



Identification of uncertainty sources in flood forecasting

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Outline

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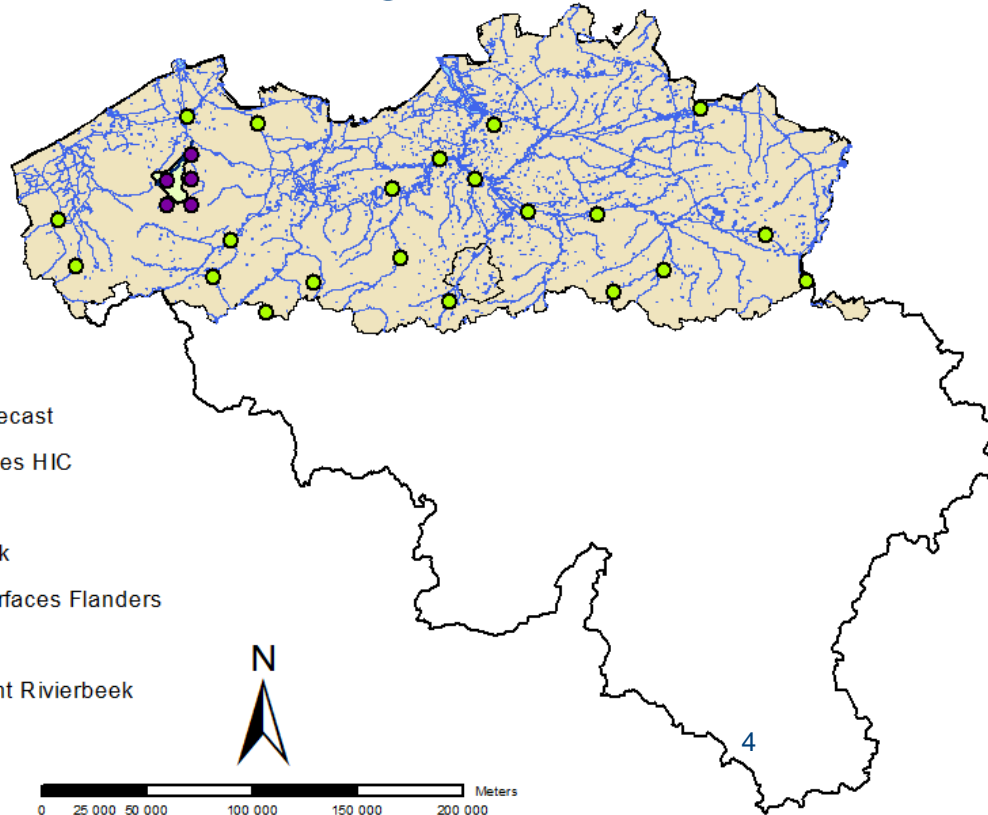
Introduction

- Flood forecasts are subject to different sources of uncertainty.
- Identification of the most important sources of uncertainty is essential in improving the flood forecasting system in an efficient way.
- Decomposition of total forecast uncertainty in its key sources by variance decomposition.



Case study

- Catchment: Rivierbeek (western part of Belgium)
 - 64 km²
 - Av. discharge: 0.6m³/s; Max. discharge: >8m³/s



