Model Skills and Uncertainties

An evaluation of model prediction performance against observations at eight LBA-MIP sites

- Latent Heat Flux -

Aaron Polhamus and Dr Joshua Fisher, NASA JPL April 18, 2011

Acknowledgements

- LBA-DMIP team
- Modeling teams
- Michel Muza
- Grayson Badgley
- Biosphere 2

Data

- Hourly observations for multiple years of key meteorological/biophysical variables, with corresponding model predictions
- We average all data to the monthly scale

The challenge

- Identify which models perform well against the observations (assumption: perfect observations)
- Identify potentially problematic sites & models

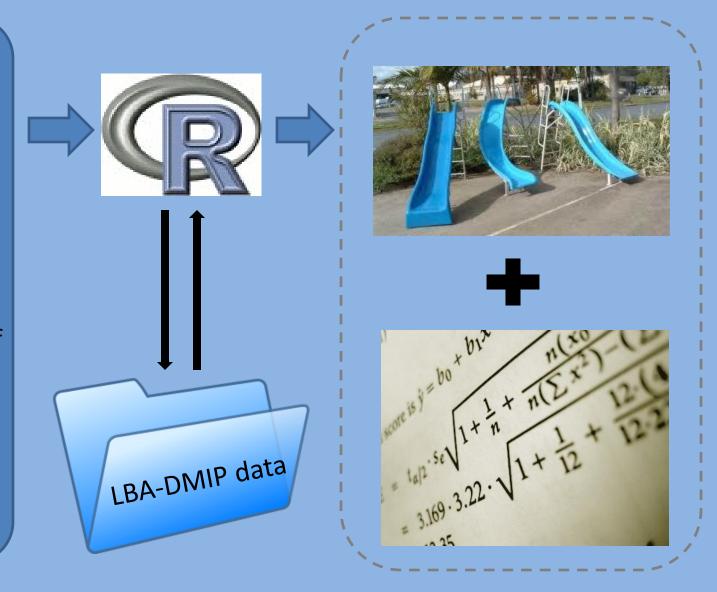
Method

- Plot time series
- Bias (MBE), spread (RMSE), and agreement (R²) on a persite/per-model basis

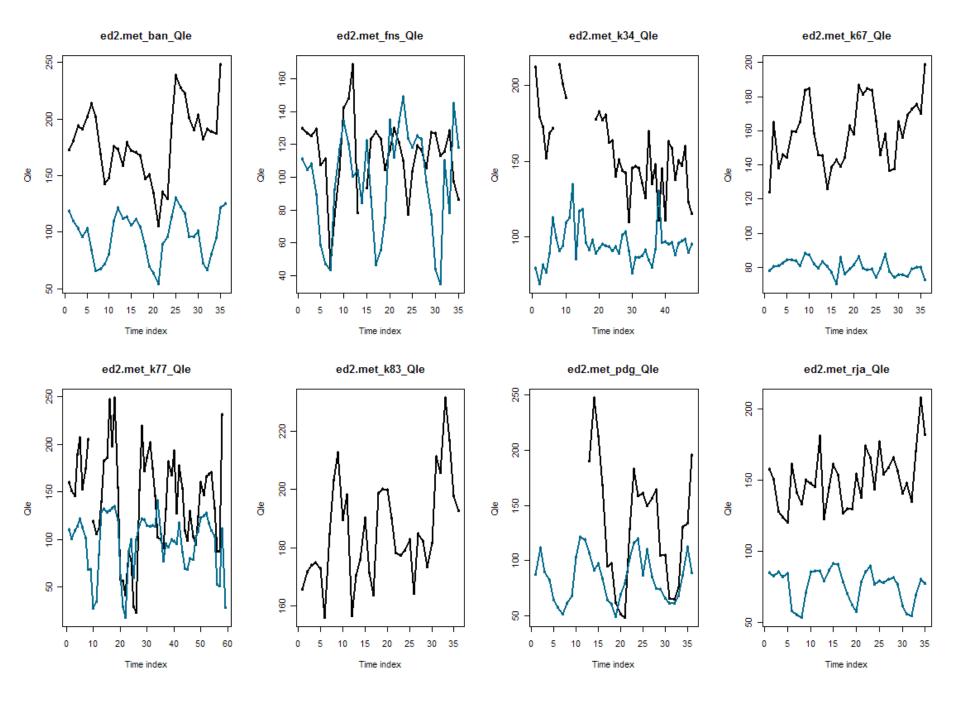
Broad objective: a protocol for automated analysis

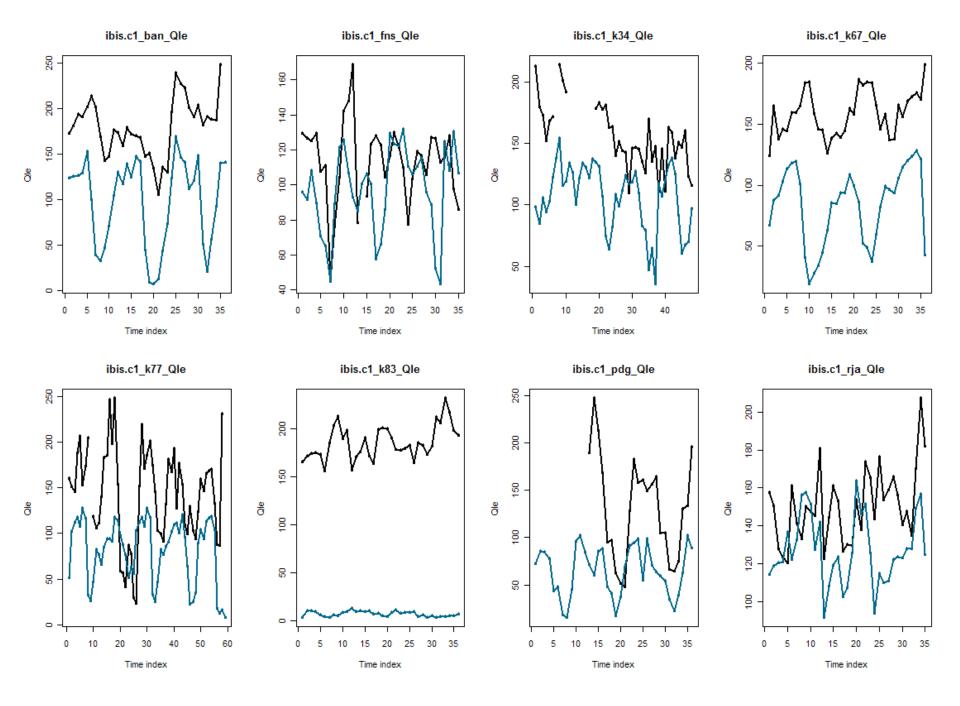
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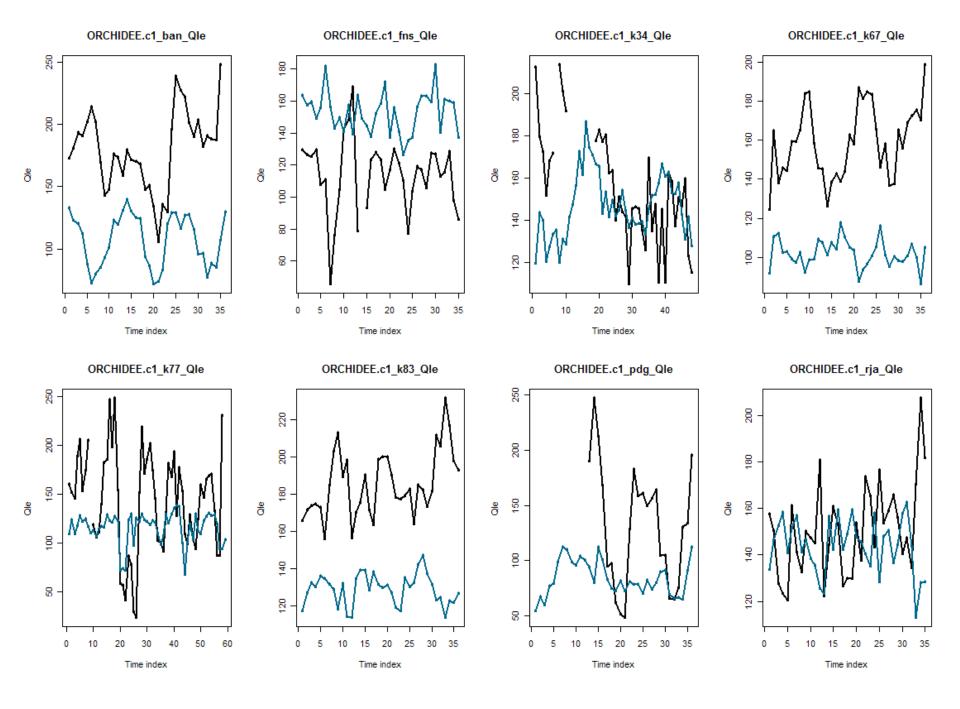
- Variables of interest and corresponding model output
- •Whether daily or monthly averages of hourly data are desire
- Daytime or daily averages

