

# The rank histogram

A versatile tool for forecast verification

Zied BEN BOUALLÈGUE

# Outline

1. The ensemble rank histogram
2. A short literature review
3. The observation rank histogram
4. The 2D ensemble rank histogram

# **1. The ensemble rank histogram**

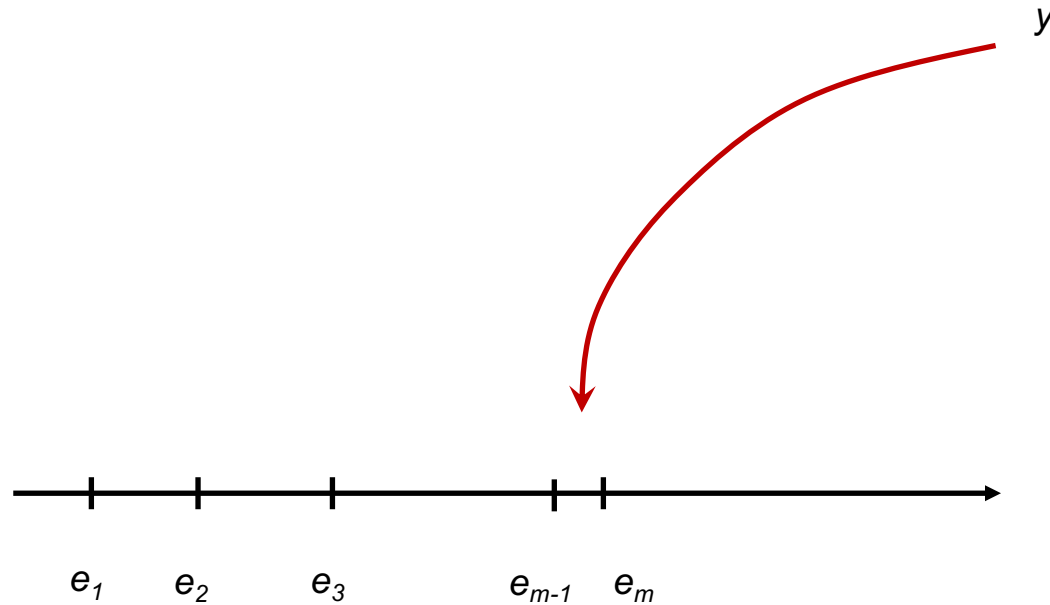
## **The rank histogram**

*a tool for comparing samples from 2 distributions.*

## The ensemble rank histogram

$e_1, e_2, \dots, e_m$ : sorted forecasts over  $m$  ensemble members at point  $(t,i,j)$   
 $y$ : observation corresponding to forecast  $e$

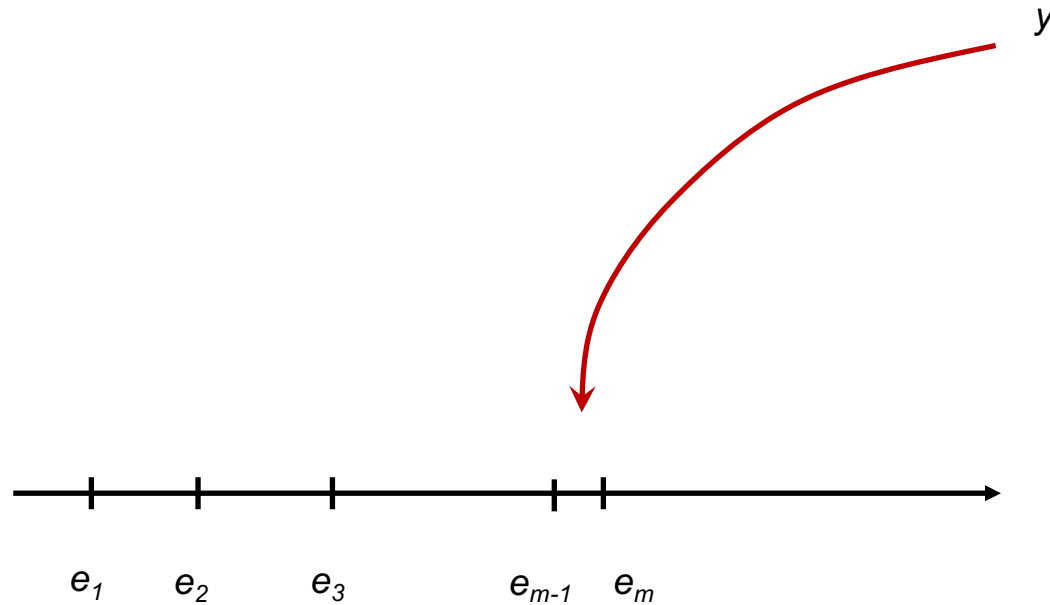
for all  $t,s$



## The ensemble rank histogram

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 $y$ : observation corresponding to forecast  $e$

for all  $t,s$



Complementary tool:  
spread/skill diagram

## **2. A short\* literature review**

**\* and subjective**

Hamill, T. M., **2001**: Interpretation of Rank Histograms for Verifying Ensemble Forecasts. *Mon. Wea. Rev.*, 129, 550–560

