

# A Probabilistic View on Winter Storm Damages

Uwe Ulbrich (1), Tobias Pardowitz (1,2), Robert Osinski (3)

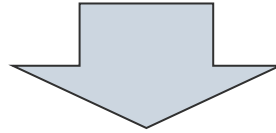
(1) Freie Universität Berlin

(2) Hans-Ertel-Zentrum für Wetterforschung, Berlin, Germany

(3) Centre National de Recherches Météorologiques

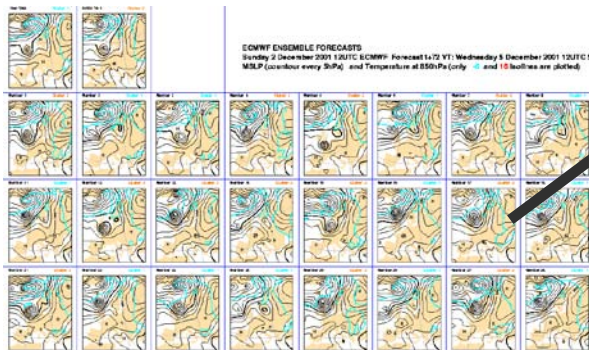
# Introduction

- Operational forecasting using *Ensemble Prediction Systems (EPS)*



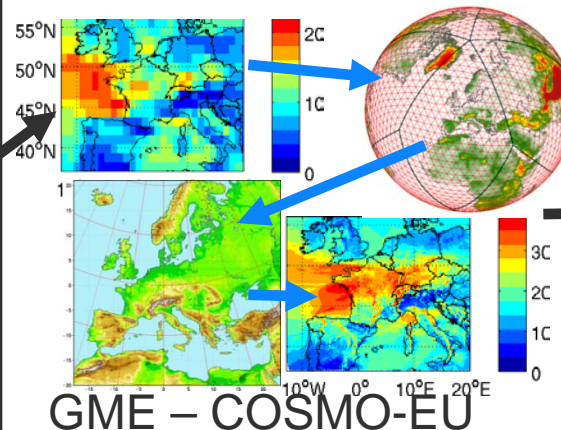
- Potential developments of storms and estimation of likelihoods of severities
- Downscaling provides higher resolution for individual simulation results

## EPS-storms from ECMWF



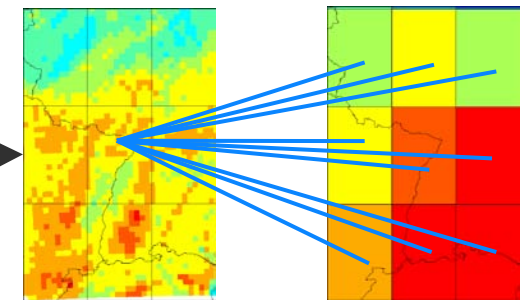
Years 2000-2010, T399

## Dynamical downscaling using a subset of storm events (from EPS & reanalyses)



GME – COSMO-EU

## Statistical transfer function for gusts



$$y = \sum_i a + b_i \cdot x_i + c_i \cdot x_i^2$$

(T. Kruschke)

# Deterministic loss model: Regionally trained wind-loss ratio relationship

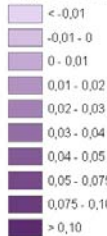
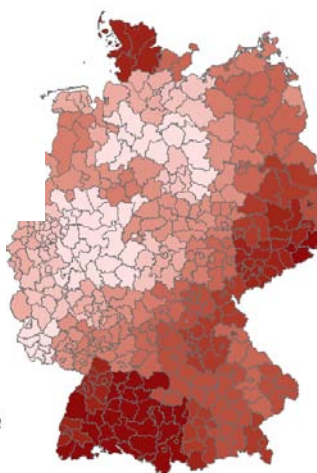
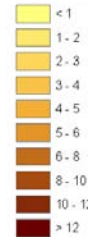
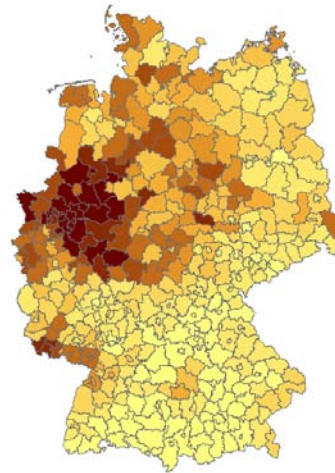
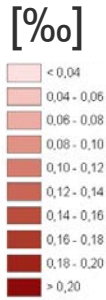
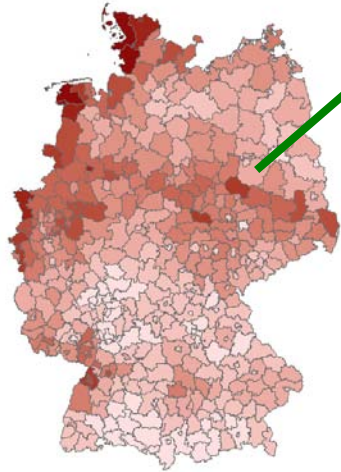
Schadenmodell: ERA-Interim  
Regression je Landkreis

$$\text{Loss ratio (region)} = B(\text{region}) * \sum_{\text{Tage}} \left( \frac{v_{\max}(\text{region}, \text{day})}{v_{98}(\text{region})} - 1 \right)^3 + A(\text{region})$$

