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Rank Regression for Analyzing Environmental Data

You-Gan Wang & Liya Fu

**CSIRO Mathematics, Informatics and Statistics
120 Meiers Road, Indooroopilly, QLD 4068, Australia**



Acknowledgements

Data were kindly provided by Seqwater, Queensland, Australia.



Outline

- Background
- Descriptive Analysis
- Linear Mixed-Effects Model
- Rank Regression Model
- Results



Two digging tools



Which one to use?

Data Description

- **Data Collection**

Wivenhoe Dam, 1997- 2002

- **Indicators (Responses)**

Chlorophyll.a (continuous data), Total Cyanophytes (count data)

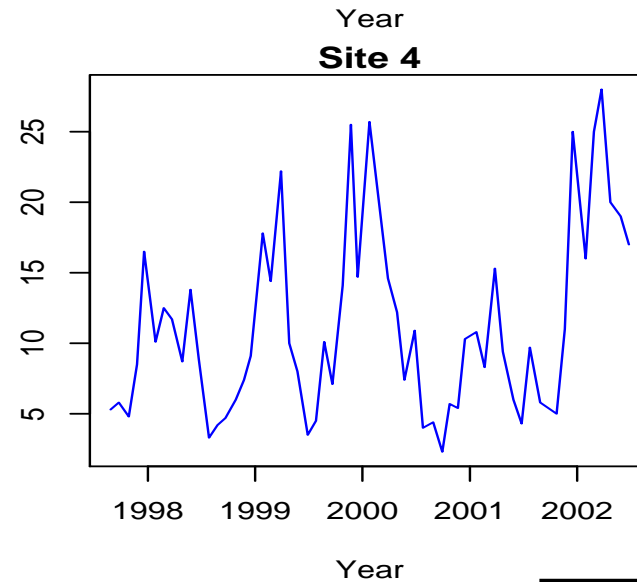
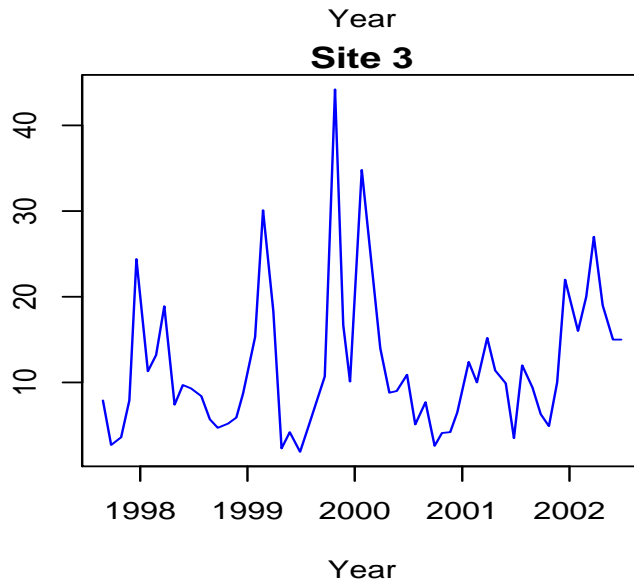
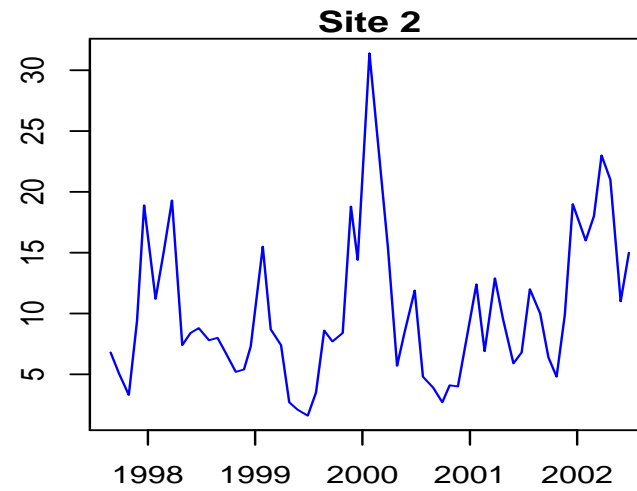
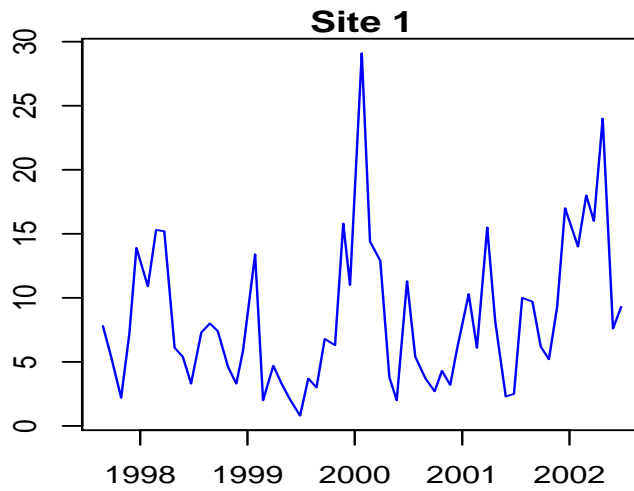
- **Covariates**