Keller, J.D., Hense, A., **2011** : A new non-Gaussian evaluation method for ensemble forecasts based on analysis rank histograms. *Met. Zet.* 20, 107-117

Bröcker J, Ben Bouallègue Z., **2020**: Stratified rank histograms for ensemble forecast verification under serial dependence. *QJR Meteorol Soc.* 146: 1976-1990.

Allen, S., Ziegel, J., Ginsbourger, D., **2024**: Assessing the calibration of multivariate probabilistic forecasts. *QJR Meteorol Soc.*, 1–21.



Hamill, T. M., 2001: Interpretation of Rank Histograms for Verifying Ensemble Forecasts.  
*Mon. Wea. Rev.*, 129, 550–560

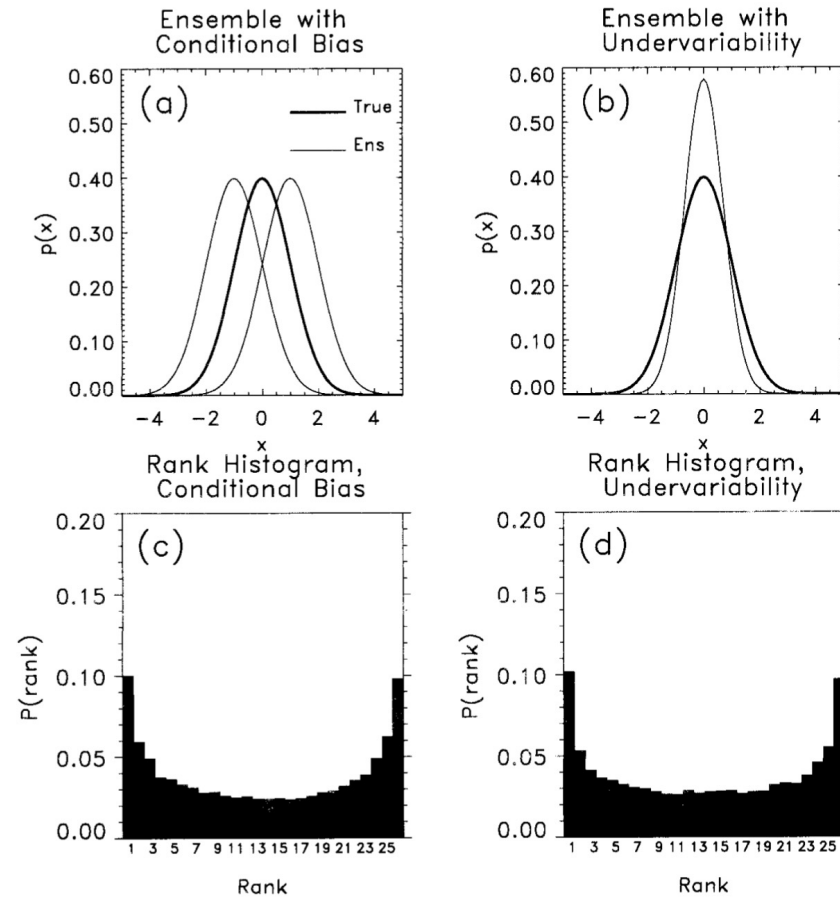
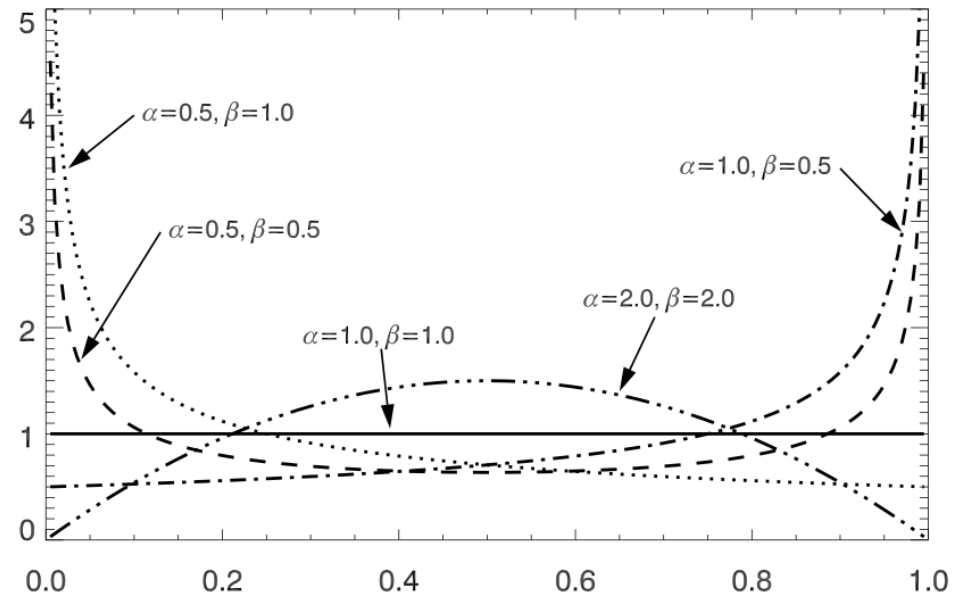


