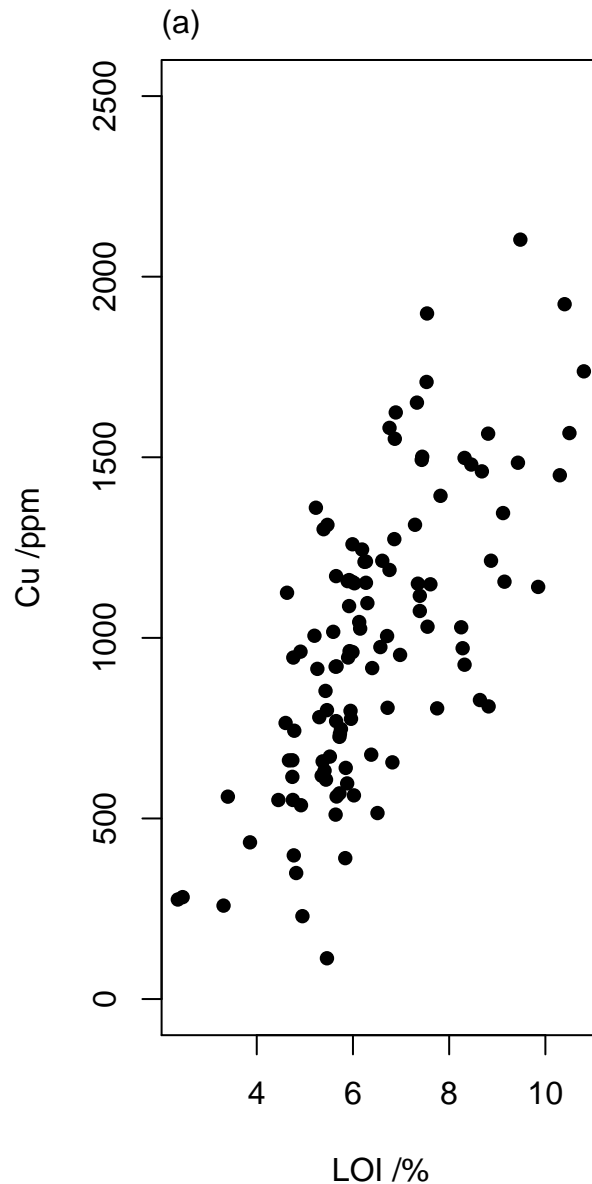
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Optimization: Objective function

Given variance components $\boldsymbol{\psi} = \{\sigma_1^2, \sigma_2^2, \dots, \sigma_m^2\}^T$, and a set of design matrices $\mathbf{U}_1, \mathbf{U}_2, \dots, \mathbf{U}_m$, we can specify the elements $\{i, j\}$ of the expected information matrix, \mathbf{M} for the variances:

$$\mathbb{E} [\mathbf{M}_{i,j}] = \mathbb{E} \left[-\frac{\partial^2 \ell_{\mathbf{R}}(\boldsymbol{\psi}|\mathbf{z})}{\partial \sigma_i^2 \partial \sigma_j^2} \right] = \frac{1}{2} \text{Tr} \{ \mathbf{P} \mathbf{V}_i \mathbf{P} \mathbf{V}_j \}, \quad (1)$$

where

$$\mathbf{V}_i = \frac{\partial}{\partial \sigma_i^2} \mathbf{V} = \mathbf{U}_i \mathbf{U}_i^T.$$

In this study the objective function was $S = \text{Tr} \{ \mathbf{M}^{-1} \}$.