Days, Dam Level, Level Change, Rainfall

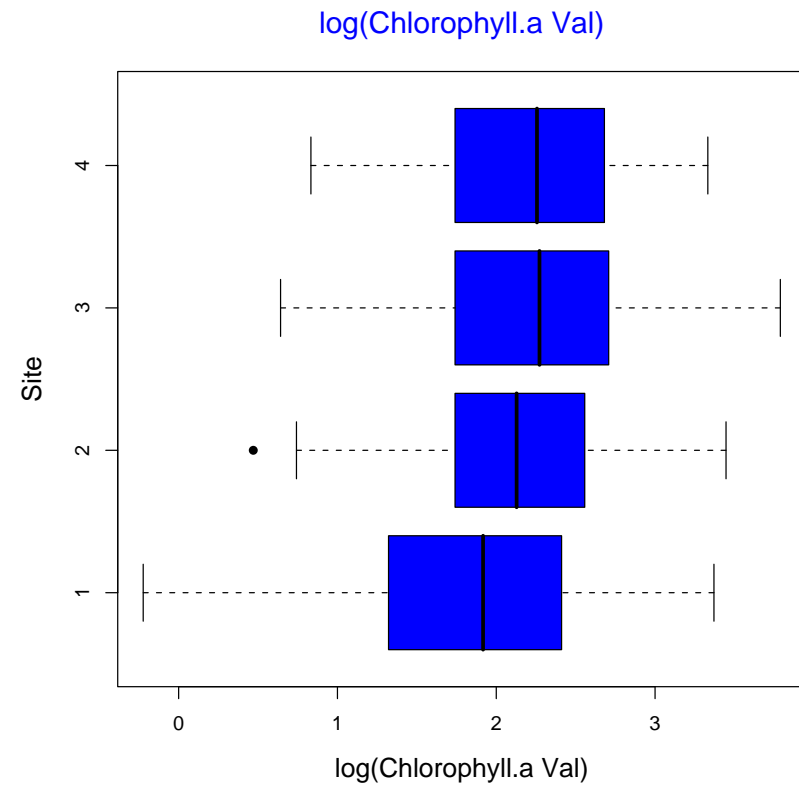
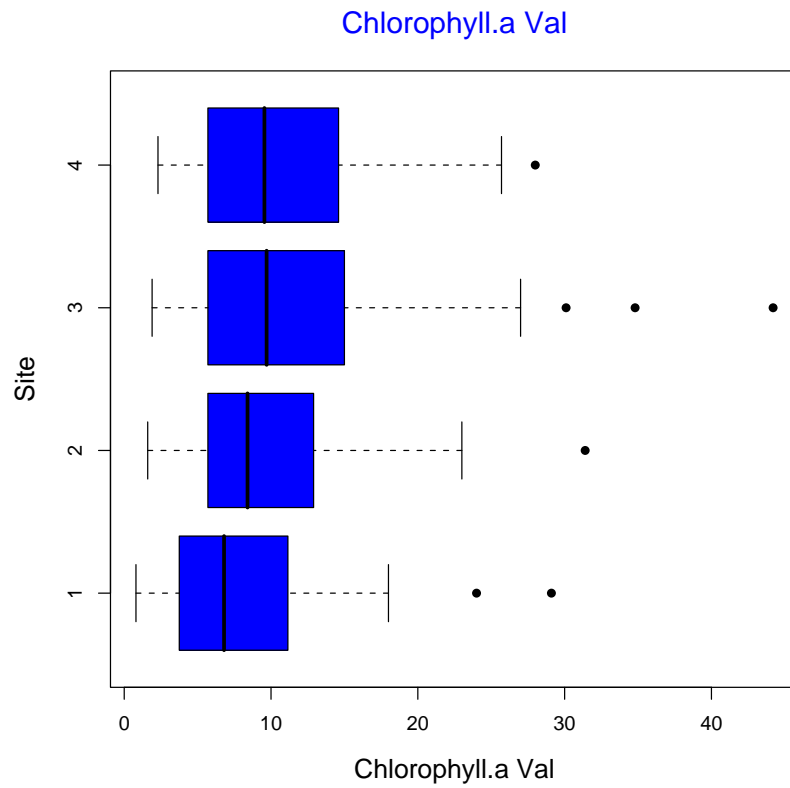
- **Purpose of this talk**