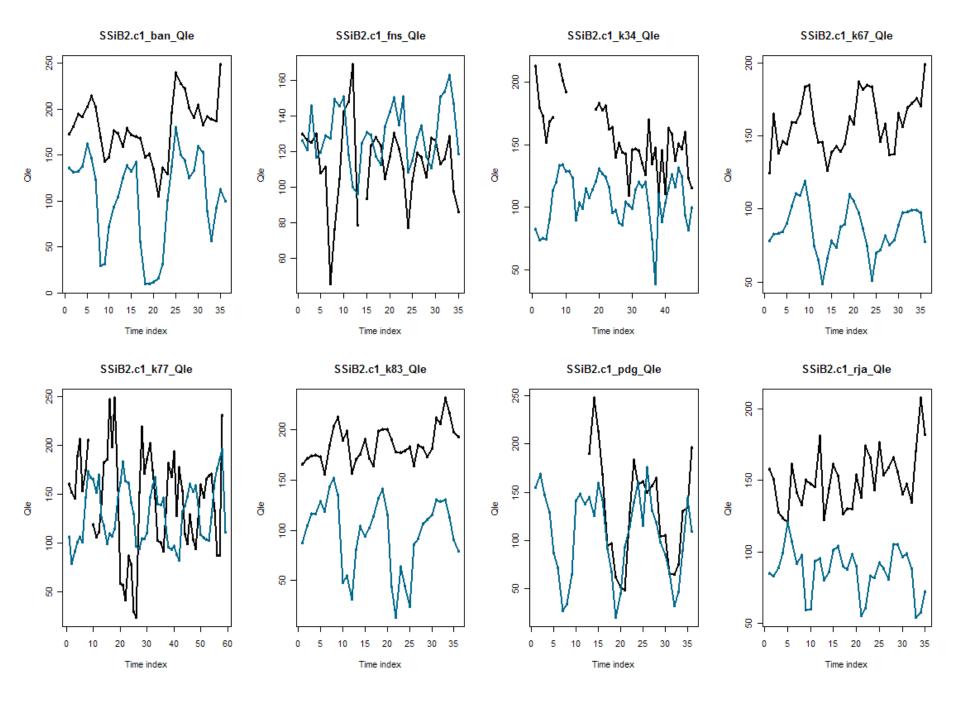


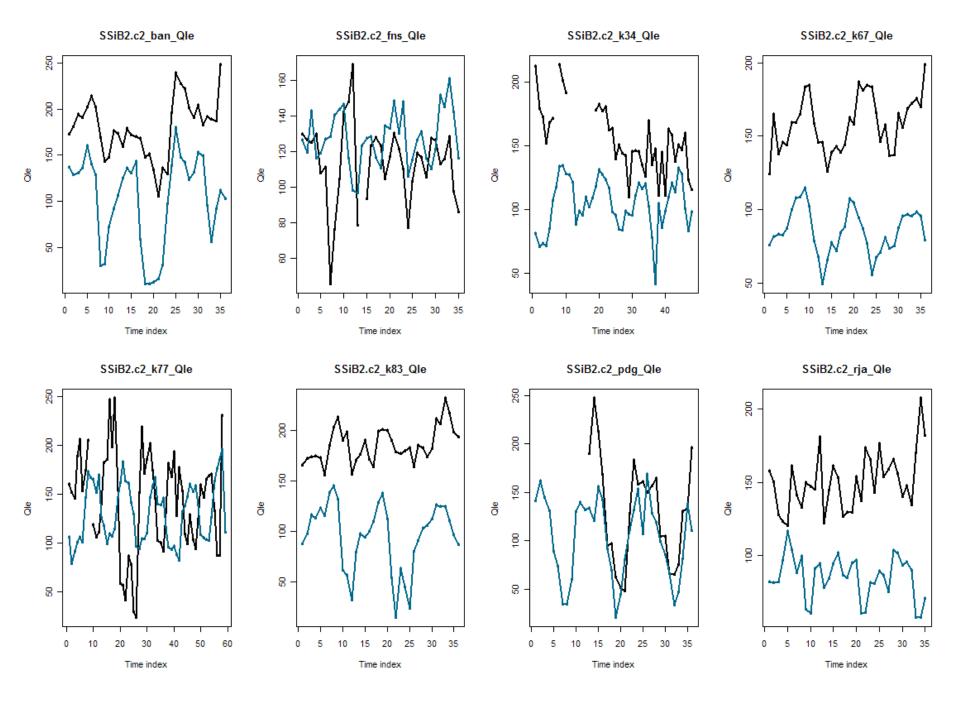
Monthly time series plots

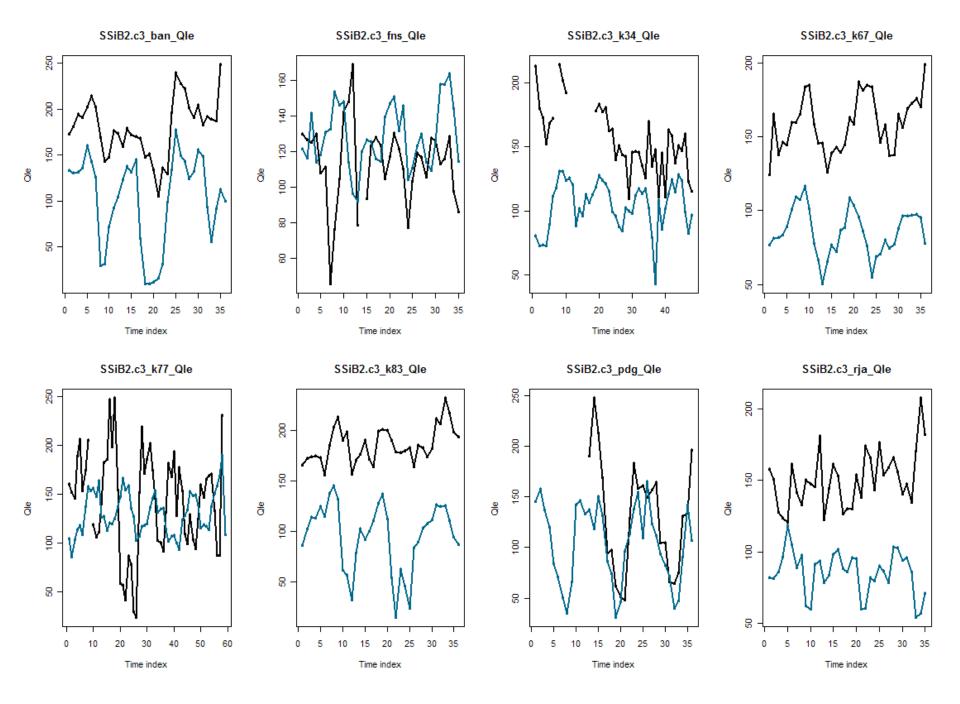


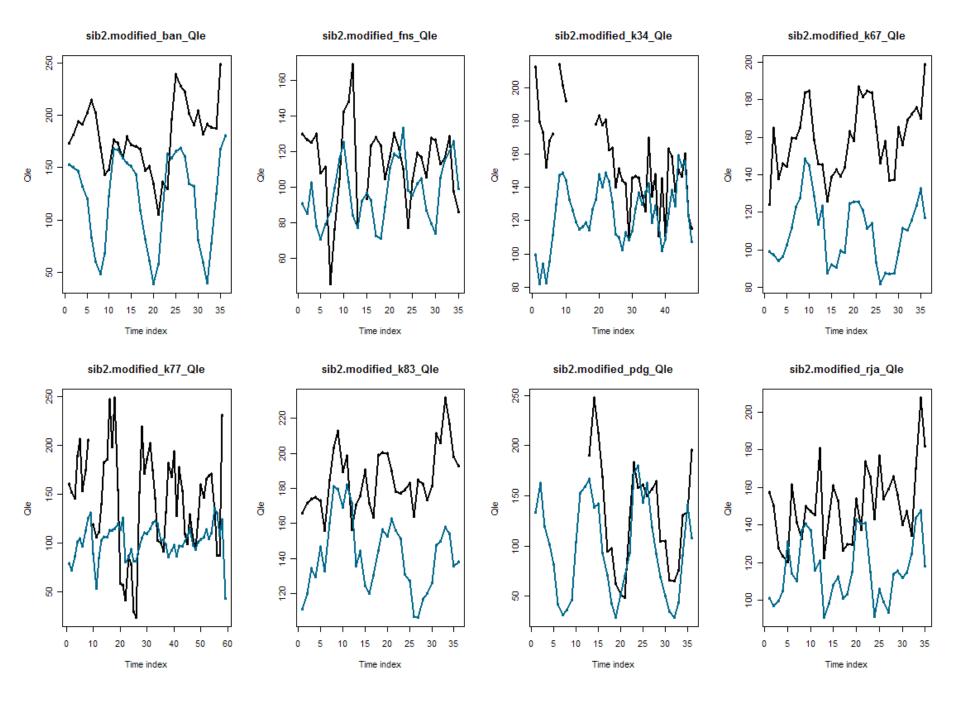


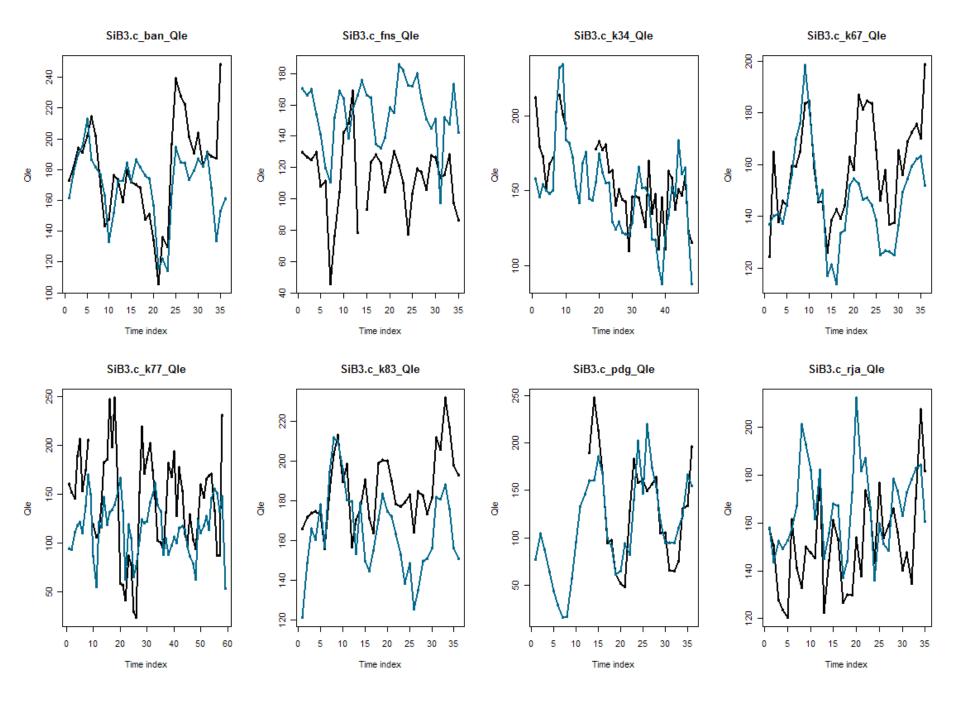


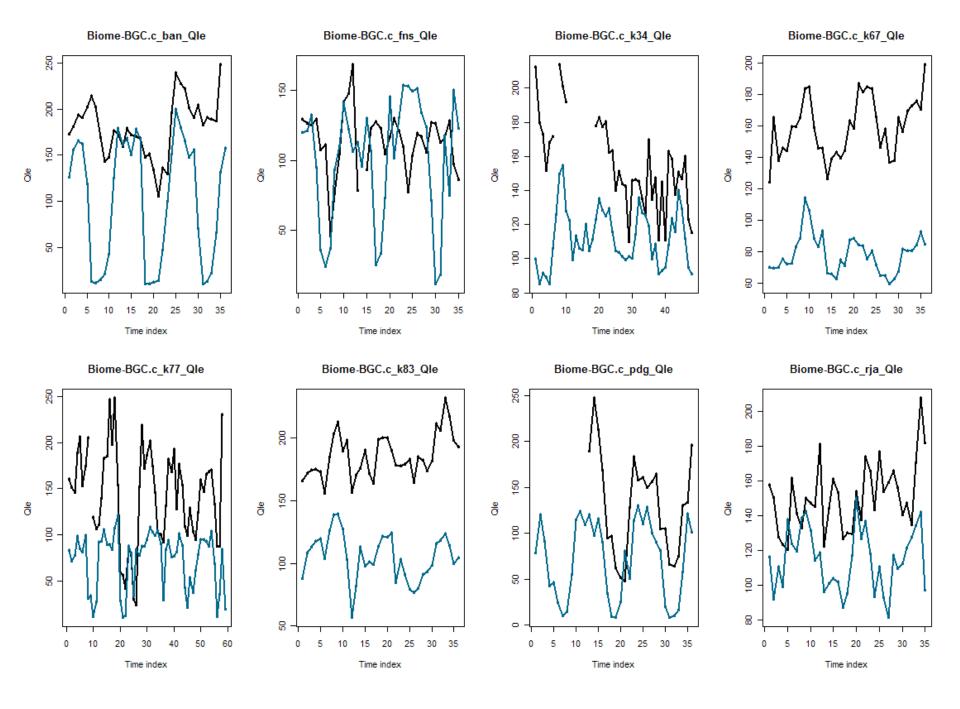


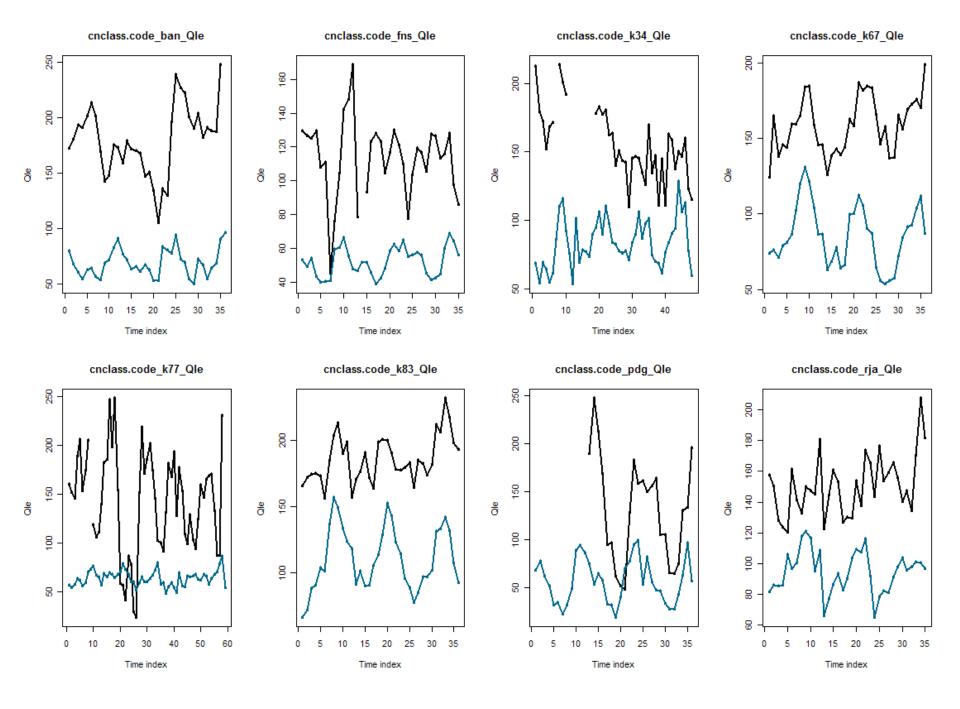