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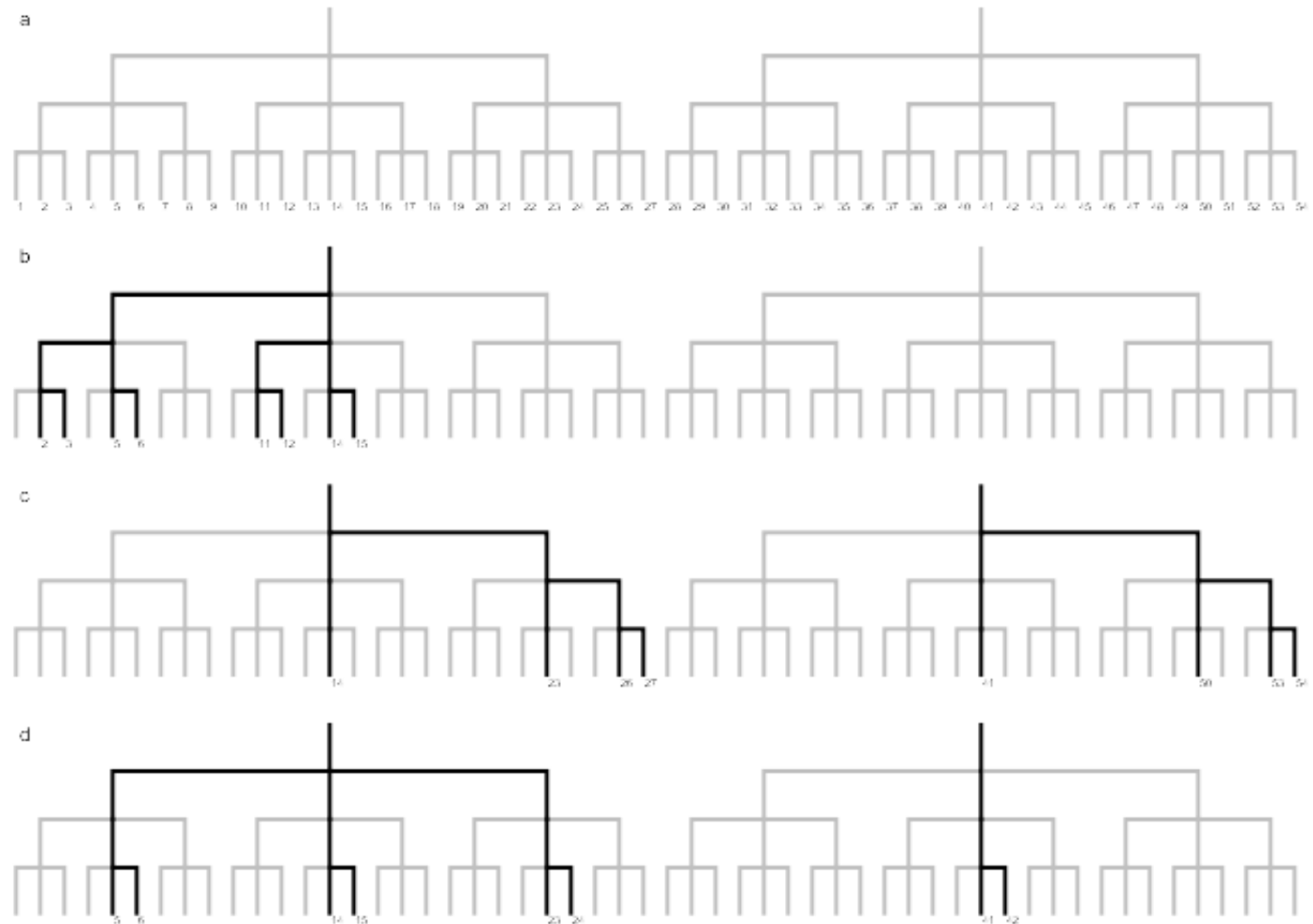
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