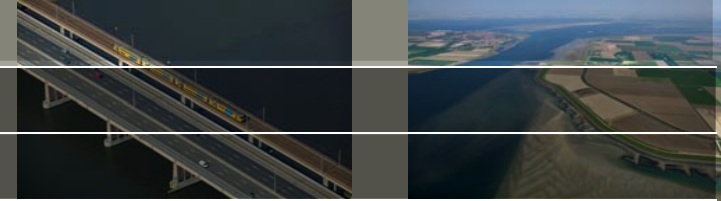




Estimating predictive hydrological uncertainty using Quantile Regression

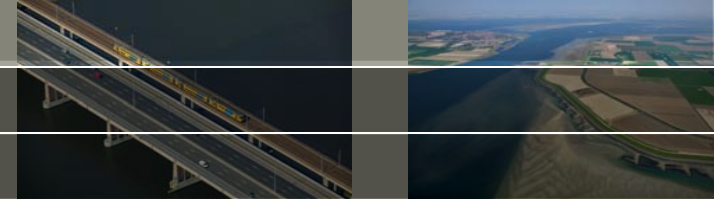
Jan Verkade
Albrecht Weerts
Steven Weijs

June 2011



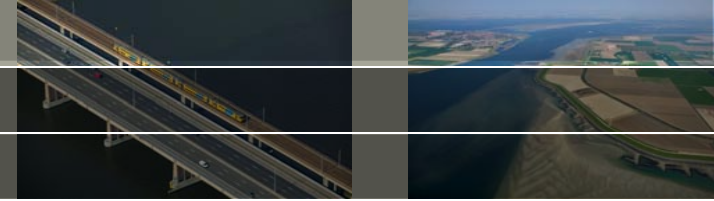
This presentation is also available on slideshare.net

Visit twitter.com/janverkade for the address



Quantile Regression

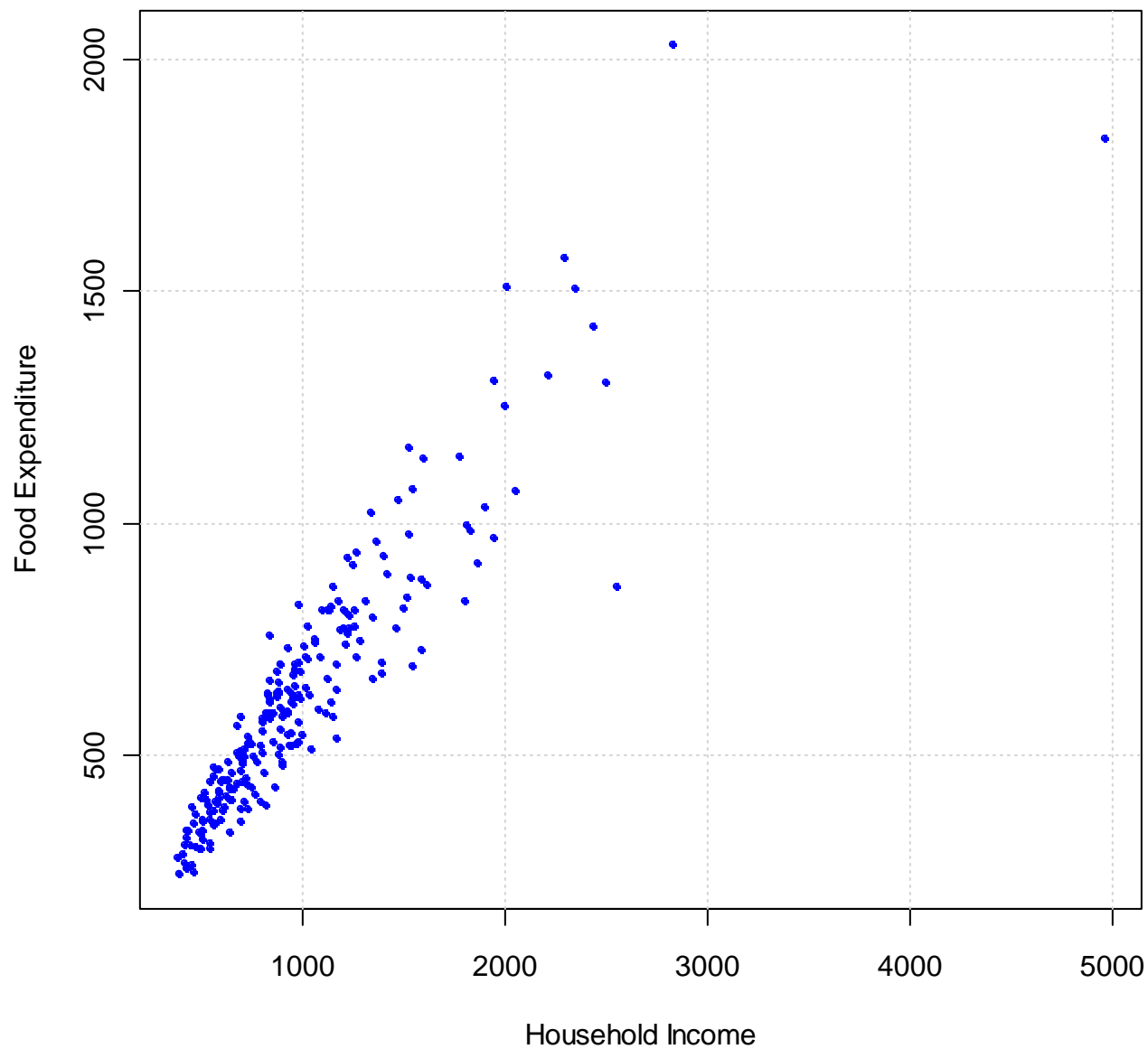
- Origin in econometrics (Koenker et al.)
- Brought to our attention by Andy Woods @ 2008 HEPEX meeting
- First research implementations:
 - ~25 NFFS basins (England and Wales)
 - White Cart (Scotland)
 - Ovens River (Australia)
 - HEPEX-basins (USA) (in progress)
- Now operationally used in National Flood Forecasting System (England and Wales)



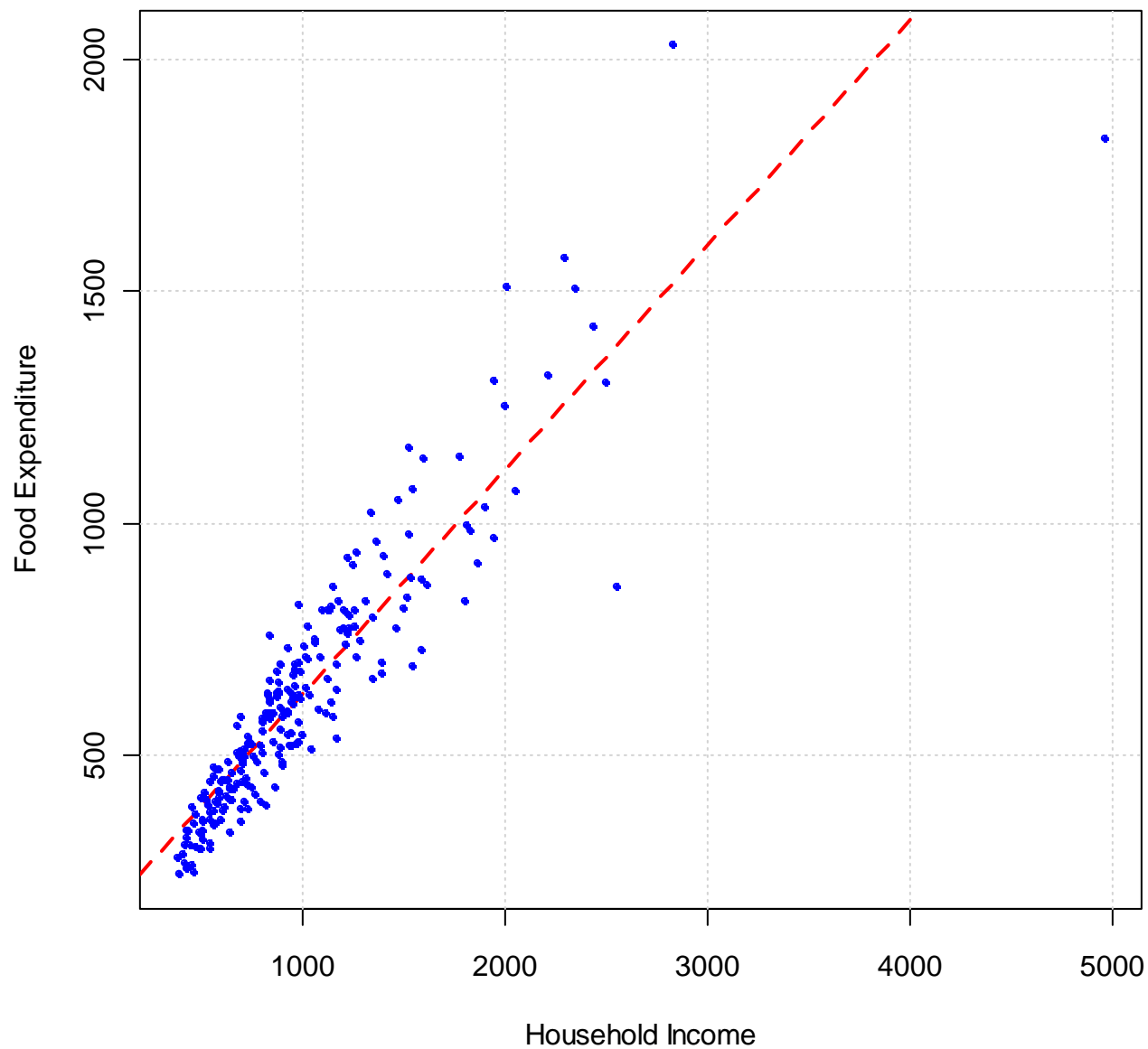
Quantile Regression: principles

- QR is a method for describing conditional quantiles
- Rather than minimising the mean *squared* error (MSE)
- QR is based on minimising the mean *absolute* error (MAE)
- This yields not the sample *mean* but the sample *median*
- Other quantiles may be derived by adding weights to errors
- E.g. weight = .1 for positive errors and .9 for negative errors
- Fitting models may be done in transformed space to account for heteroscedasticity