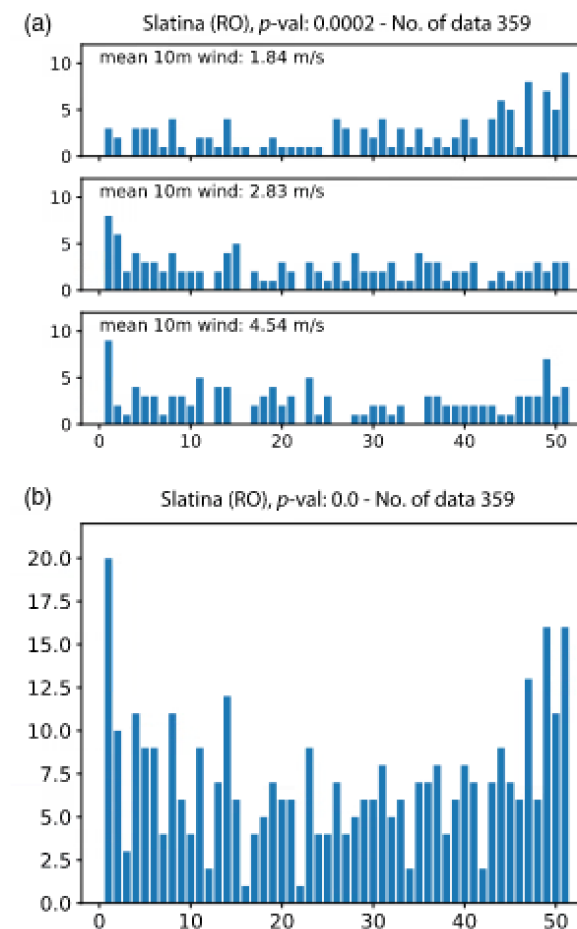
FIG. 4. (a) As in Fig. 2a, but where ensemble is selected with equally likely probability from one of the two biased distributions, a  $N(-1, 1)$  or  $N(1, 1)$  distribution, with the verification tallied 10 000 times for each distribution. (b) As in (a), but where ensemble forecasts are selected from a probability distribution with a lack of variability,  $N(0, 0.69)$ . (c) Rank histogram corresponding to (a). (d) Rank histogram corresponding to (b). Verification rank tallied 20 000 times.

Keller, J.D., Hense, A., 2011 : A new non-Gaussian evaluation method for ensemble forecasts based on analysis rank histograms.  
*Met. Zet.* 20, 107-117



**Figure 2:** Examples of the probability density of the  $\beta$ -function for different combinations of the parameters  $\alpha$  and  $\beta$ .

Bröcker J, Ben Bouallègue Z., 2020: Stratified rank histograms for ensemble forecast verification under serial dependence. *QJR Meteorol Soc.* 146: 1976-1990.



**FIGURE 3** (a) Stratified rank histograms, (b) unstratified rank histogram, and (c) corresponding covariance matrix  $\Upsilon$  for Salla (Finland). Stratification is based on averaged forecast and observed 2 m temperature. The average of this quantity within the stratum is indicated in each sub-panel of (a). The  $p$ -value of the reliability test and the sample size (number of forecast–observation pairs) are indicated above (a) and (b). The unstratified histogram shows a warm forecast bias, and the stratified histograms indicate that this is confined to cold conditions [Colour figure can be viewed at [wileyonlinelibrary.com](http://wileyonlinelibrary.com)]

Allen, S., Ziegel, J., Ginsbourger, D., 2024: Assessing the calibration of multivariate probabilistic forecasts. *QJR Meteorol Soc.*, 1–21.