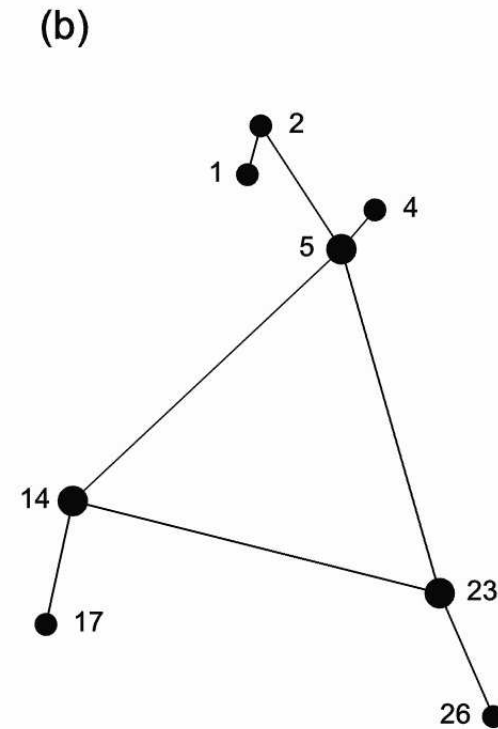
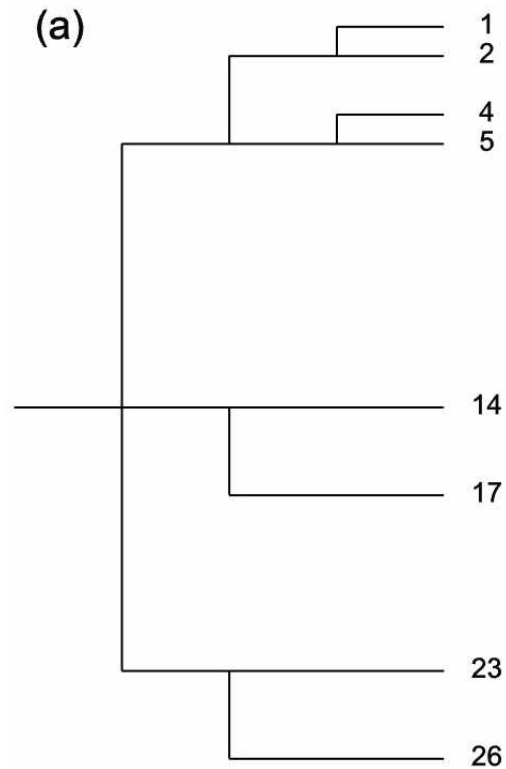
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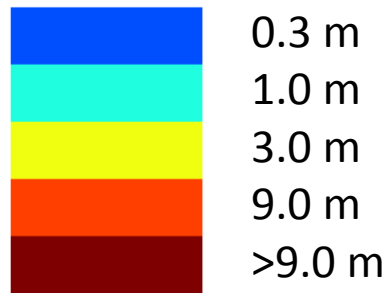