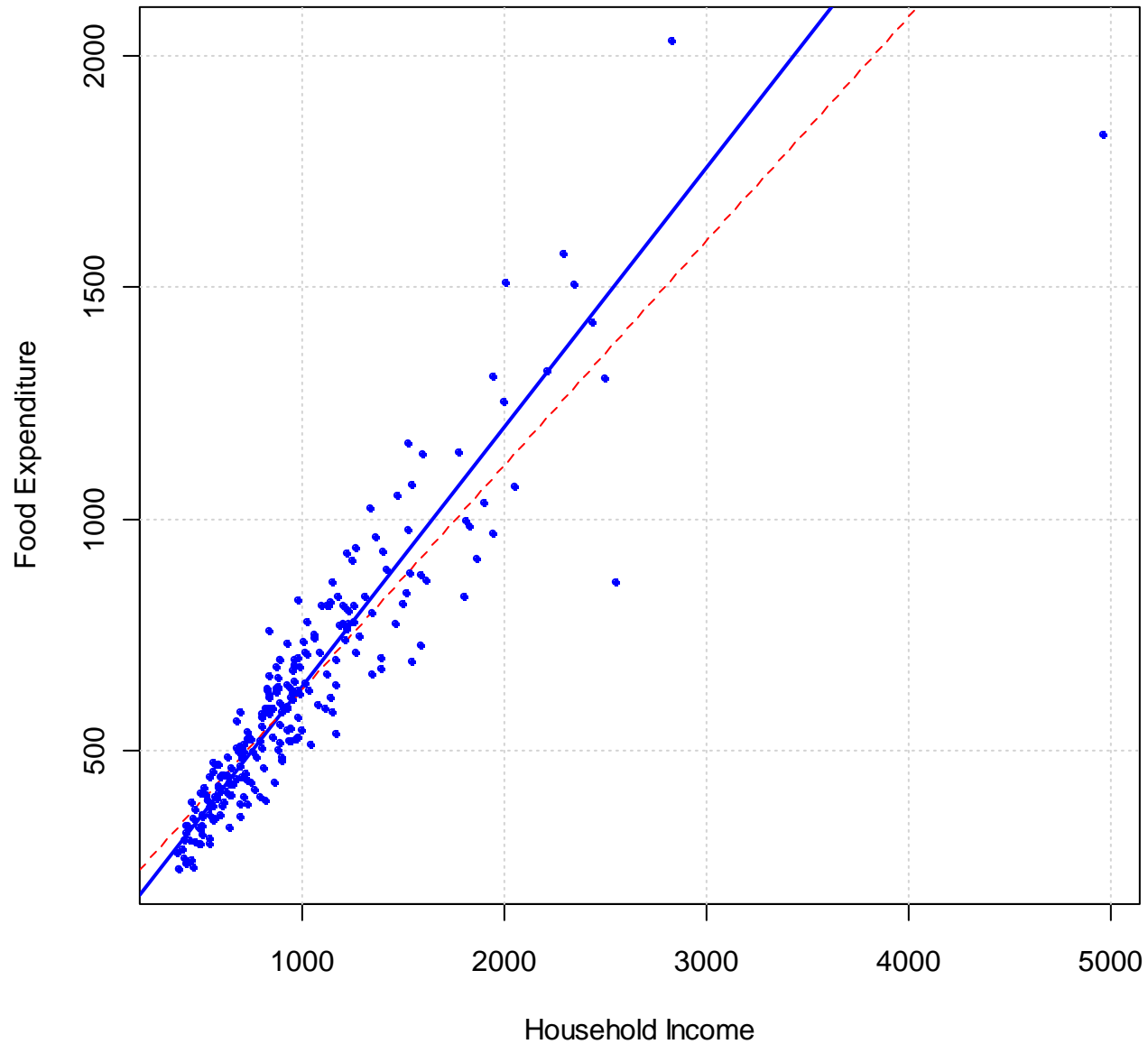
Engel data, 19th century, Belgium



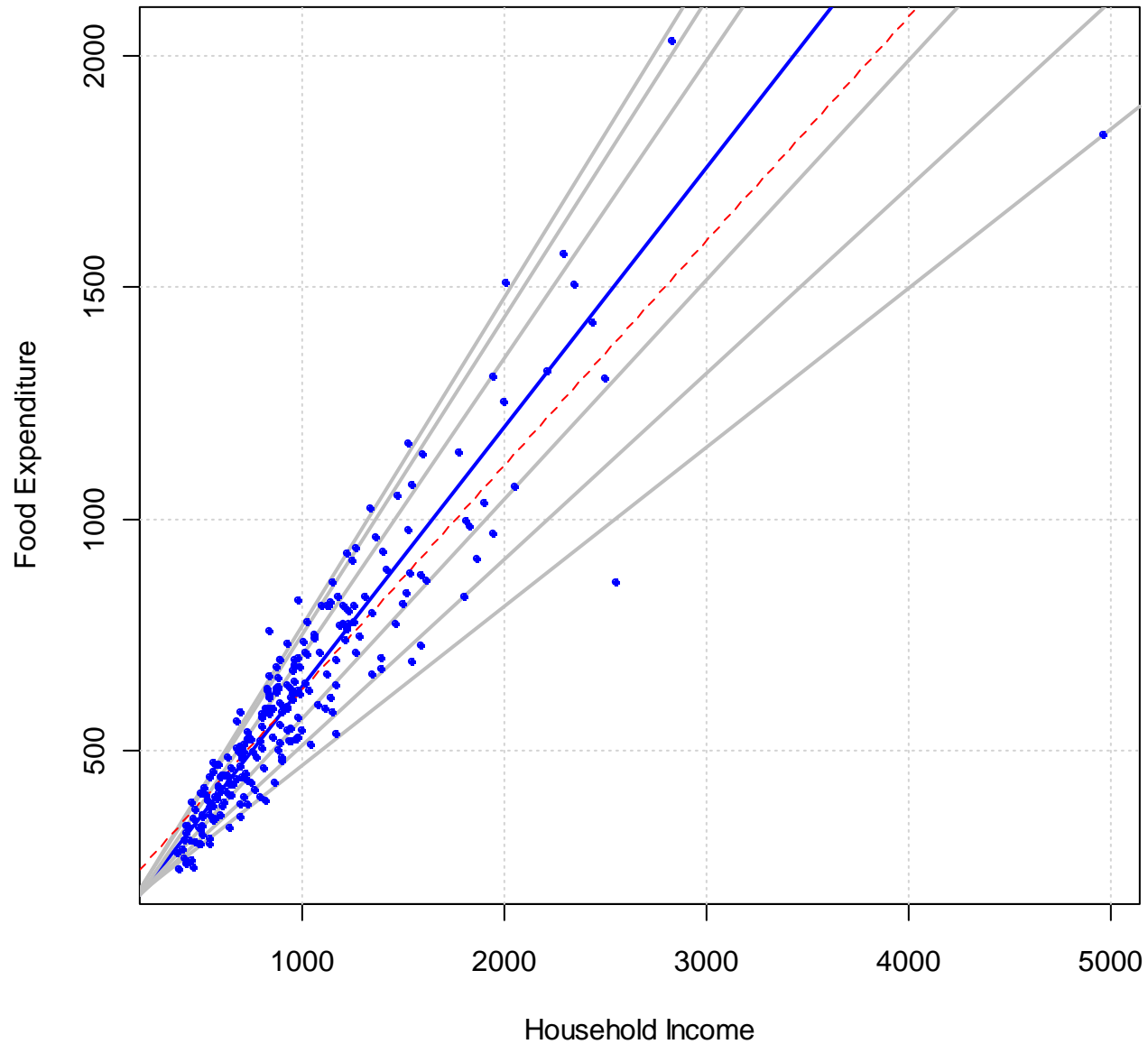
Least Squares Estimate



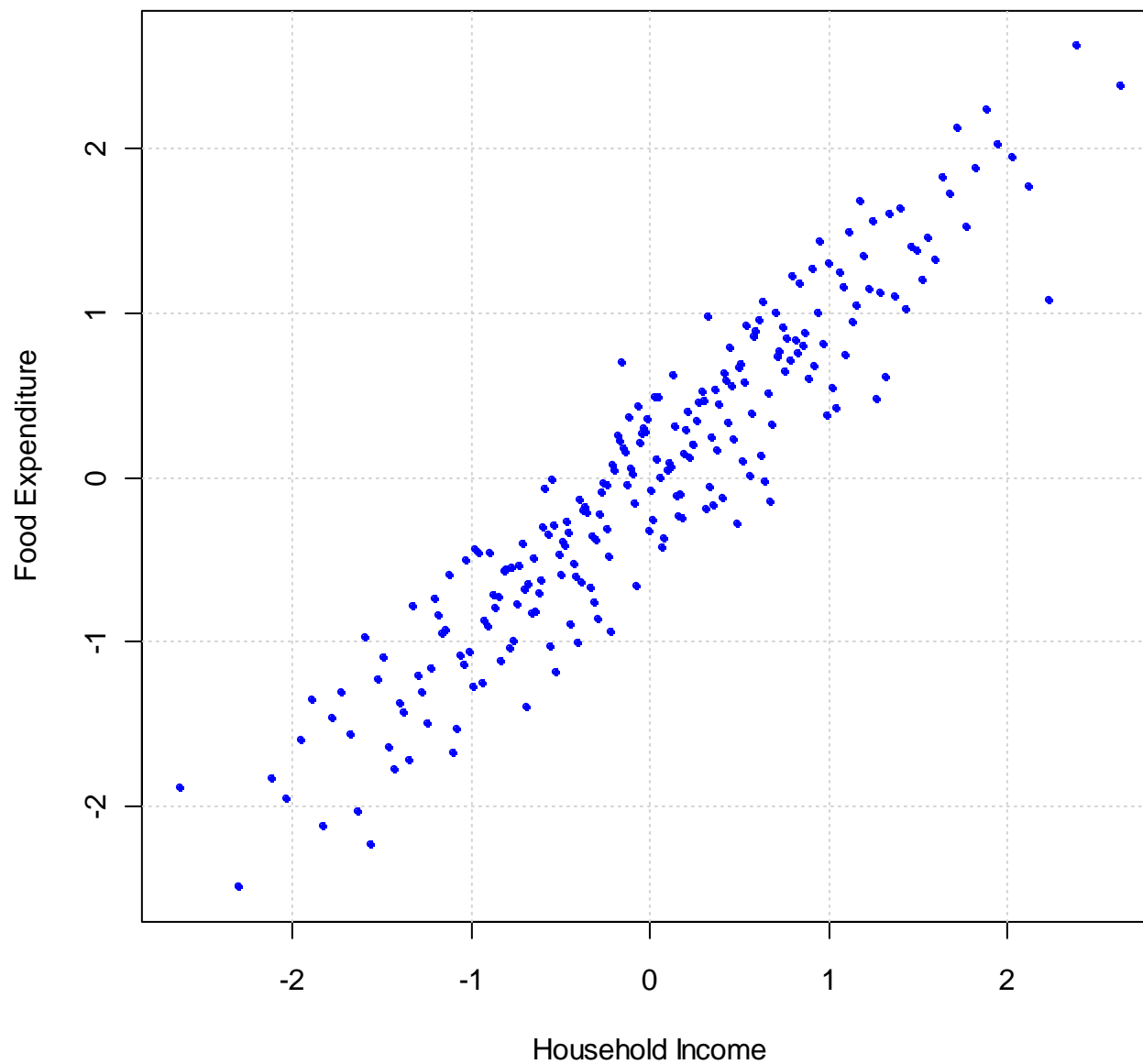
Quantile Regression - median



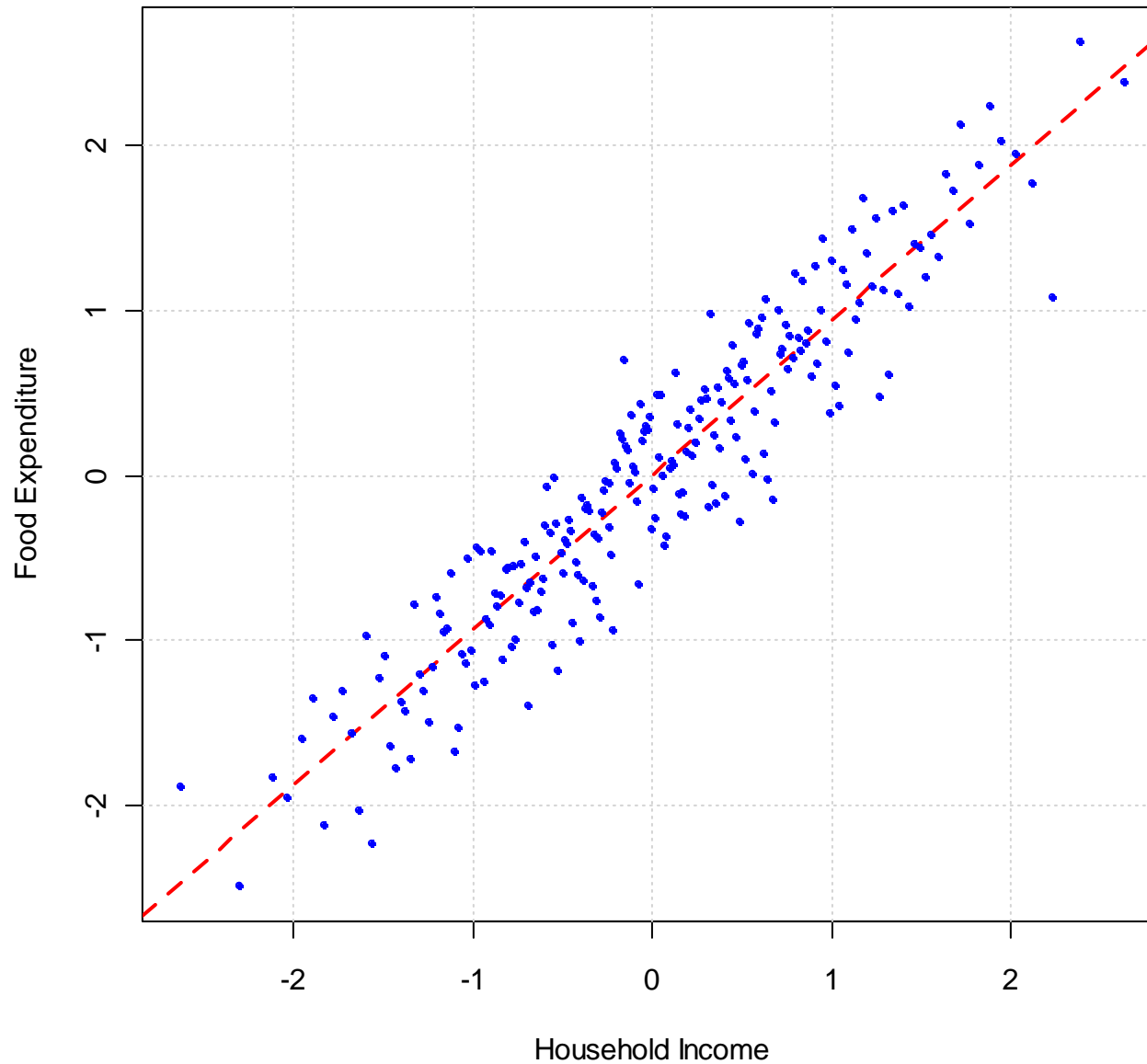
Quantile Regression, multiple quantiles



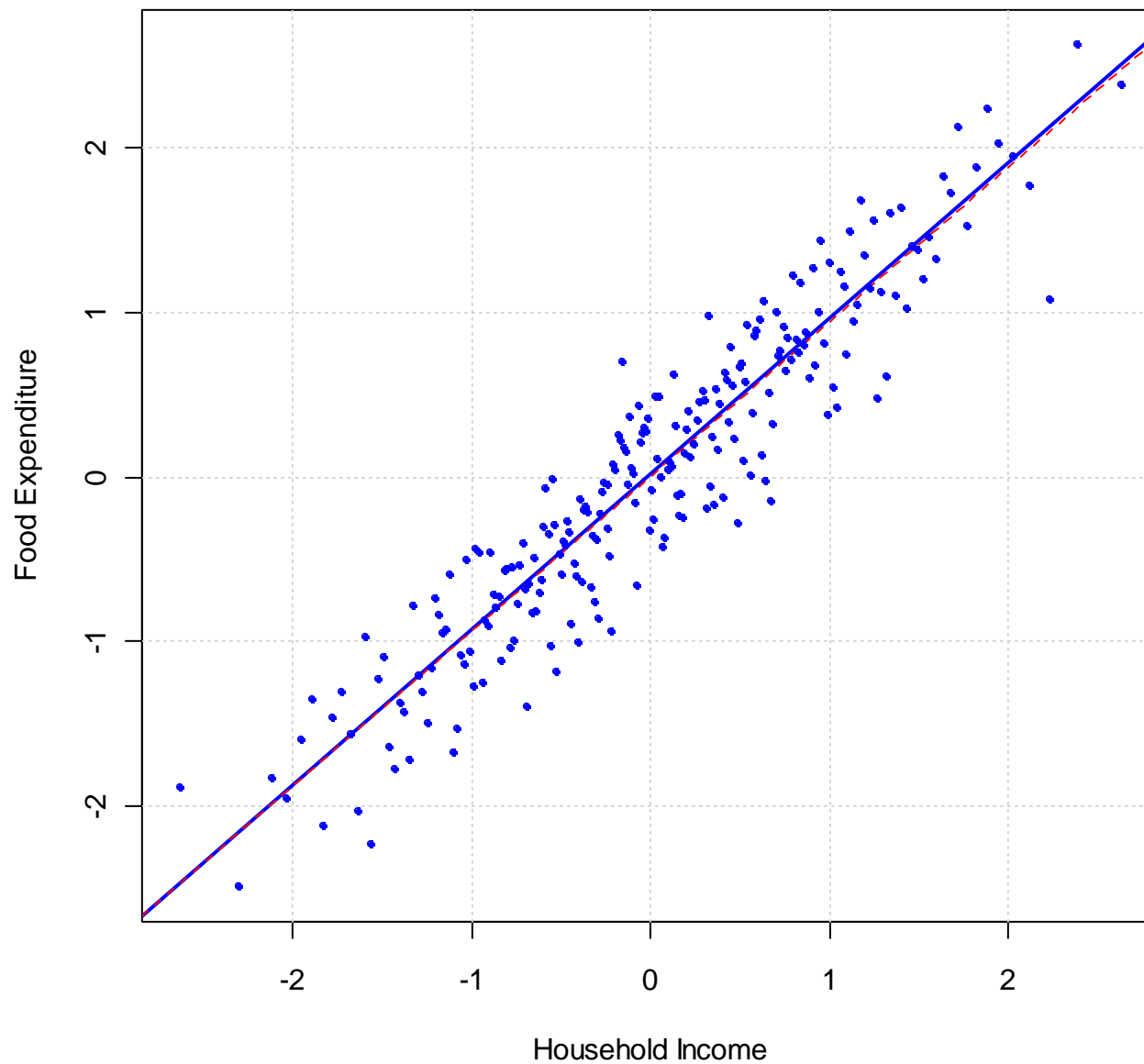
