

Verification of hydrological ensemble forecasts

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Joint COST Action 731 and NetFAM
Workshop on

**Uncertainty in High-Resolution
Meteorological and Hydrological Models**

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Overview

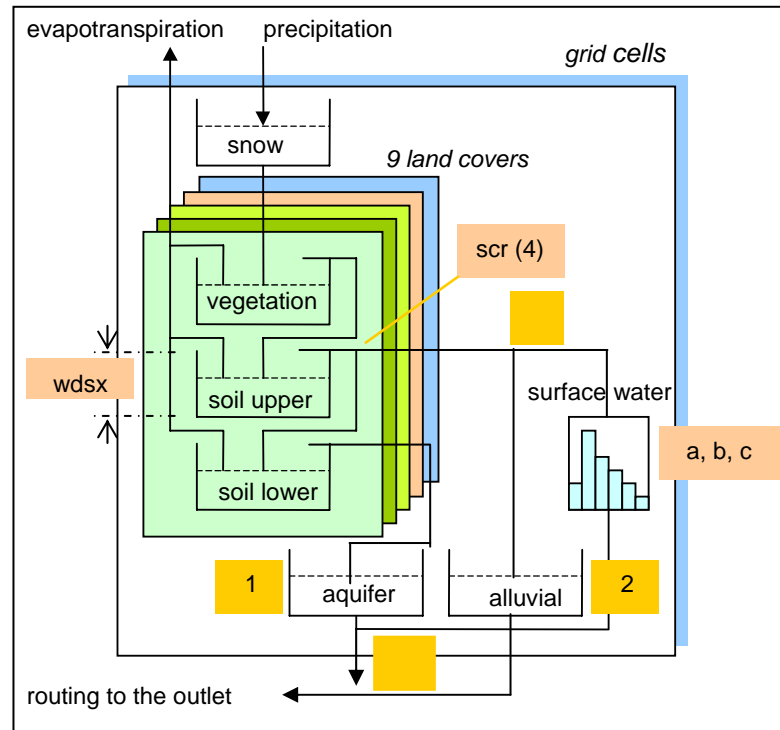
- Hydrological ensemble forecasts
- Verification methods
- Operational procedure
- Conclusions

Hydrological ensemble forecasts

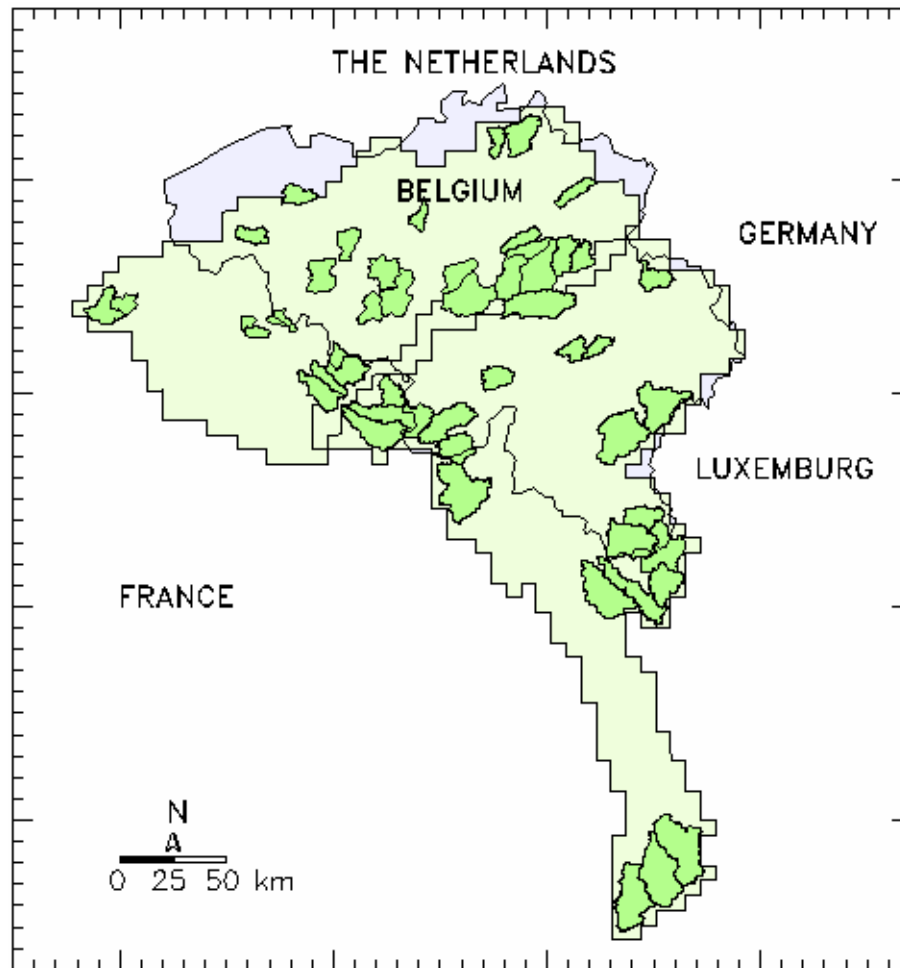
- Hydrological model « SCHEME » (SCHEIde & MEuse River Basins)
- Ensemble Prediction System ECMWF
- Probability forecasts
- Two test catchments

- Hydrological model « SCHEME »
 - Based on “IRMB” conceptual model (Bultot and Dupriez, 1976) applied on a grid with 7 7 km²
 - Routing module based on “width function”
 - Optimization of model parameters (SCE-UA)
 - Regionalization: ANN models between parameter values and physiographic indices
 - Daily time step

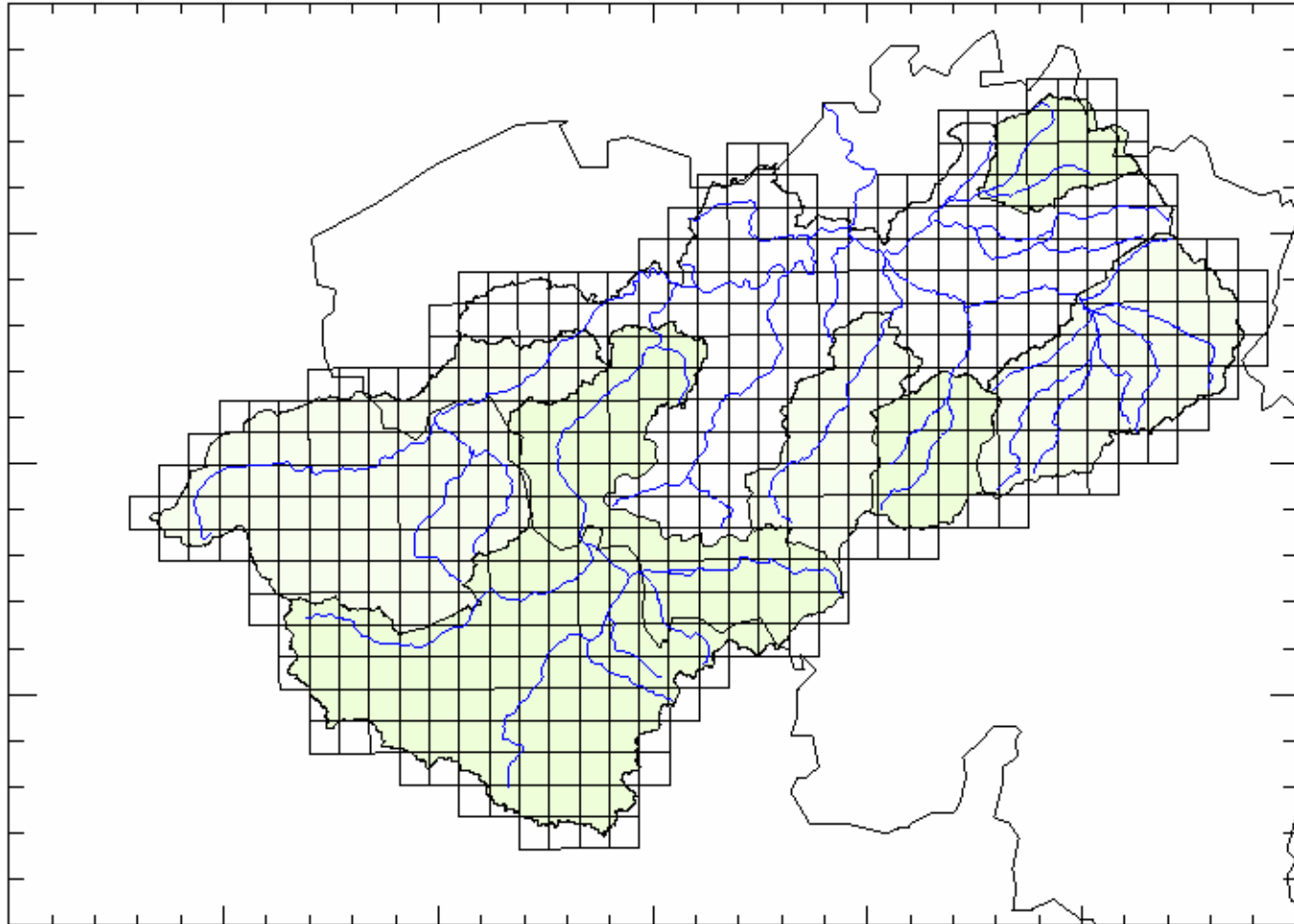
– “IRMB” conceptual model applied on 7 7 km²