Legend

- Alaro Forecast
- Raingauges HIC

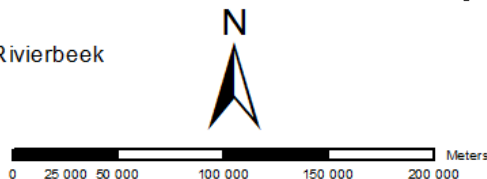
— Rivierbeek

Water Surfaces Flanders

Catchment Rivierbeek

Flanders

Belgium



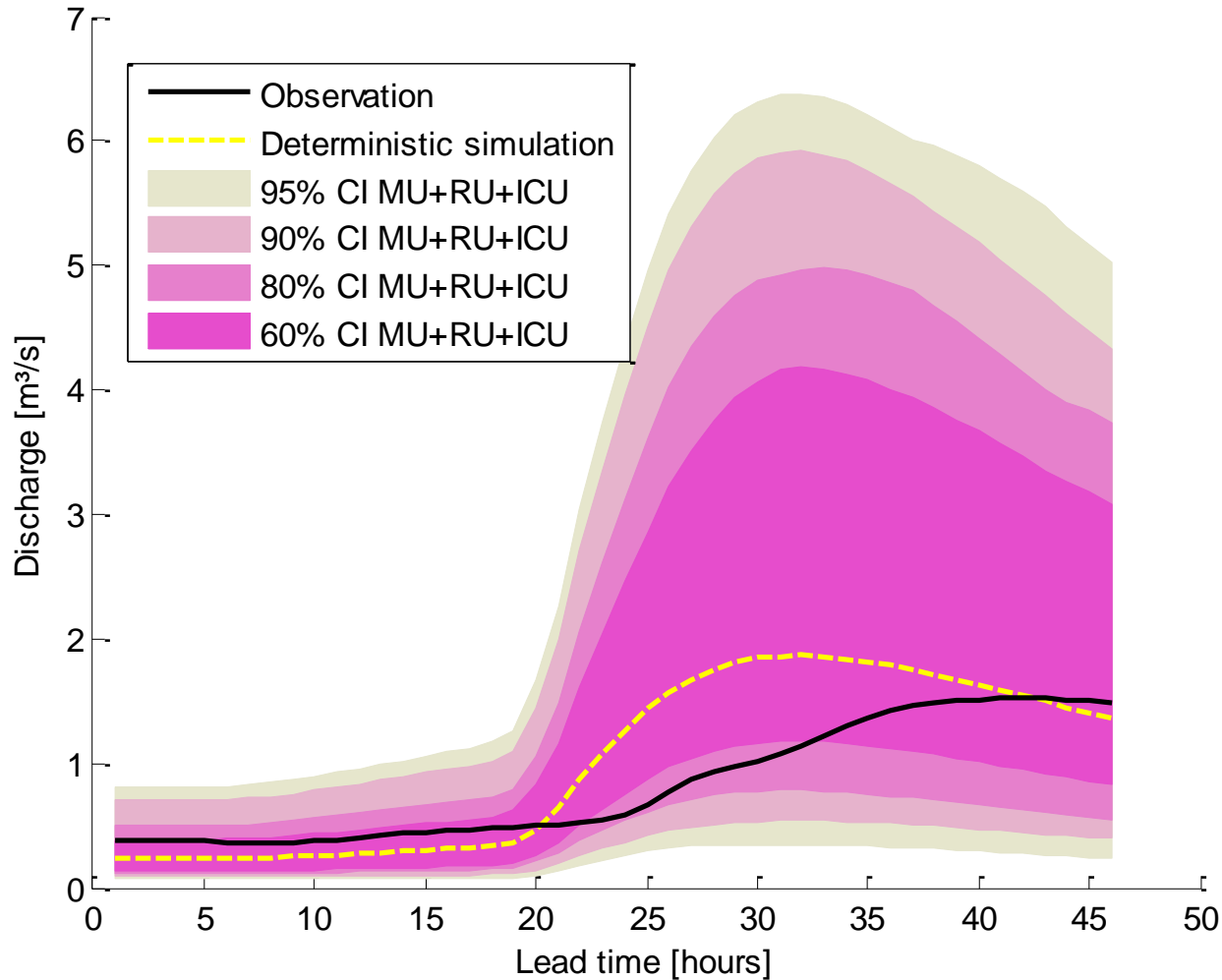
Case study

- Flood forecasting system:
 - Operational system of Flanders Hydraulics Research
 - Navigable waterways in Flanders
 - Lead time of 48h
 - Floodwatch system of DHI (incl. Mike11 & NAM)
 - Rainfall forecasts: NWP Alaro (7x7km grid, 3h cum.)
 - Data assimilation on Hydrodynamic output not on Hydrological output
 - In this research only the hydrological part of the FF-system is considered.