Engel data, 19th century, Belgium



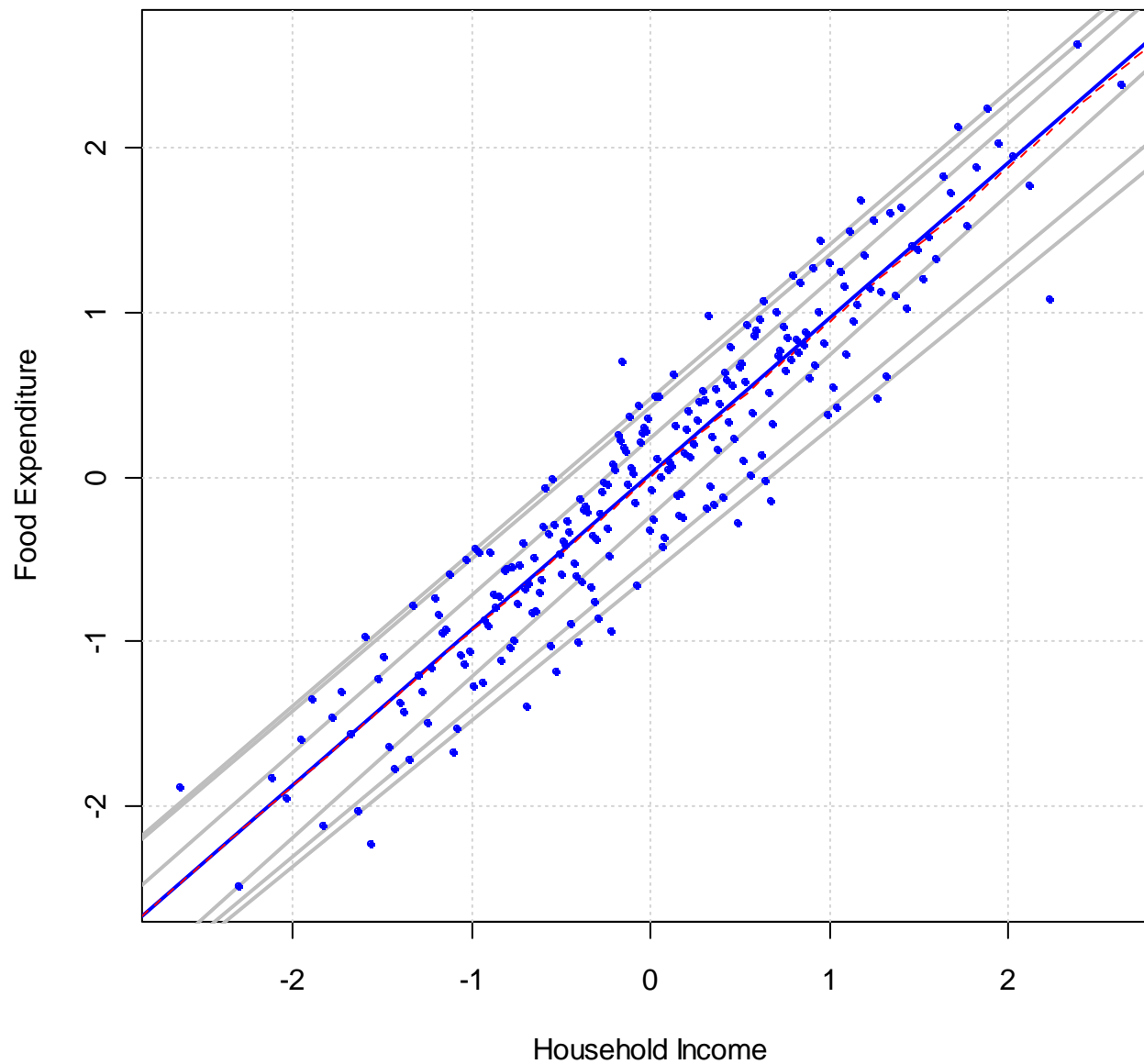
Least Squares Estimate

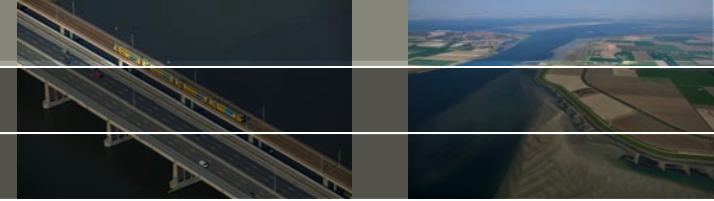


Quantile Regression - median



Quantile Regression, multiple quantiles

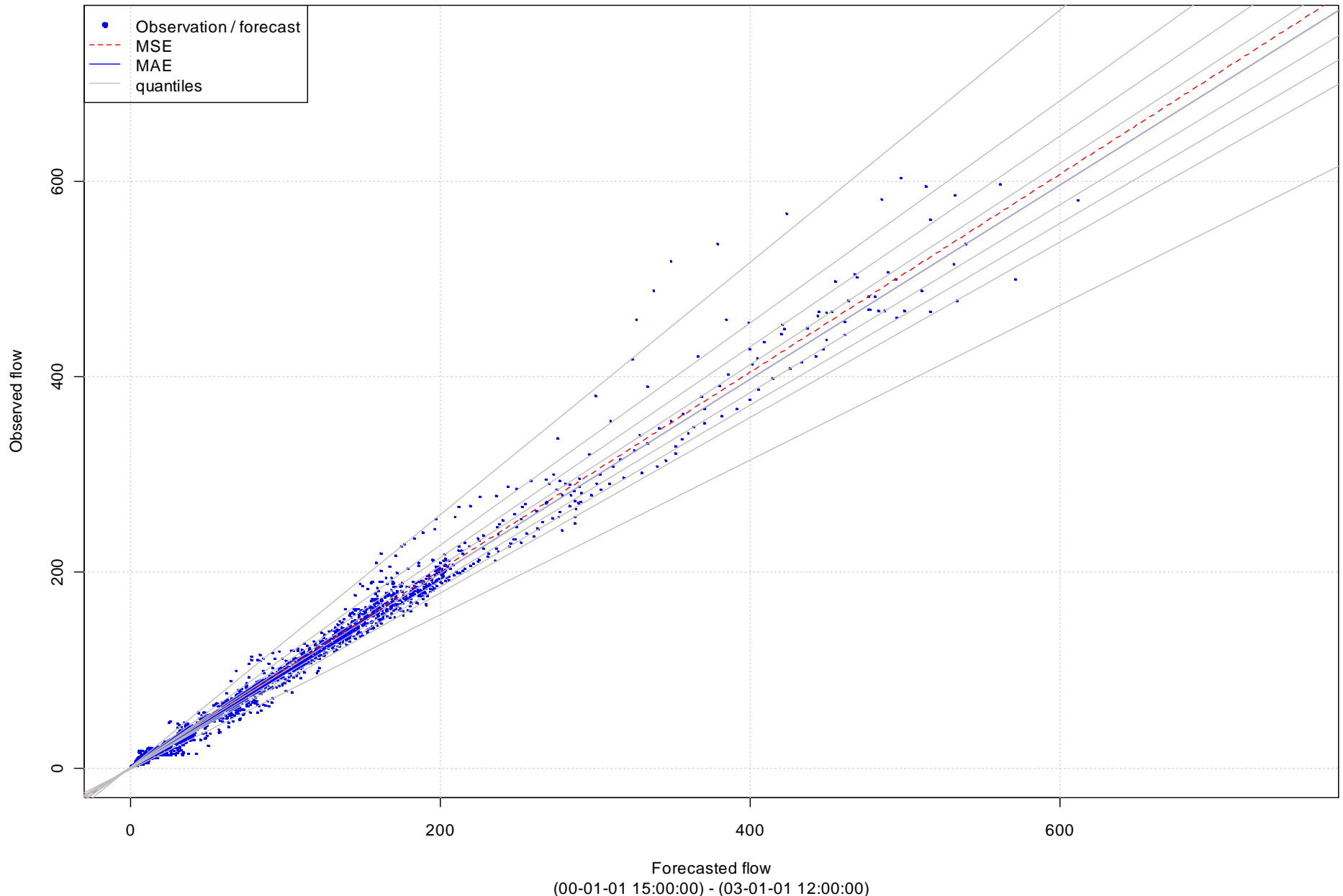




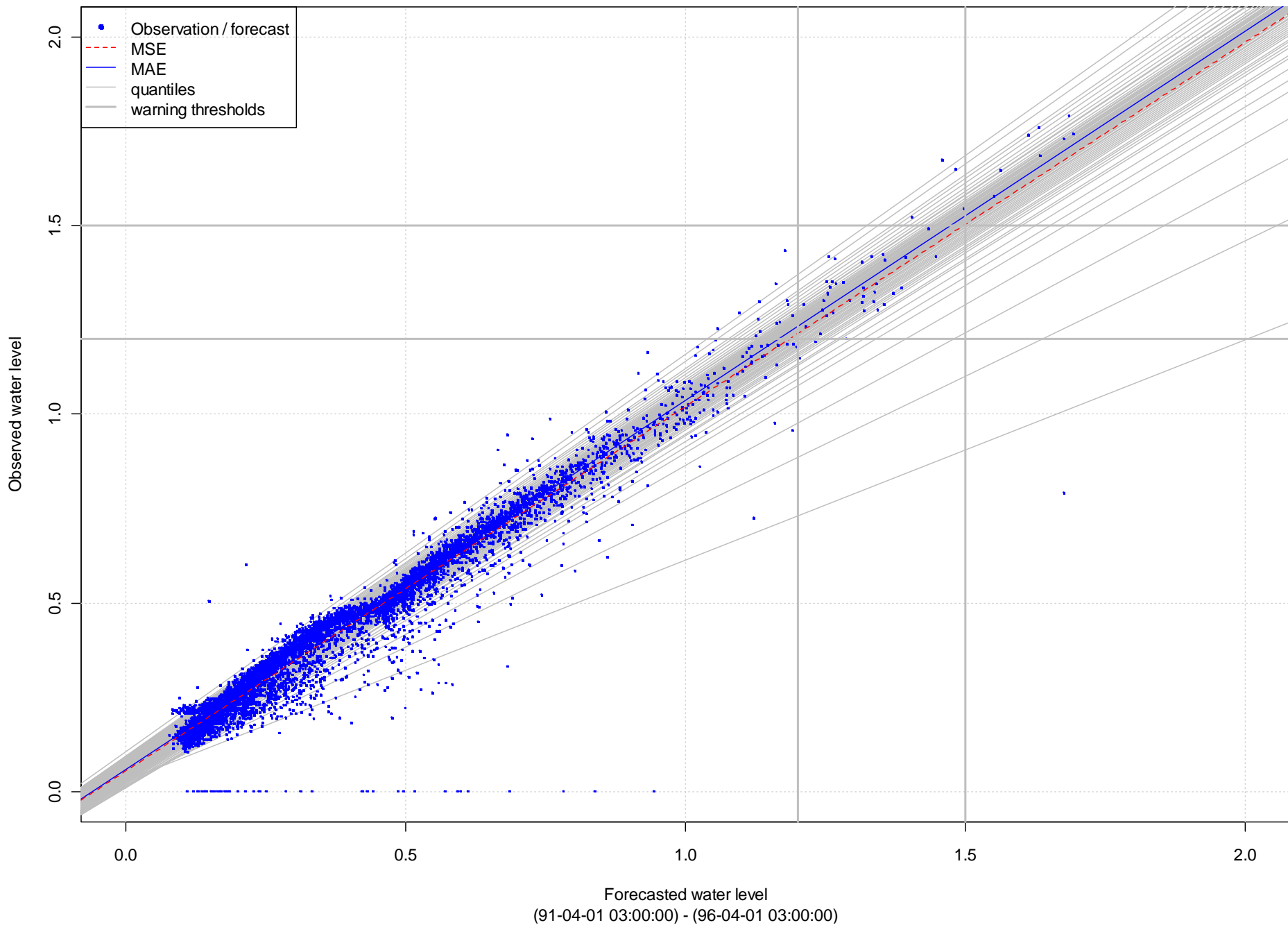
Estimating predictive hydrological uncertainty