- Model parameters optimized & regionalized



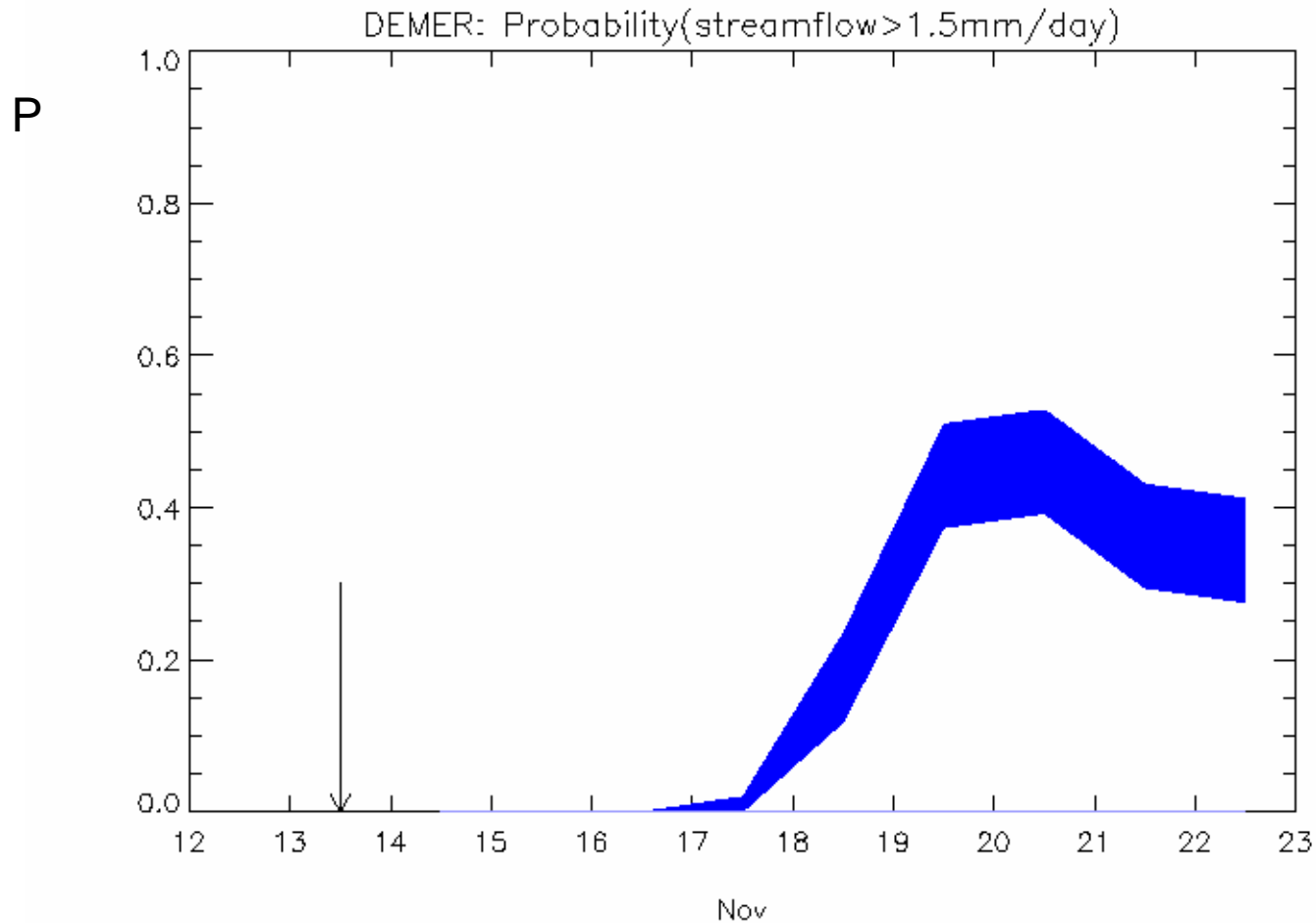
- Routing parameters on catchments $> 500 \text{ km}^2$



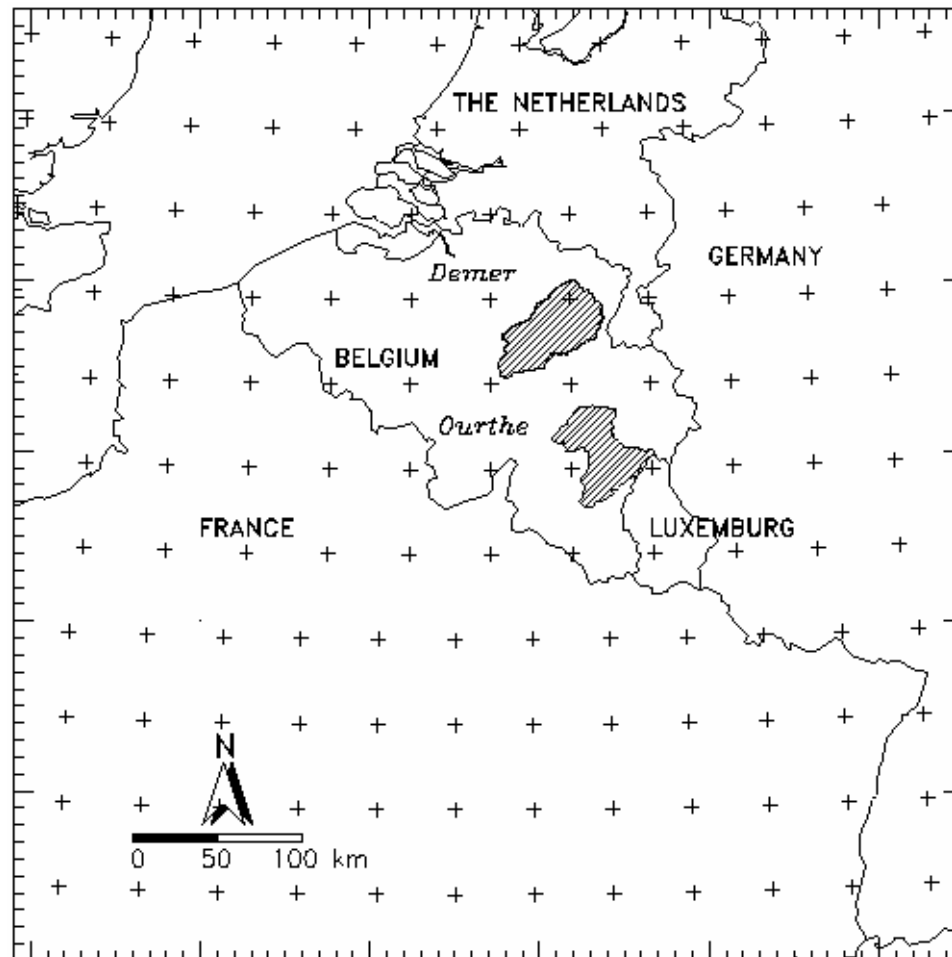
- **Ensemble Prediction System ECMWF**
 - Use of archived forecasts since December 1996
 - Horizontal resolution: 120 km, 80 km (Nov. 200), 50 km (Feb. 2006)
 - Direct use of precipitation

- Probability forecast

- EPS + SCHEME = 51 hydrograms probability to exceed a threshold (e.g. P95)



- Two test catchments
 - Démer at Diest (1775 km²), Ourthe at Tabreux (1616 km²)



Verification methods

- Rank histogram
- Distribution Oriented approach
- Relative economic value

- Rank histogram

- Rank histogram (e.g. Talagrand *et al.*, 1997)

- event x and each forecast f_i equally plausible*

- Correction (Hamill and Colucci, 1997)

- $$P(x < f_i) = \frac{R_j}{j}$$

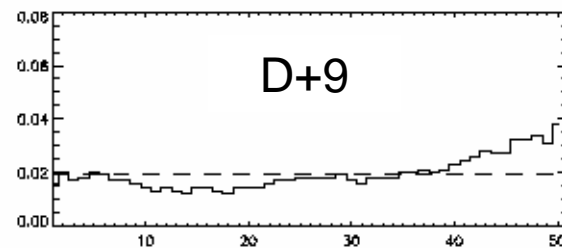
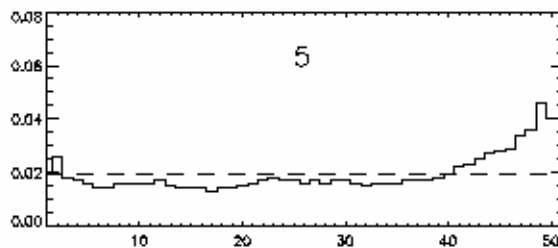
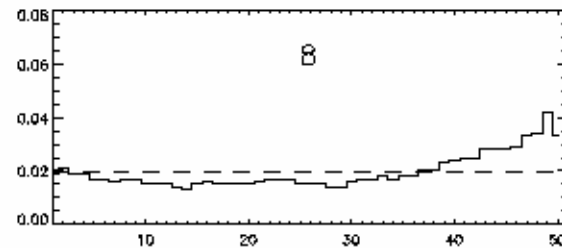
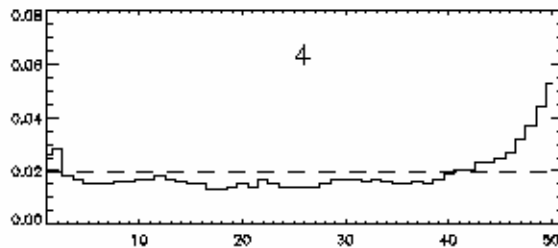
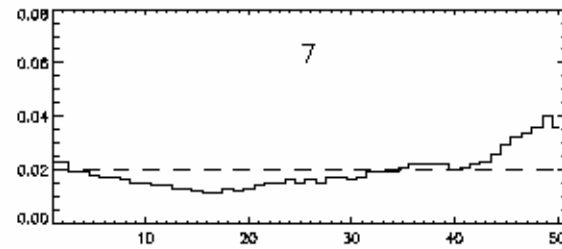
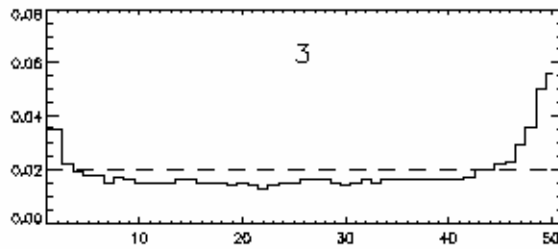
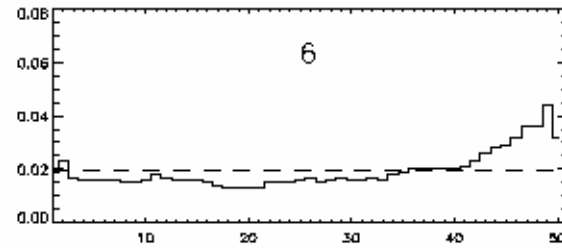
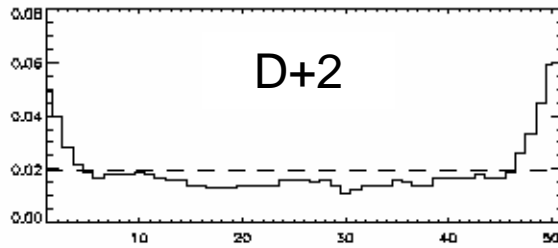
- below the lowest ensemble member : uniform