Method to assess total forecast uncertainty

- Non-parametric Data-based Approach (NDA)
 - Uncertainty analysis of historical flood forecasts
 - Calculation of residuals between observed and forecasted discharge
 - Classification of residuals according to lead time and forecasted discharge
 - For each combination of lead time class and forecasted discharge class, different percentiles are calculated and stored in 3D-error matrix
 - By interpolation in error matrix confidence intervals can be calculated
 - Residual = $(Q_{\text{obs}} - Q_{\text{for}})/Q_{\text{for}}$

Method to assess total forecast uncertainty



Variance Decomposition

- Identification of three uncertainty sources: Model Uncertainty (MU), Forecasted Rainfall Uncertainty (RU), Initial Conditions Uncertainty (ICU)
- Resimulation of historical forecasts with optimal initial conditions, based on a long term simulation run => MU + RU
- Resimulation with observed rainfall instead of rainfall forecasts => MU
- Original historical forecasts => MU + ICU + RU

Variance Decomposition

- $w_{1,i}$ = mean width of confidence interval MU+RU+ICU
- $w_{2,i}$ = mean width of confidence interval MU+RU
- $w_{3,i}$ = mean width of confidence interval MU
- i = lead time class

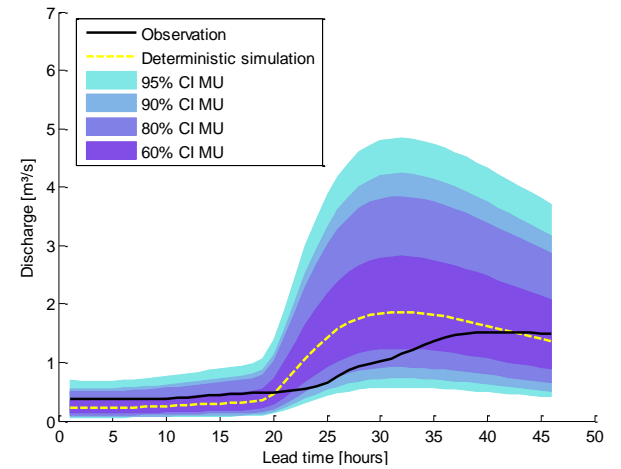
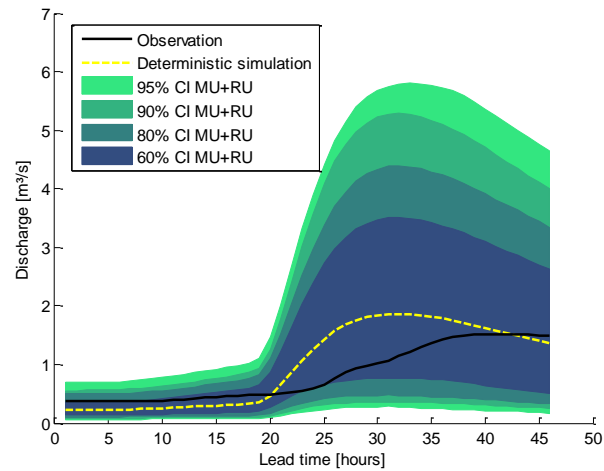
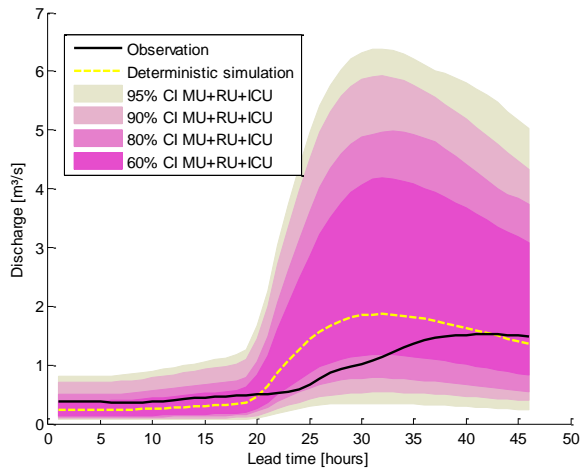
- Relative contribution of MU = $\frac{w_{3,i}^2}{w_{1,i}^2}$