Find robust and efficient parameter estimation

Time Series of Chlorophyll.a

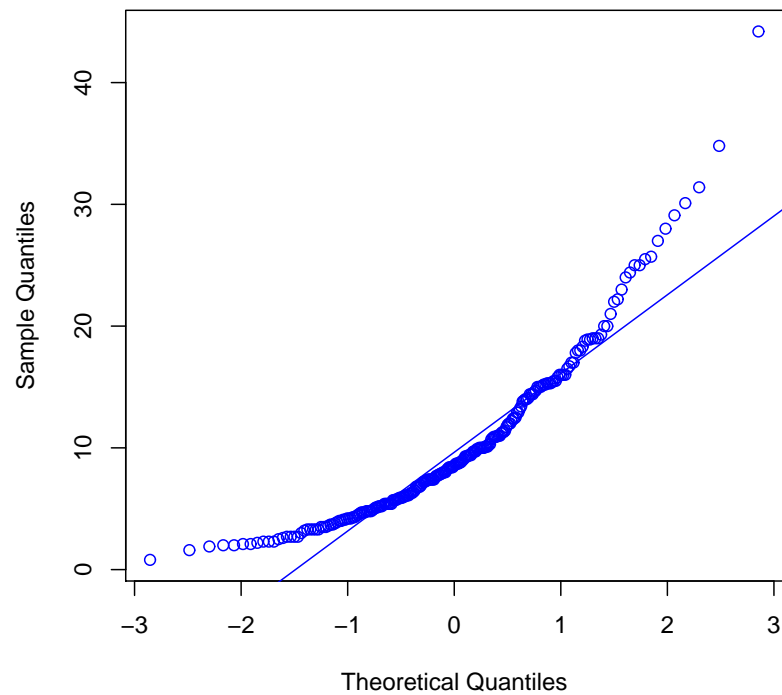


Box-Plots Chlorophyll.a

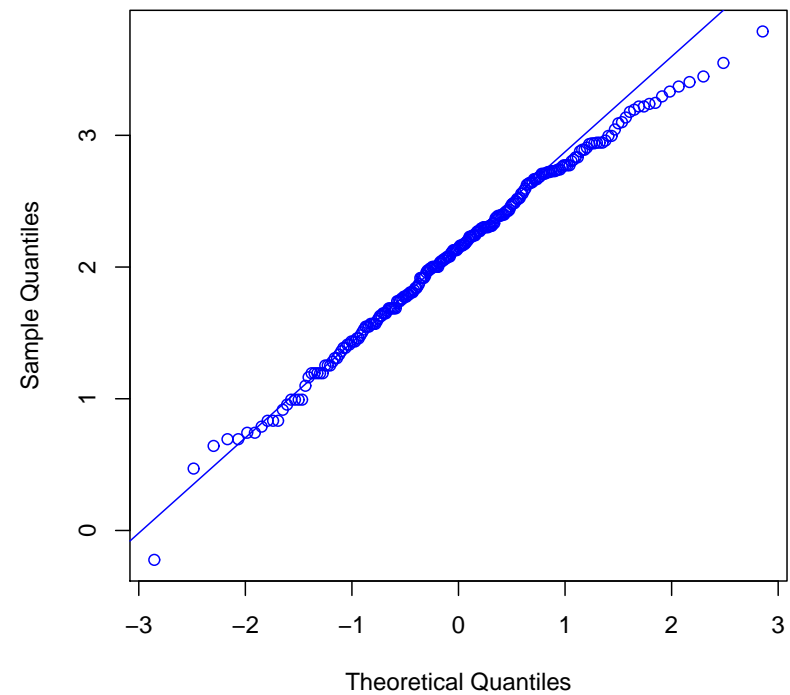


Q-Q Plots Chlorophyll.a

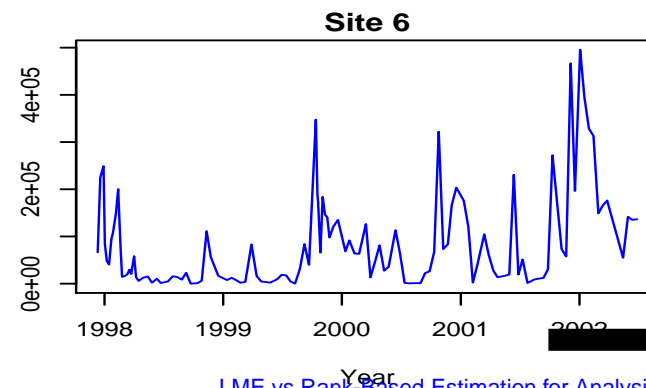