- beyond the highest ensemble member : Gumbel distribution

- Example: Ourthe (Nov 2000 – Mar 2004)

* Here, the reference streamflow is simulated using observed precipitation.

– Example: Ourthe, threshold 6 mm, winter



- **Distribution Oriented approach**
 - Joint distribution of forecast and observations
 - Brier Skill Score
 - Calibration-refinement factorization
 - Likelihood-base rate factorization

- Joint distribution of forecast and observations
 - f : forecast = probability that P or $Q >$ threshold
 = (# members with P or $Q >$ threshold) / size of ensemble
 - x : observation = 1 if P or $Q >$ threshold,
 otherwise, $x = 0$
 - Thresholds based on [1971-2000], in mm day⁻¹

	P80	P90	P95
P Demer	3.6	6.5	9.8
P Ourthe	4.7	8.3	12.2
Q Demer	0.8	1.1	1.5
Q Ourthe	1.8	2.8	3.9

– Brier Skill Score

- Brier Score or mean square error

$$\text{MSE}(f,x) = E(f - x)^2$$

- Brier Skill Score relative to the “sample climate”
or $\sigma_x^2 = p(1 - p)$

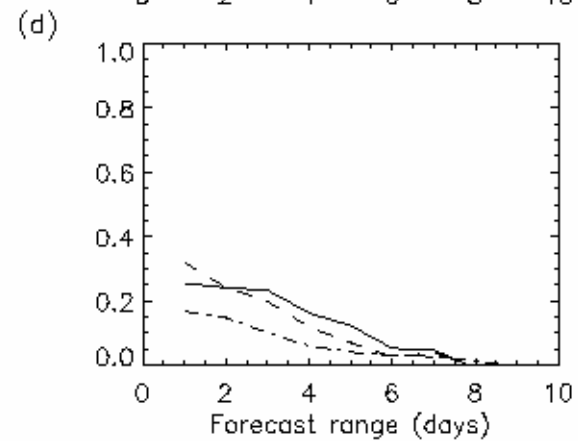
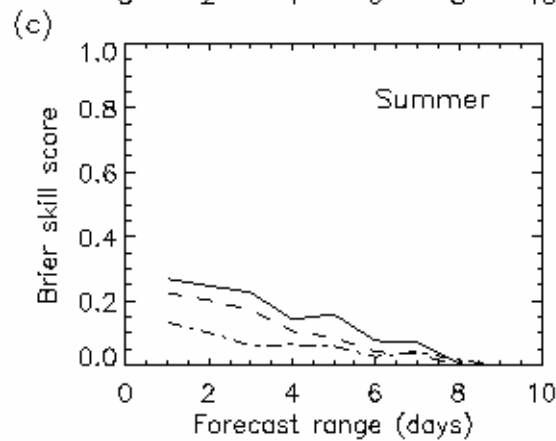
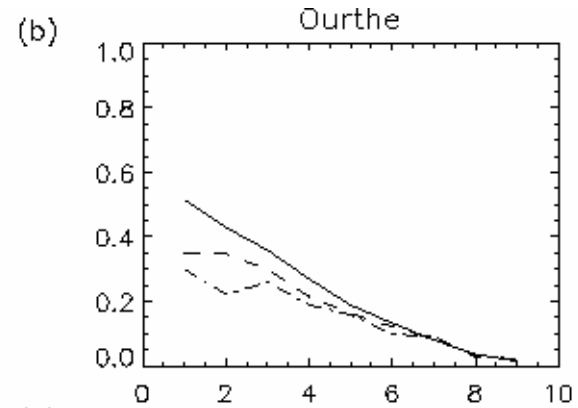
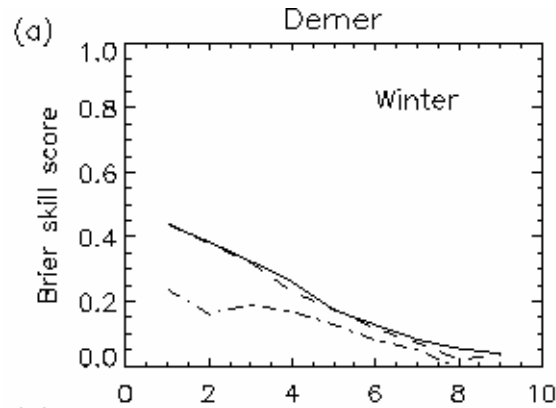
$$\text{BSS} = 1 - \text{MSE}(f,x) / \sigma_x^2$$

– BSS: EPS precipitation

P80

P90

P95

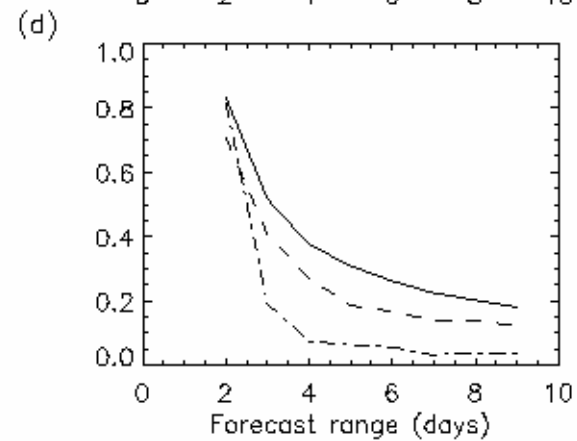
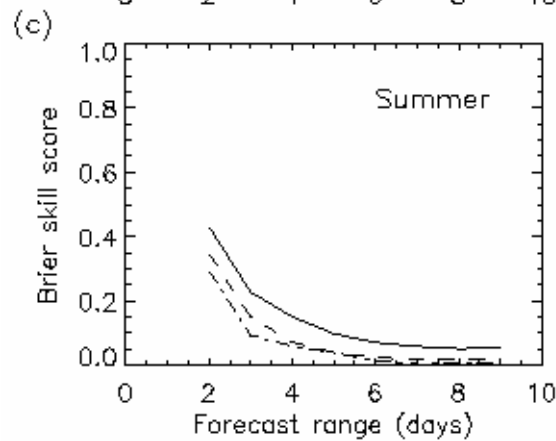
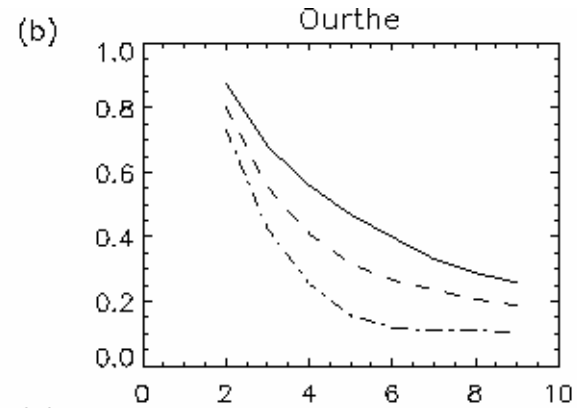
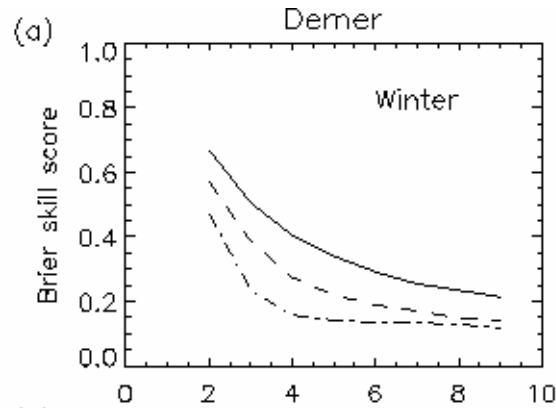