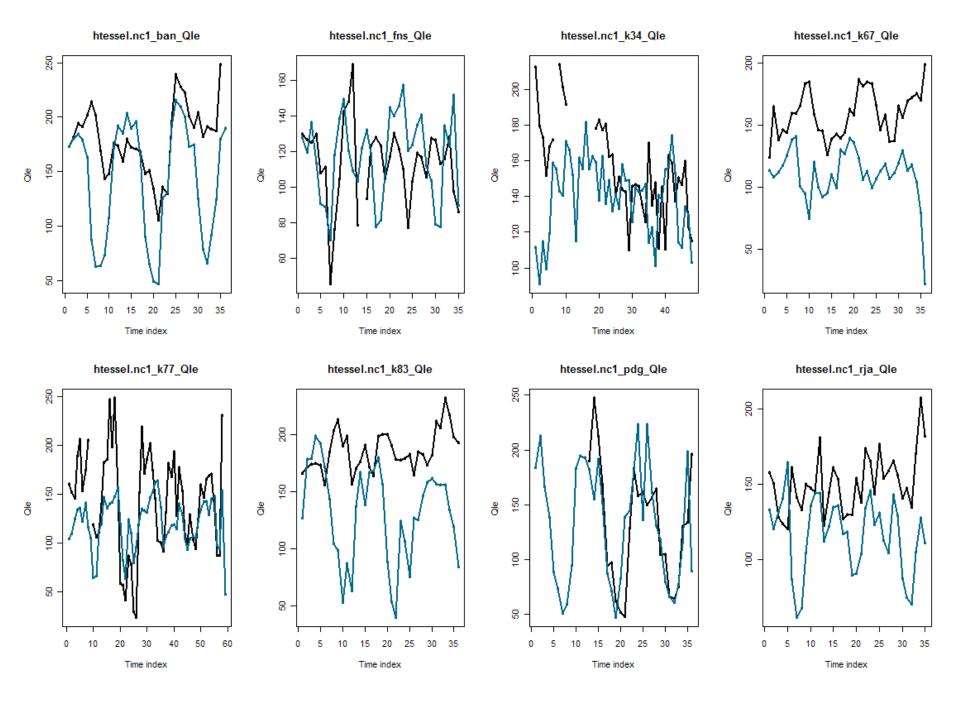


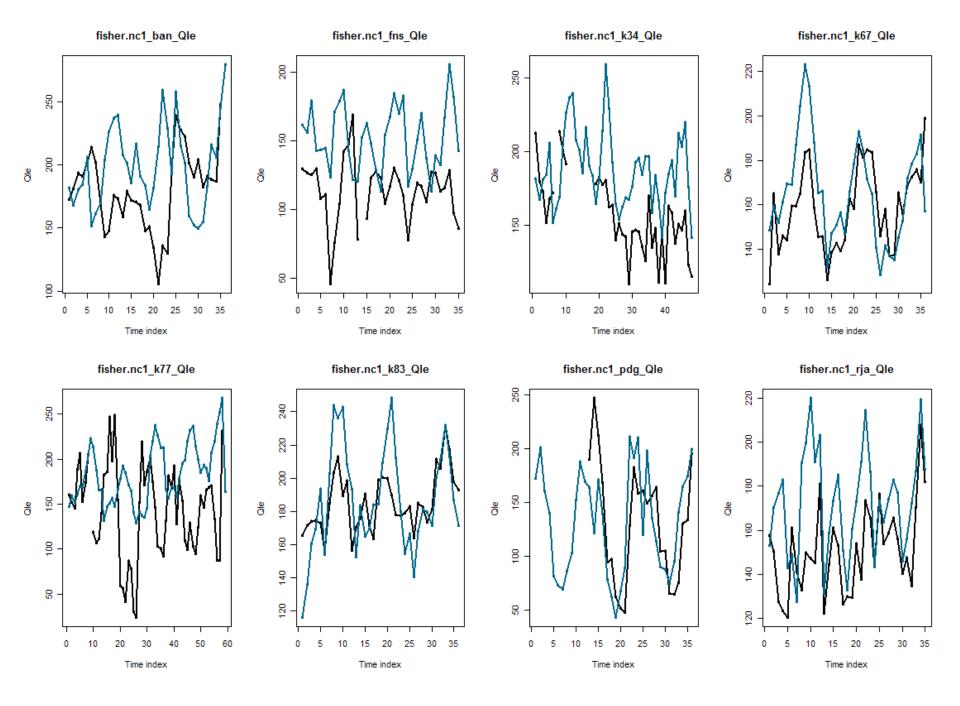


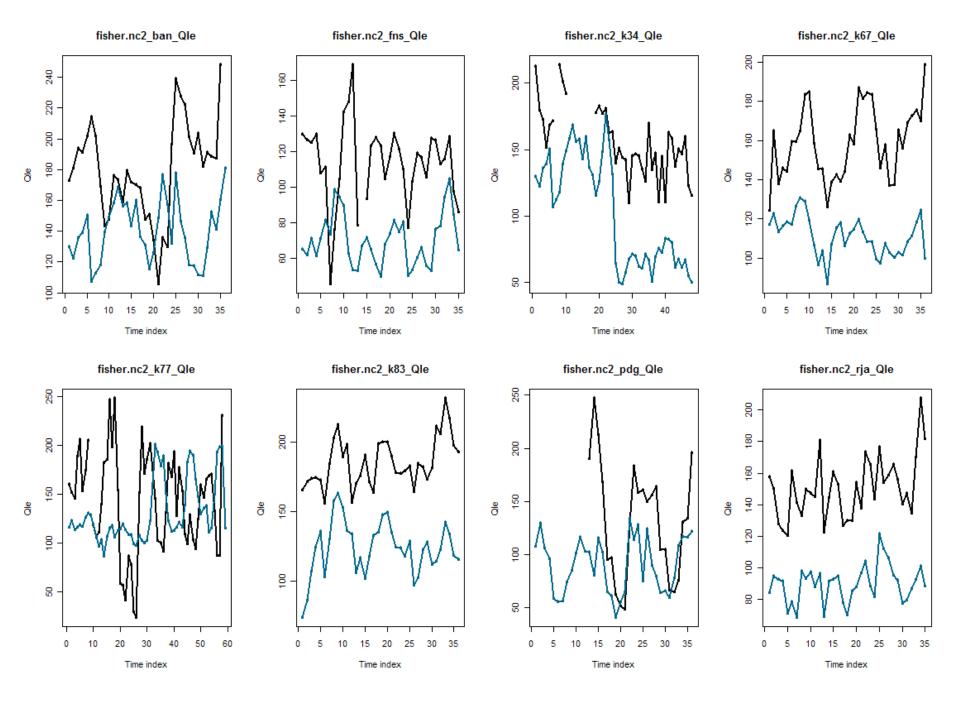


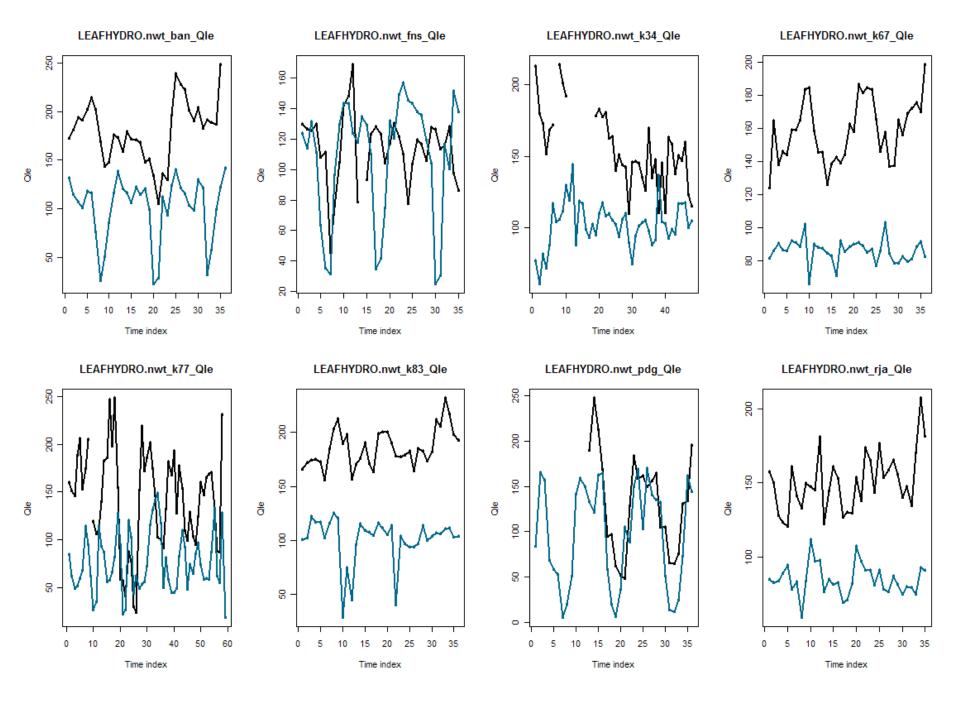


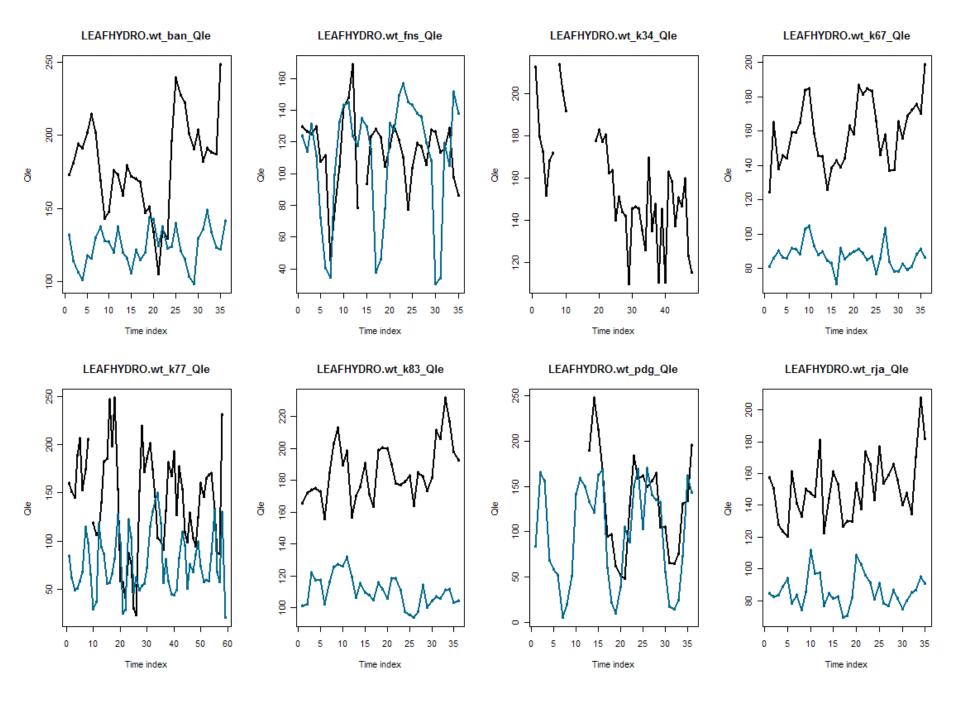


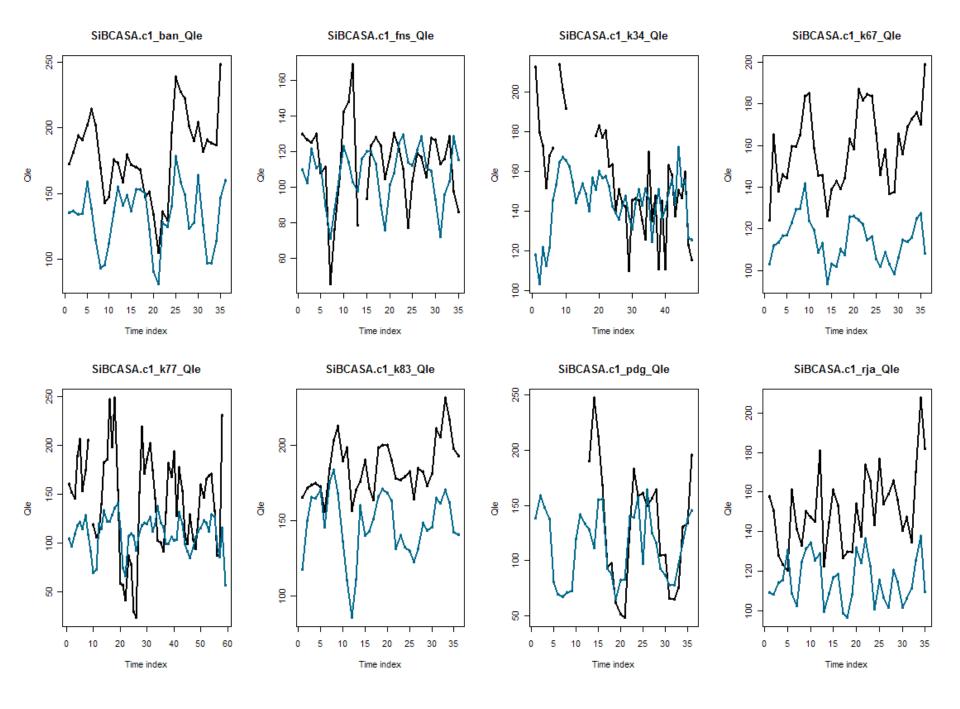


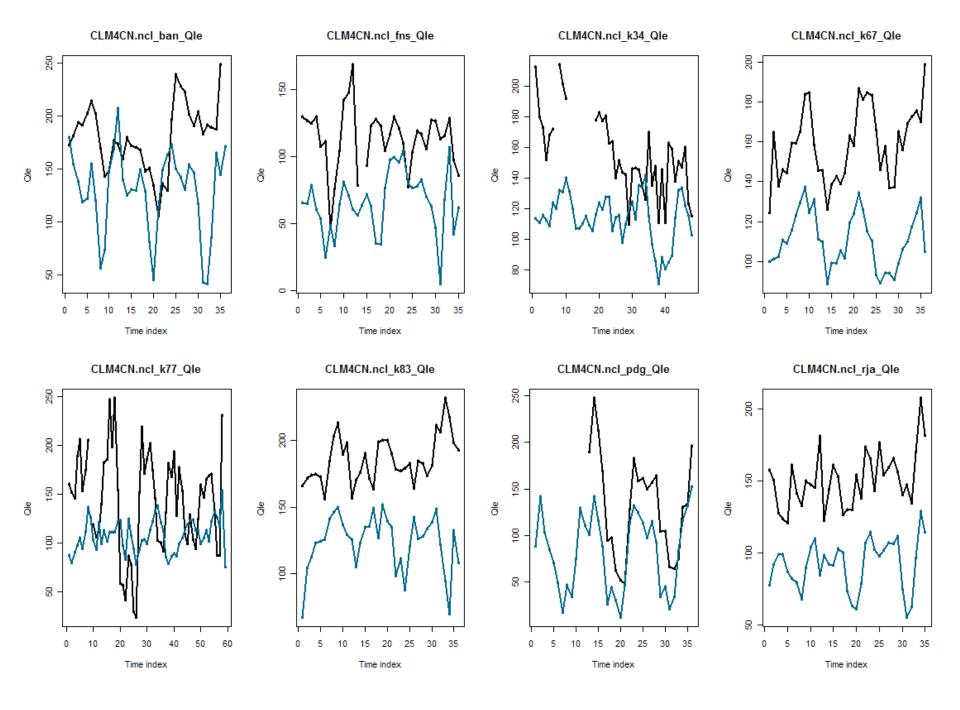