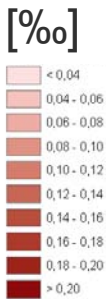
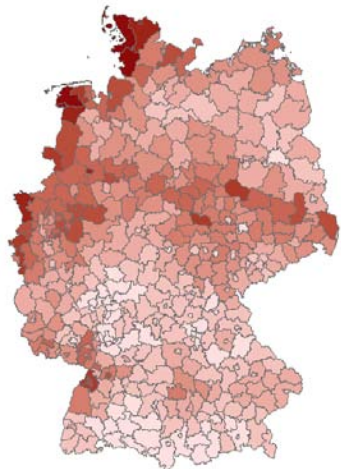
Regression  $y = B*x + A$ : Koeffizient B

Schadenpotential ERA-Interim  
( $v/v_{98} - 1$ )<sup>3</sup>

Regression  $y = B*x + A$ : Koeffizient A



GDV Schadensatz  
Winter ONDJFM



Observed loss ratios from GDV

However,  
Grid-point wind – damage relation  
is **probabilistic** !



Source: GDV



Deutsches Komitee  
Katastrophenvorsorge e.V. (DKKV)

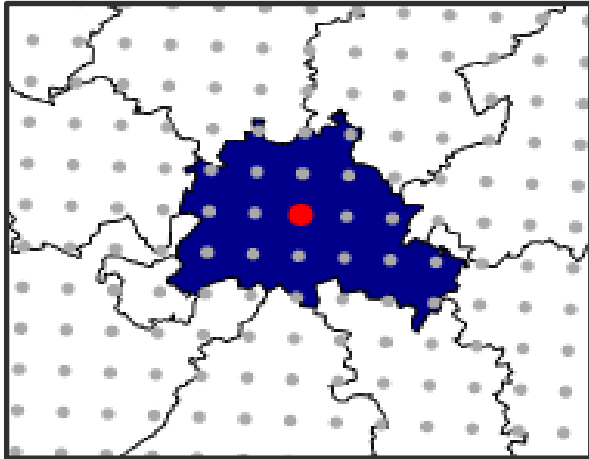
German Committee for Disaster Reduction  
within the International Strategy for Disaster Reduction (ISDR)



Jahresbericht 2006

# A Probabilistic Model for Loss Occurrences

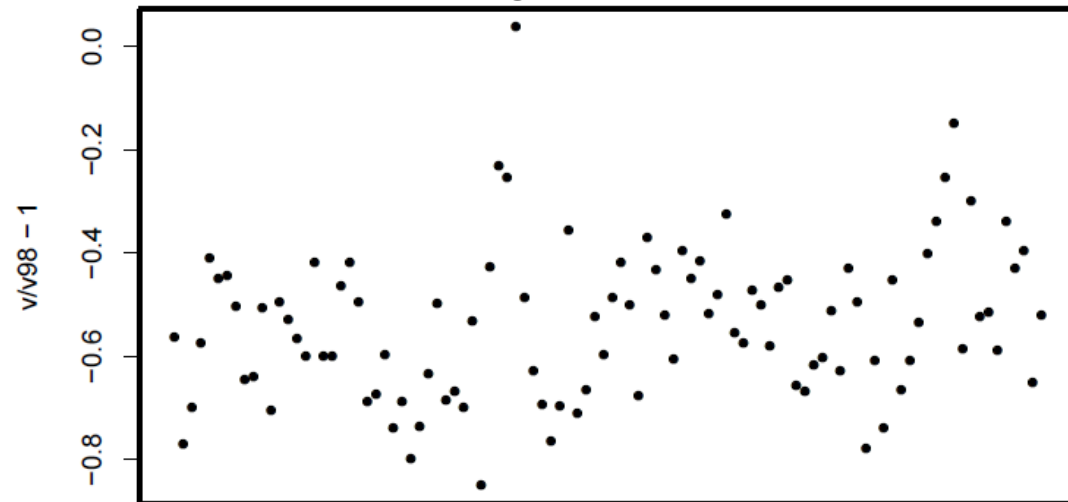
Relate normalized daily max. gusts at district center ...