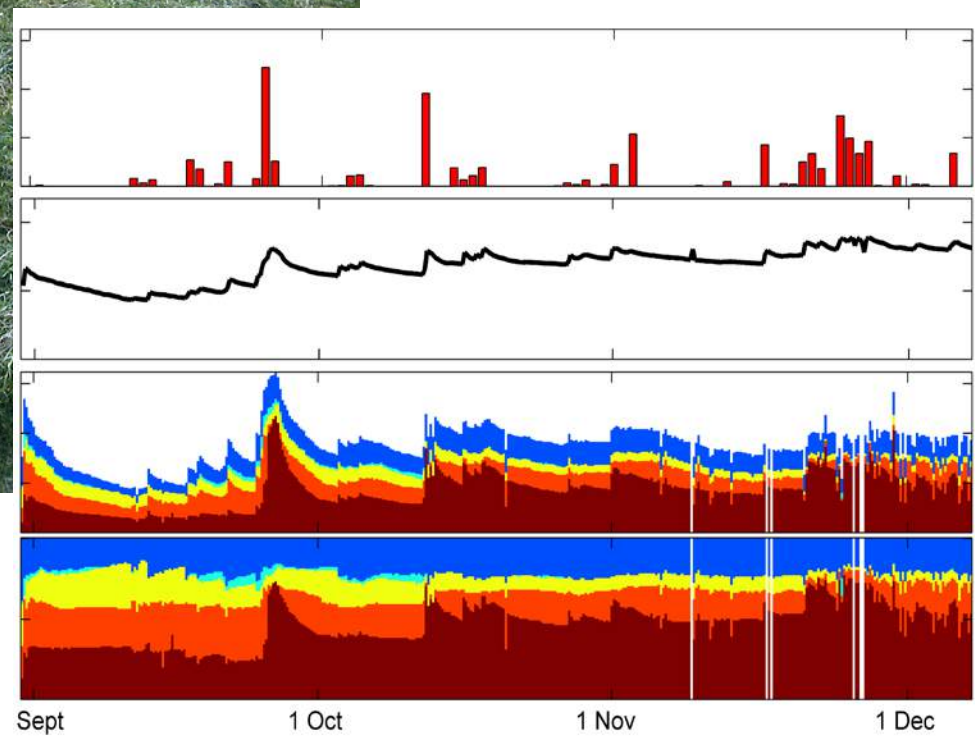
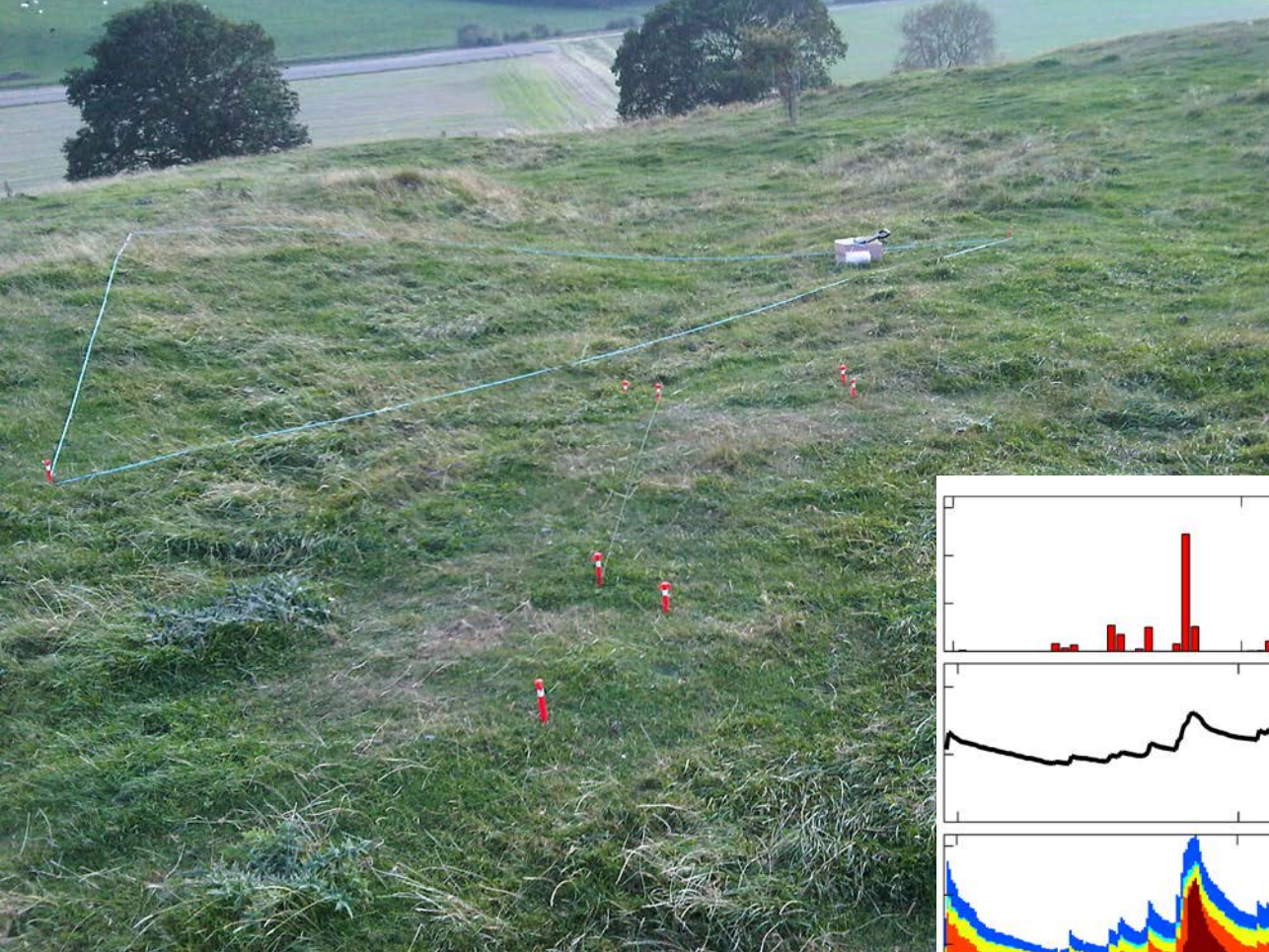
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Methodologies and uncertainty