- Relative contribution of ICU = $1 - \frac{w_{2,i}^2}{w_{1,i}^2}$

- Relative contribution of RU = $\frac{w_{2,i}^2 - w_{3,i}^2}{w_{1,i}^2}$

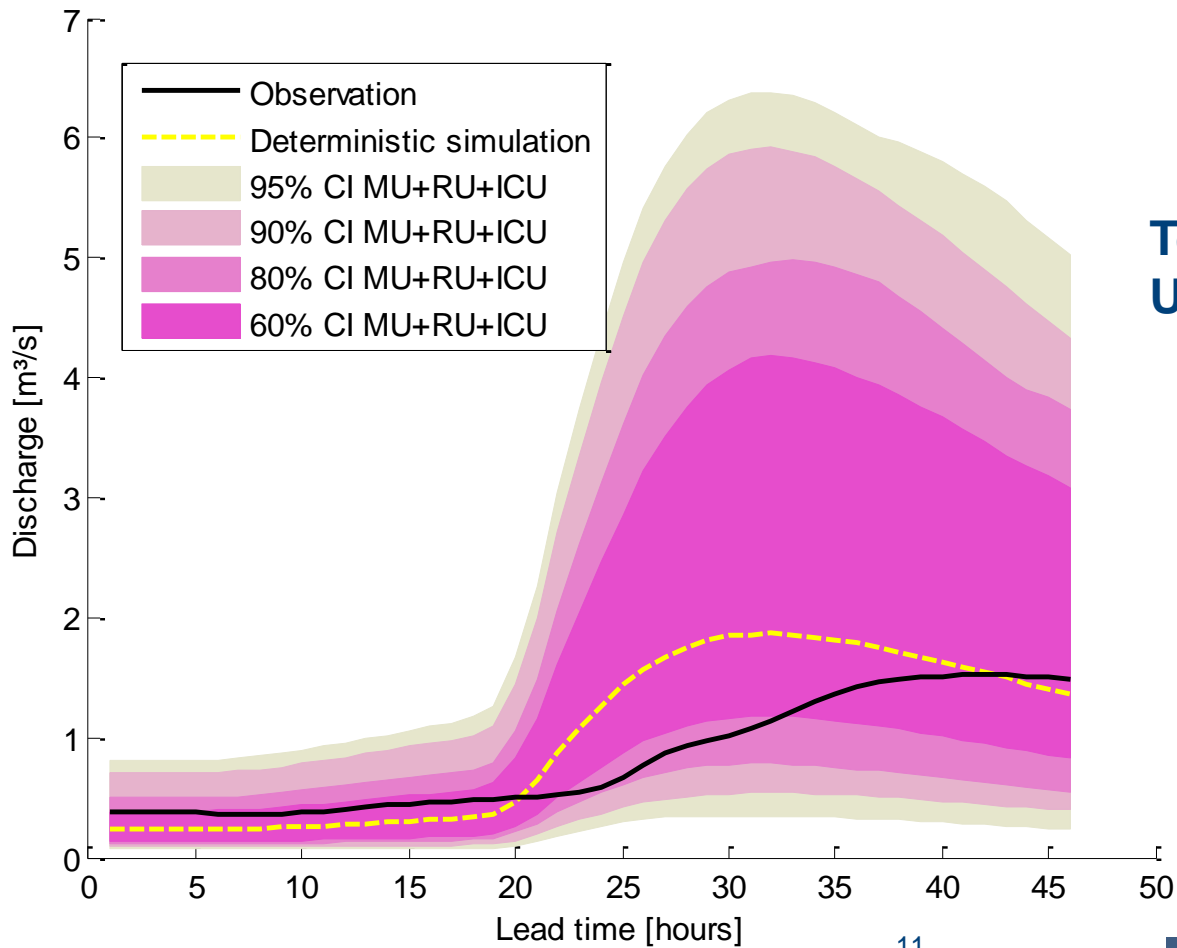
Results

Single Event: Time of forecast 05/11/2012 06:00h.