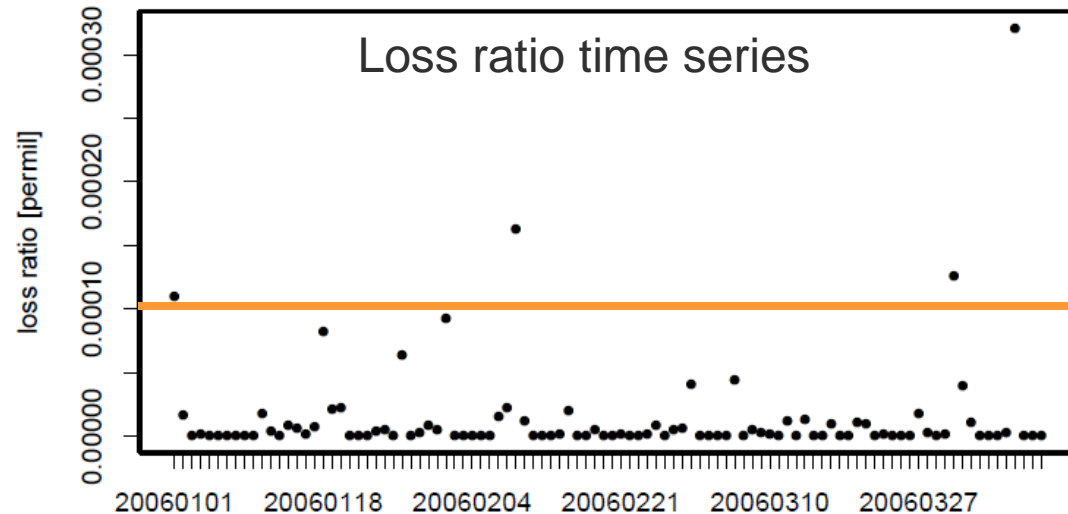
Define threshold according to user requirements,