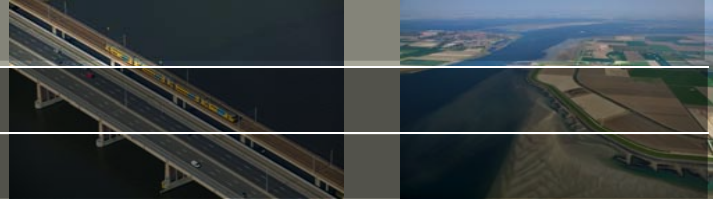
- Fit QR-models where observation is ‘a function of’ a forecast
- Alternative model: *forecast error* as ‘a function of’ a forecast
- Other predictors may be added to further constrain uncertainties
- QR-models can be polynomials, don’t have to be lines
- Derive QR-models on calibration set, then hope that the error structure will remain unchanged

Observation v. forecast, flow @ Wangaratta, lead time = 012h



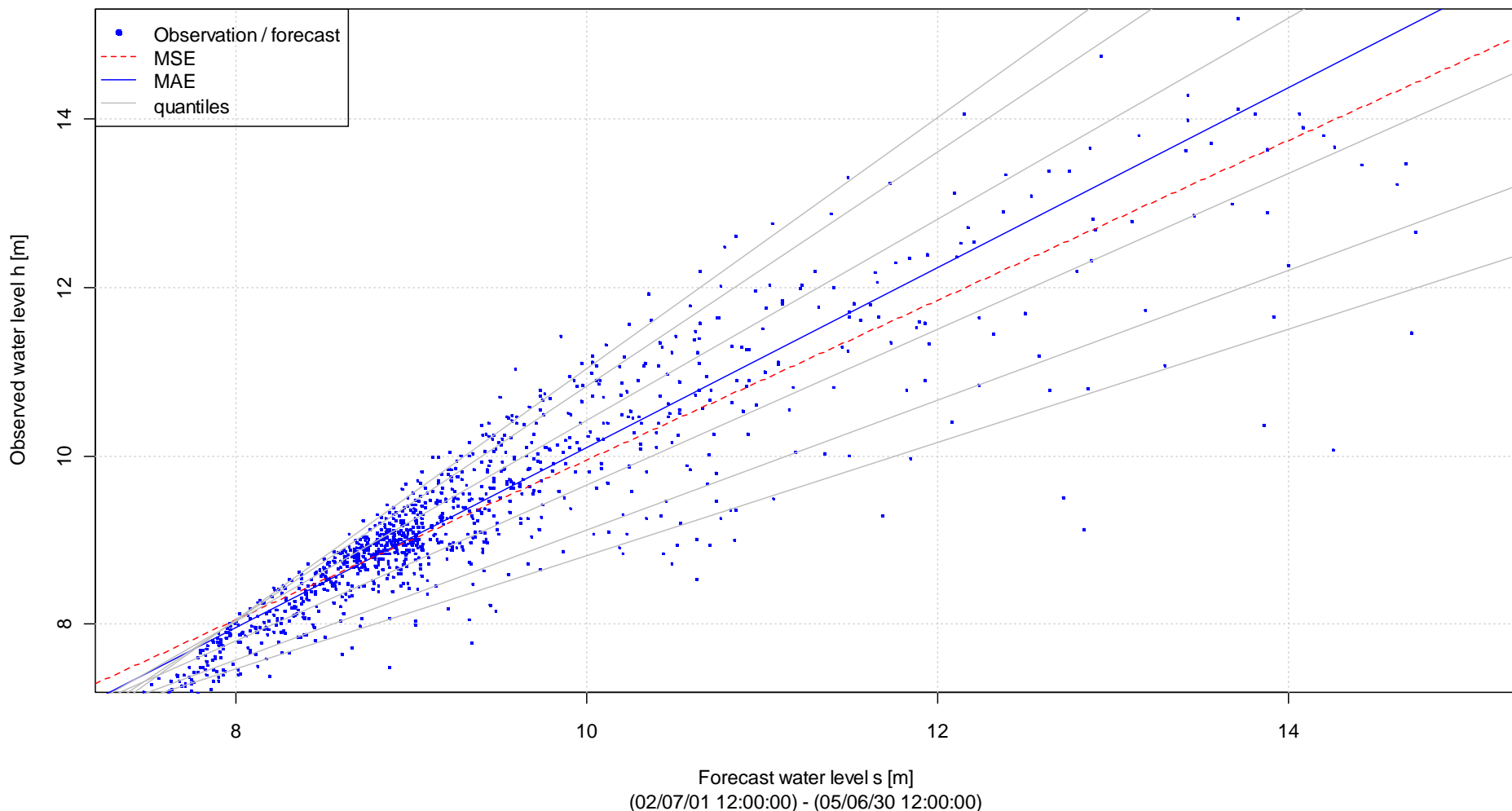
Observation v. forecast, water level @ Overlee, lead time = 003h
NB: QR was derived for forecasts >= .45m only!

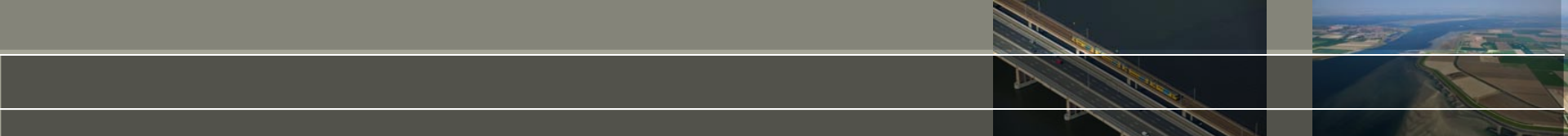




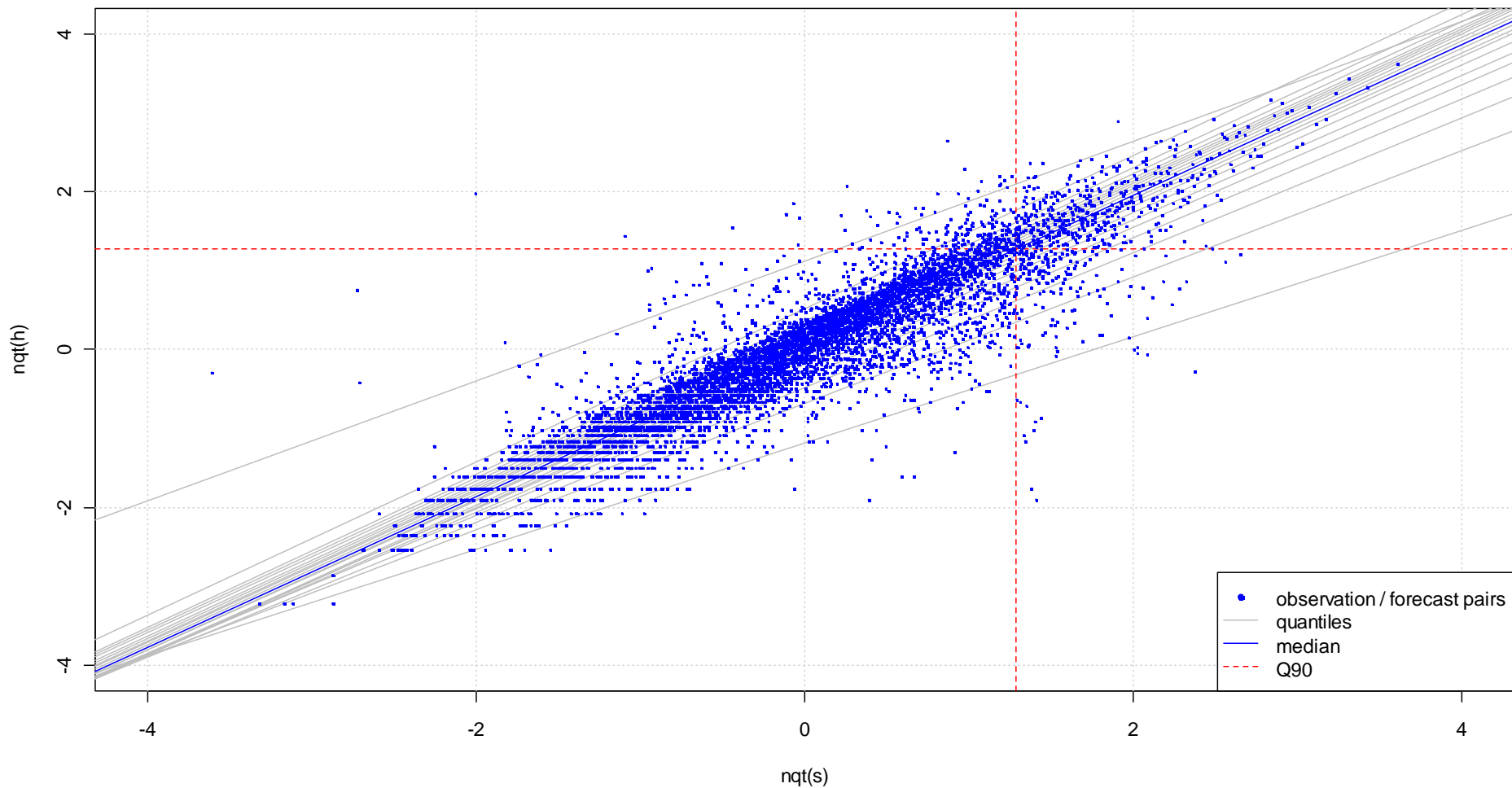
Watch out for quantile crossing

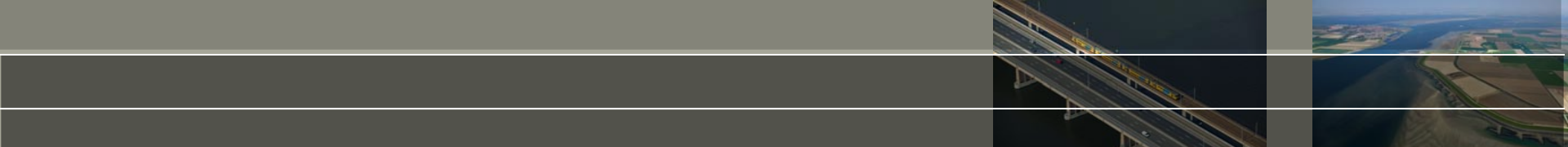
Water level observation v. forecast, Rhine @ Lobith, lead time = 72h