Results

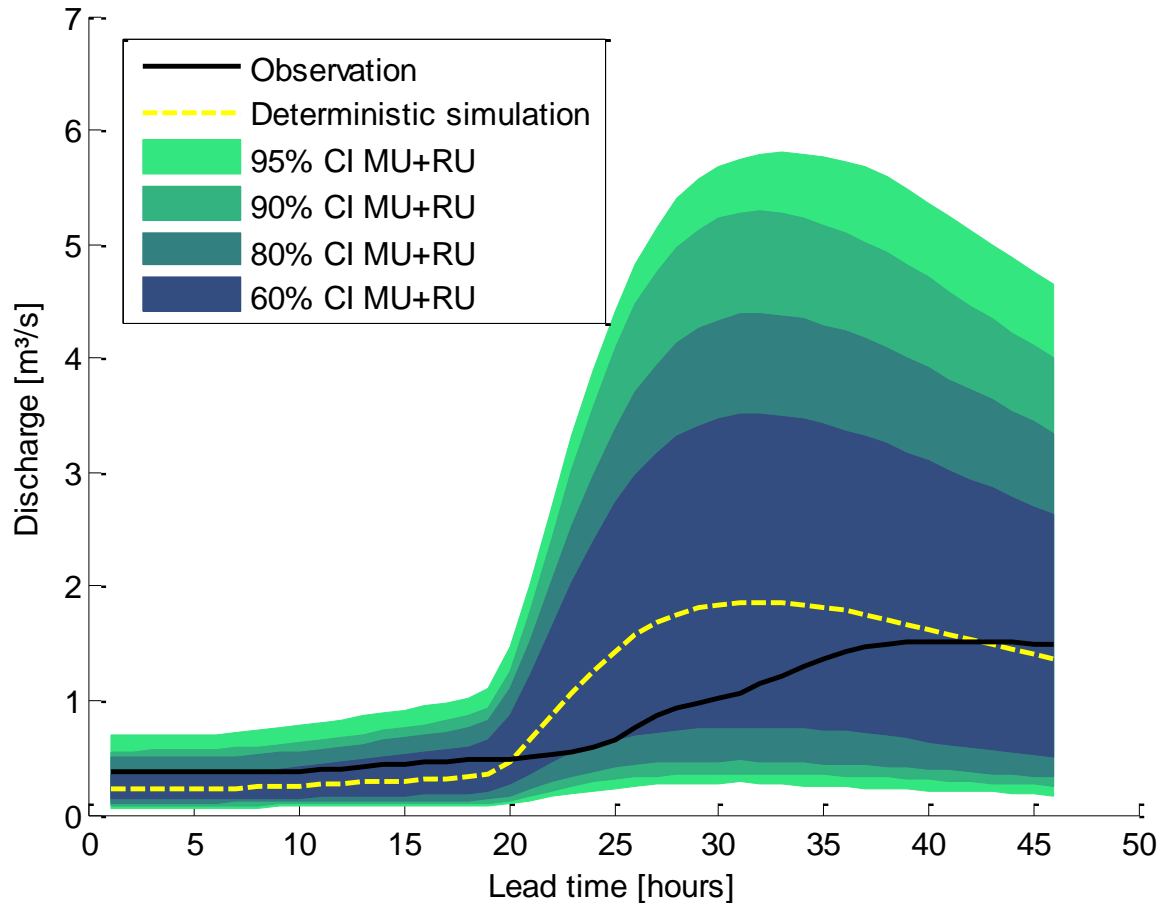
Single Event: Time of forecast 05/11/2012 06:00h.