– BSS: Streamflow with precipitation from climatology

P80

P90

P95

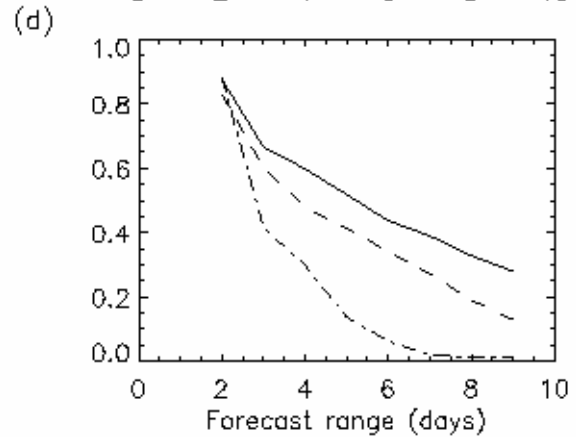
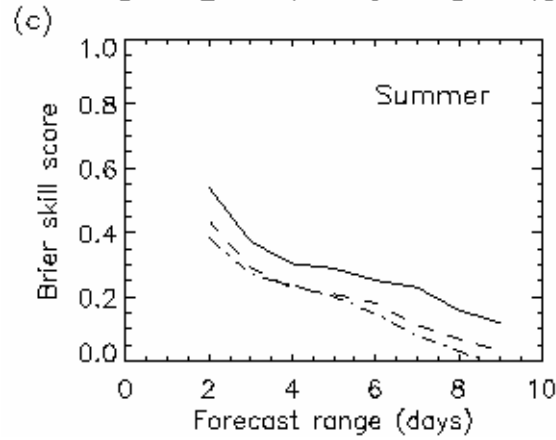
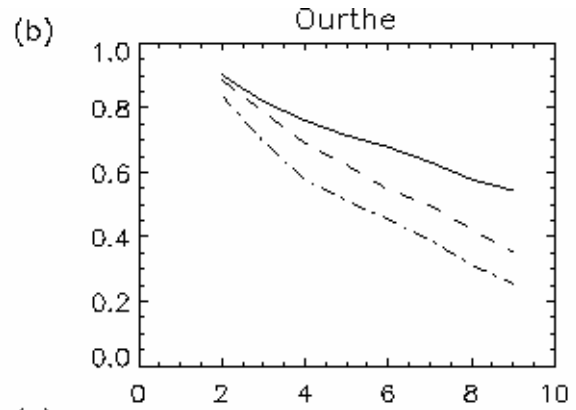
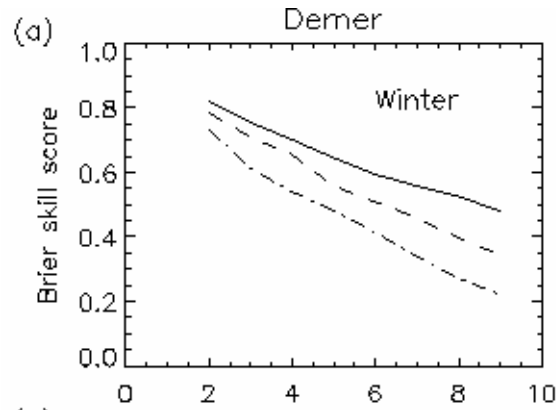


– BSS: Streamflow with EPS precipitation

P80

P90

P95



– Calibration-refinement factorization

- $p(f,x) = p(x|f) p(f)$

- Decomposition of the mean square error:

$$\text{MSE}(f,x) = E(f - x)^2 = \sigma_x^2 + E(x|f - f)^2 - E(x|f - x)^2$$

- σ_x^2 : *uncertainty*

- $E(x|f - f)^2 = \text{REL}$: *reliability (conditional bias)*

- $E(x|f - x)^2 = \text{RES}$: *resolution*

- Skill score relative to the “sample climate” (σ_x^2)

- $\text{BSS} = 1 - \text{MSE}(f,x) / \sigma_x^2$

- $\text{BSS} = \text{RRES} - \text{RREL}$

– Likelihood-base rate factorization

- $p(f,x) = p(f|x) p(x)$

- Decomposition of the mean square error:

$$\text{MSE}(f,x) = E(f - x)^2 = \sigma_f^2 + E(f|x - x)^2 - E(f|x - x)^2$$

- $\sigma_f^2 = \text{SH} : \textit{sharpness}$

- $E(f|x - x)^2 = \text{TY2} : \textit{type II conditional bias}$

- $E(f|x - x)^2 = \text{DIS} : \textit{discrimination}$

- Skill score relative to the “sample climate” (σ_x^2)

- $\text{BSS} = 1 - \text{MSE}(f,x) / \sigma_x^2$

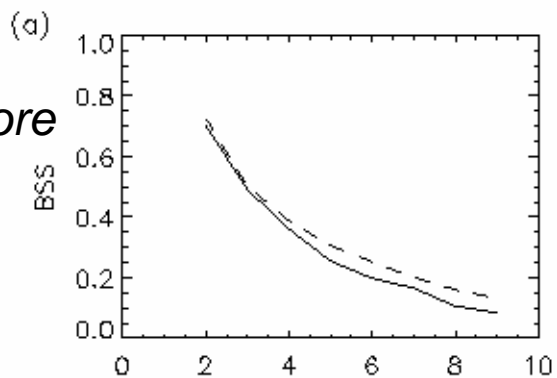
- $\text{BSS} = 1 + \text{RDIS} - \text{RSH} - \text{RTY2}$

- Example 1: corrected ensemble (see above, Ourthe during winter, from Nov 2000 to Jul 2004, threshold: streamflow above 6 mm)
- Example 2: comparison of ensemble streamflow with streamflow forecast using precipitation from the ECMWF deterministic, the EPS ensemble average and the EPS control (threshold 4 mm)
- Example 3: uncertainty associated to the value of hydrological parameters (in progress: uncertainty about only 1 parameter) (Demer)

