

Understanding Water Quality Measurements Collaboration work with Seqwater



Sarah Lennox, You-Gan Wang & Ross Darnell CSIRO Mathematical and Information Sciences 19th February 2009



Abstract

The Queensland Bulk Water Supply Authority (trading as Seqwater) provides and monitors water for the South-East Queensland, Australia region. To help understand the factors that affect the measurement of water quality within the reservoirs, data was collected for several sites within a reservoir. This dataset focuses on 7 water quality variables taken at a variety of depths. Issues which arise include the multivariate nature of the problem, non-linear trends in both depth and time, correlations in both depth and time, interactions between these variables, instrument error and heterogeneous variances. Of particular interest to the environmentalists is the diurnal variability.



Acknowledgements







Australian Government



• Sponsor



Queensland Bulk Water Supply Authority

- Trading as Seqwater
- Began operating on 1 July 2008 as a new service provider
- Provides water for South-East Queensland
- 3 major reservoirs around Brisbane
 - Wivenhoe 1,165,240 ML @ full capacity
 - Somerset 379,850 ML @ full capacity
 - North Pine 214,960 ML @ full capacity
- Water is used for:
 - Recreation
 - Irrigation
 - Industry
 - Drinking Water



Collaboration between CMIS and Seqwater

- Analysis of existing datasets
- Gain greater understanding of the variability within and between reservoirs
 - Development of a Report Card for the reservoirs
 - Spatial and Temporal Analyses
 - Daily variability
- Investigating monitoring design network
 - System design
 - Space and time representativeness
 - Spatial optimisation
 - Gaps



Lake Wivenhoe

Photograph: Andrew Watkinson (Seqwater)





Aim of current work

- Quantify impact of taking samples at different times throughout the day
 - Long-term trend
 - Seasonal trend
 - Daily trend
 - Daily trend differences between seasons (ie. Interactions)
- When do maximum/minimum levels occur throughout the day?
- Does the daily trends and maximum/minimums change throughout the year?
- What are the impacts on the current monitoring design?
- How can future sampling designs take account of the daily variability?



Monitoring Program data

- 6 historical sites
- 8 more recent drought sites
- Surface and bottom measurements
- Samples taken from 1 to 4+ times per month
- Some event monitoring
- Numerous water quality variables measured
- Sampling occurs at different times of the day between sites and within each site



Profiler Data

• 3 main sites

- Dam wall (28m @ deepest)
- Upstream site (shallow)
- Site -mid-way upstream (deep)
- Record from surface to bottom ~1m
- Each depth is recorded every 2 hours
- Dam wall site starts recording in August 2007 (554 days)
 - The other two sites start recording sometime later
- Dam Wall site: 137, 500 observation
- Multiple variables:
 - Water Temperature, pH, Specific Conductivity, Turbidity, Chlorophyll a, Blue Green Algae, Dissolved Oxygen



Location





Challenges

- Too much rather than too little data (100, 000+ observations at a single site for each variable)
- Abrupt changes due to instrument calibration or replacement
- Correlation in time
- Correlation in depth
- Non-homogeneous variance
- Non-linear trends across depth and time.
 - Seasonality
 - Stratification (separation of surface and bottom layers)
- Interactions between depth and time of day
- Interactions between month & time of day
- Multivariate nature of variables



Instrument changes



Changing variability across depths





Changing variability within depth



CSIRO. Understanding water quality measurements

Example: stratification during low variability





Mid-stream site



Interactions between time of day and month



Assumptions for Simplified Model to extract daily trend

- Correlations in time
 - Lagged residual terms
- Each depth is independent
- Each variable is modelled independently by a univariate design
- Non-linear trends in time
 - Smooth terms.



Model process (adapted from Rigby & Stasinopoulos, 1996)

• Assume $y_i \sim N(\mu_i, \sigma_i^2)$

• Mean is modelled by: $g_1(\mu) = \sum s_j(x_j)$

• Variance is modelled by: $g_2(\sigma^2) = \sum t_k(m_k)$ where

 s_j and t_k are linear or smooth terms x_j and m_k are explanatory variables for mean and variance g() is the link function



Iterative process

- Assume prior weights $\omega_1 = 1$ 1.
- $g_1(\mu) = \sum s_i(x_i); \ \omega_1 = 1$ Fit GAM model for mean 2
- 3. Calculate residuals $\mathcal{E}_t = y_t \hat{\mu}_t; \ \eta = \mathcal{E}_t / \sigma_t$ $\eta_{t-1} = rac{\mathcal{E}_{t-1}\sigma_t}{\sigma_{t-1}}$
- Calculate lagged residuals 4.
- 5. Fit GAM model to ε^2 $g_2(\varepsilon_t^2) = \sum t_k(m_k) = g(\sigma_t^2)$
- $\omega_1 = 1/\hat{\varepsilon}^2$ Calculate new prior weights 6.
- $g_1(\mu_{full}) = \sum s_i(x_i) + \rho \eta_{t-1}; \ \omega_1 = 1/\hat{\varepsilon}^2$ 7. Refit GAM model with weights and lagged residuals covariate
- Calculate $\hat{\mu}_t = \hat{\mu}_{full} \rho \eta_{t-1}$ 8.
- 9. Repeat steps 3 to 8 until convergence