- There is considerable interest in understanding Africa's groundwater resources, in particular their recharge and contributing factors.
- Colleagues at BGS have consolidated results from studies across the continent but:
 - They use different methodologies
 - They are far from independently distributed in space

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Methodologies and uncertainty

Methodologies and uncertainty

- Mean annual rainfall
- Number of wet days
- Mean annual SD of rainfall
- Leaf Area Index

also

- Hydrogeological units (5 aquifer domains)
- Reference Soil Groups (16 usable, World Reference Base classes)

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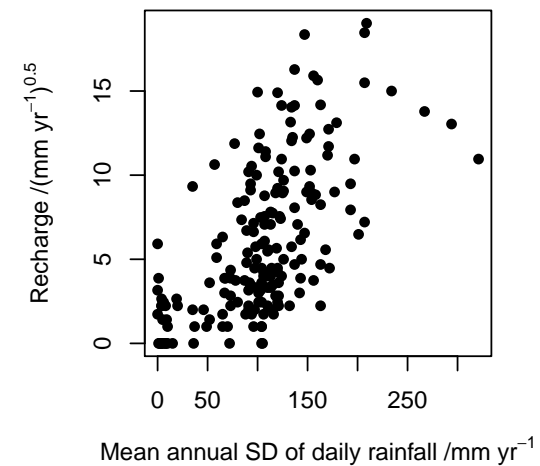
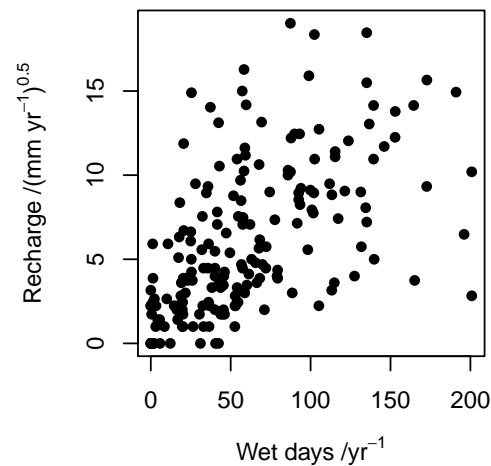
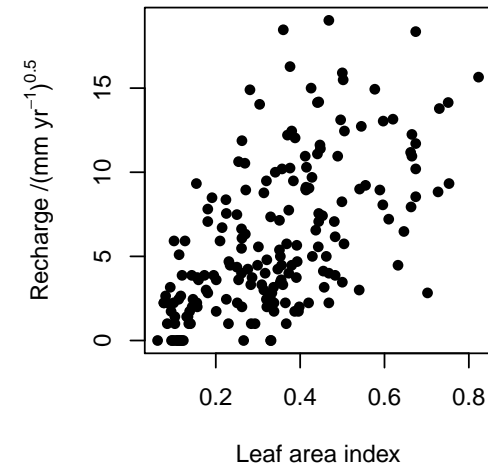
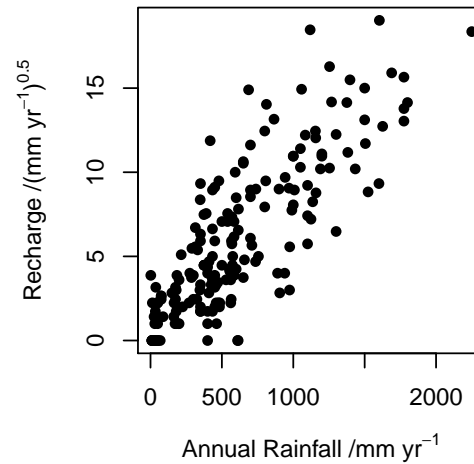
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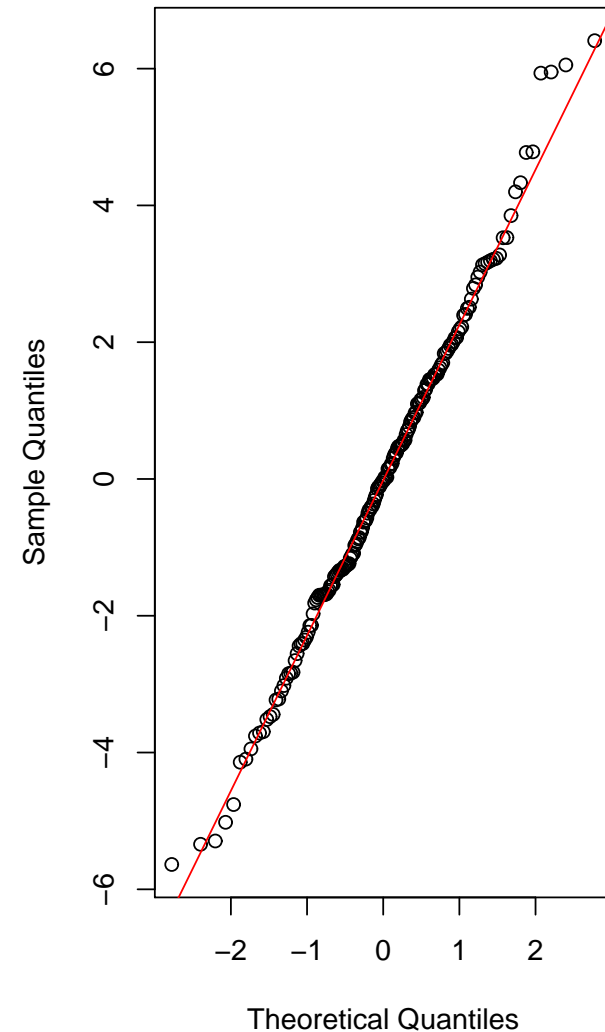
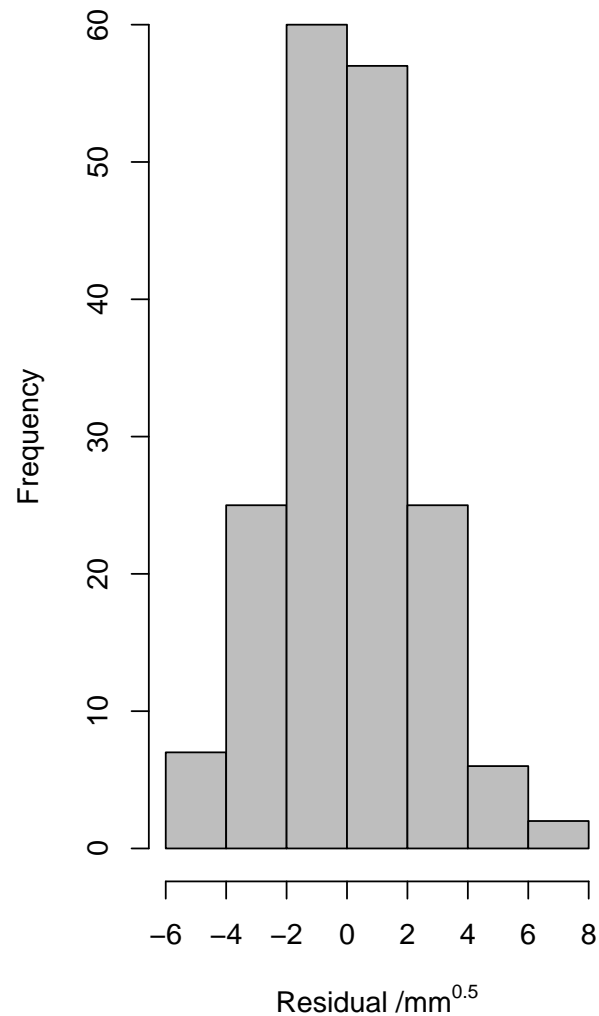
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- Alternative random effects structures
- Correlated random effect (Matérn function) and ‘nugget’ term
- Estimation by REML.