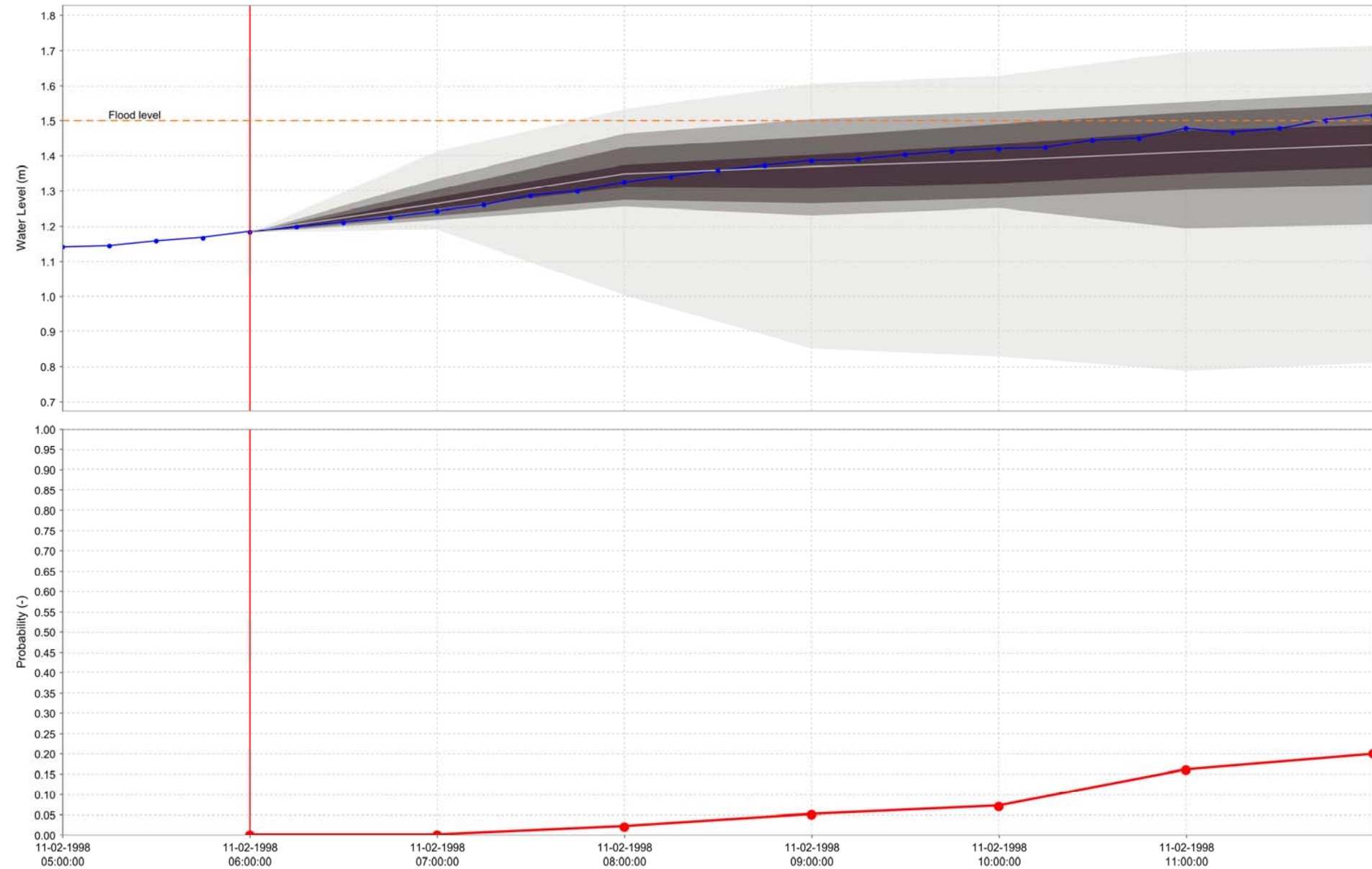


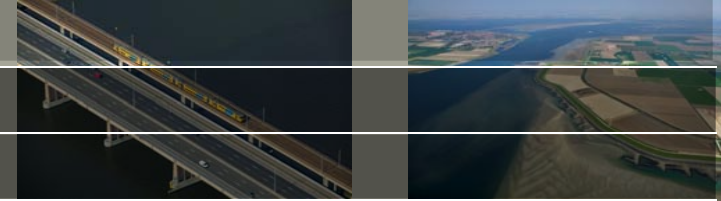
nqt(h) v. nqt(s)
01643000_sac_ARMA02, 02d, n=6561 range 01/03/62 - 01/02/80





Overlee

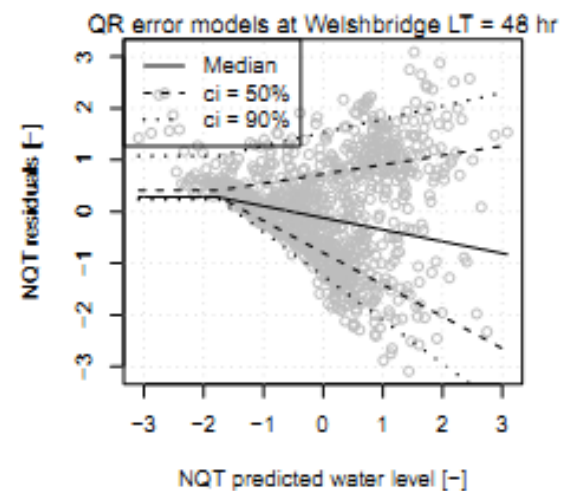
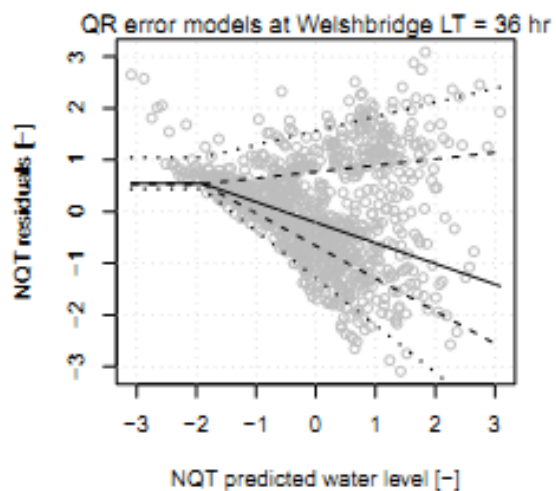
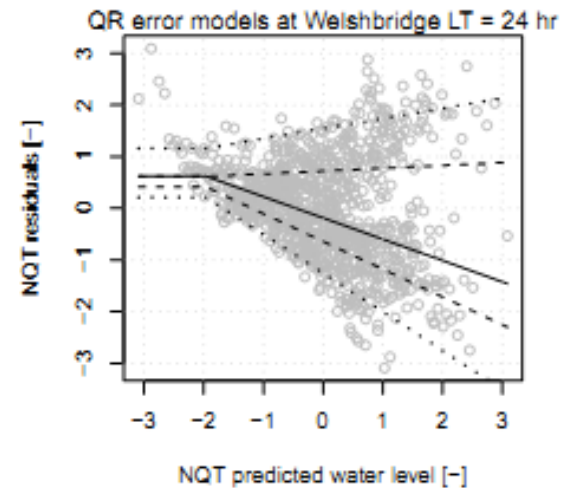
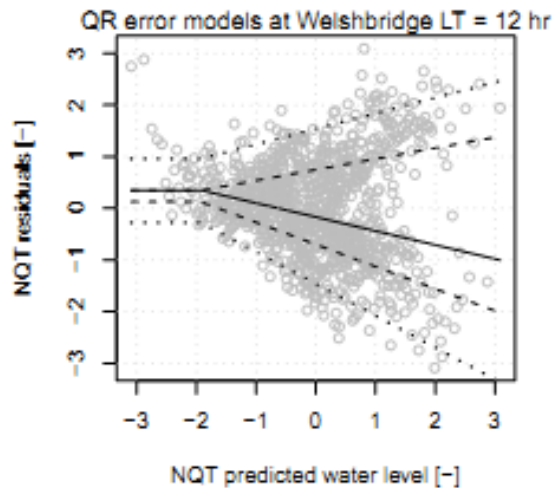


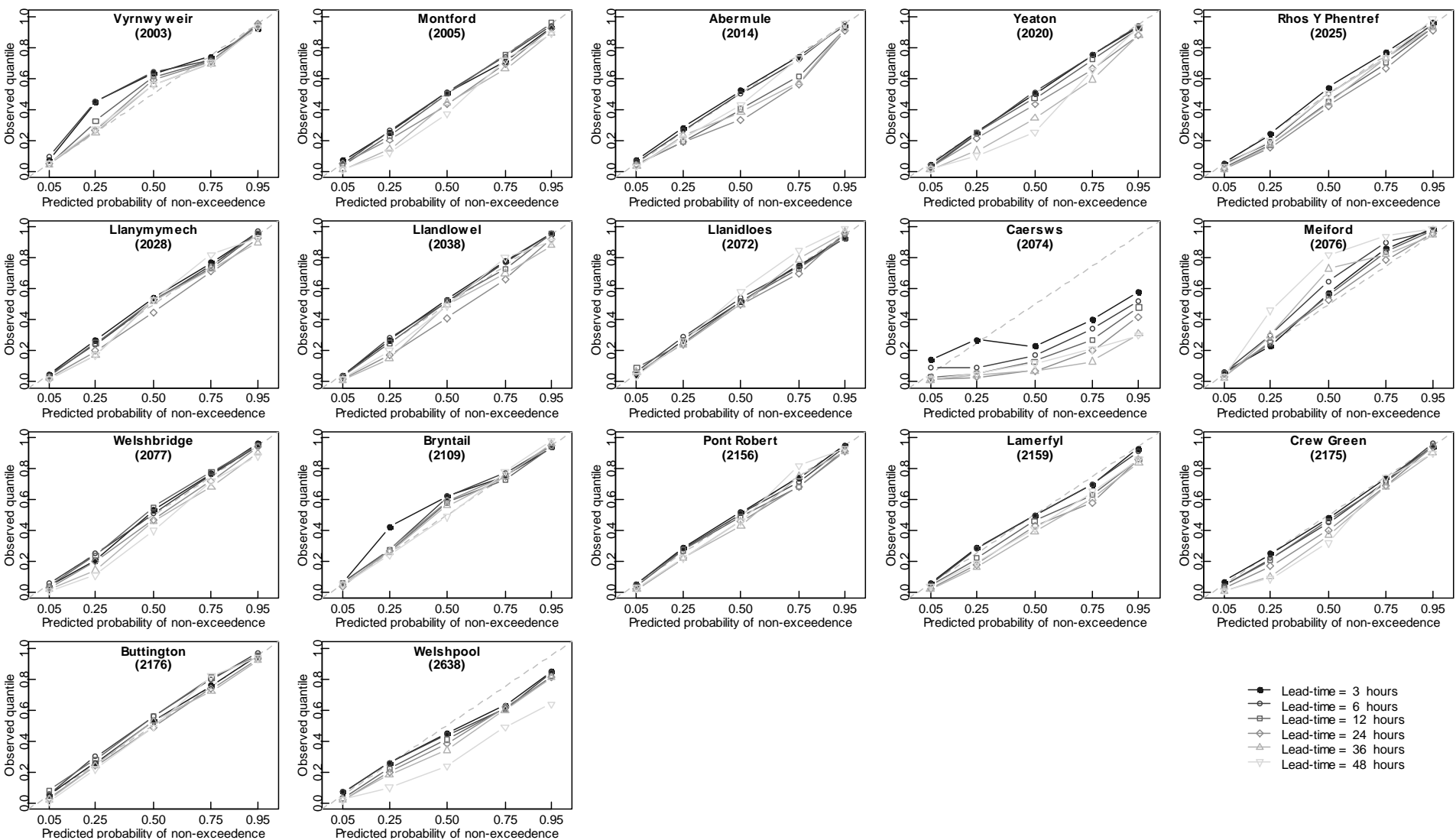


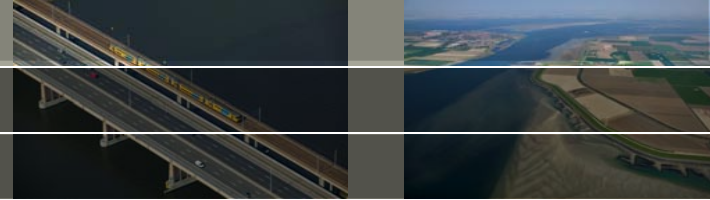
Applications this far

- ~25 NFFS basins (England and Wales)
- White Cart (Scotland)
- Ovens River (Australia)
- HEPEX-basins (USA) (in progress)