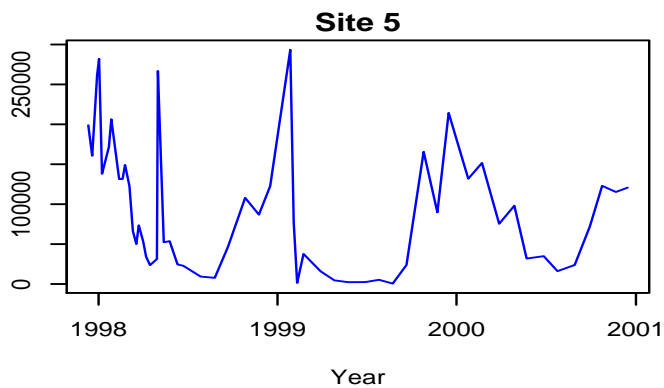
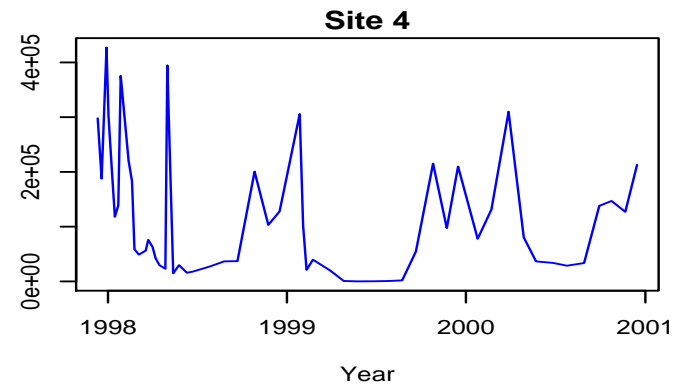
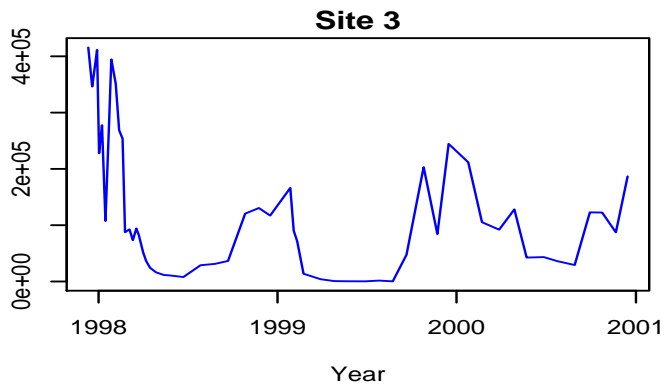
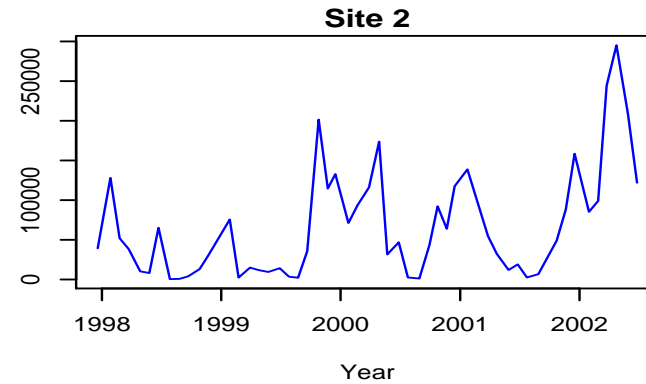
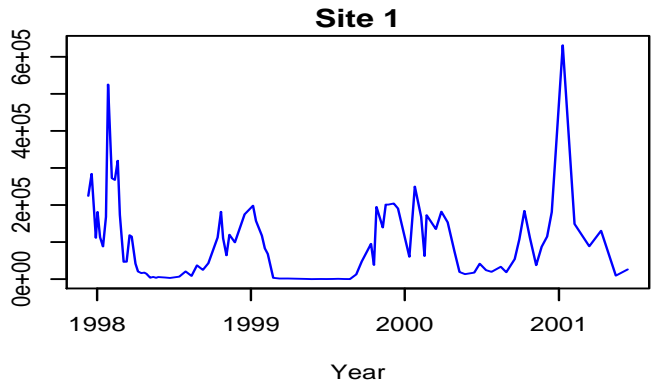
Chlorophy.a Val Normal Q-Q Plot