Determine probability of exceedance dependent on wind speed

gust time series



Loss ratio time series

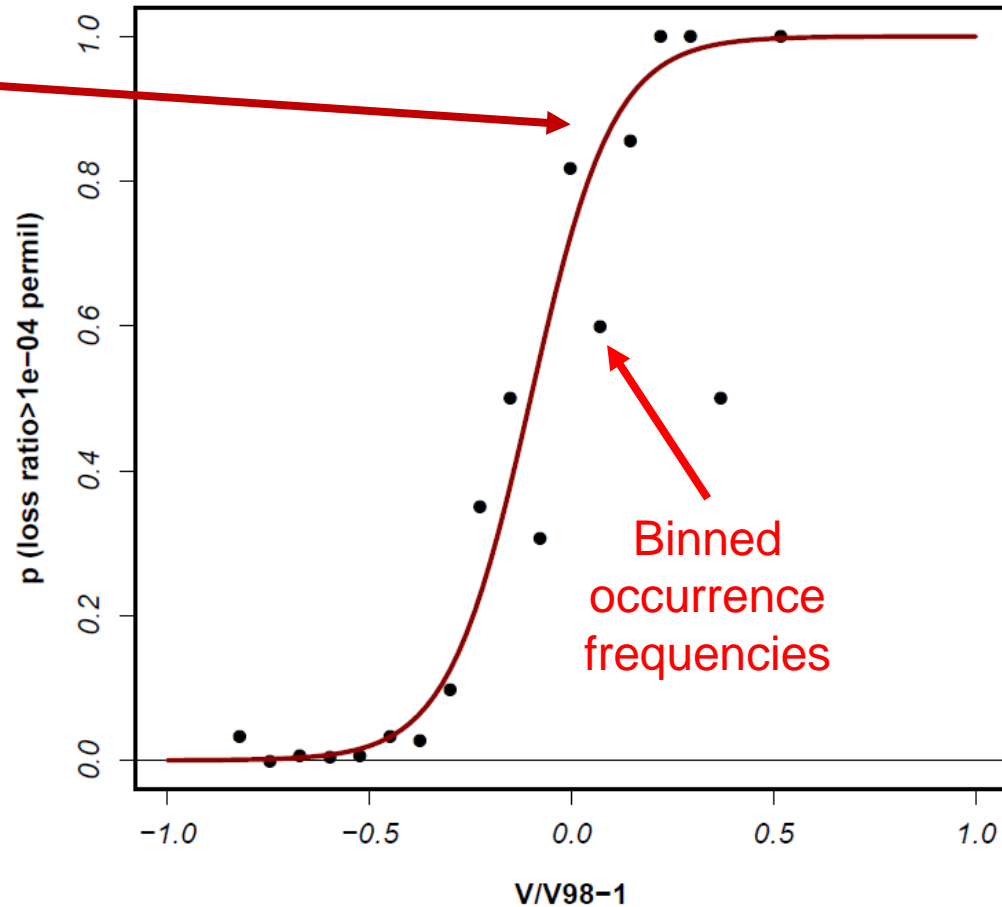


# A Probabilistic Model for Loss Occurrences

Logistic Regression

$$p(event) = \frac{\exp(a + bx)}{1 + \exp(a + bx)}$$

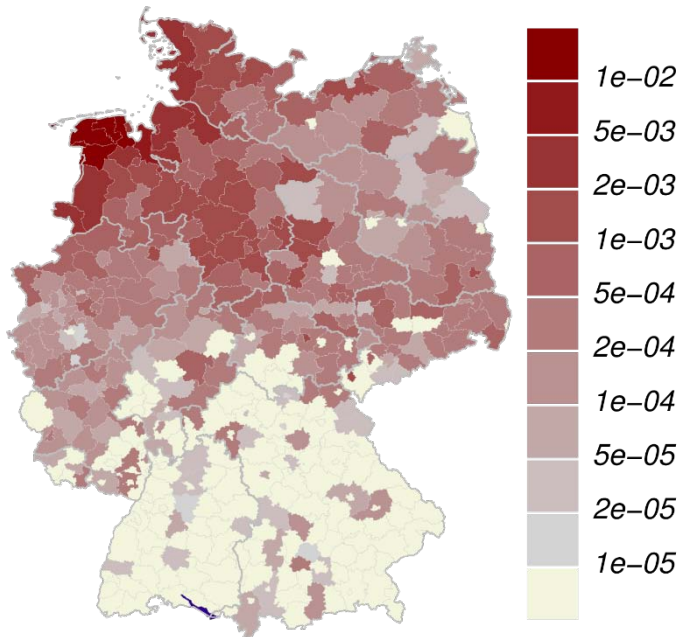
Probability of a loss ratio larger than chosen value



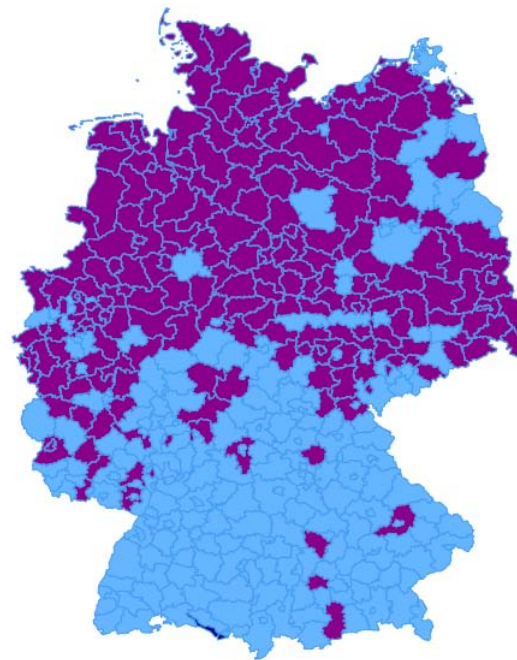
Normalized gust wind speed

# Example – Storm Britta (2006-10-31)

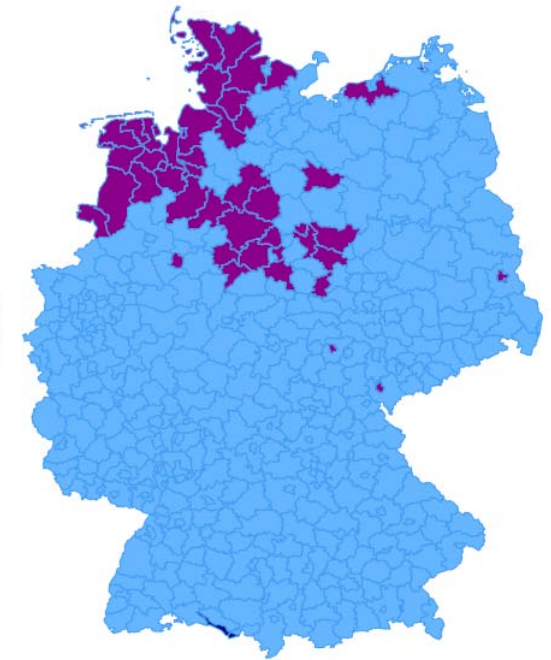
Loss ratio [‰]