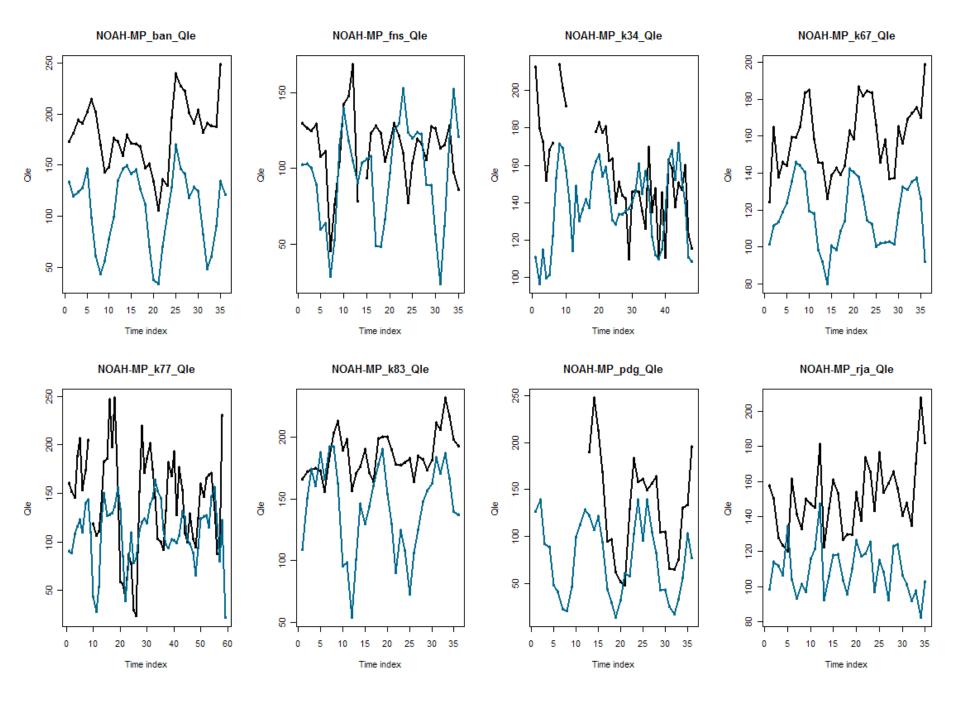


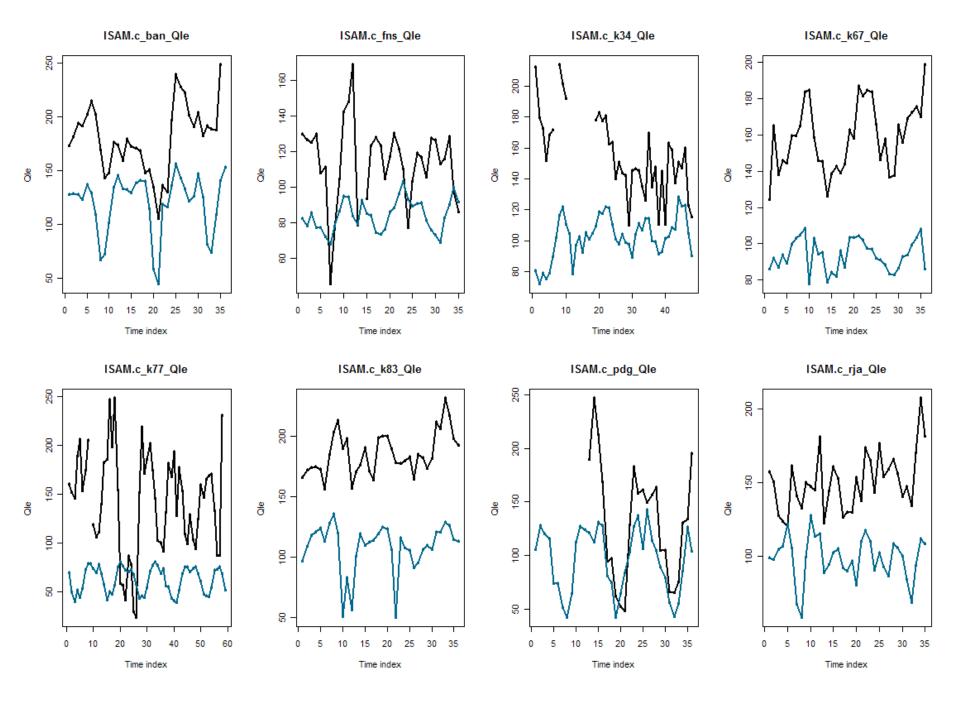


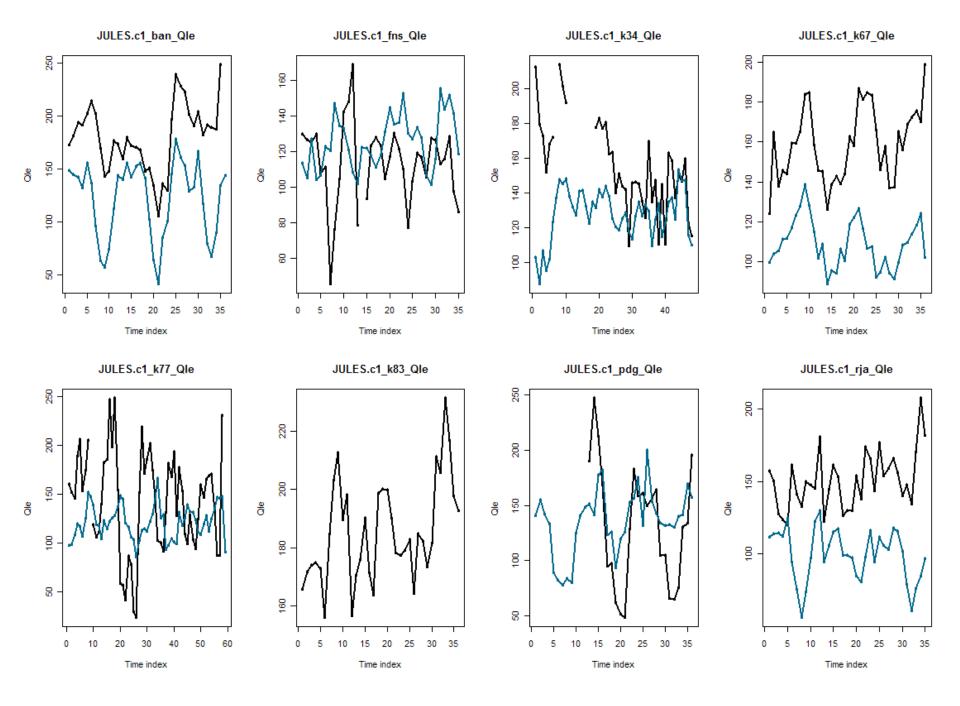


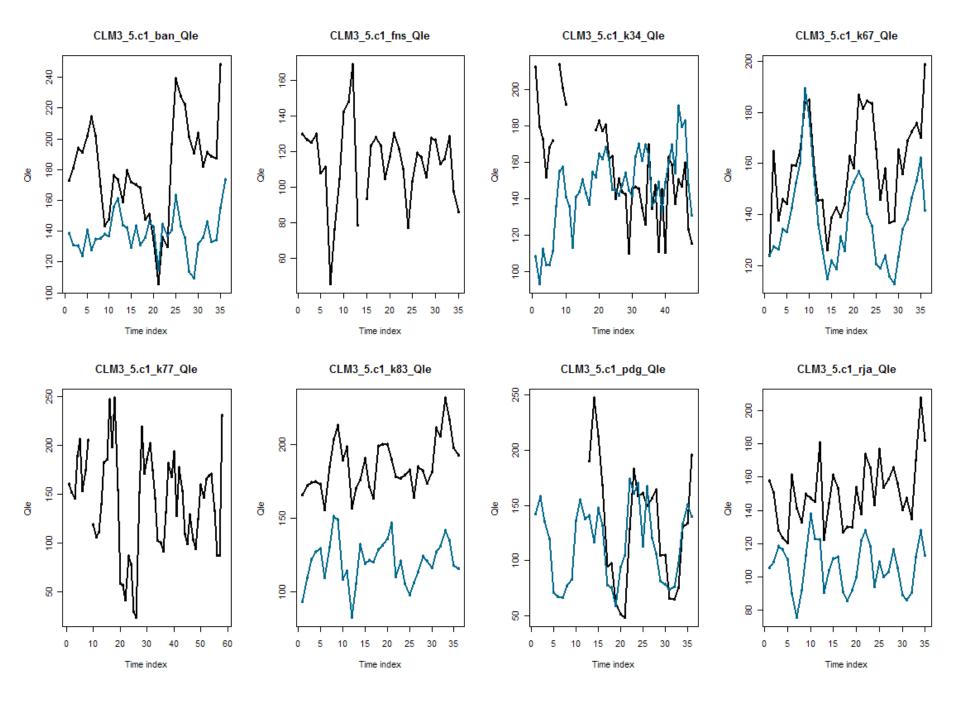


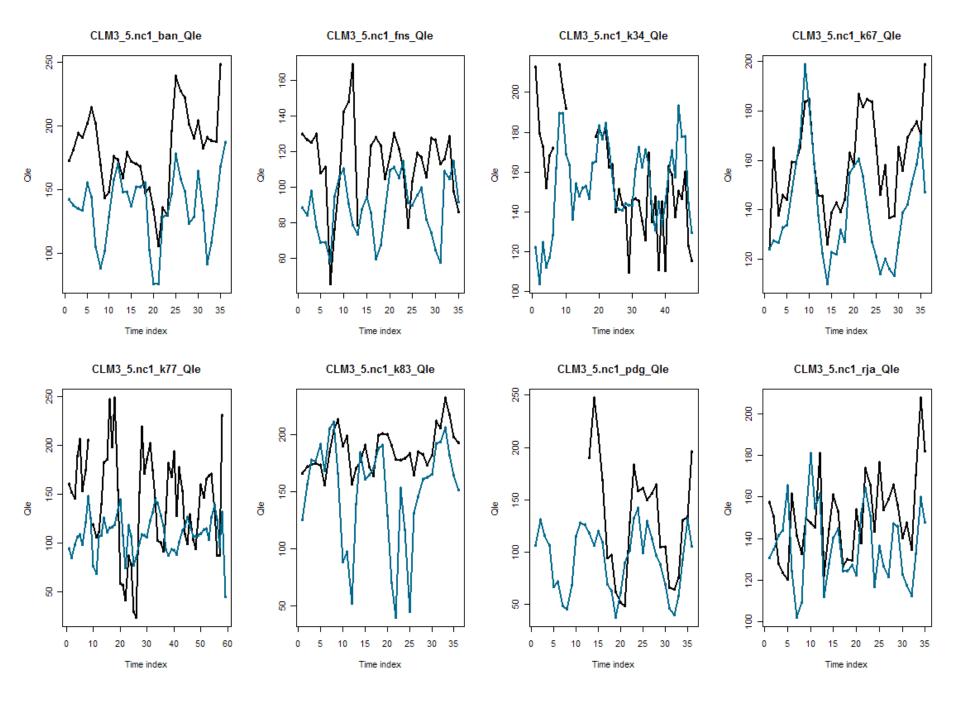




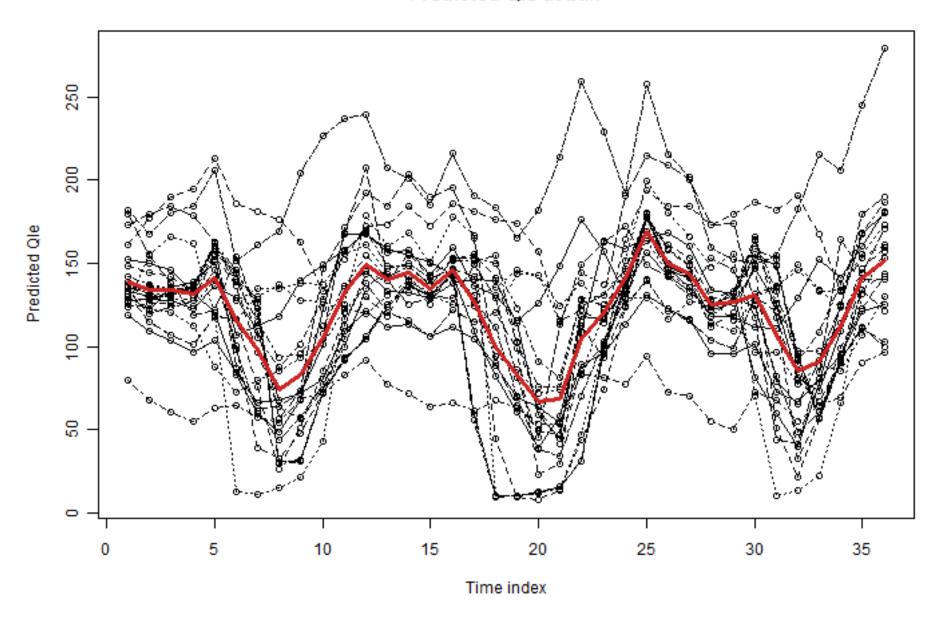




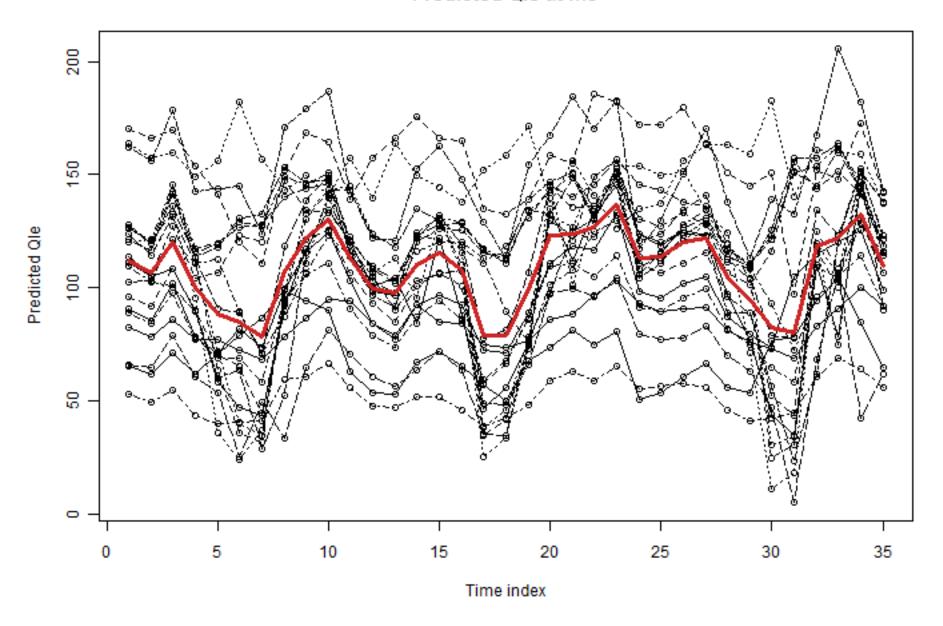