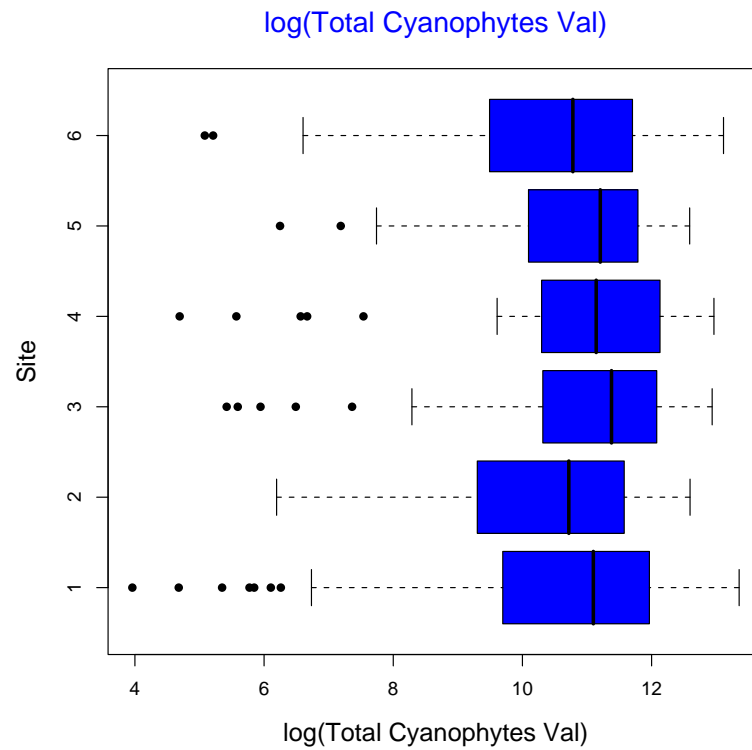
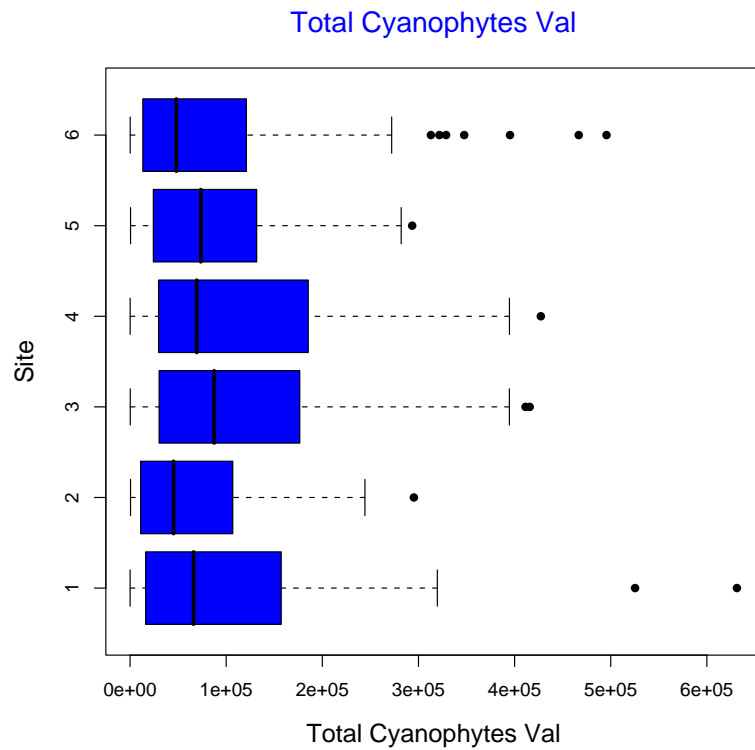
Chlorophy.a log(Val) Normal Q-Q Plot