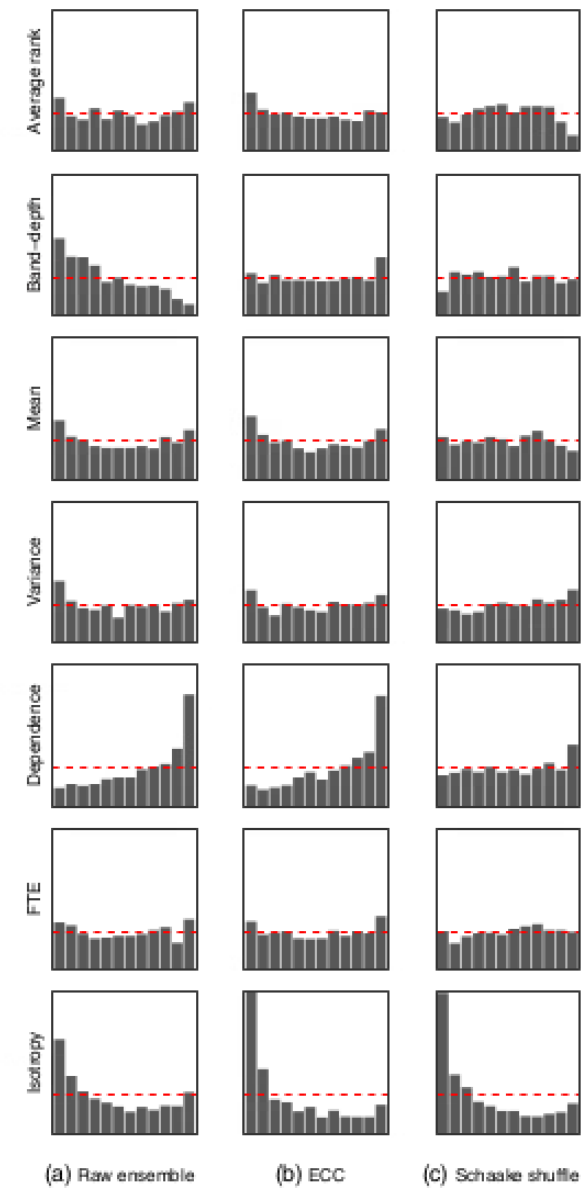


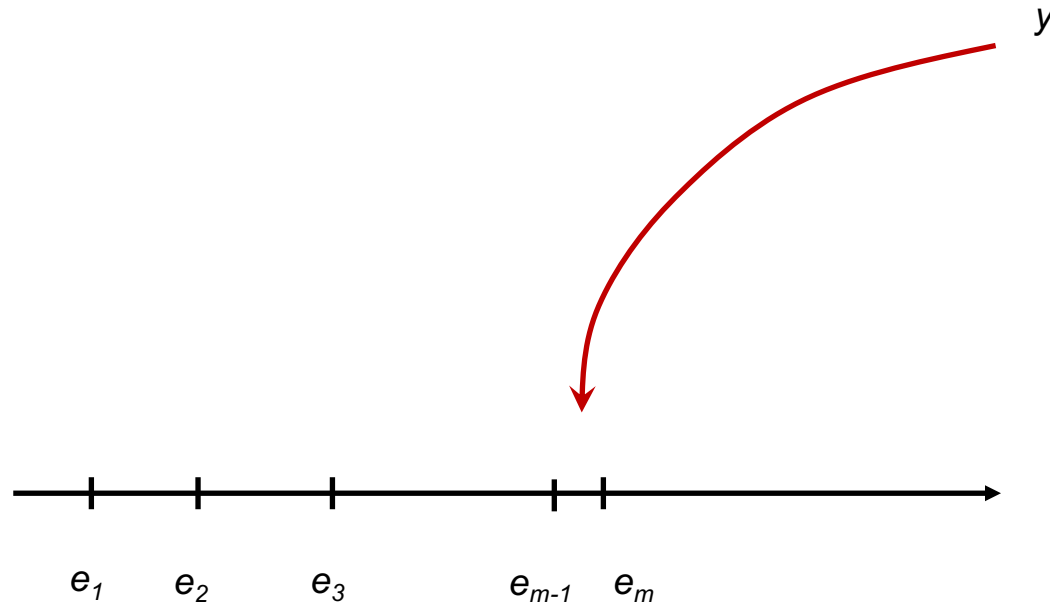
FIGURE 5 Multivariate rank histograms constructed using seven pre-rank functions for (a) the raw ensemble forecasts and (b, c) the post-processed forecasts reordered using (b) ensemble copula coupling (ECC) and (c) the Schaake shuffle. The dashed line indicates a uniform histogram. The y-axis limits of all plots are the same. [Colour figure can be viewed at [wileyonlinelibrary.com](https://onlinelibrary.wiley.com)]