NFFS basins: Welsbridge

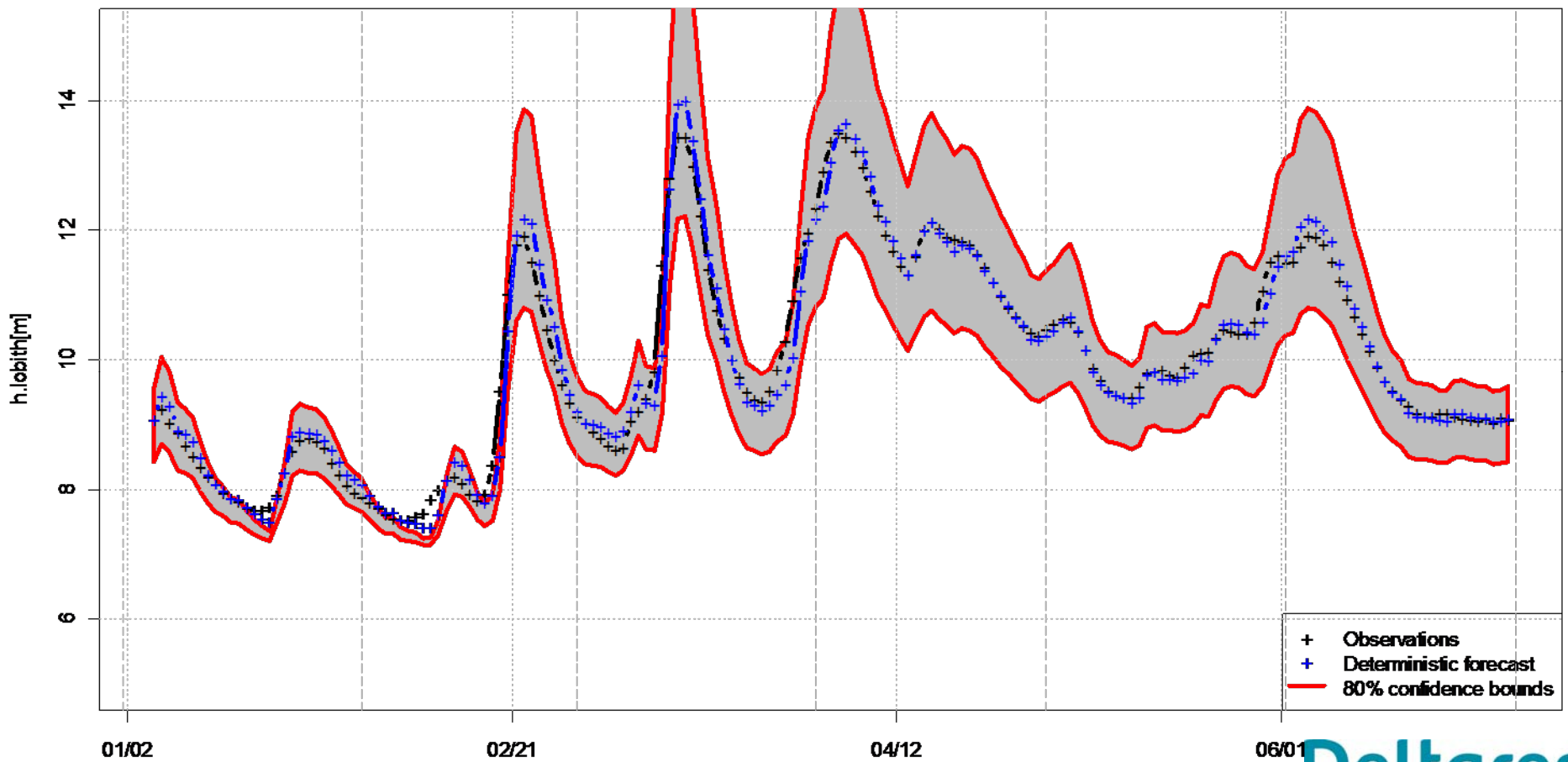






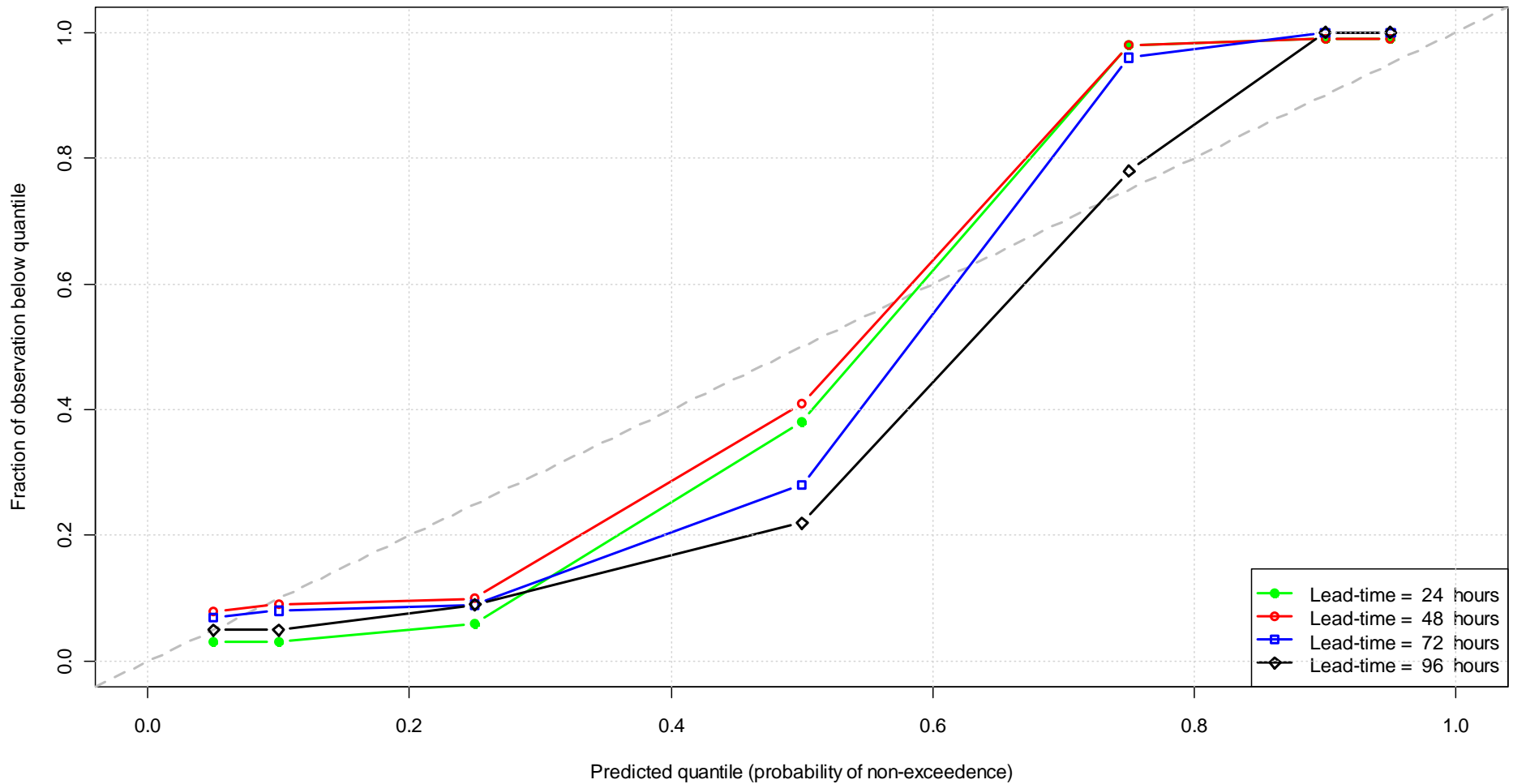
Application: Lobith forecasting system

QR 80% Confidence intervals, Lead time:3day(s)



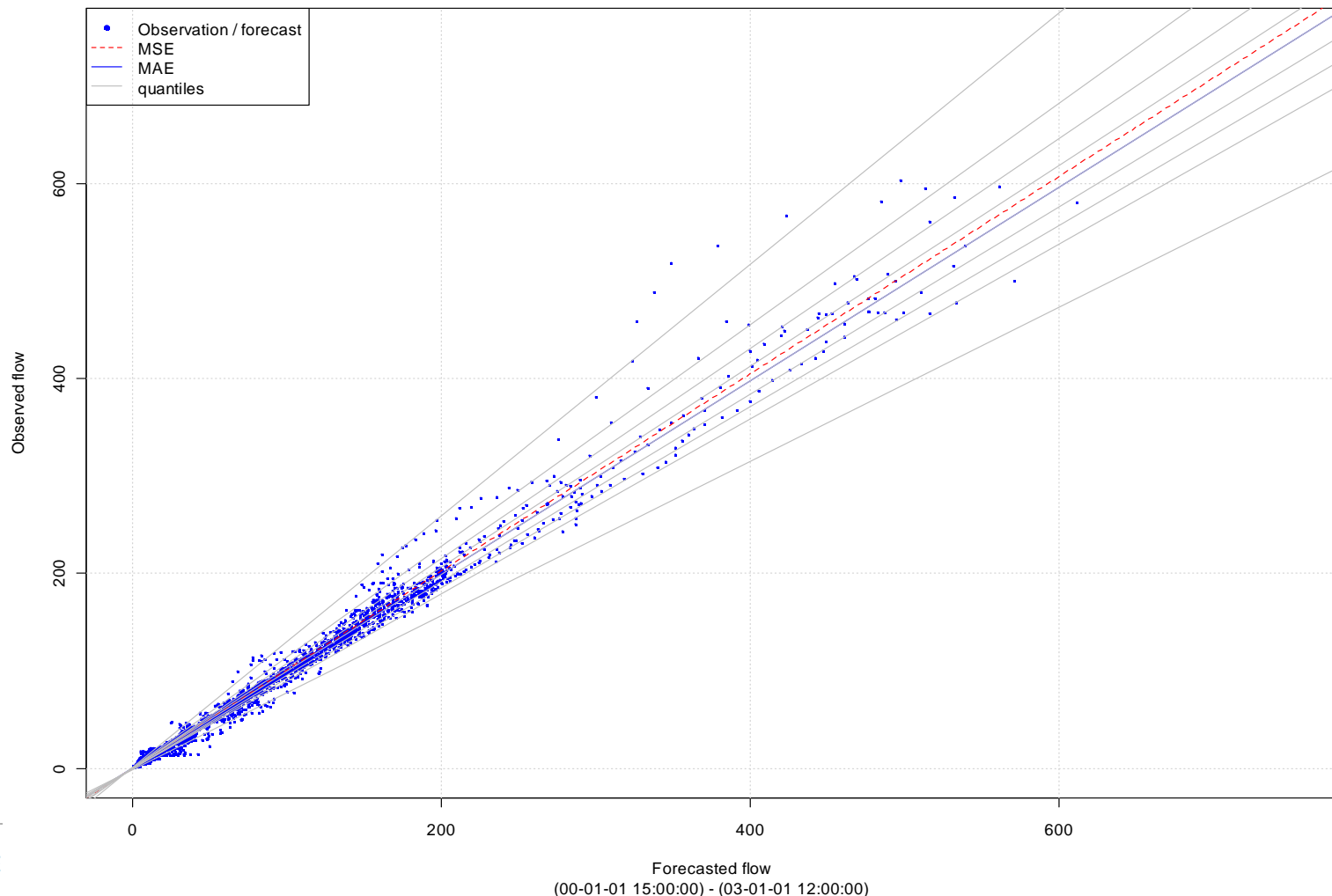


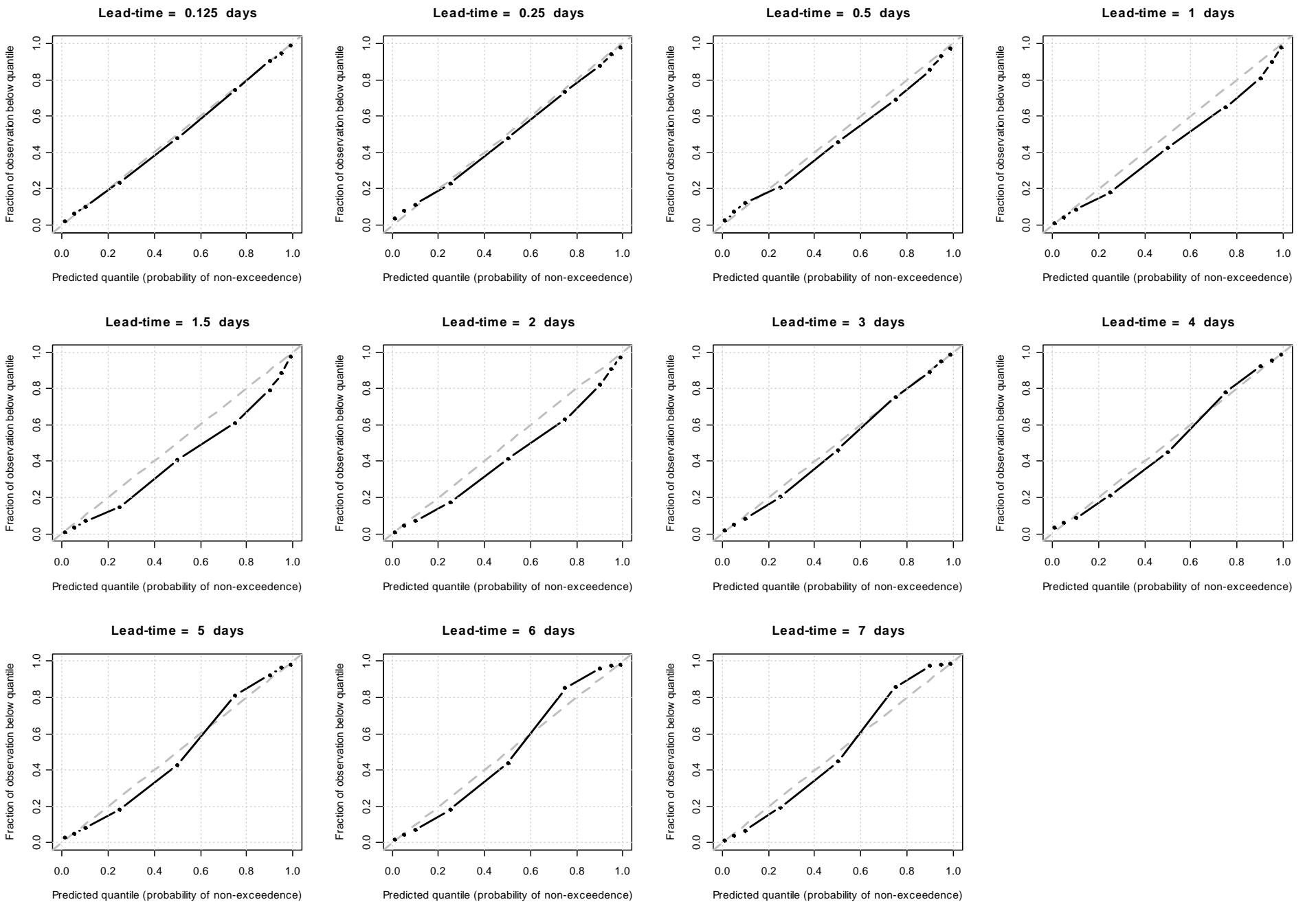
QR Lobith, Quantile plot



Ovens River @ Wangaratta

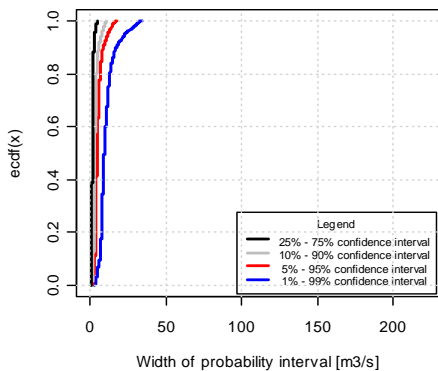
Observation v. forecast, flow @ Wangaratta, lead time = 012h