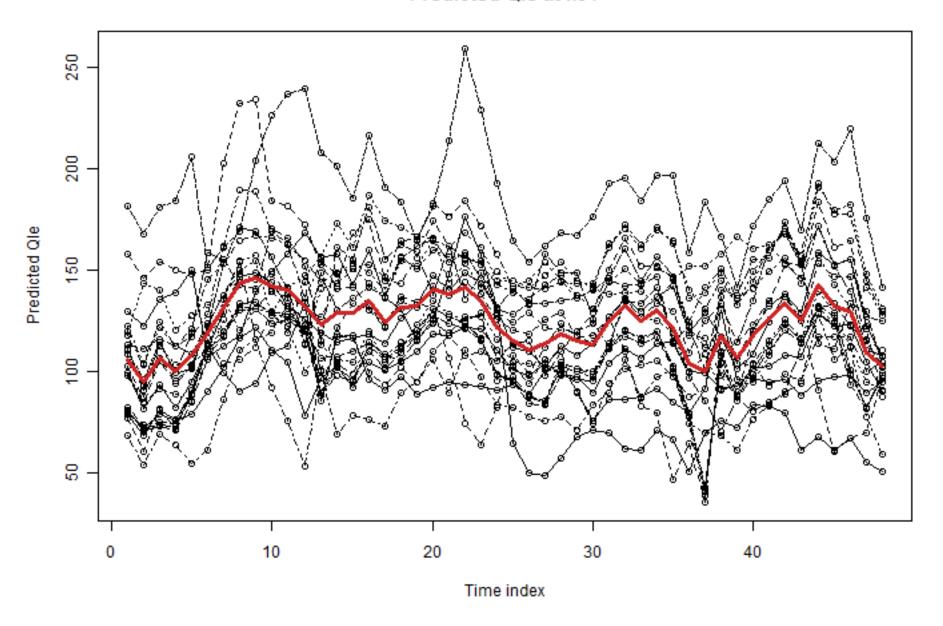


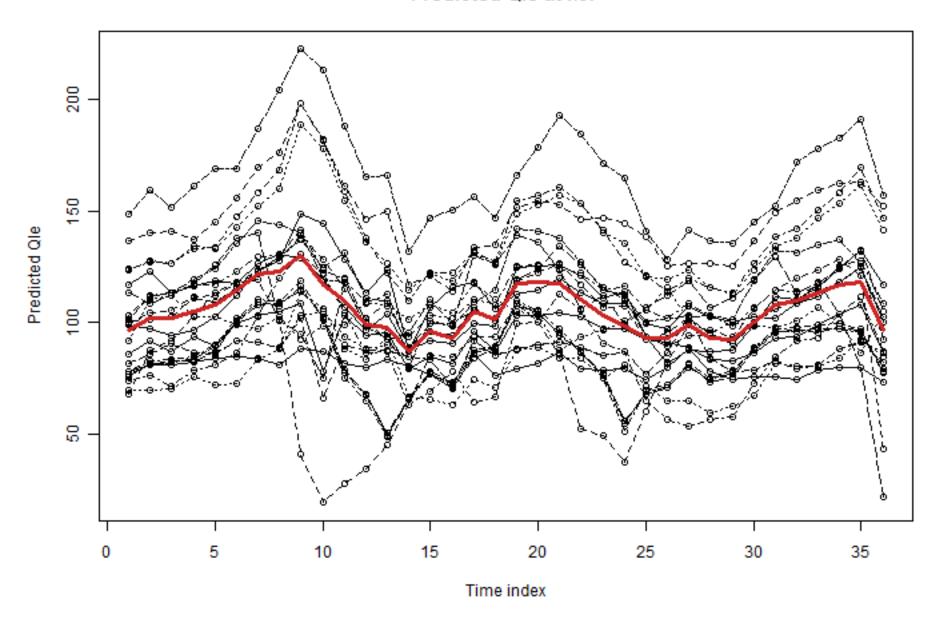
Predicted Qle at ban

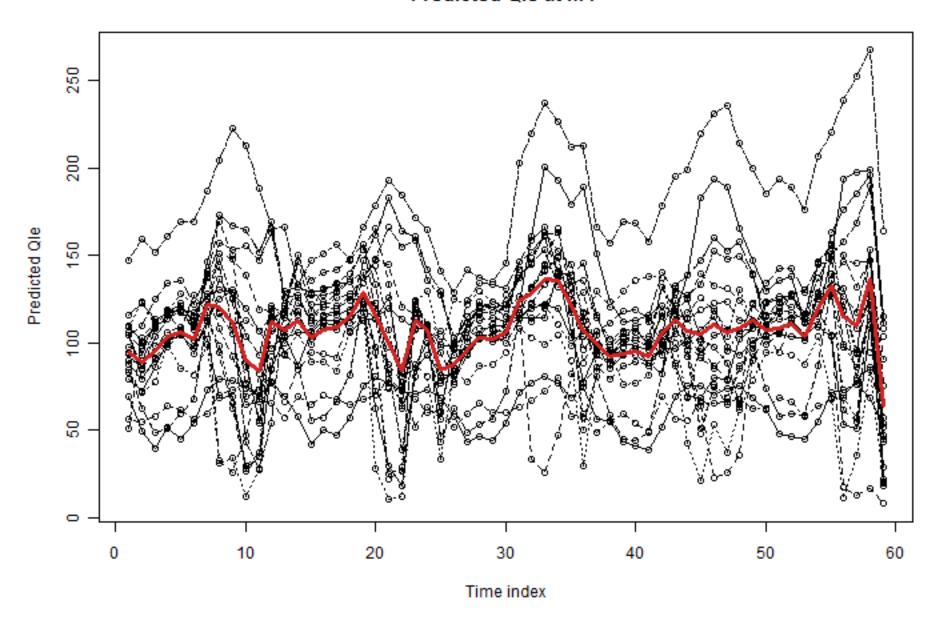


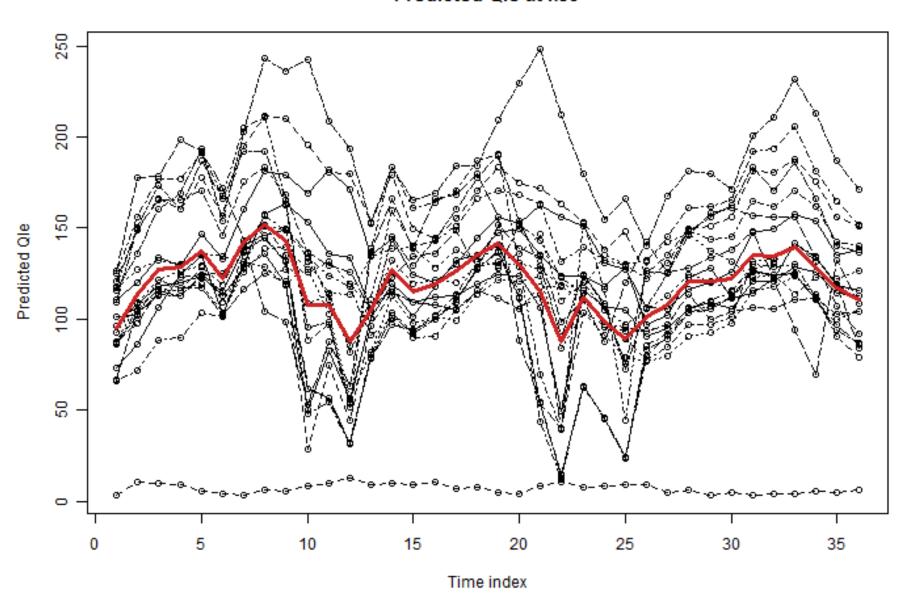
Predicted Qle at fns



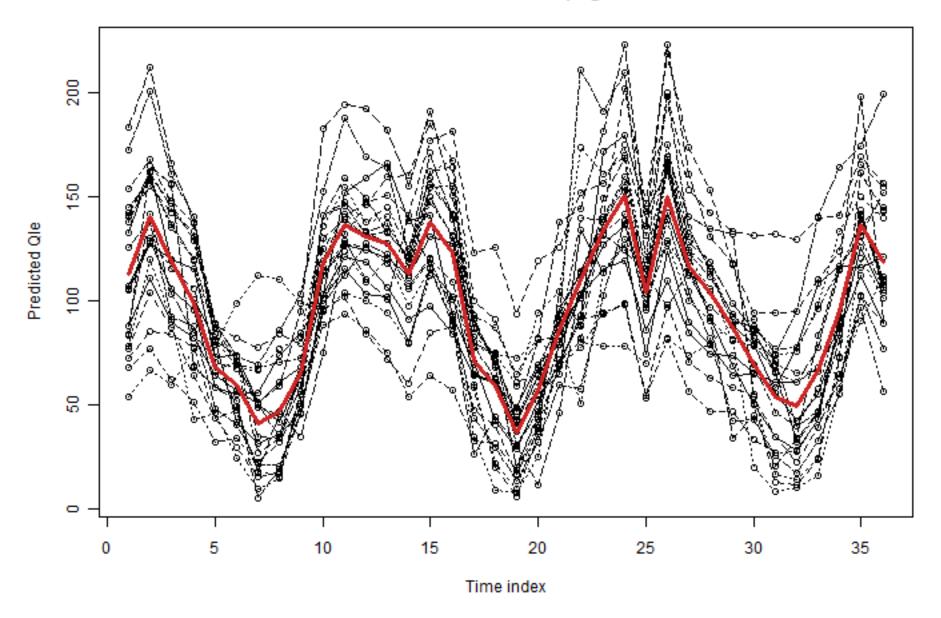




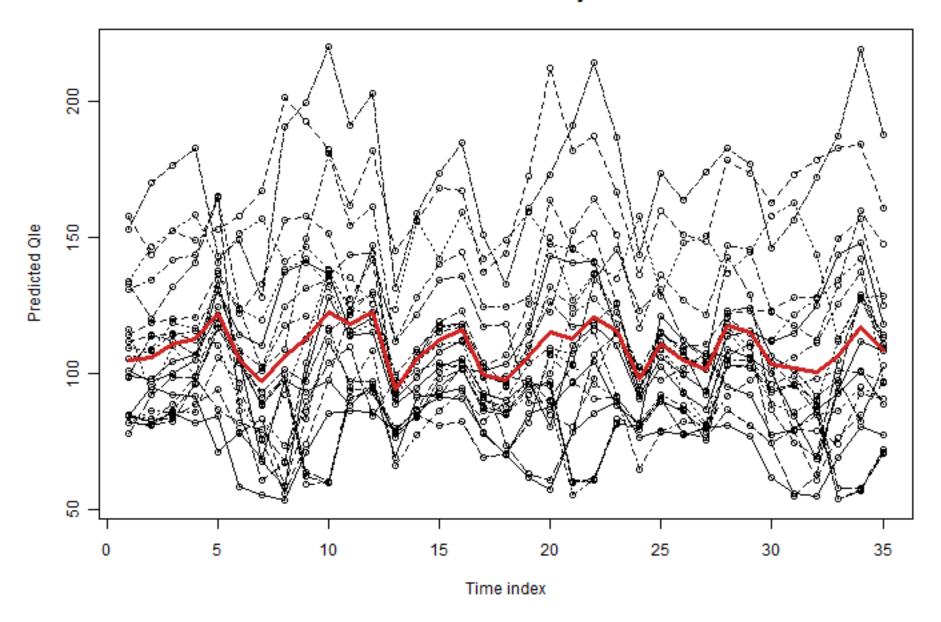




Predicted Qle at pdg