**Total Flood Forecast
Uncertainty**

Results

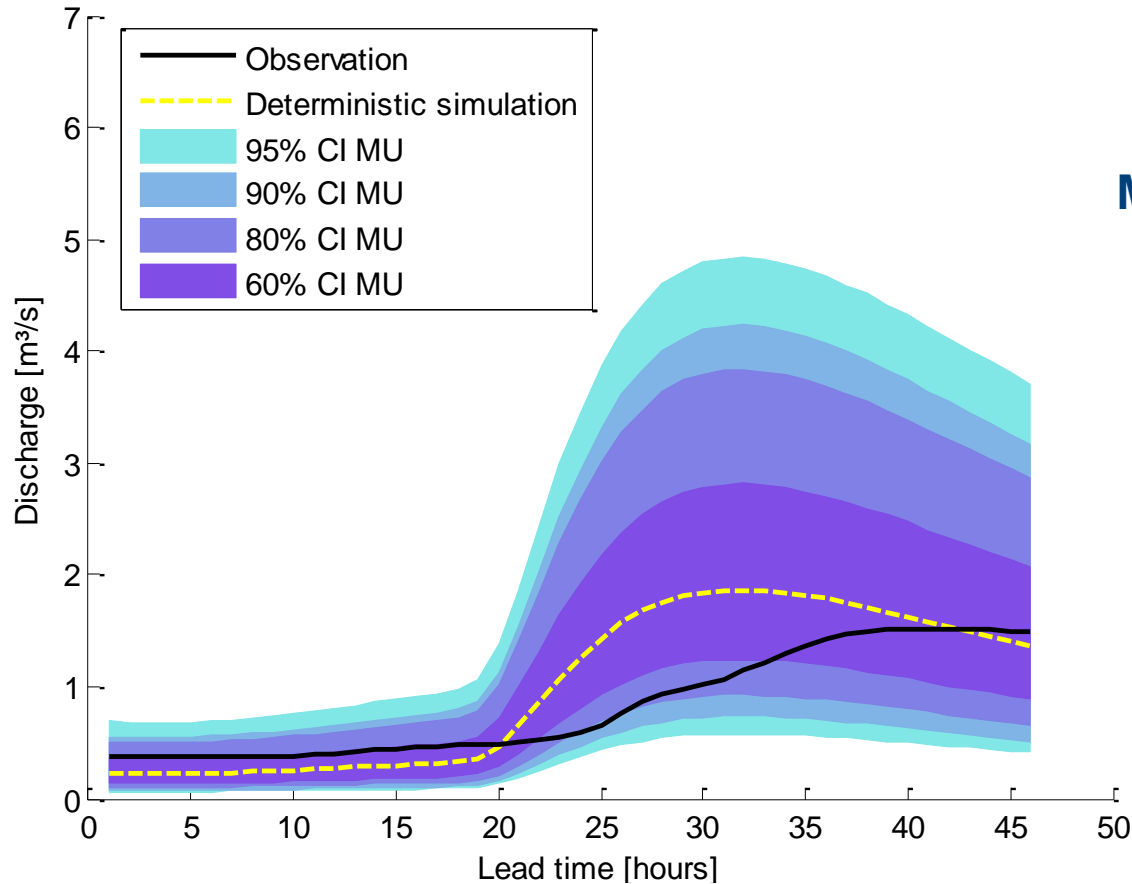
Single Event: Time of forecast 05/11/2012 06:00h.