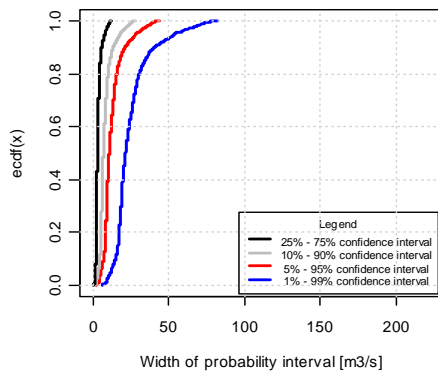




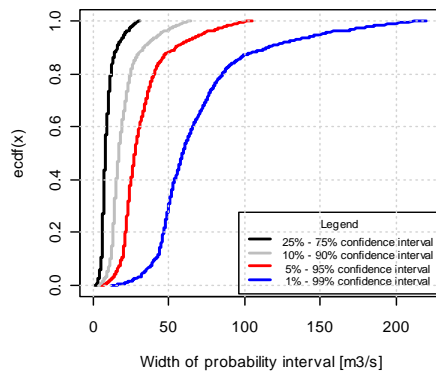
Lead time 0.125 days



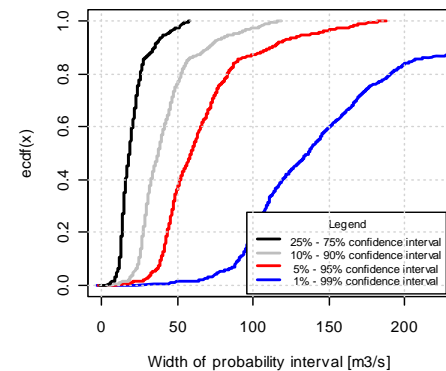
Lead time 0.25 days



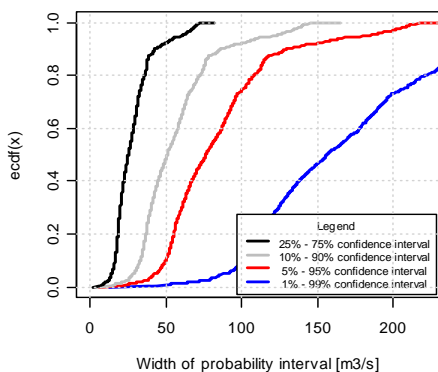
Lead time 0.5 days



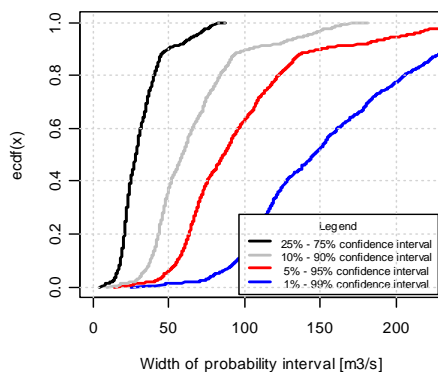
Lead time 1 days



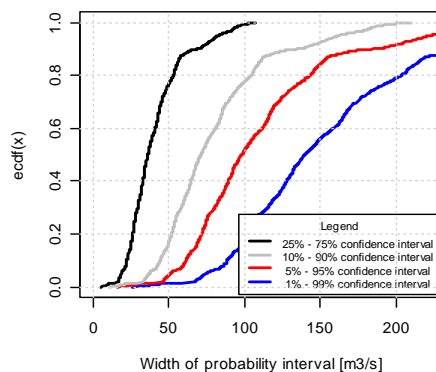
Lead time 1.5 days



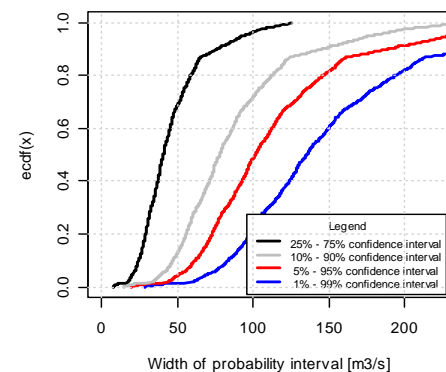
Lead time 2 days



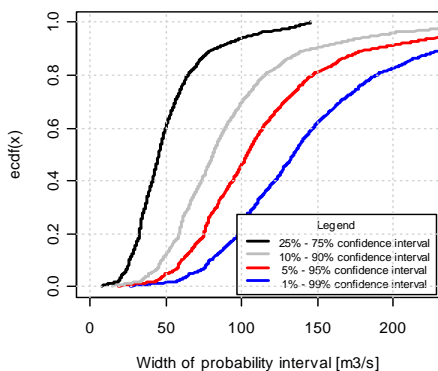
Lead time 3 days



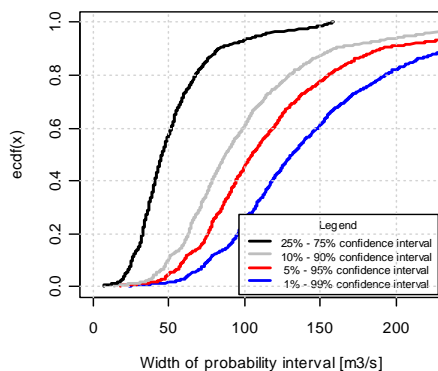
Lead time 4 days



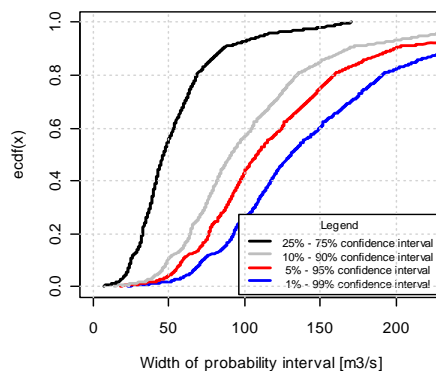
Lead time 5 days

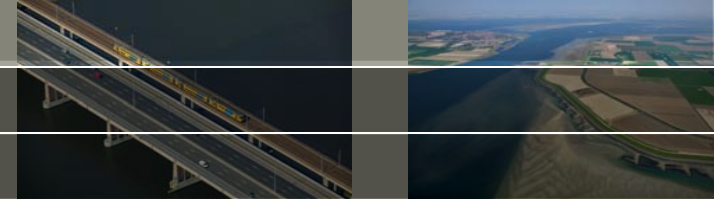


Lead time 6 days



Lead time 7 days

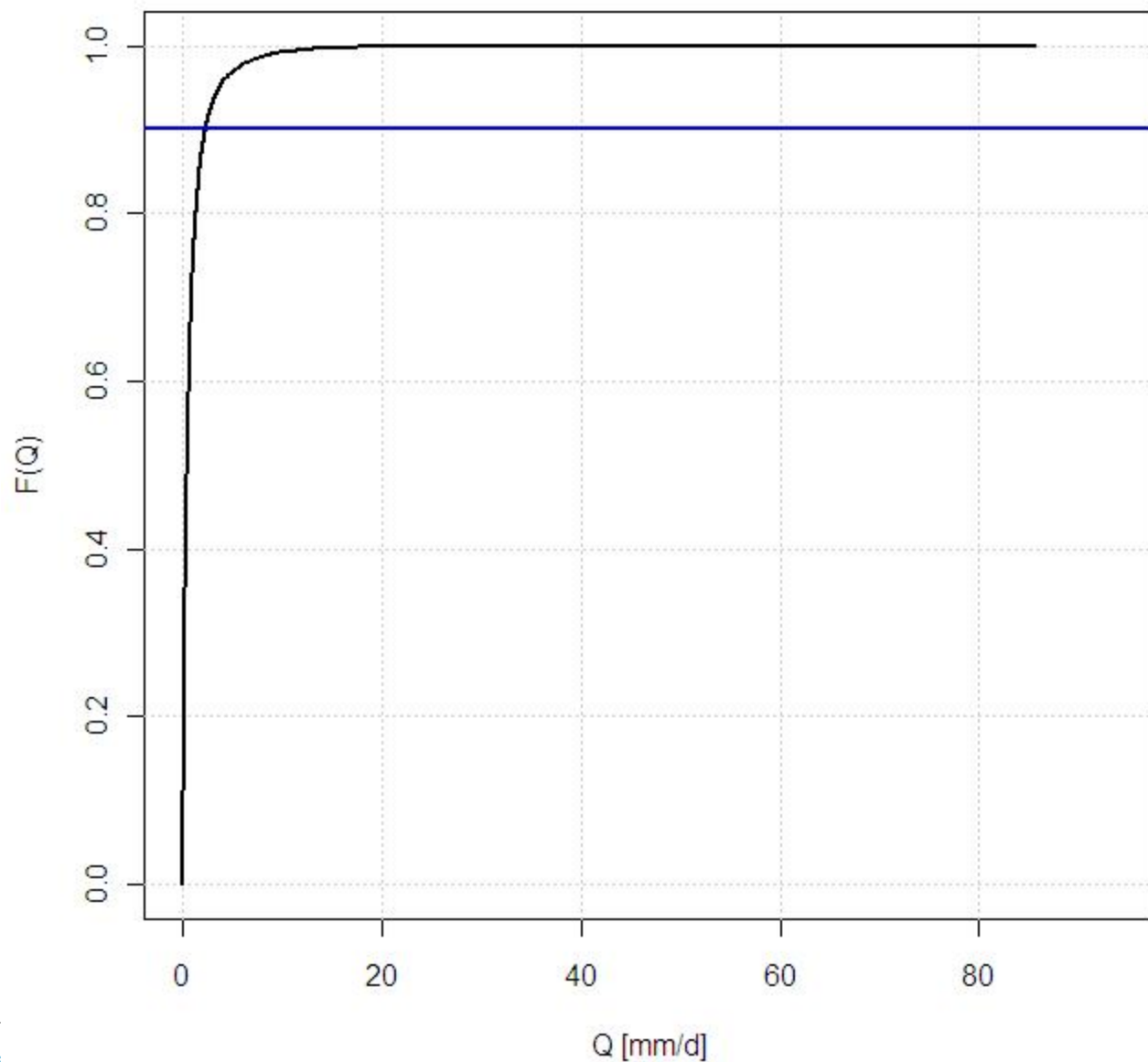


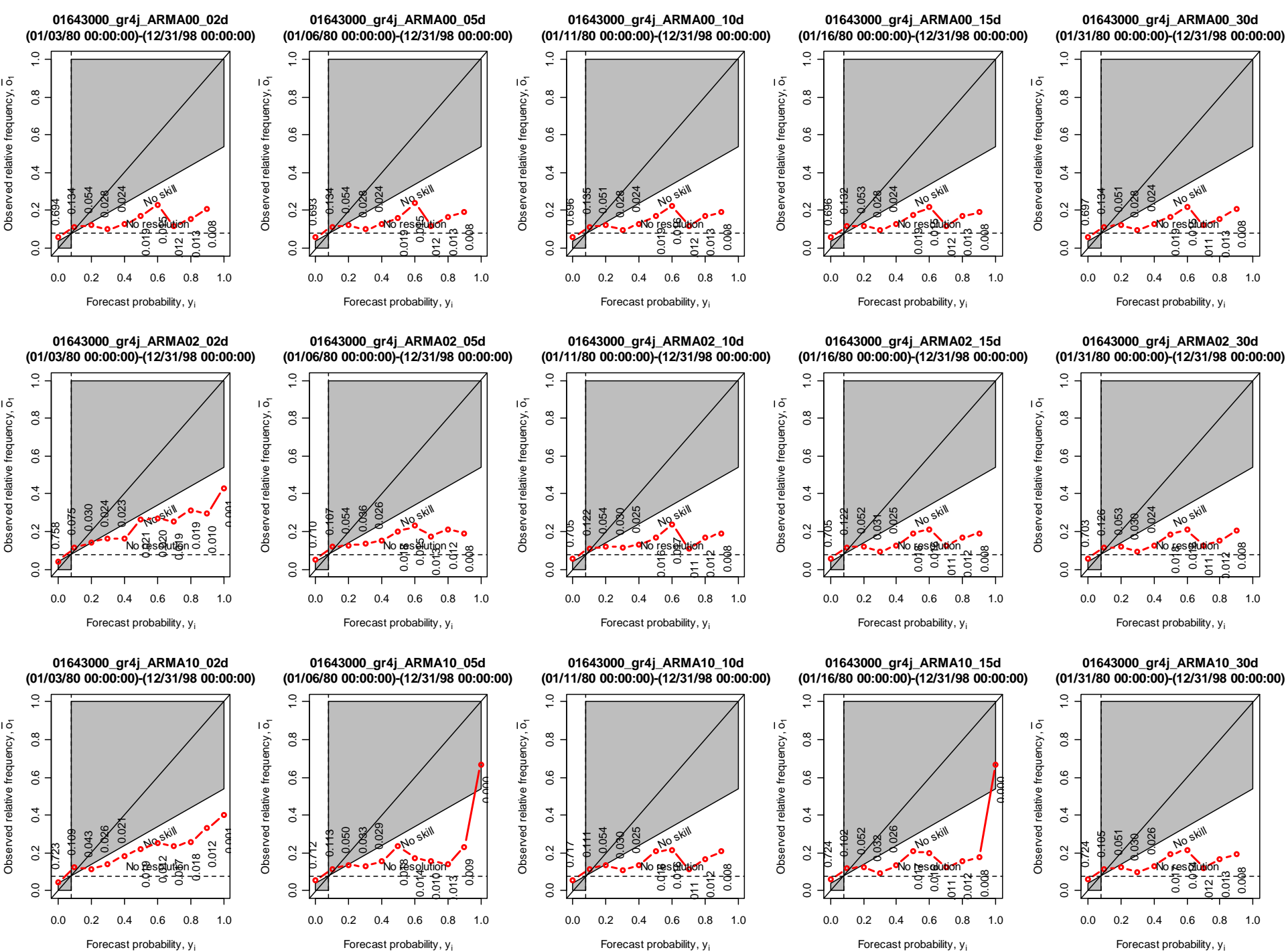


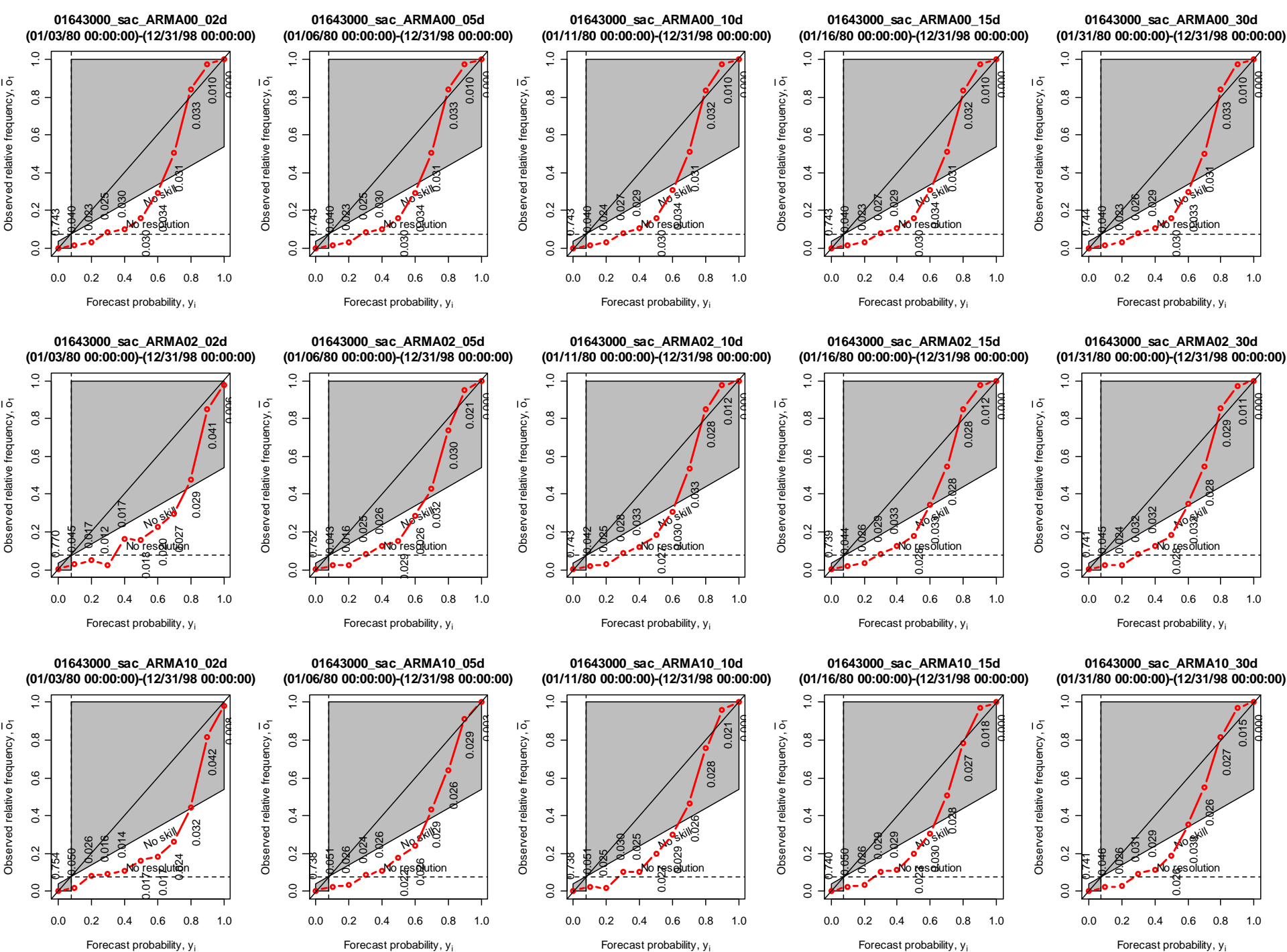
HEPEX scenario 1: hydrol. model simulations

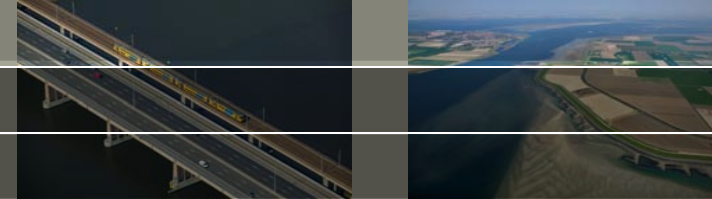
- Monocacy River @ Frederick (0164 3000)
 - daily flow, not accumulated
 - (calibrated) sac and gr4j models
 - lead times considered: 2, 5, 10, 15 and 30 days
 - 3 alternative error correction options:
 - 1. no correction: simulation = forecast
 - 2. ARMA-correction, fixed parameters
 - 3. ARMA-correction, dynamically parameterised on past 10d
- 1 location x 2 models x 3 error corr'n options x 5 leadtimes = 30 cases to be evaluated
- so far, only evaluation is that of probability of high flow

ECDF, 01643000, 1961-1978
(flooding threshold @ 2.37 mm/d)



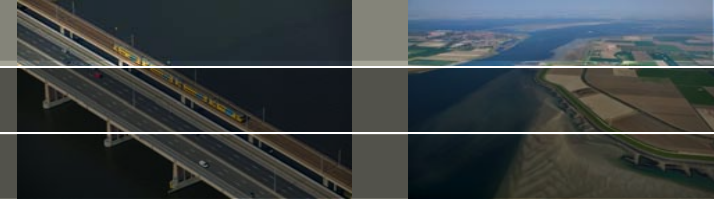






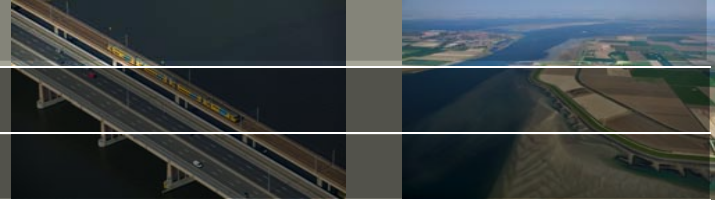
locId	1643000
-------	---------

			Data					
model	arma	leadtime	bs	bs.baseline	ss	bs.uncert	bs.resol	bs.reliability
sac	ARMA00	2	0.04	0.07	0.38	0.07	0.04	0.01
		5	0.04	0.07	0.38	0.07	0.04	0.01
		10	0.04	0.07	0.38	0.07	0.04	0.01
		15	0.04	0.07	0.39	0.07	0.04	0.01
		30	0.04	0.07	0.39	0.07	0.04	0.01
	ARMA02	2	0.04	0.07	0.36	0.07	0.04	0.01