Time Series of Total Cyanophytes

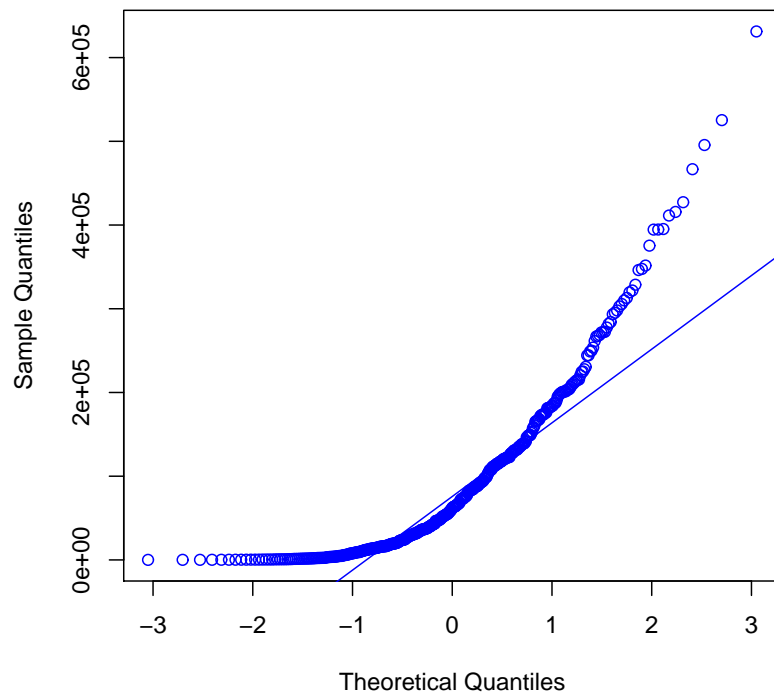


Box-Plots Total Cyanophytes

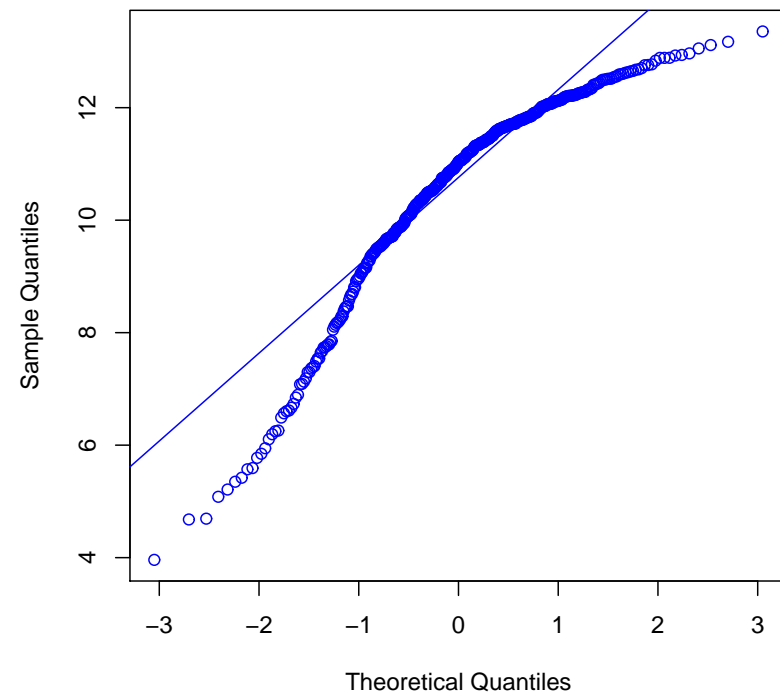


Q-Q Plots Total Cyanophytes

Total Cyanophytes Val Normal Q-Q Plot



Total Cyanophytes log(Val) Normal Q-Q Plot



Linear Mixed Effects Model

Suppose observations for the i -th site are

$$Y_i = (y_{i1}, \dots, y_{in_i}), i = 1, \dots, N,$$

taken at times $T_i = (t_{i1}, t_{i2}, \dots, t_{in_i})$. The linear mixed effects model is

$$\log(Y_i) = X_i\beta + Z_i\alpha_i + \epsilon_i,$$

where $X_i = (x_{i1}, \dots, x_{in_i})'$ and $Z_i = (z_{i1}, \dots, z_{in_i})'$ are known design matrices respectively; β are fixed effects, α_i and ϵ_i are random effects and random errors, respectively.

Linear Mixed Effects Model

- **Assumption:** $\alpha_i \sim N(0, \Psi)$ and $\epsilon_i \sim N(0, \Lambda_i)$
- **Estimation Method:** REML
- **Correlation Structure:** Gaussian spatial correlation

Rank Methods

- Robust
- Censored data (below detection limits)
- More efficient when errors have heavy-tailed distributions. To alleviate
- computational issues
- Interpretation?

Rank Regression Model

The rank regression model is $\log(Y_{ik}) = X_{ik}^T \beta + \epsilon_{ik}$.

- **Assumption:** $\text{median}(\epsilon_{ik} - \epsilon_{jl}) = 0$, for any i, j .
- **Estimation:** Residuals $e_{ik} = Y_{ik} - X_{ik}^T \beta$, Jung and Ying (2003, *Biometrika*) regarded $(Y_{i1}, \dots, Y_{in_i})$ as independent observations, and proposed minimizing the total loss function