Loss ratio > 0.0001



Loss ratio > 0.001



# Example – Storm Britta (2006-10-31)

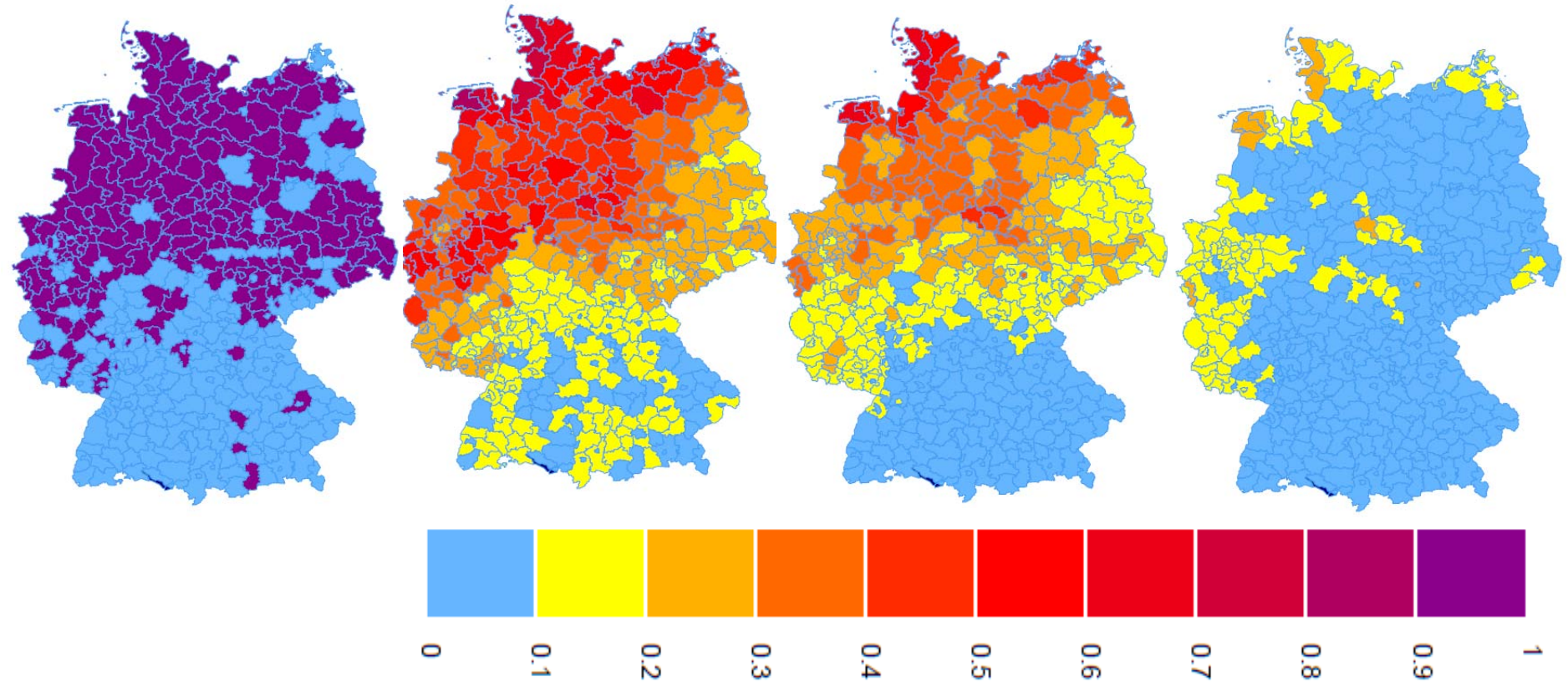
Ensemble mean wind based

Lossratio > 0.0001

Lead time 1 day

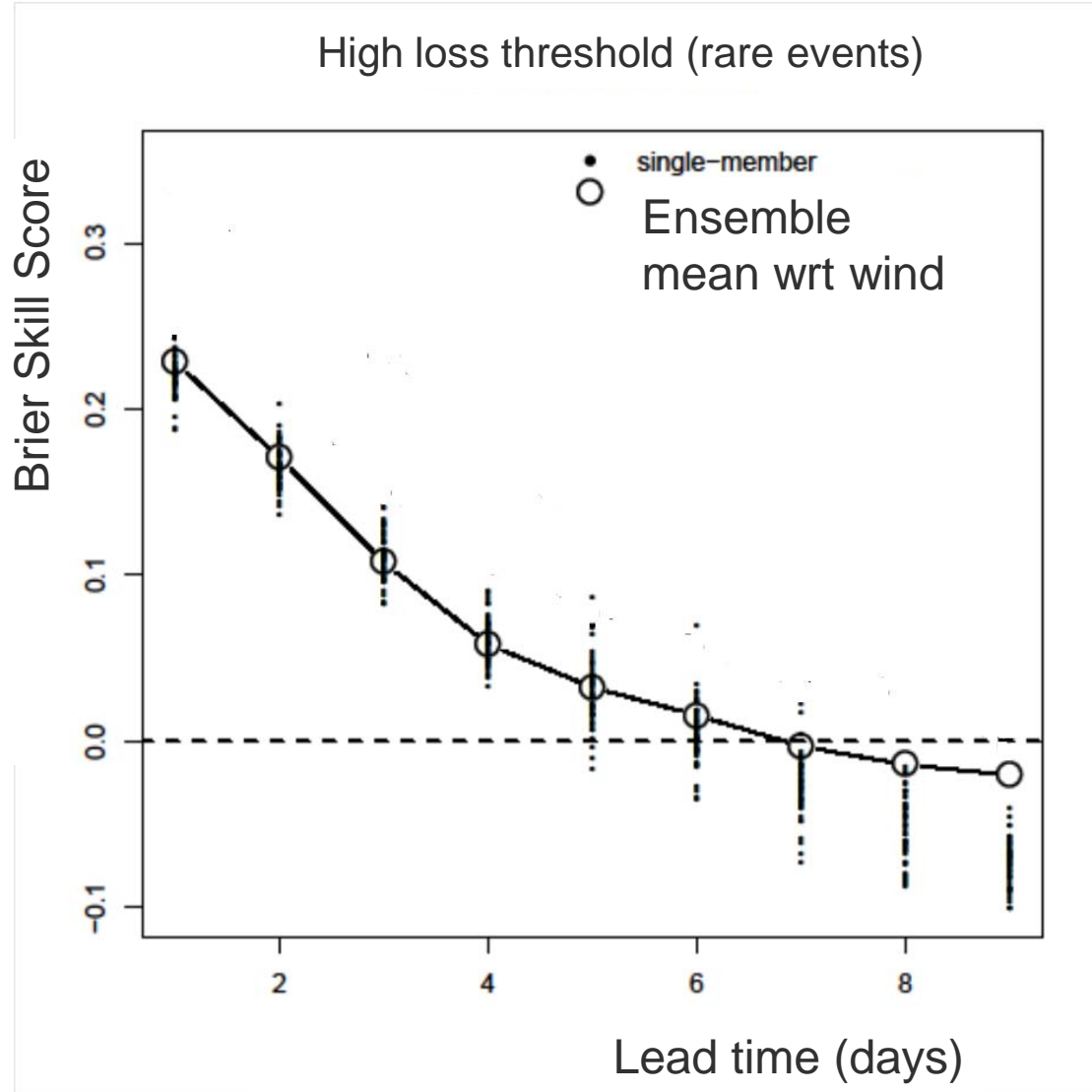
Lead time 3 days

Lead time 9 days