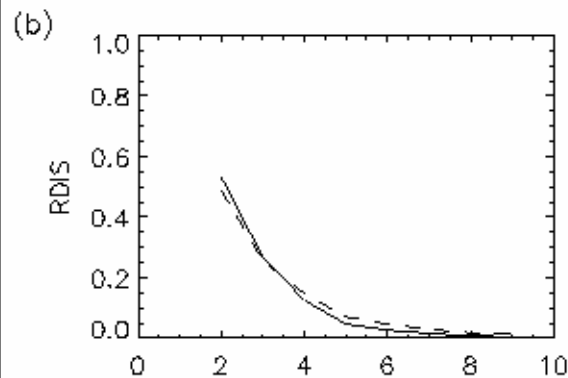
———— ensemble

----- corrected ensemble

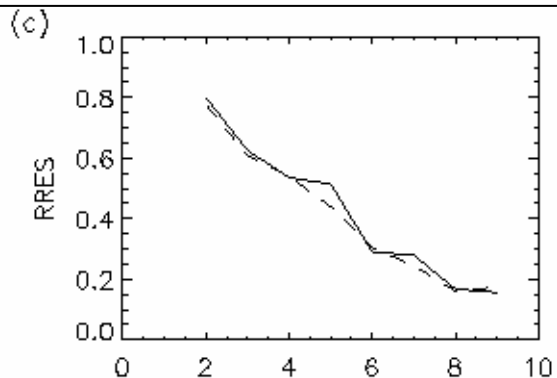
Brier skill score



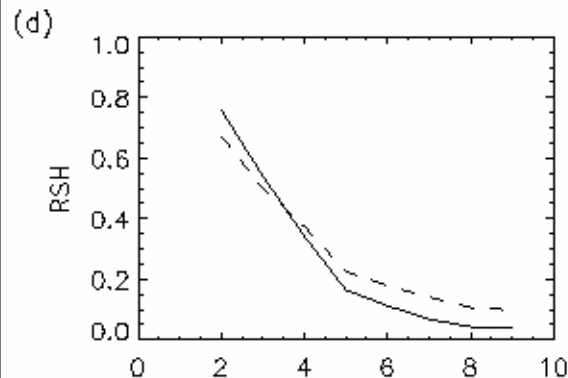
discrimination



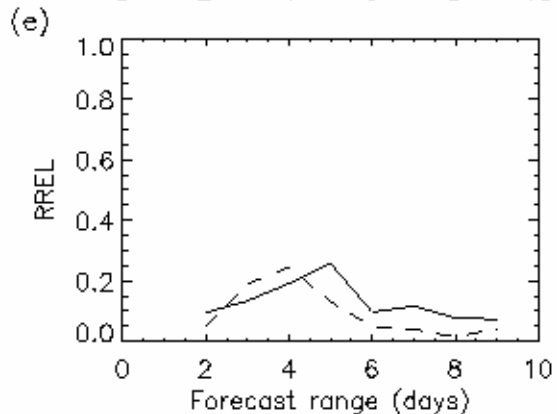
resolution



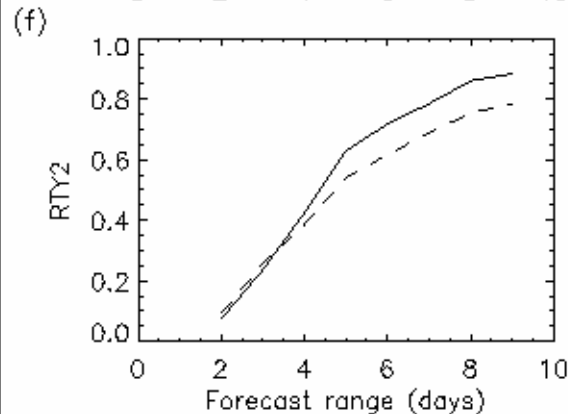
sharpness



reliability

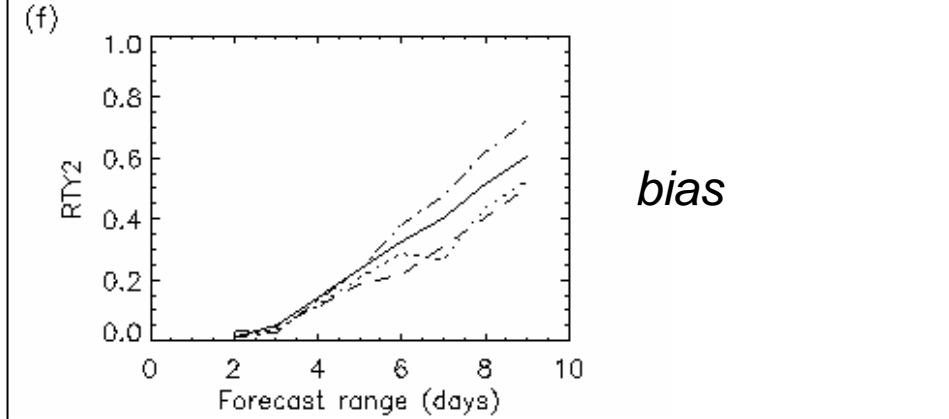
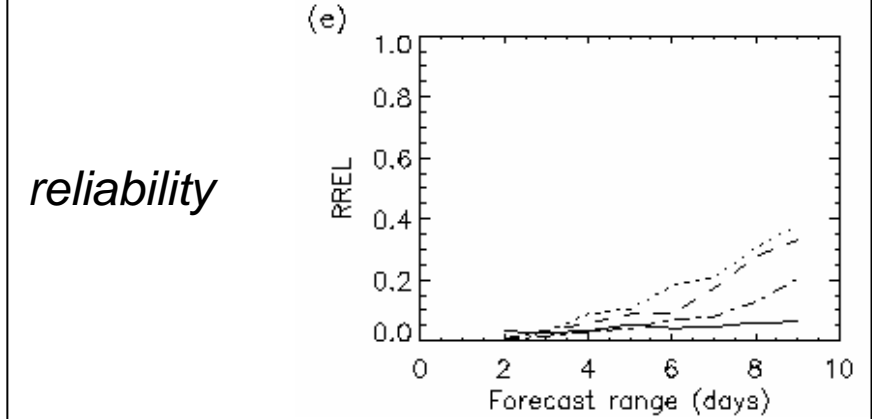
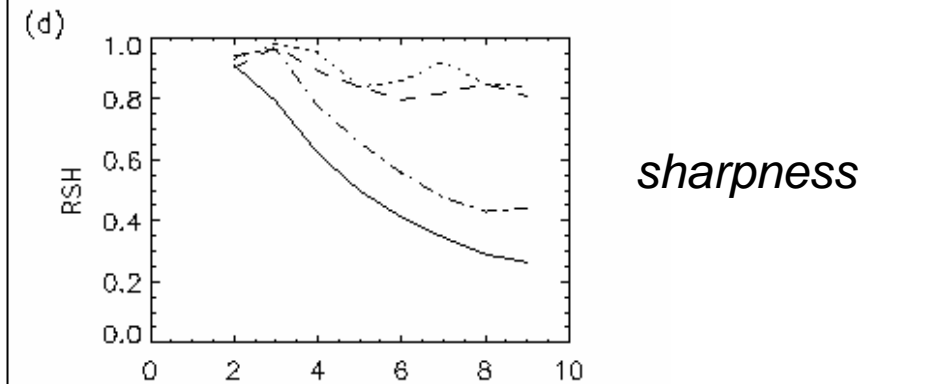
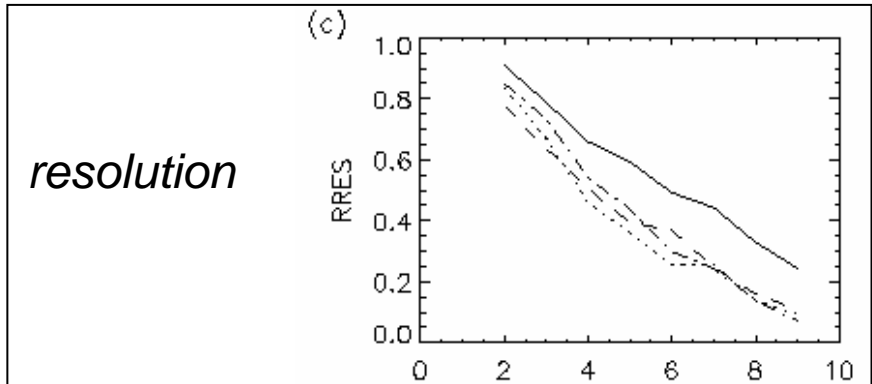
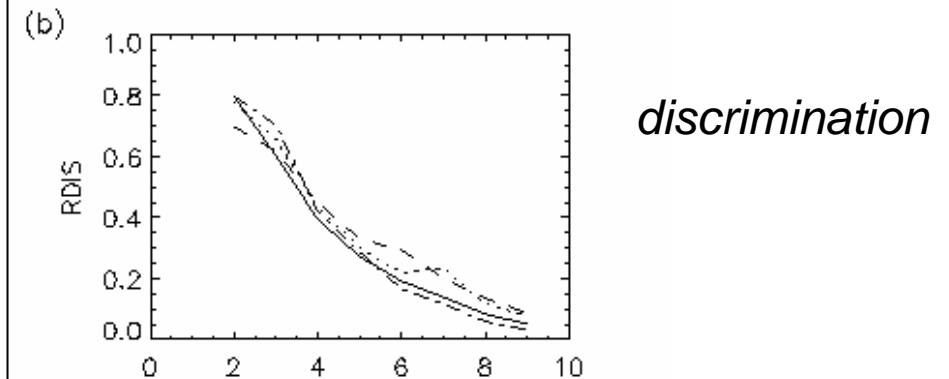
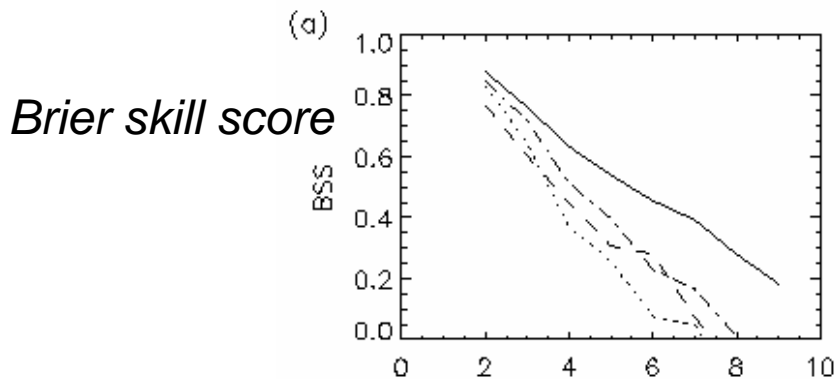


bias



———— ensemble
- - - - - average

- - - - - deterministic
..... control

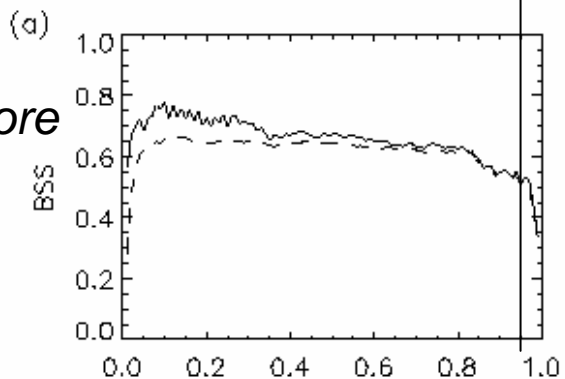


———— ensemble

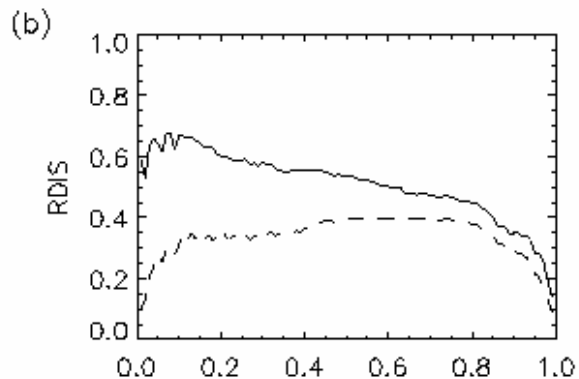
----- par. uncertainty

P95

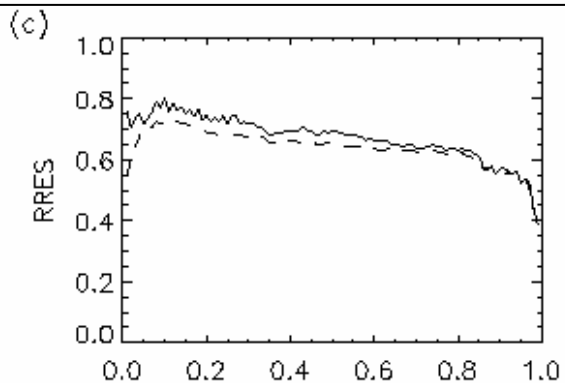
Brier skill score



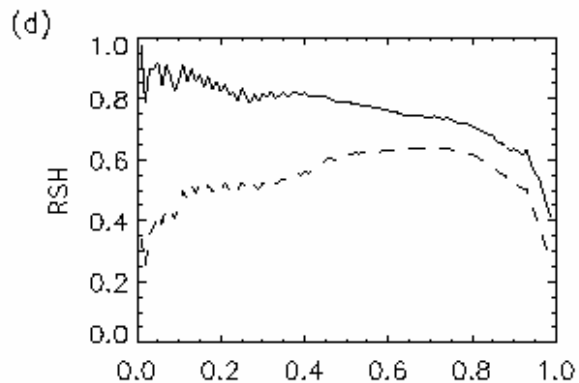
discrimination



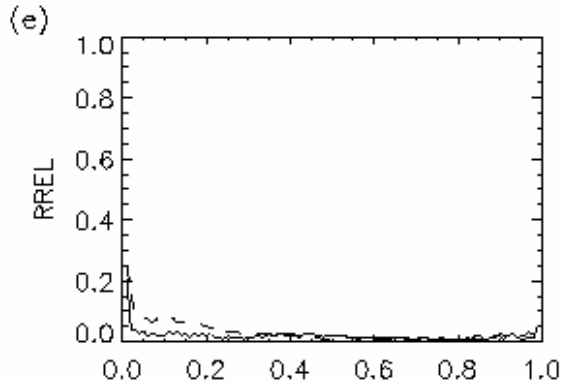
resolution



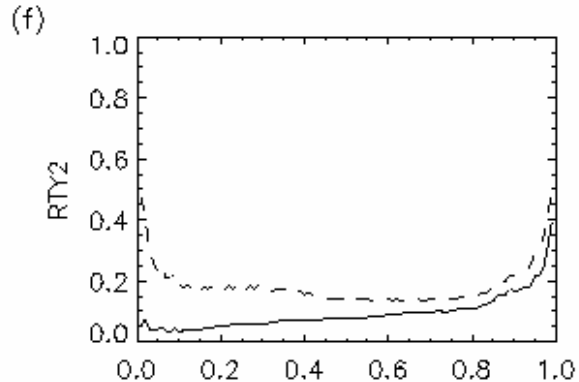
sharpness



reliability



bias



threshold

threshold

- Relative economic value

- Static cost-loss model (Richardson, 2000)