### **3. The observation rank histogram**

## The ensemble rank histogram

$e_1, e_2, \dots, e_m$ : sorted forecasts over  $m$  ensemble members at point  $(t,s)$   
 $y$ : observation corresponding to forecast  $e$

for all  $t,s$

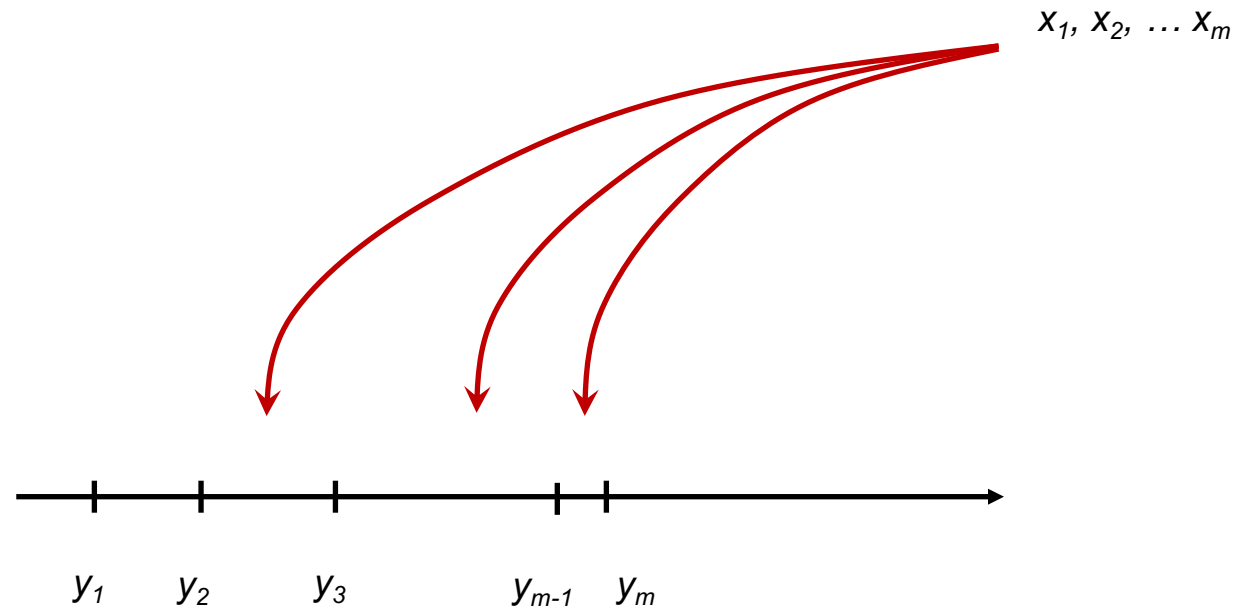


## The observation rank histogram

$y_1, y_2, \dots, y_m$ : sorted observation over  $m$  dates at station  $s$

$x_1, x_2, \dots, x_m$ : forecasts over  $m$  dates at station  $s$

for all  $s$

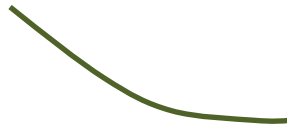


## The observation rank histogram

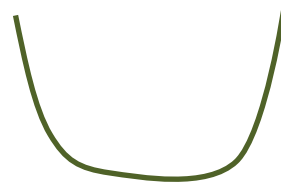
Interpretation (stand-point = observation) :



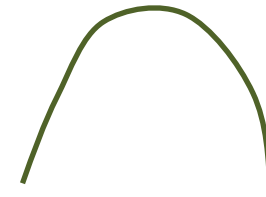
*(marginal) calibration*



*(negative) bias*



*over-activity*



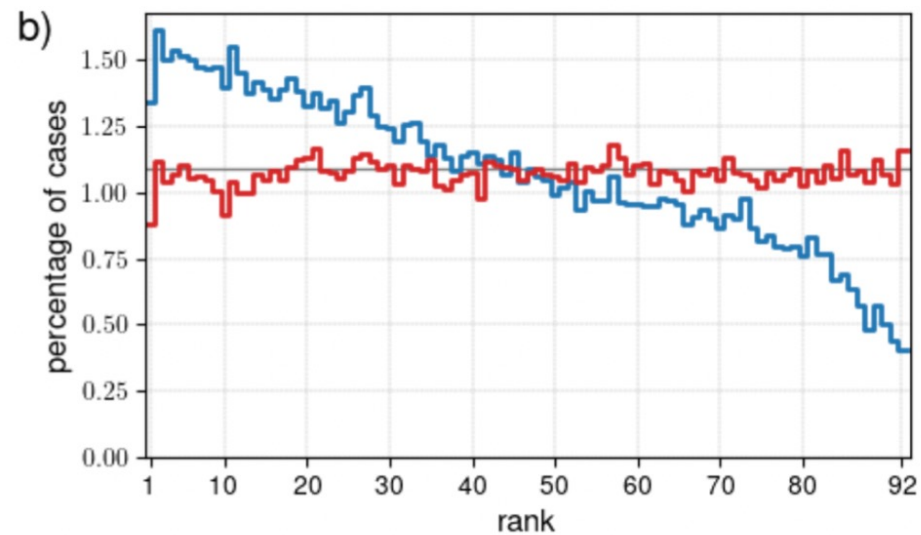
*under-activity*



## The observation rank histogram

Example:

*2m temperature, Summer 2022, day 6, Europe*  
**IFS** PanguWeather

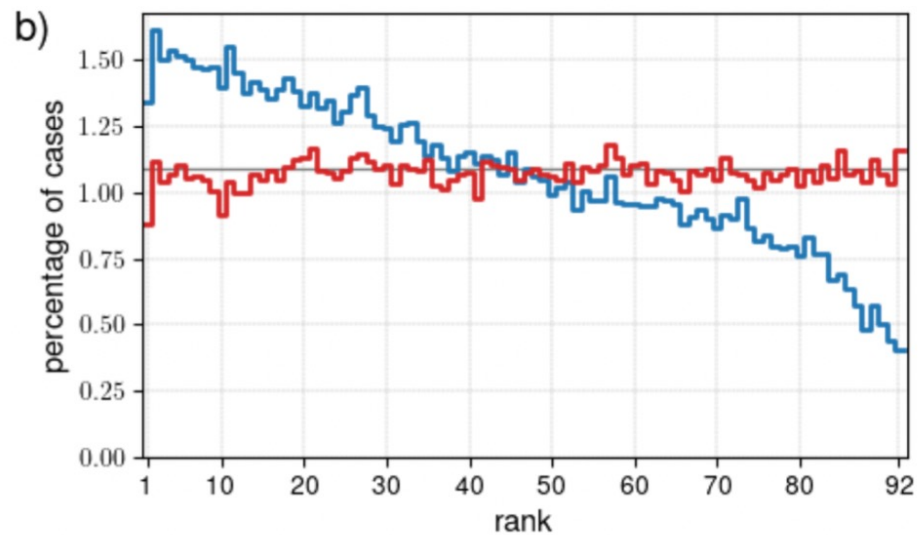


Ben-Bouallegue et al (2024)  
[10.1175/BAMS-D-23-0162.1](https://doi.org/10.1175/BAMS-D-23-0162.1)

## The observation rank histogram

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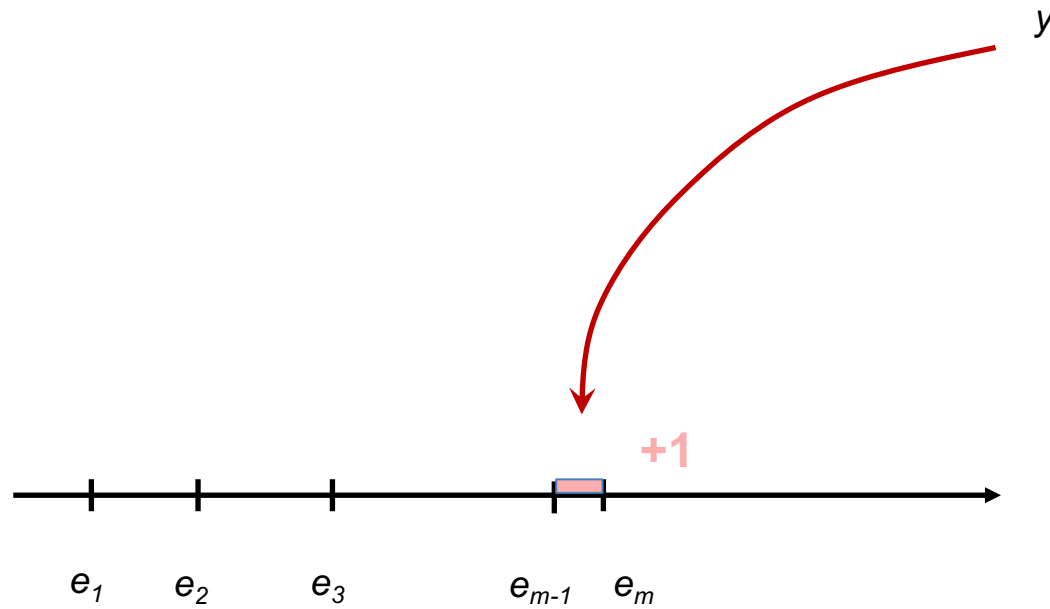
**Complementary tool:**  
Q-Q plot

## **4. The 2D ensemble rank histogram**

## The ensemble rank histogram

$e_1, e_2, \dots, e_m$ : sorted forecasts over  $m$  ensemble members at point  $(t,s)$   
 $y$ : observation corresponding to forecast  $e$