# Brier Skill Score for different lead times



Result for Germany:  
„Skillfull“ forecasts for  
lead times up to 6 days  
i.e.  
better than forecasting  
the climatological  
probability for each day

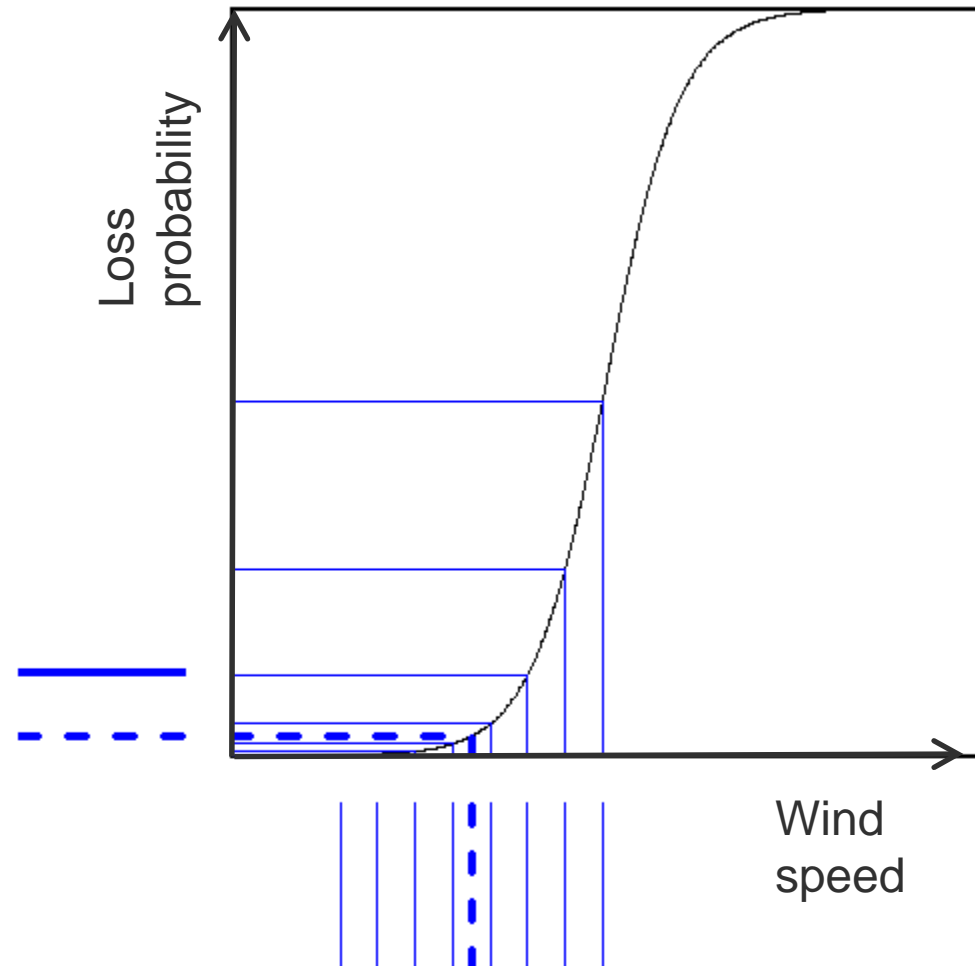
# Use ensemble mean wind speed, or ensemble mean loss probabilities?