Final Model – Dam wall site (surface)

• Mean is modelled (using Simon Wood's mgcv package for R) by: $g_1(\mu) = a + change + s(Date, k = 100)$ $+ s(time, by = month, k = 7, bs = "cc") + \rho \eta_{t-1}$

change is a categorical variable defining periods of instrument changes, η_{t-1} is a lagged residual term.

• Variance is modelled by: $g_2(\sigma^2) = b + s(Date) + s(time)$



Observed and fitted values for Dam wall site 1m Depth





Seasonal / Longer trend covariate



Instrument Change covariate



Periods between instrument changes

Model diagnostics



Model diagnostics cont.

Variance function

Diurnal variability (nicchu-no)

February

Jan

June

May

September

0.6

0.4

time.of.day

0.8

1.0

0.0

-0.4

0.0

0.2

time.of.day

0.4

July

December

Oct

0.2

80

-0.4

0.0

0.6

0.8

1.0

8

-0.4

0.0

Seasonal / Longer term trend for Up Stream site 1m Depth

CSIRO

Water Temperature for Upstream Site

Variance Function

Diurnal variability at 1m depth for Upstream site

June

4

ö

0.0

4

ó

0.0

0.2

February

May

September

9

8

Ψţ

Ó,

0.0

0.2

0.4

time.of.dav

0.6

0.8

1.0

time.of.day

0.4

August

time.of.day

October

0.6

0.8

1.0

0.4

80

4

o

December

1.0

Site comparison

	Dam wall site		Up-Stream site	
Time of day	Min	Max	Min	Max
January	5:11am	1:32pm	6:32am	5:57pm
February	6:35am	1:21pm	9:39am	10:57pm
March	6:35am	2:56pm	8:27am	4:36pm
April	6:46am	2:21pm	7:29am	4:25pm
May	7:09am	3:59pm	7:44am	4:36pm
June	7:32am	4:25pm	9:33am	12:20am
July	7:29am	2:15pm	7:44am	4:16pm
August	6:40am	3:36pm	7:32am	5:40pm
September	7:15am	5:17pm	5:57am	6:40pm
October	6:29am	1:38pm	7:21am	5:54pm
November	5:46am	1:03pm	7:26am	5:54pm
December	6:23am	2:12pm	7:21am	7:58pm

Example 2 – Bottom of Dam Wall site

• Covariates for mean:

- Factor for periods of change due to instrument changes
- Date
- Time of day curve for each month
- Covariates for variance
 - Date
 - Time of day

Water Temperature (bottom)

Seasonal/Longer term trend

CSIRC

Model diagnostics

CSIRO

o o ο. 8. 0 8. 0 0.0 0 0.0 0.0 0.0 variance variance variance 0.4 0 4 0 4 00 8 ó ο. è. 0 0 ο. 0.2 0.2 0.2 0 0.0 0.0 0.0 16 18 20 22 24 0.6 0.0 0.4 0.8 0.2 1.0 10/07 05/08 08/08 02/09 01/08 11/08 Time of day Fitted value Date

Variability modelled

Results - bottom

June

July

May

0.06

.ddy

December

September

Conclusions

- Daily water temperature variability changes throughout the year.
- Daily water temperature trends at the surface differ between sites
- Models developed for water quality parameters need to account for autocorrelation and non-constant variance
- Water Temperature variability at the Dam Wall site
 - Is greatest at the surface during September and October
 - Is greatest at the bottom during December
 - Is lowest at the surface during July
 - Is lowest at the bottom during March (no significant trend)

Future Work

• Models for depth covariate

- Possible interactions between depth and time, depth and date
- Correlations between depths
- Stratification separation between surface and bottom layers
- Alternative methods for modelling autocorrelation and nonconstant variance
- Investigate Time Series alternatives
 - Generalized Autoregressive Conditional Heteroscedastic process

vw.csiro.au

Division/Unit Name

Sarah Lennox Research Scientist

Phone: +61 7 3214 2717 Email: <u>sarah.lennox@csiro.au</u> Web: <u>www.csiro.au/CMIS</u>

Thank you

Contact Us Phone: 1300 363 400 or +61 3 9545 2176

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