$$\hat{\beta}_{JY} = \arg \min_{\beta} \left\{ N^{-2} \sum_{i=1}^N \sum_{j=1}^N \sum_{k=1}^{n_i} \sum_{l=1}^{n_j} |e_{ik} - e_{jl}| \right\},$$

Incorporating Cluster Correlations

Wang and Zhu (2006, Biometrika) suggested decomposing **ranks** into **between-** and **within-site ranks**, and hence obtained two types of estimates.

$$\hat{\beta}_B = \arg \min_{\beta} \left\{ N^{-2} \sum_{i \neq j=1}^N \sum_{k=1}^{n_i} \sum_{l=1}^{n_j} |e_{ik} - e_{jl}| \right\},$$
$$\hat{\beta}_W = \arg \min_{\beta} \left\{ N^{-1} \sum_{i=1}^N \sum_{k=1}^{n_i} \sum_{l=1}^{n_i} |e_{ik} - e_{il}| \right\}.$$

Incorporating Cluster Correlations

Combine corresponding between- and within-site estimating functions $U_B(\beta)$ and $U_W(\beta)$,

$$U_C(\beta) = (D_B, D_W)\Sigma^{-1} \begin{pmatrix} U_B(\beta) \\ U_W(\beta) \end{pmatrix},$$

where

$$\Sigma = \begin{pmatrix} U_B(\beta) \\ U_W(\beta) \end{pmatrix}.$$

How to Obtain $\hat{\Sigma}$

- Method 1: Perturbation method of Wang & Zhu (2006, Biometrika)

$$\tilde{U}_B(\beta) = N^{-2} \sum_{i \neq j} \sum_k \sum_l \omega_i \omega_j (X_{ik} - X_{jl})(e_{ik} - e_{jl}),$$

$$\tilde{U}_W(\beta) = N^{-1} \sum_i \sum_{k \neq l} \omega_i (X_{ik} - X_{il})(e_{ik} - e_{il}),$$

- Method 2: $\hat{\Sigma} = ??$ (analytic expression)

How to Obtain $\hat{\beta}_C$

Brown and Wang (Biometrika, 2005) put forward induced smoothing method. Here we investigate this approach for rank regression.