**Model + Rainfall Forecast
Uncertainty**

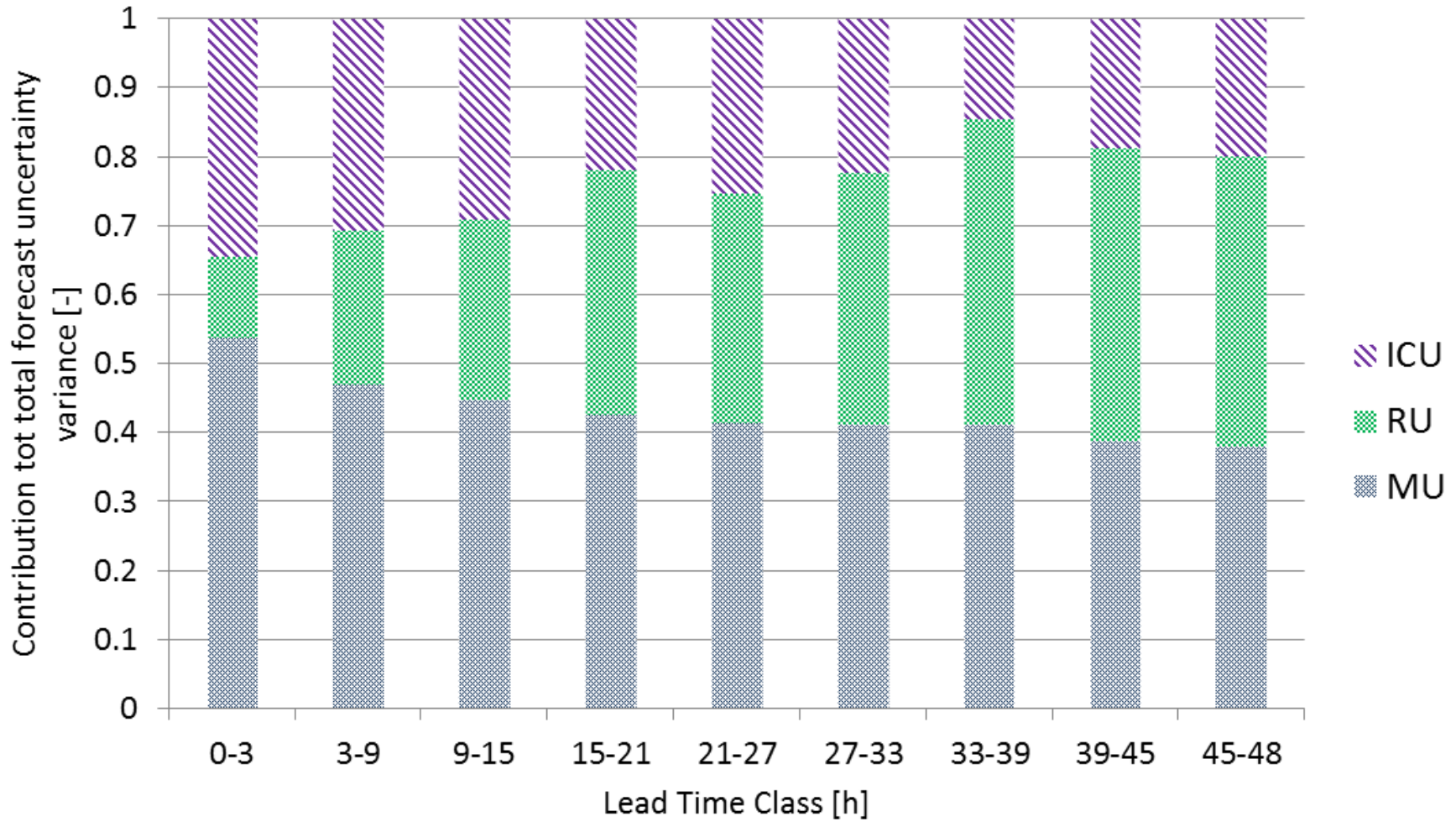
Results

Single Event: Time of forecast 05/11/2012 06:00h.