Predicted Qle at rja



Statistical output

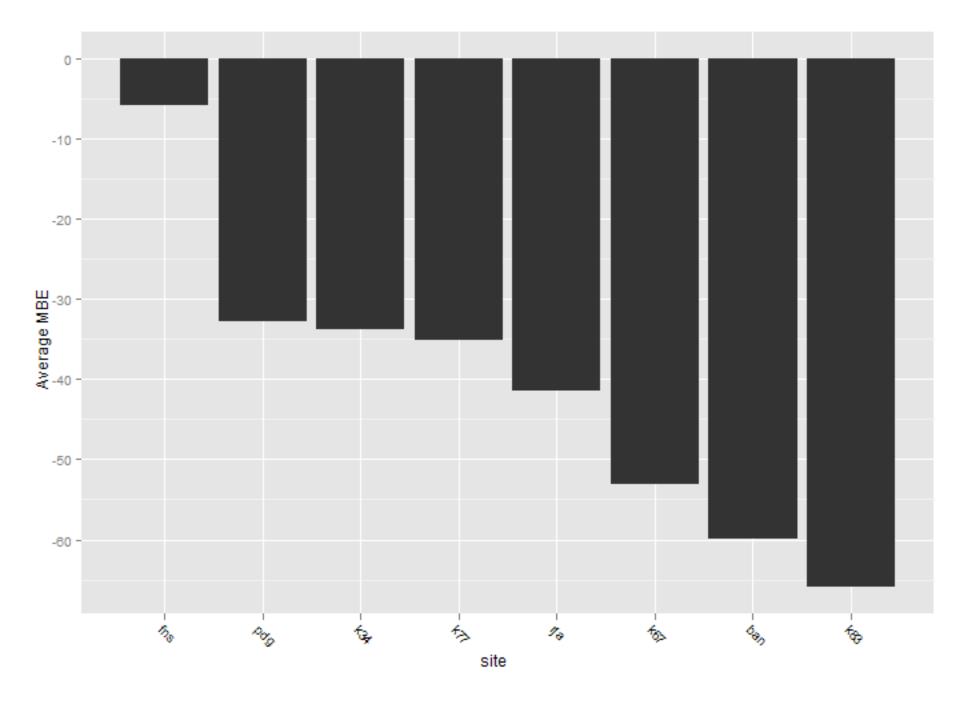
$$MBE = \frac{\sum_{i=1}^{n} (P_i - O_i)}{N}$$

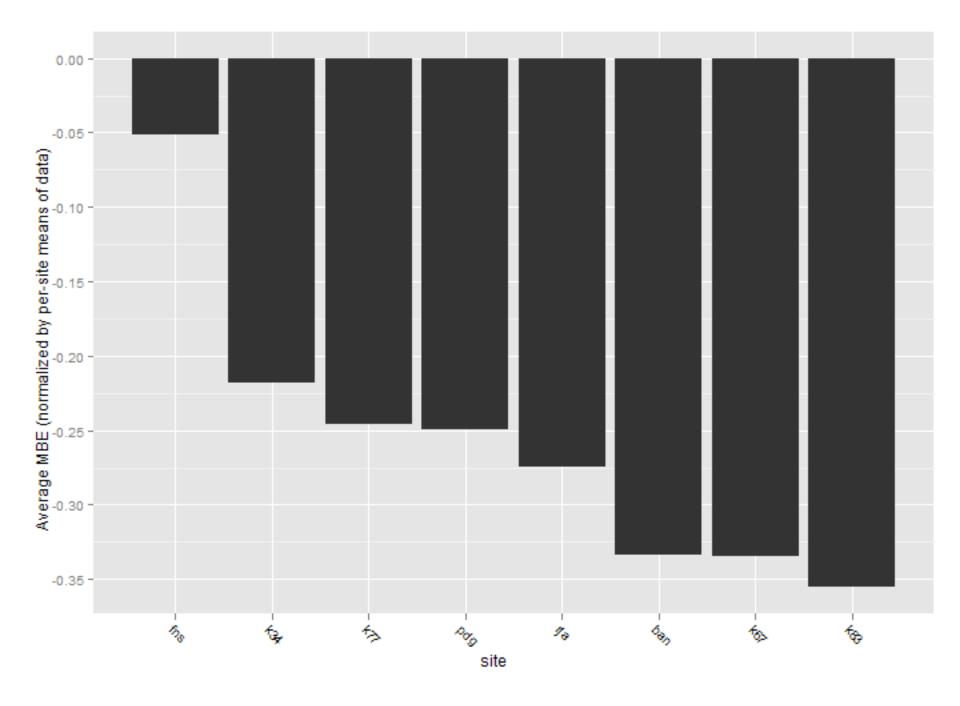
$$MBE_{norm} = \frac{\sum_{i=1}^{n} (P_i - O_i)}{N}$$

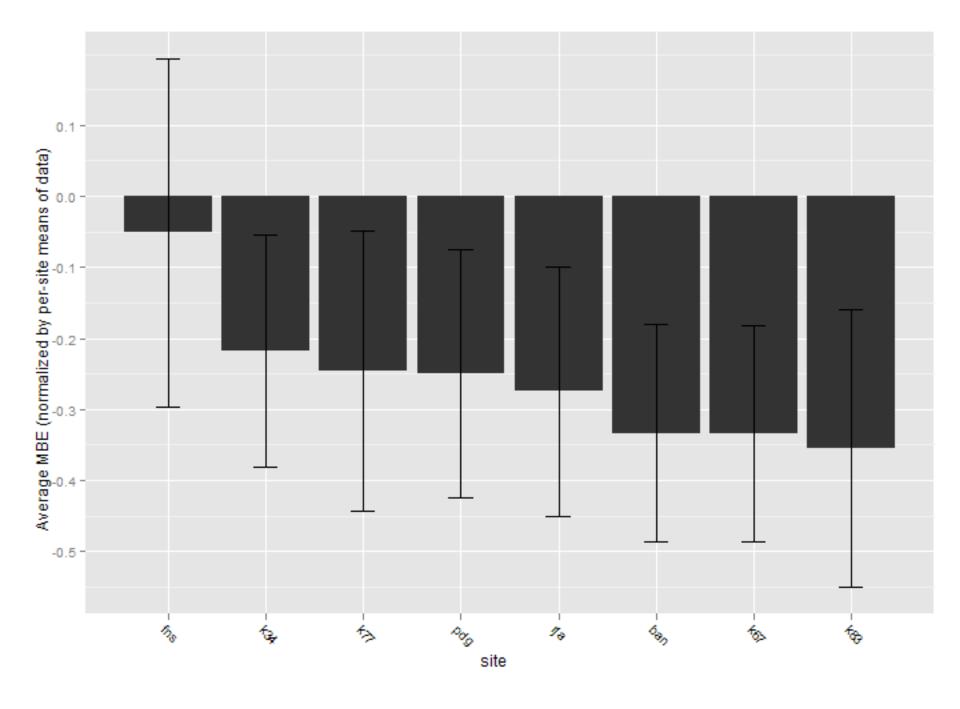
$$RMSE = \sqrt{\frac{\sum_{i=1}^{n} (P_i - O_i)^2}{N}}$$