		5	0.04	0.07	0.38	0.07	0.04	0.01
		10	0.04	0.07	0.4	0.07	0.04	0.01
		15	0.04	0.07	0.41	0.07	0.04	0.01
		30	0.04	0.07	0.41	0.07	0.04	0.01
	ARMA10	2	0.05	0.07	0.33	0.07	0.04	0.02
		5	0.04	0.07	0.38	0.07	0.04	0.01
		10	0.04	0.07	0.41	0.07	0.04	0.01
		15	0.04	0.07	0.41	0.07	0.04	0.01
		30	0.04	0.07	0.42	0.07	0.04	0.01



Next steps in HEPEX experiment

- Thoroughly check intermediate results
- Drop gr4j dataset in favour of vic model simulations
- Try constrain uncertainties by either adding predictors or by stratifying:
 - initial conditions / antecedent precipitation
 - flow at t_0
- Extend forecast evaluation
 - full distribution in addition to probability of flooding
- Hopefully compare with other post-processors



Further reading

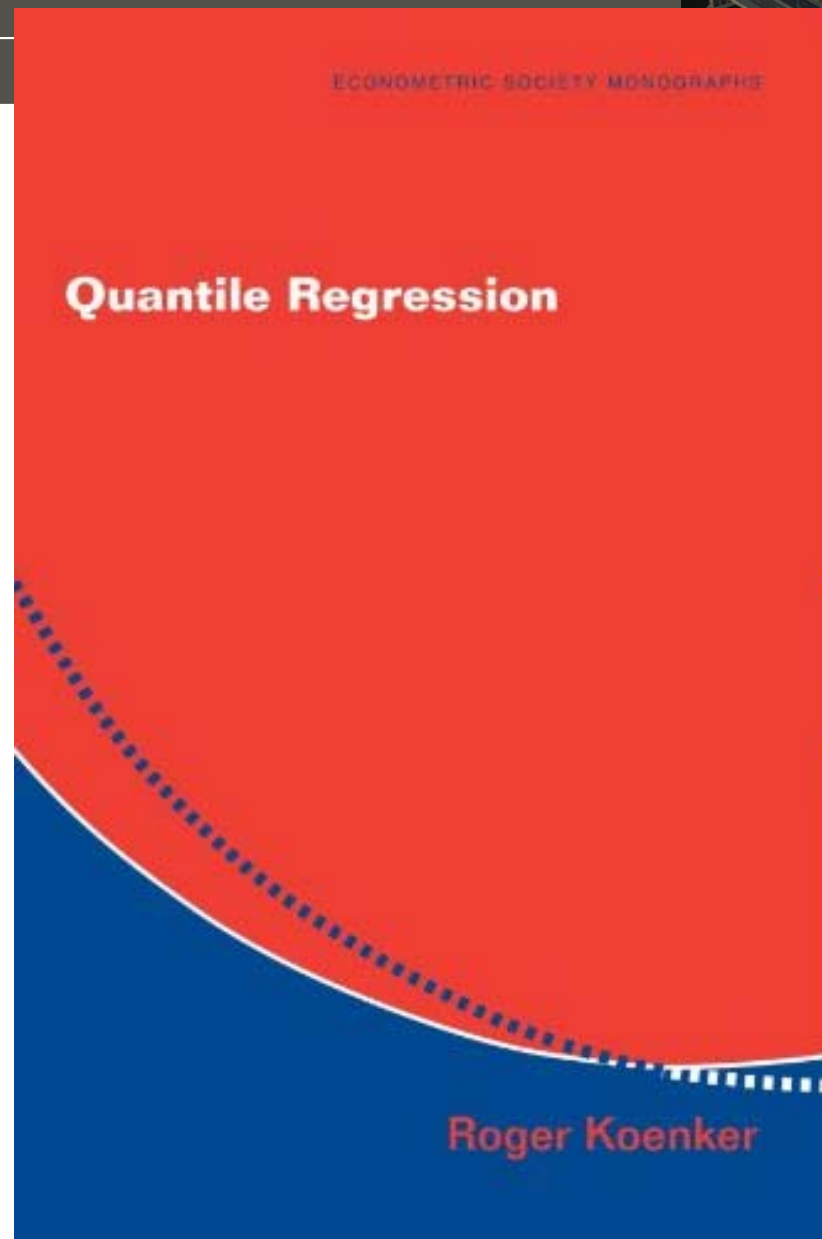
QUANTILE REGRESSION IN R: A VIGNETTE

ROGER KOENKER

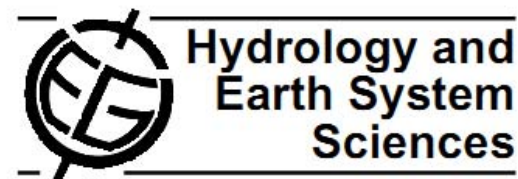
ABSTRACT. Quantile regression is an evolving body of statistical methods for estimating and drawing inferences about conditional quantile functions. An implementation of these methods in the R language is available in the package `quantreg`. This vignette offers a brief tutorial introduction to the package. R and the package `quantreg` are open-source software projects and can be freely downloaded from CRAN: <http://cran.r-project.org>.

1. INTRODUCTION

Beran's (2003) provocative definition of statistics as “the study of algorithms for



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www.hydrol-earth-syst-sci.net/15/255/2011/
doi:10.5194/hess-15-255-2011
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Estimation of predictive hydrological uncertainty using quantile regression: examples from the National Flood Forecasting System (England and Wales)

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Questions?

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