- C: cost of action

- L: loss if the event occurs and no action is taken

- L1: part of the loss prevented by taking action

- o: fraction of occasions when the event occurs

- Minimize the expense E:

- $E_{clim} = \min \{ C + o (L - L1), o L \}$

- $E_{perf} = o (C + L - L1)$

- Relative value:

- $V = (E_{clim} - E_{for}) / (E_{clim} - E_{perf})$

- Relative economic value

- Deterministic forecast system

- H : hit rate
- F : false-alarm rate
- $\alpha = C / L1$, cost-loss ratio

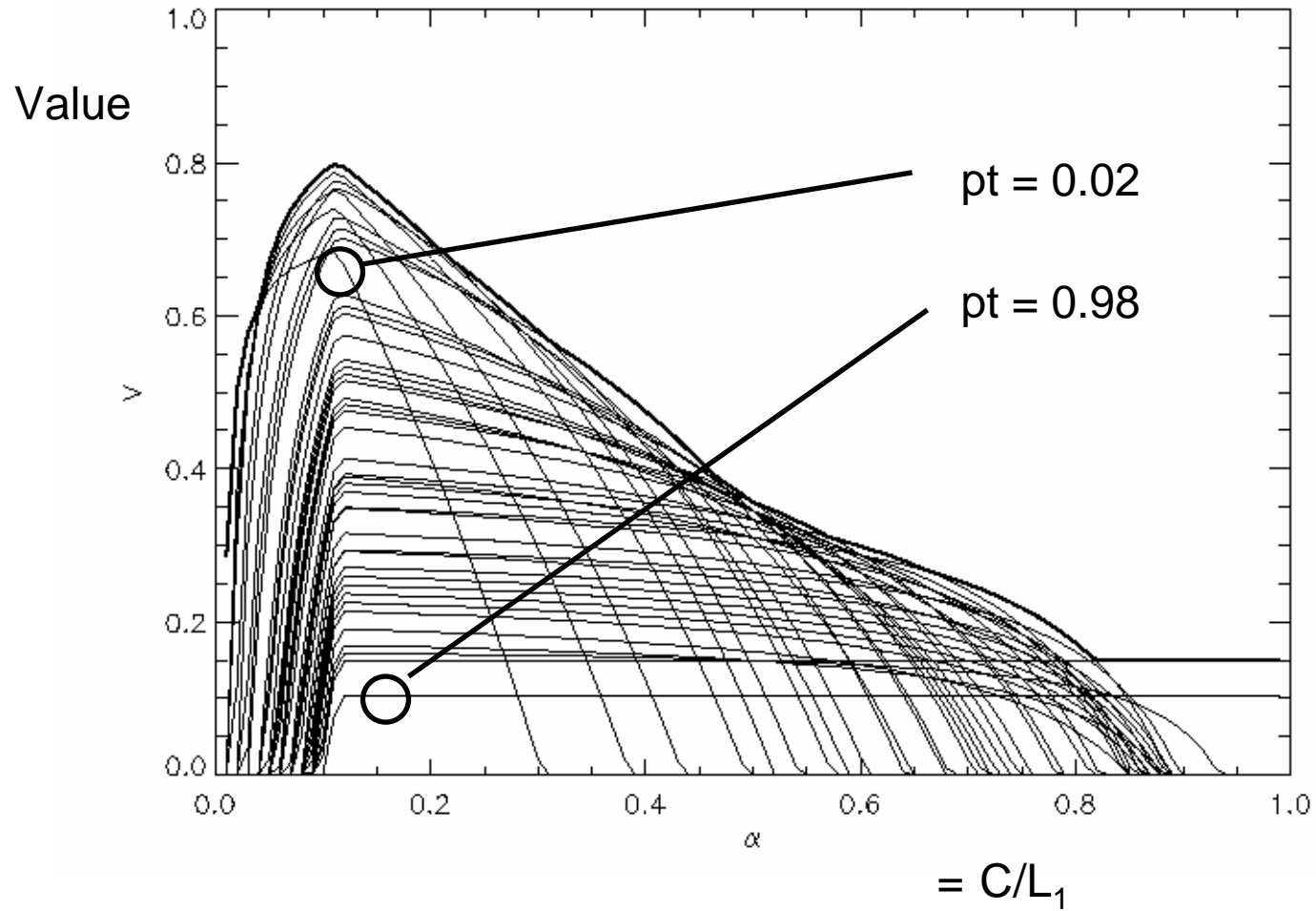
- Relative value

- $$V = \frac{\min(C, L) - F(1 - \alpha) + H\alpha(1 - C) - C}{(\min(C, L) - C)}$$