The versions of D_B and D_W :

$$|\tilde{D}_B - D_B| \xrightarrow{a.s.} 0 \quad \text{and} \quad |\tilde{D}_W - D_W| \xrightarrow{a.s.} 0$$

- **Parameter Estimation:**

$$(\tilde{D}_B, \tilde{D}_W) \hat{\Sigma}^{-1} \begin{pmatrix} U_B(\beta) \\ U_W(\beta) \end{pmatrix} = 0$$

Model of Water Quality Data

$$\log(\text{Val}) \sim \text{Intercept} + H(\text{Days}, k = 2) + (\text{Level}) \\ + (\text{Cha.Level}) + (\text{Rain})$$

- **Days**: the number of days (27/08/1997– 26/06/2002)
- **Level**: the dam level when the observation is collected
- **Cha.Level**: 30 days change on the dam level
- **Rain**: 14 days cumulative rainfall;
- **H(Days, k)**: is a harmonic function, and defined by following:
$$H(x, 2) = \sum_{k=1}^2 (\sin(2k\pi x/365.25) + \cos(2k\pi x/365.25)).$$

Comparison of parameter estimation for Chlorophyll.a

	$\hat{\beta}_{time}$	$\hat{\beta}_C$
H(Days, 2)1	-0.387	-0.439
(SE)	(0.070)	(0.017)
H(Days, 2)2	-0.352	-0.258
(SE)	(0.071)	(0.009)
H(Days, 2)3	0.207	0.207
(SE)	(0.058)	(0.036)
H(Days, 2)4	0.059	0.103
(SE)	(0.056)	(0.011)
Level	-0.0141	-0.015
(SE)	(0.053)	(0.012)
Cha.Level	-0.112	-0.156
(SE)	(0.040)	(0.022)
Rain	0.026	0.103
(SE)	(0.049)	(0.008)

Comparison of parameter estimation for Total Cyanophytes

	All Data		Outliers Removed	
	$\hat{\beta}_{lme}$	$\hat{\beta}_C$	$\hat{\beta}_{lme}$	$\hat{\beta}_C$
H(Days, 2)1	-0.100	-0.025	-0.087	-0.028
(SE)	(0.109)	(0.067)	(0.087)	(0.055)
H(Days, 2)2	-1.667	-1.472	-1.345	-1.265
(SE)	(0.110)	(0.055)	(0.090)	(0.041)
H(Days, 2)3	-0.342	-0.167	-0.298	-0.116
(SE)	(0.106)	(0.059)	(0.086)	(0.056)
H(Days, 2)4	-0.017	-0.073	0.081	0.019
(SE)	(0.107)	(0.028)	(0.086)	(0.017)
Level	-0.136	0.015	0.089	0.045
(SE)	(0.083)	(0.044)	(0.067)	(0.041)
Cha.Level	-0.098	-0.319	-0.184	-0.270
(SE)	(0.081)	(0.051)	(0.072)	(0.036)
Rain	-0.308	-0.096	-0.229	-0.063
(SE)	(0.079)	(0.030)	(0.066)	(0.021)

Conclusions

- LME model is not always appropriate, although it is good when the data are generated from normal distributions.
- LME is much more sensitive to the outliers than rank estimation.
- Rank method is robust, and produces smaller standard errors.
- Rank methodology is computationally more intensive, but very doable in practice.

CSIRO Mathematics, Informatics and Statistics
120 Meiers Road, Indooroopilly, QLD 4068, Australia

You-Gan Wang

Phone: +61 7 3214 2816

Email: you-gan.wang@csiro.au

Web: www.cmis.csiro.au

www.csiro.au

Thank you

Contact Us

Phone: 1300 363 400 or +61 3 9545 2176

Email: enquiries@csiro.au Web: www.csiro.au