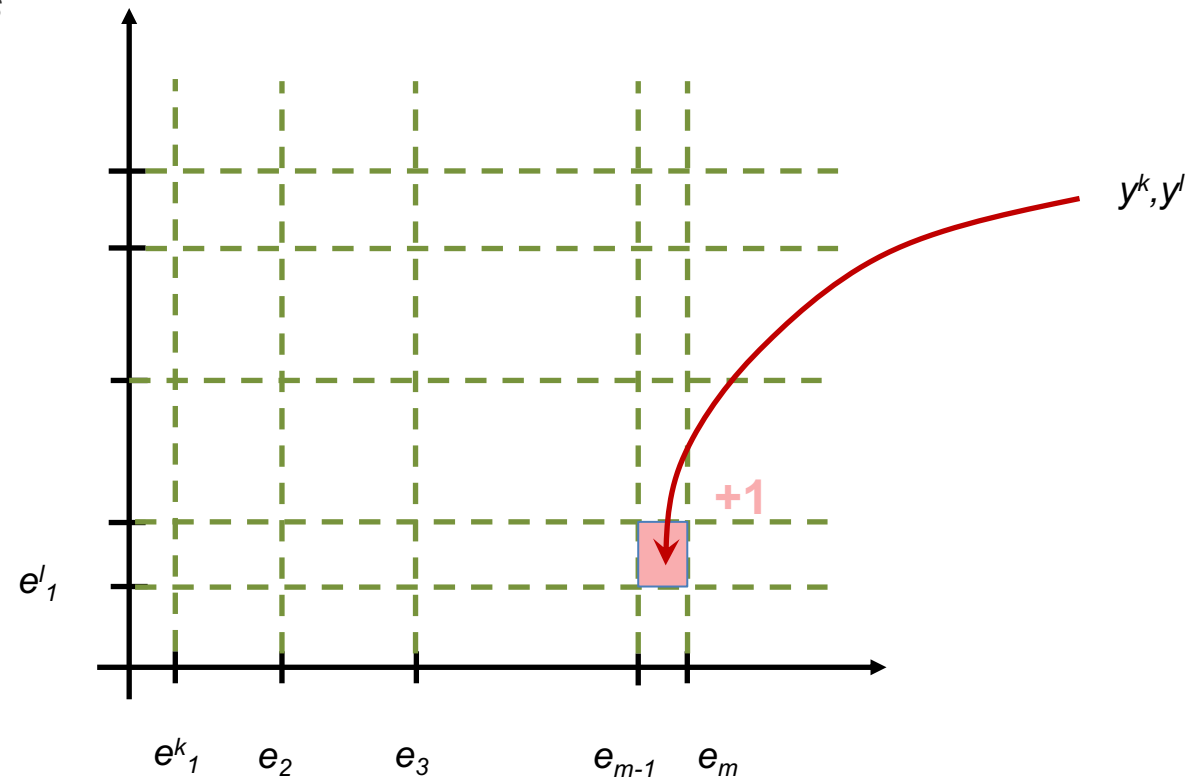
for all  $t,s$



## The 2D ensemble rank histogram

$e_1, e_2, \dots, e_m$ : sorted forecasts over  $m$  ensemble members at point  $(t,s)$   
 $y$ : observation corresponding to forecast  $e$

for all  $t,s$



## The 2D ensemble rank histogram

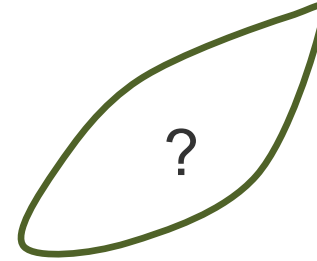
Interpretation

**in 1D**



*calibrated*

**in 2D**



*it depends! (not known a-priori)*

## The 2D ensemble rank histogram

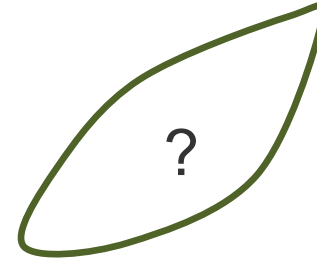
Interpretation

in 1D



*calibrated*

in 2D



*it depends! (not known a-priori)*

First, we need to get a reference doing the test with the ensemble (**only**) :

$\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_{m-1}$ : sorted forecasts over  $m$  ensemble members at point  $(t,s)$

$\mathbf{e}_m$ : pseudo-observation corresponding to member  $m$

for all  $t,s$

## The 2D ensemble rank histogram

Interpretation

**Sklar's theorem:**

***F***: Marginals

$$G(y_1, \dots, y_L) = C(F_1(y_1), F_L(y_L))$$

***G***: Multivariate  
cumulative distribution  
function

***C***: Copula



## The 2D ensemble rank histogram

### Examples:

1. Z500 at a zonal distance of 10 deg. ----- **in space**
2. Z500 at day 5 and day 6 ----- **in time**
3. U850 and V850 ----- **intervariable**