- Selected model includes annual rainfall and Reference Soil Groups (additive effects only)
- Nugget variance is 80% of the random variation
- Correlated component has an effective range of ~ 650 km

Linear Mixed Model

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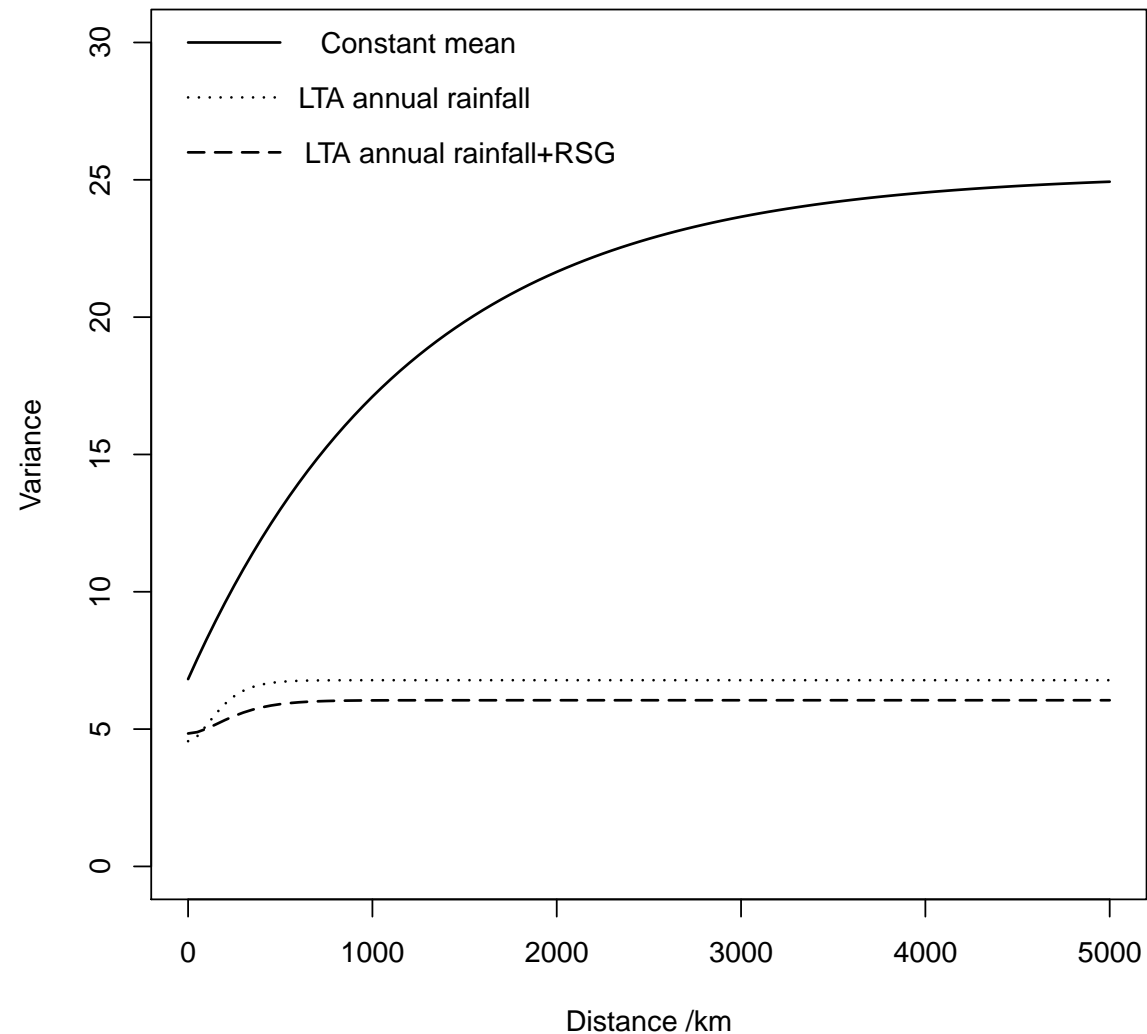
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- Recharge was determined by a range of methods including chloride mass balance, soil moisture deficit, water table fluctuation
- A confidence class was allocated to each value, depending on method and context
- Not necessarily a linear scale, so treated as a categorical variable (5 levels, 5 is least confidence)
- There was a significant improvement by modelling the random variance as a function of the confidence level:

Confidence level	1	2	3	4	5
σ	1.81	1.92	2.55	2.75	5.27

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