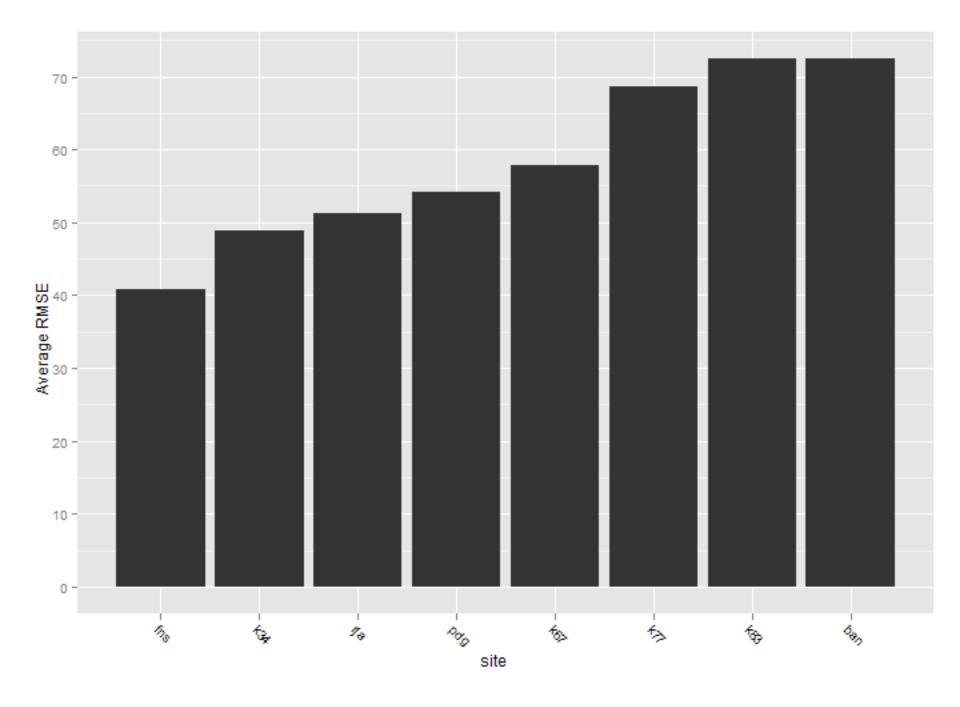
$$RMSE_{norm} = \frac{\sqrt{\frac{\sum_{i=1}^{n} (P_i - O_i)^2}{N}}}{N}$$

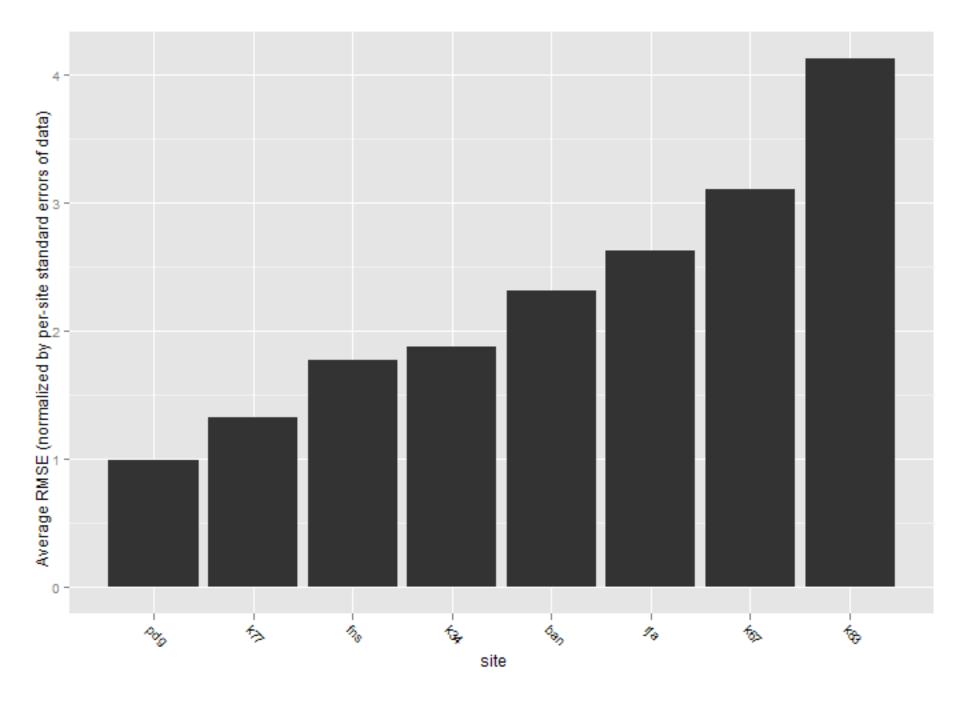
$$RMSE_{norm} = \frac{\sqrt{\frac{\sum_{i=1}^{n} (P_i - O_i)^2}{N}}}{N}$$