## „Wind averaging“

Apply transfer function to ensemble averaged wind speed

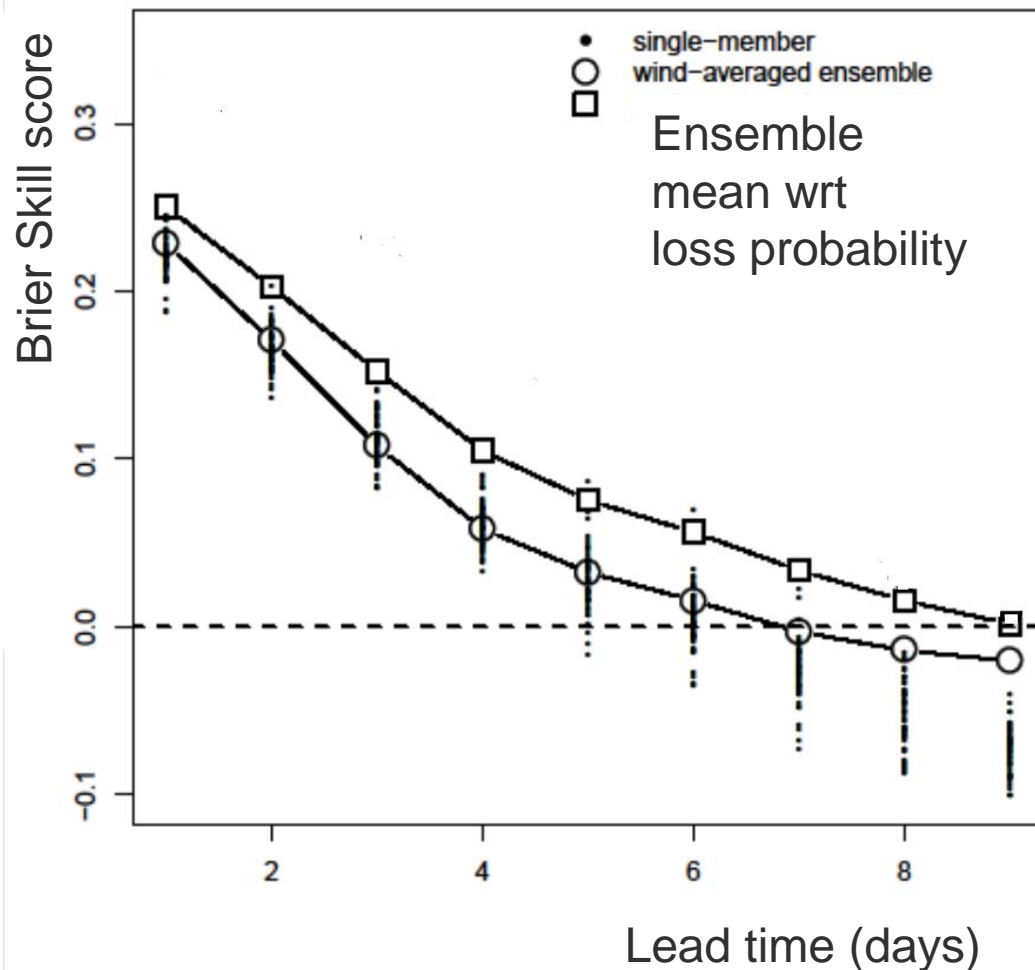
## vs „Loss Probability averaging“

Applying transfer function, then calculating the ensemble average of resulting probabilities



# Brier Skill Score for different lead times

High loss threshold (rare events)

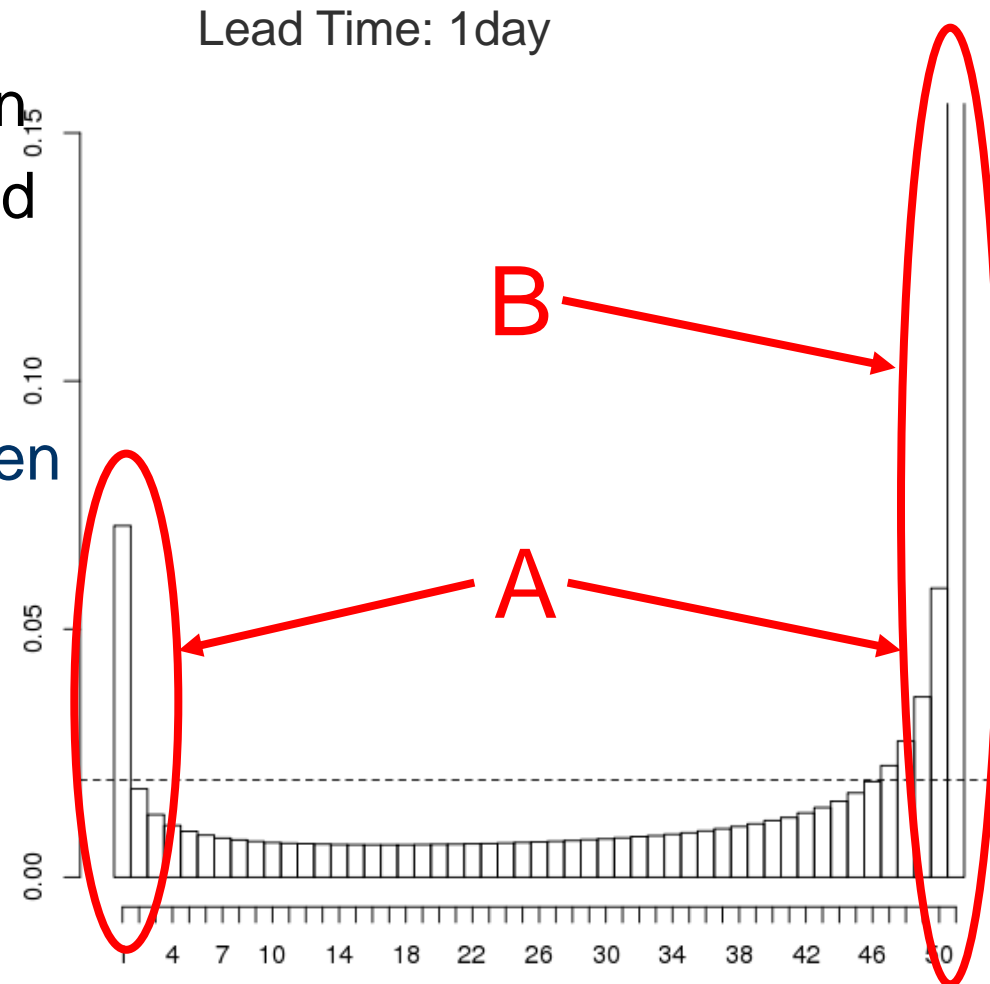


Improved score  
for ensemble mean  
using loss probability  
averaging!