Model Uncertainty

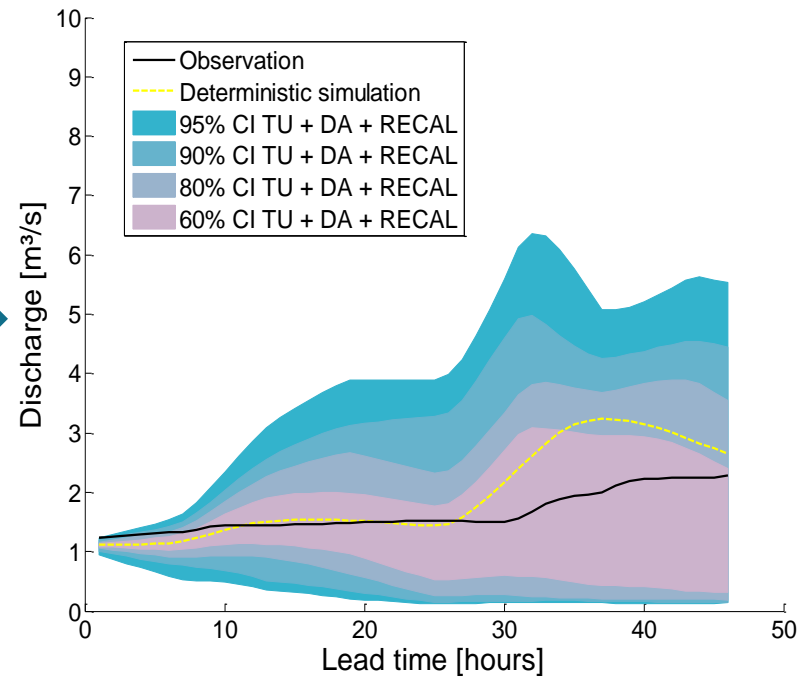
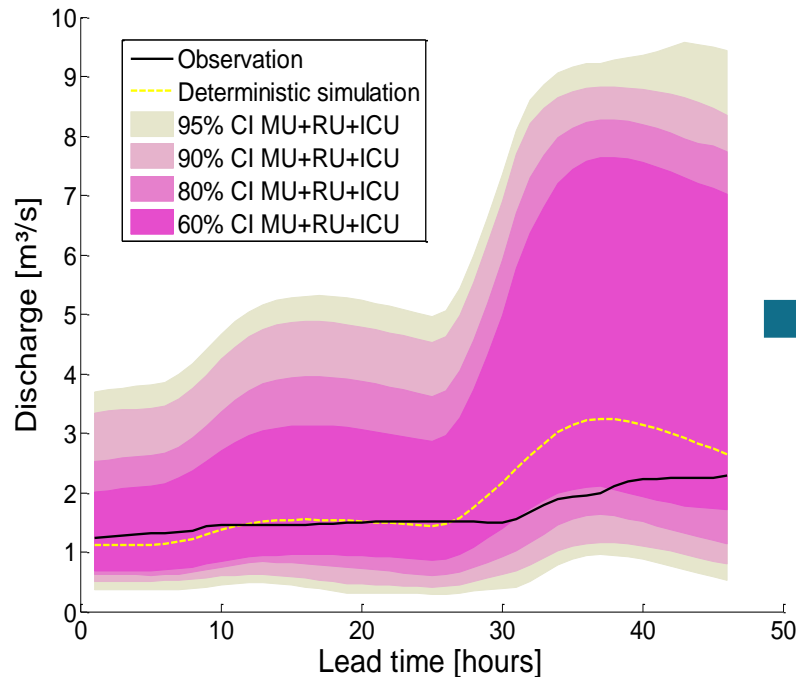
Results



Next steps

- Recalibrating NAM model to reduce model parameter uncertainty
- Improve initial conditions by DA and updating soil moisture state

TOF: 10/11/2010 06:00h



Van Steenbergen, N., Willems, P. (in revision). Uncertainty decomposition and reduction in river flood forecasting: Belgian case study. *Journal of Flood Risk Management*.

Conclusions

- Methodology to identify most important sources of uncertainty in (hydrological) flood forecasting
- Easy to apply methodology
- Model uncertainty and uncertainty in initial conditions were found to be most important uncertainty sources.
- Model recalibration
- Application of DA on hydrological output
- => Identification of most important improvement actions

Thanks for your attention !

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