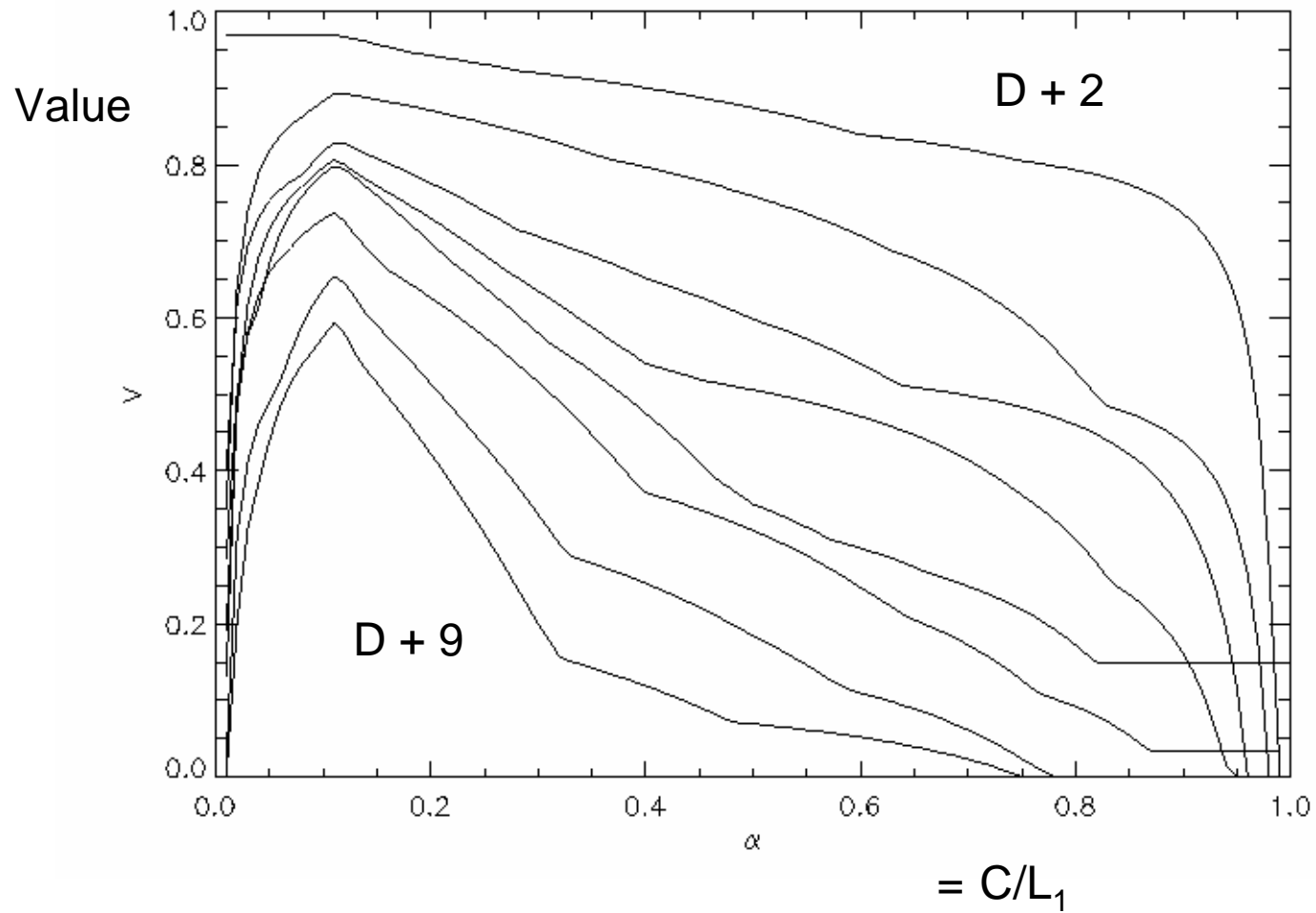
- Probabilistic forecast

- p_t : threshold probability

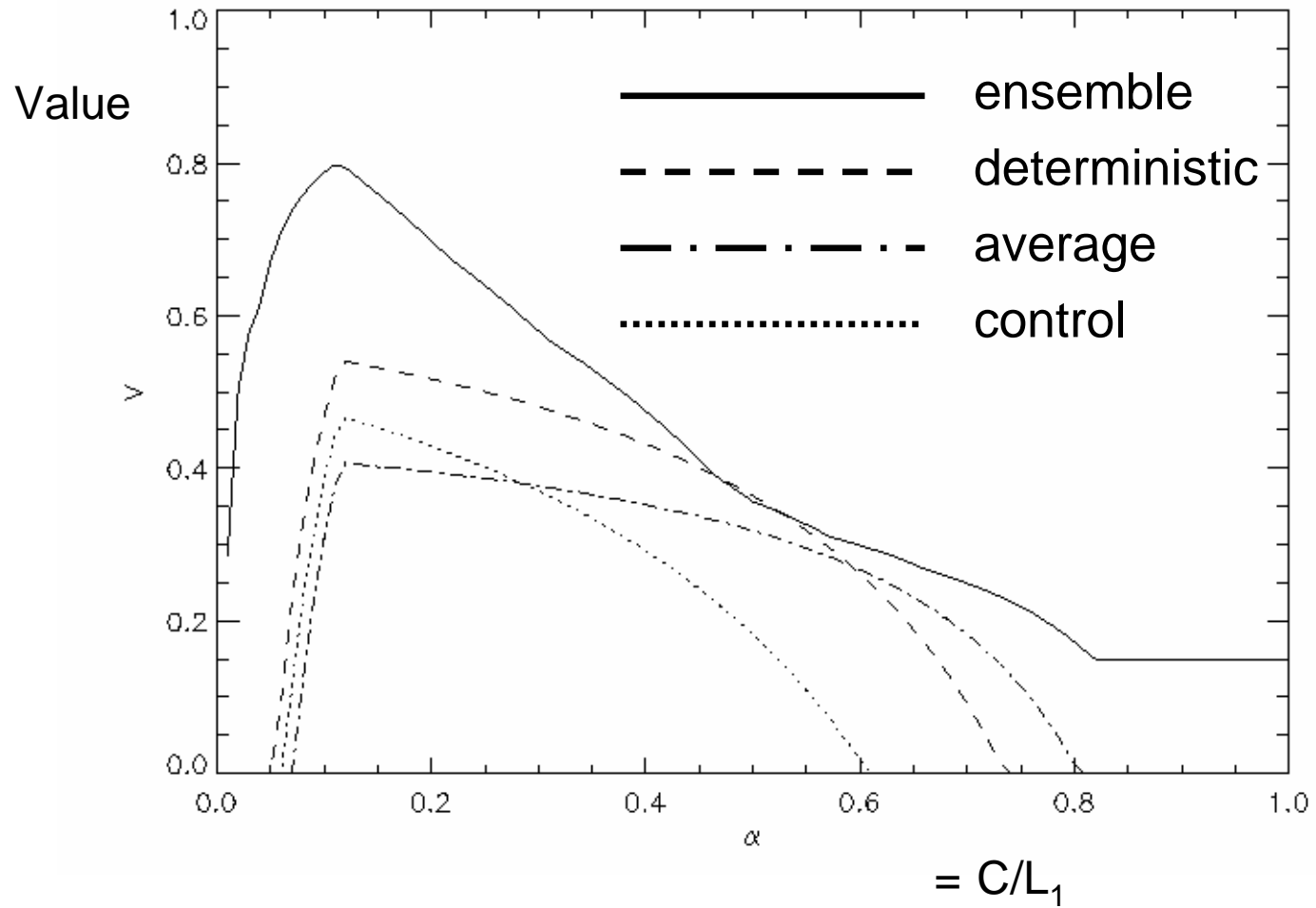
– Hydrological ensemble, 4 mm (P95), D+6



– Hydrological ensemble, 4 mm (P95), D+2 D +9



– 4 mm (P95), D+6



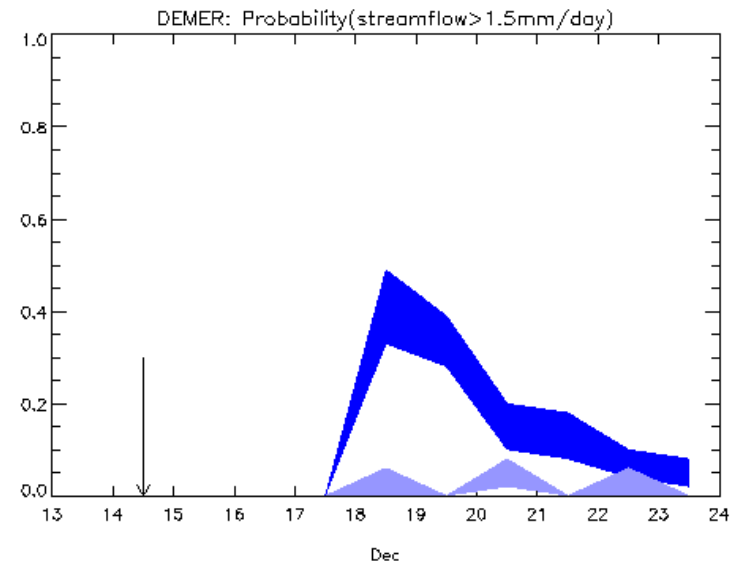
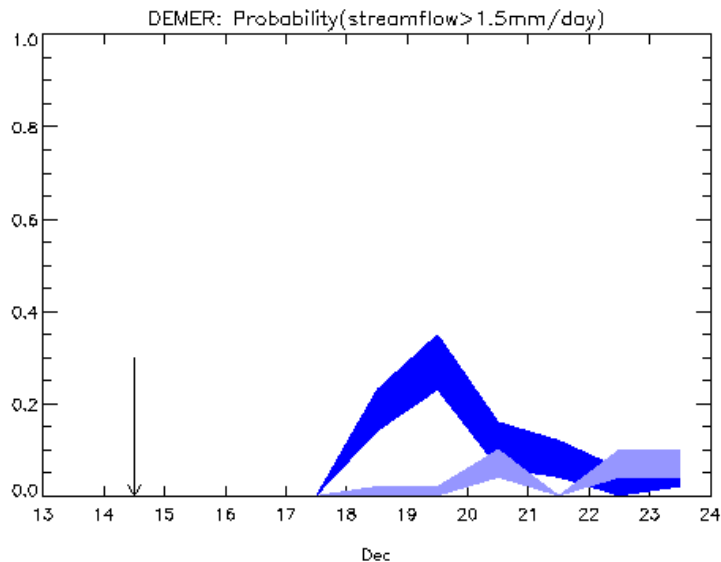
Operational procedure

- HEPDO project
- Validation
- Verification

- HEPDO project

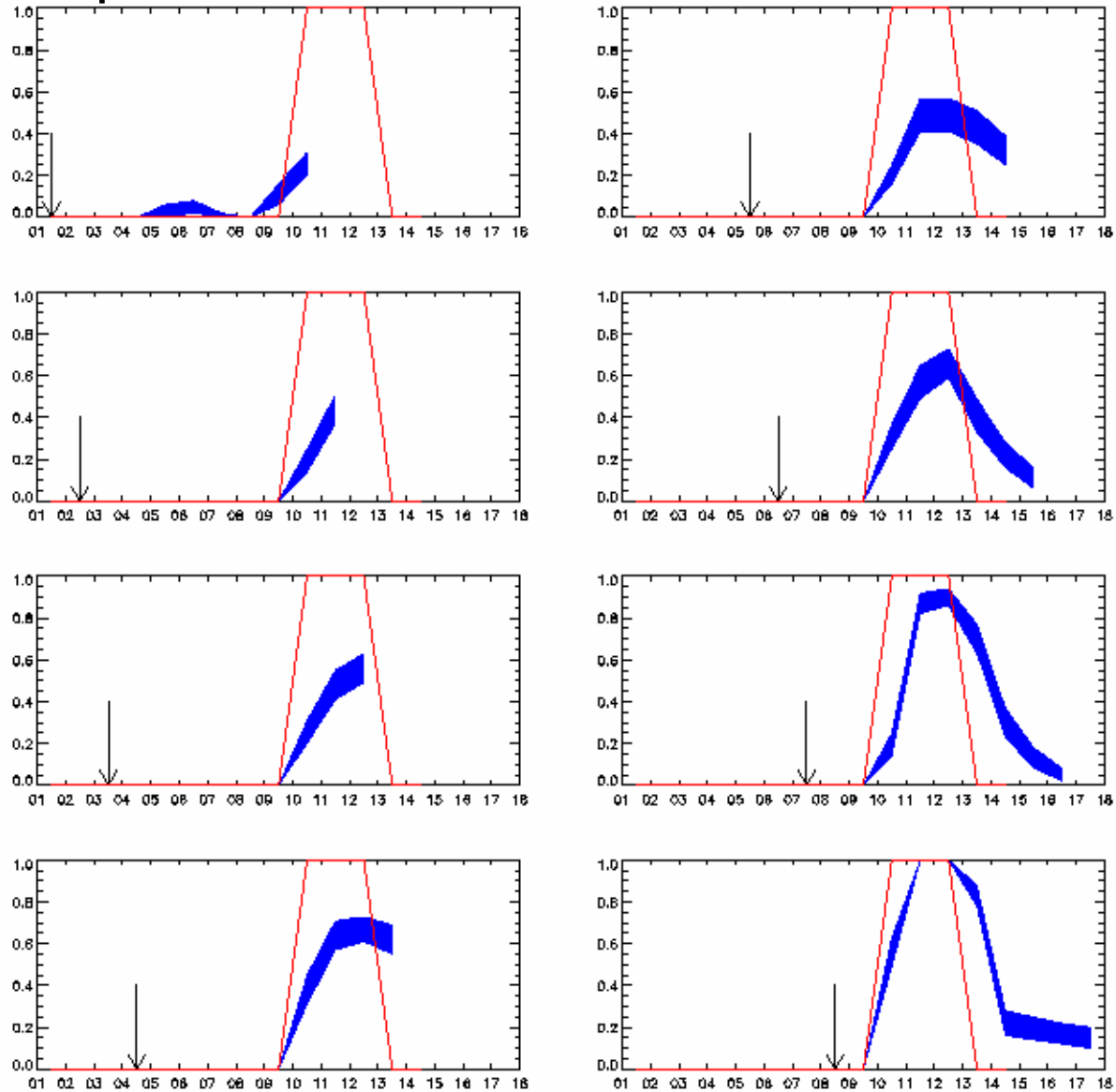
- “Hydrological Ensemble Prediction for the Demer and the Ourthe”: setting-up an automatic operational procedure
- EPS precipitation but also temperature, wind speed etc. to estimate potential evapotranspiration and account for snow accumulation and melting
- Use of precipitation data available operationally: weather radar and automatic weather stations to update the water content of the conceptual reservoirs