# Underdispersion of downscaled EPS

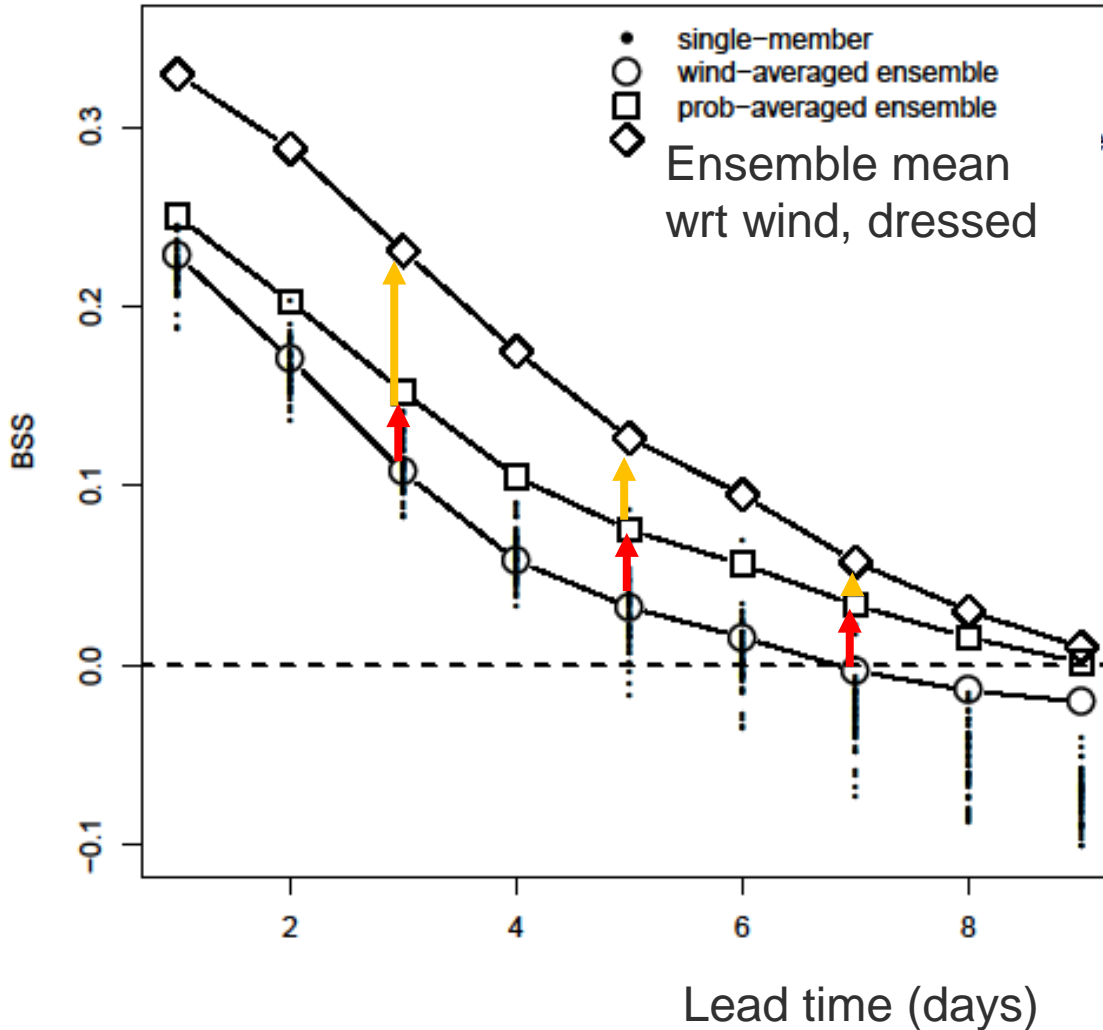
Rank histogram (Talagrand diagram): Where is observation (analysis) wrt the value-ordered ensemble ranks

- Analyses („observations“) often outside of ensemble range  
→ **Underdispersion**
- More frequently above all ensemble members  
→ **Bias**



# Brier Skill Score with ensemble dressing

High loss threshold (rare events)



# Summary

- Estimate relationship of (gridpoint) wind and likelihood of damage in excess of a threshold from observations
- Forecasts using the ensemble mean damage result in better skill wrt damage than forecasts using the wind damage
- Ensemble post processing (“dressing”) yields further gain, up to 2 days

**Thank you for your attention**