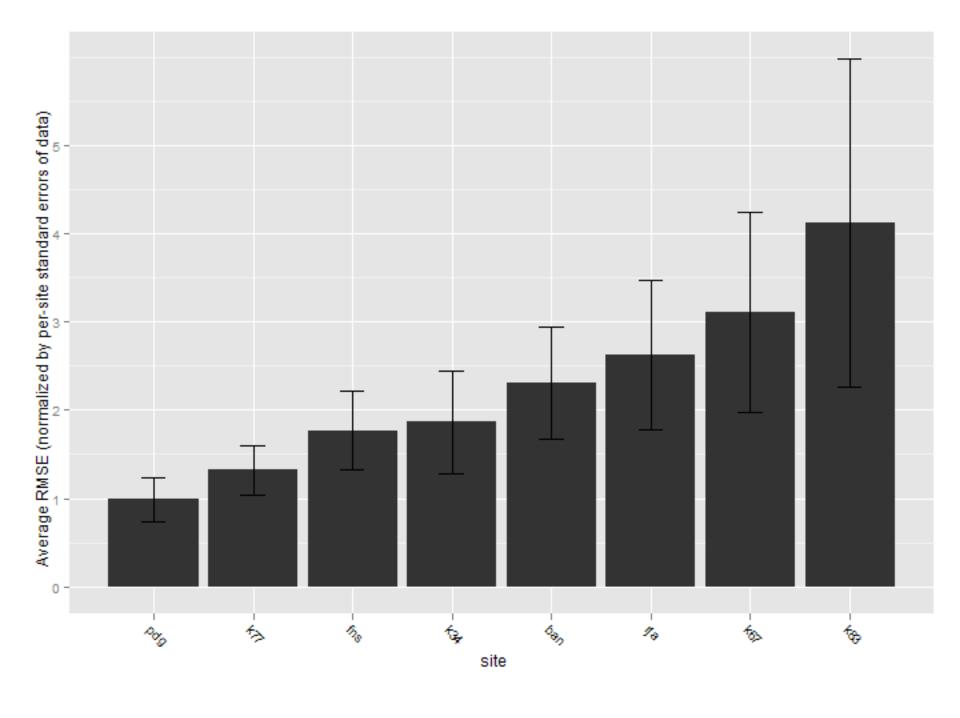


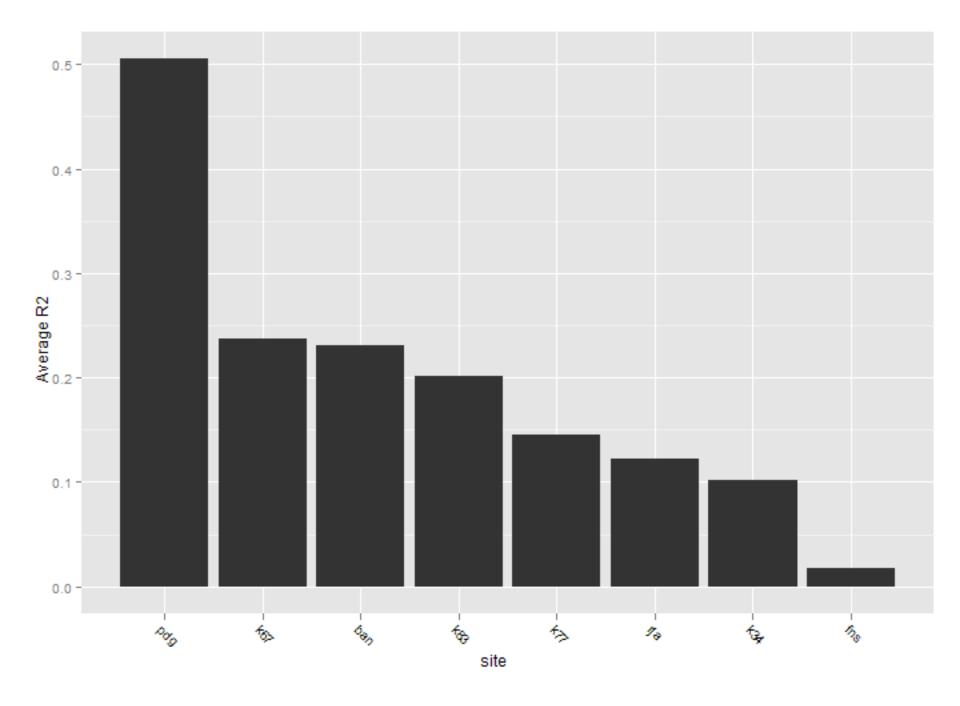


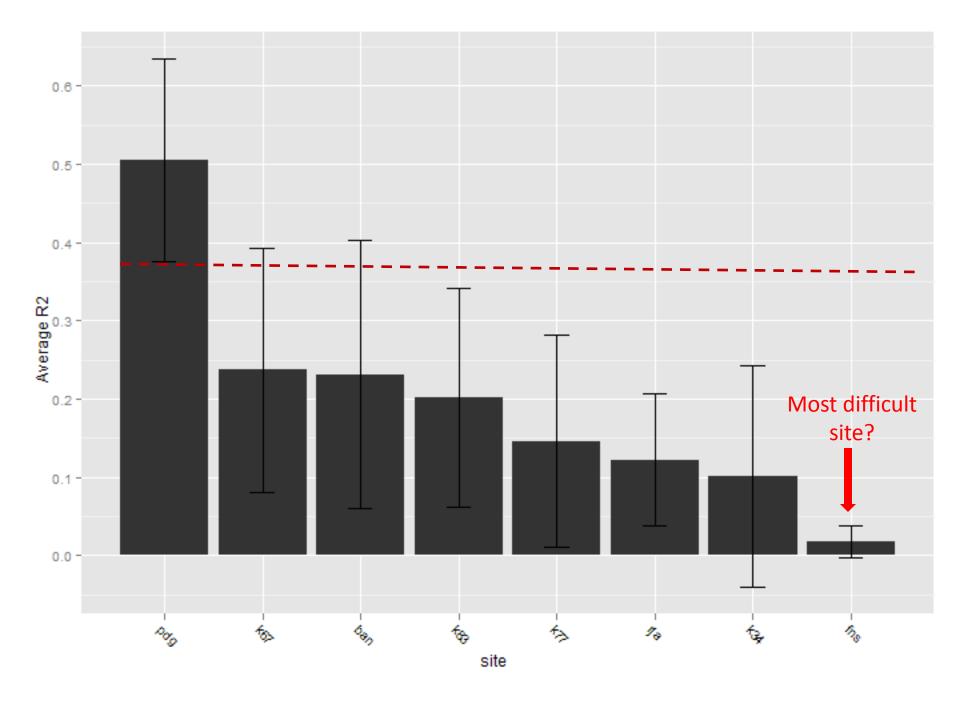


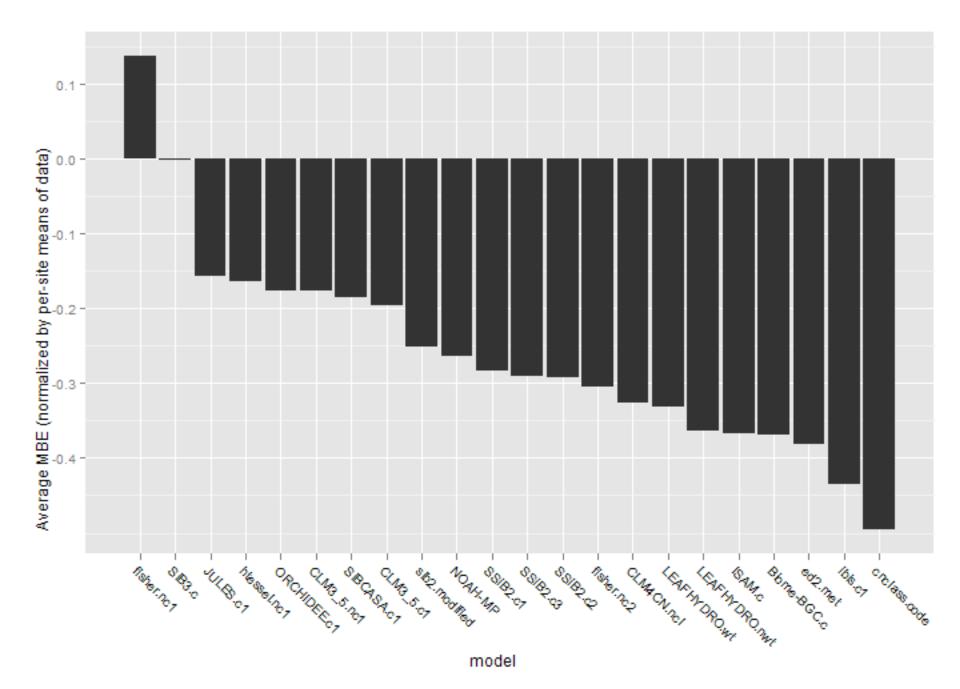


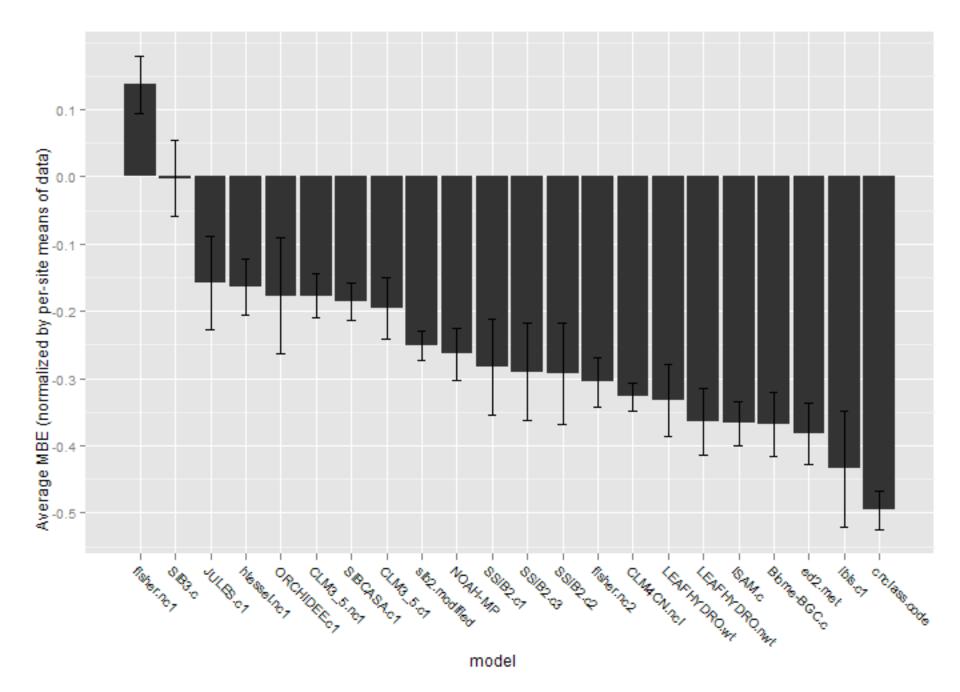


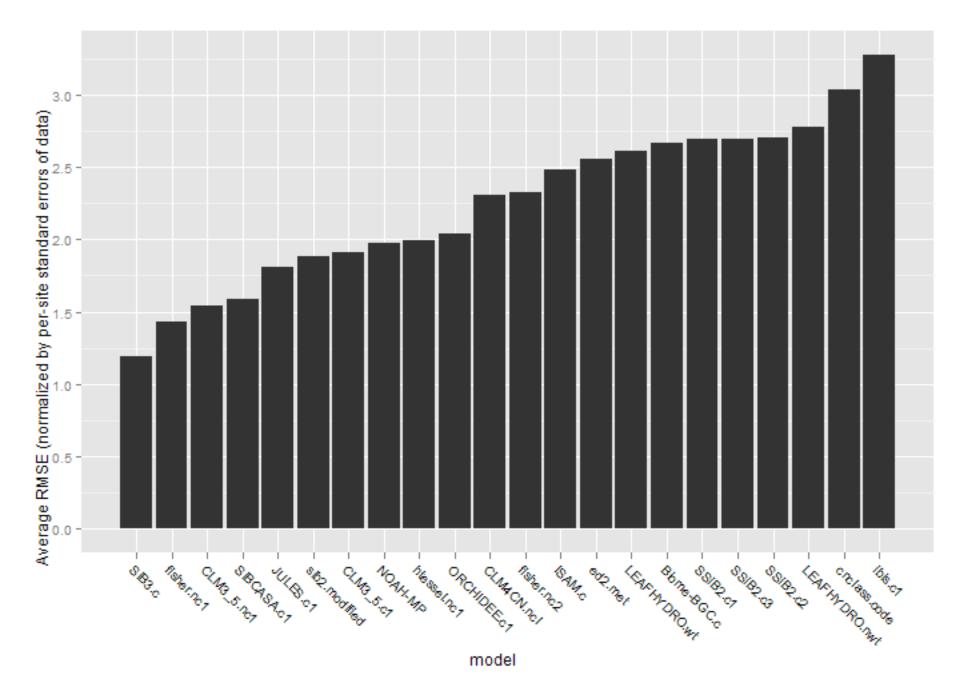


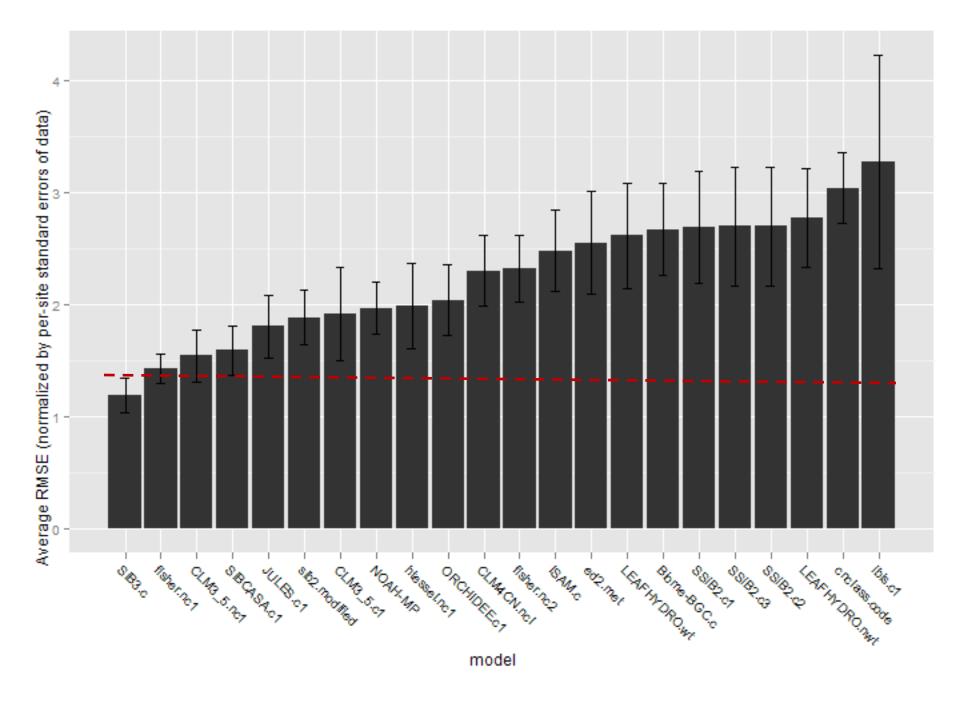


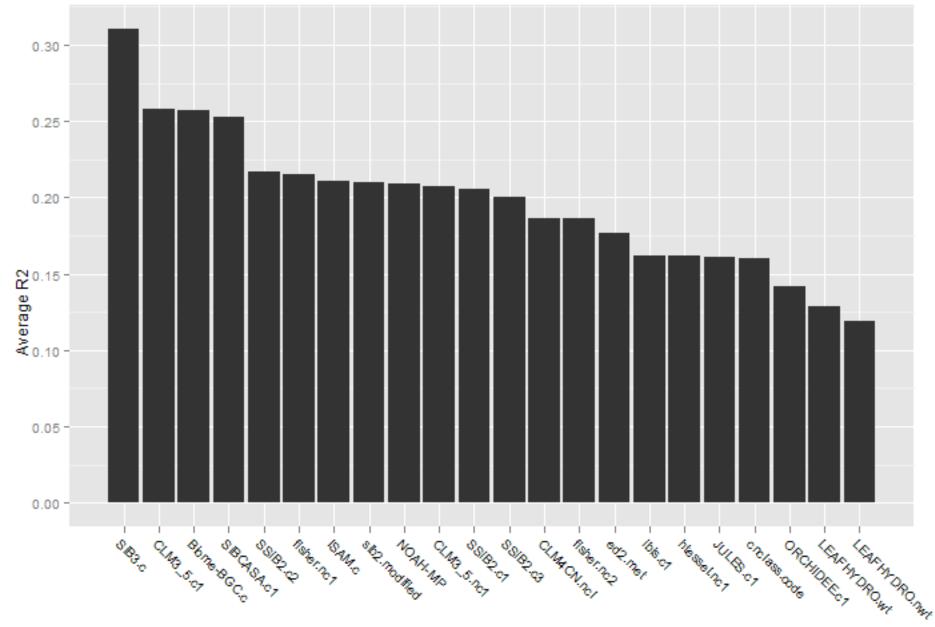




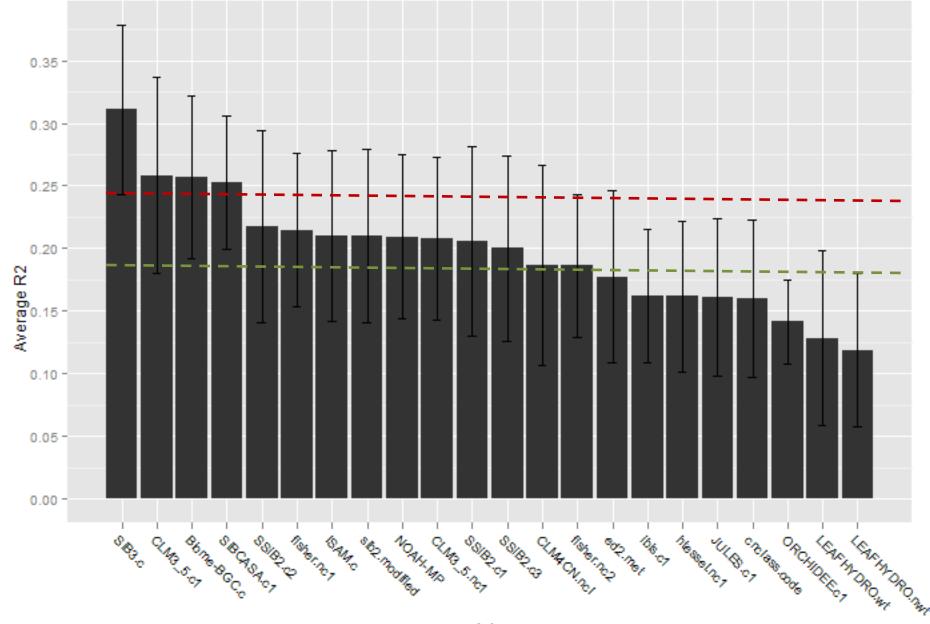








model



model

Discussion

- SiB3 emerges as overall 'strongest performer,' though with a large degree of variation.
- Models were in the right range at the FNS site, but lacked agreement with the observed data as measured by R².
- PDG emerges at the best overall site, but with a tendency toward model under—prediction.
- Seasonality dominates performance metrics much more at some sites than others.
- Observed data is very stochastic for some sites—how confident are we that these values are 'true' as opposed to noisy?
- No model or site is the clear `winner'—at this point standard errors are too large.
- What is observed data uncertainty? May need to think about whether predictions fall within error range of observations.

Next steps...

 Protocol for error attribution—a mix of summary stats, OLS, and machine learning algorithms

Thanks! Questions? Suggestions?