- Demer in December 2005; left: with radar data only; right: with raingauge data during November and then, with radar data; in progress: with combined radar data and available raingauge data



- Proposed for validation to:
 - Weather forecasters : maps with probability of precipitation
 - Regional authorities in charge of water management : web page with ensemble precipitation and probability of streamflow > P95

– Example: Ourthe, March 2006

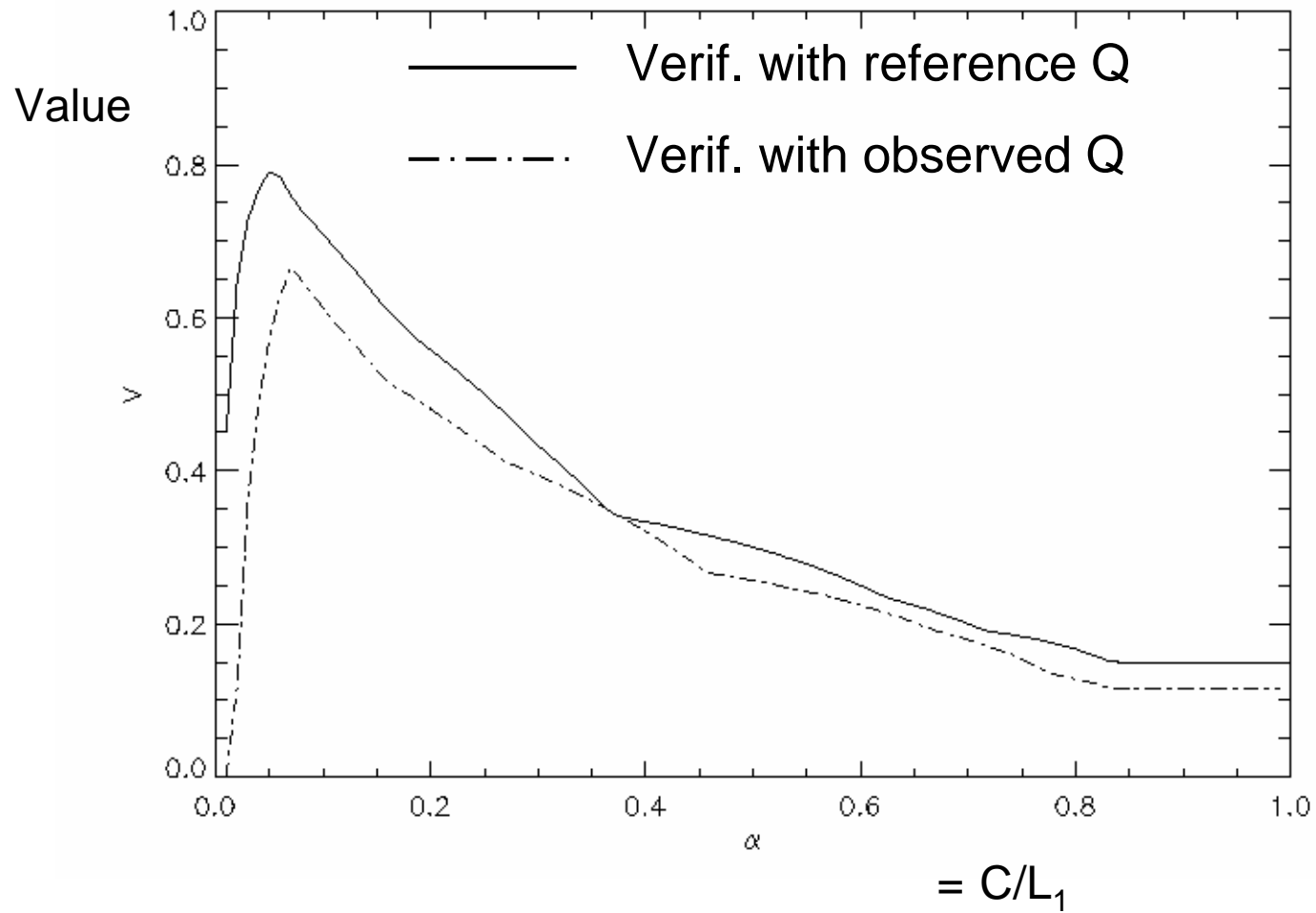


- Verification

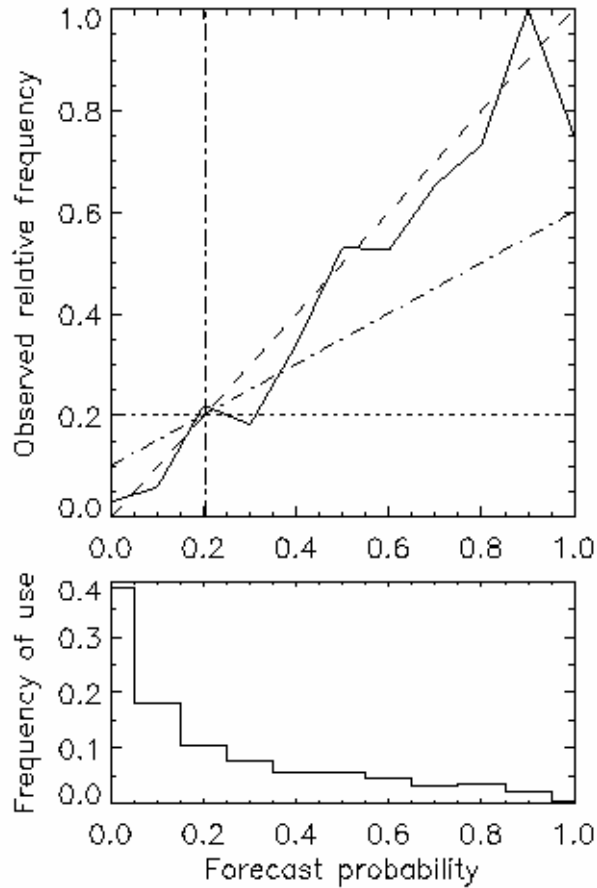
- Comparison with measured streamflow

- New EPS resolution: small sample size

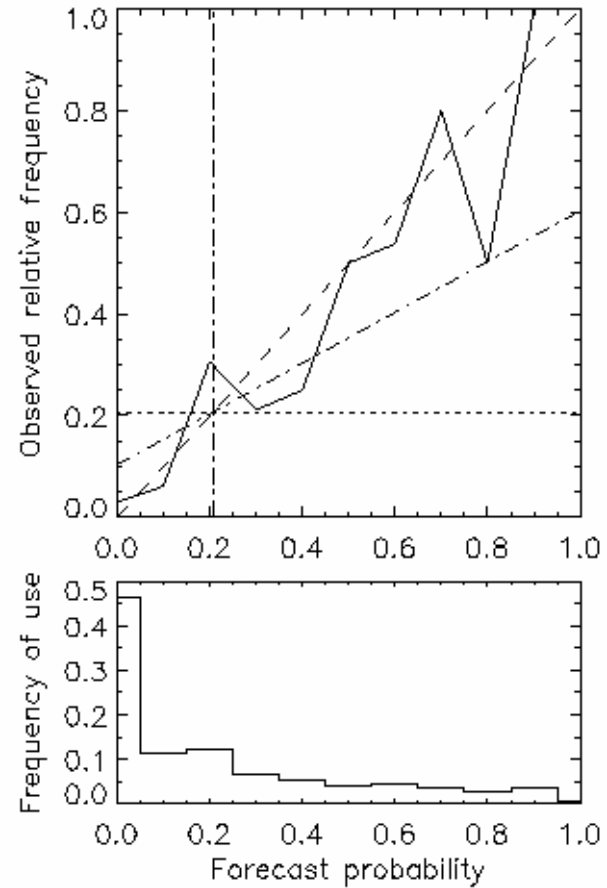
– Threshold 5mm, D+6, Nov. 2000 – Dec 2002



– Reliability diagram for precipitation (Demer), P80



Winter 2001-2005



Feb-March 2001-2005

Conclusions

- Hydrological ensemble predictions have skill and value for early warning
- Need to define the probability threshold for a specific management situation
- Improve verification methods for rare events and small data sample
- Investigation on effect of uncertainties on the aspects of forecasts quality