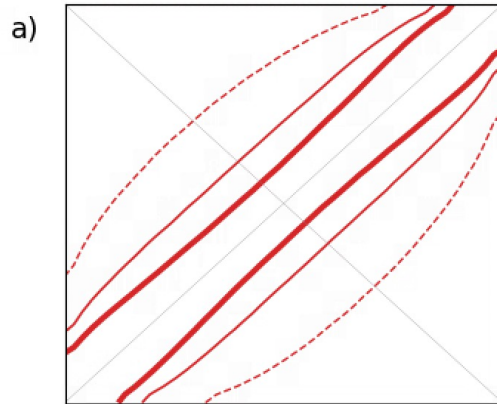
**IFS ensemble, JJA 2023**

## The 2D ensemble rank histogram

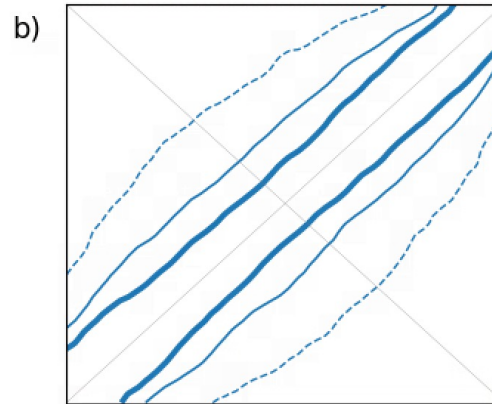
[stand-point = 2D ensemble]

1. Z500 at a zonal distance of 10 deg. (S. Hem.)

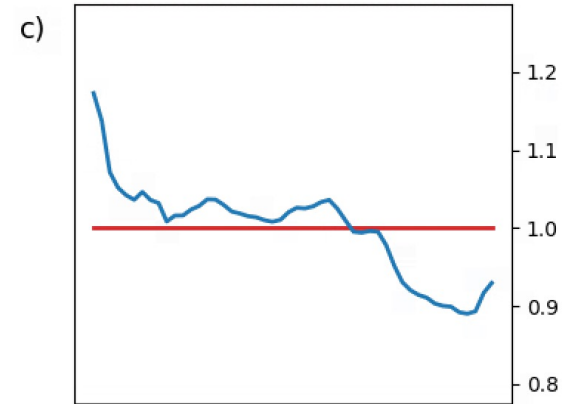
Ensemble copula



2D rank histogram



1D rank histogram



**Red:** reference

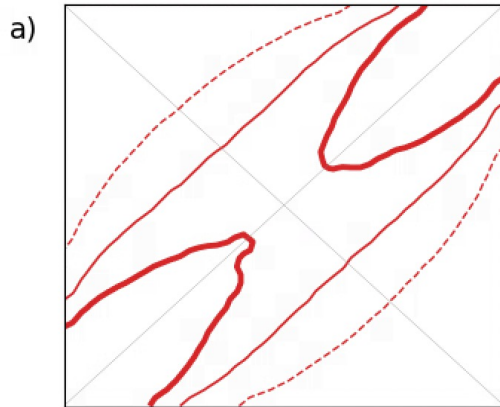
**Blue:** observation

## The 2D ensemble rank histogram

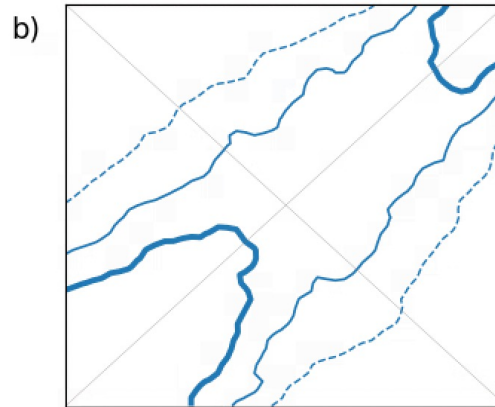
[stand-point = 2D ensemble]

### 2. Z500 at day 5 and day 6 (N. Hem.)

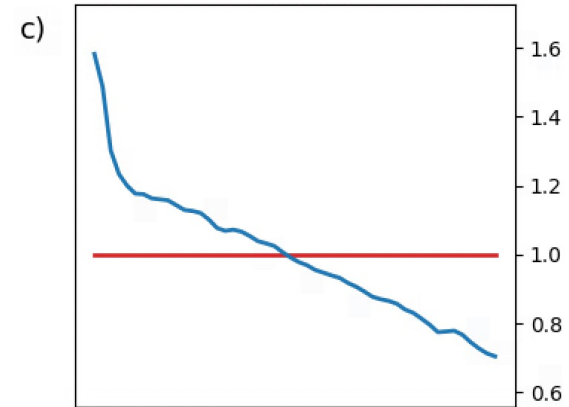
Ensemble copula



2D rank histogram



1D rank histogram



**Red:** reference

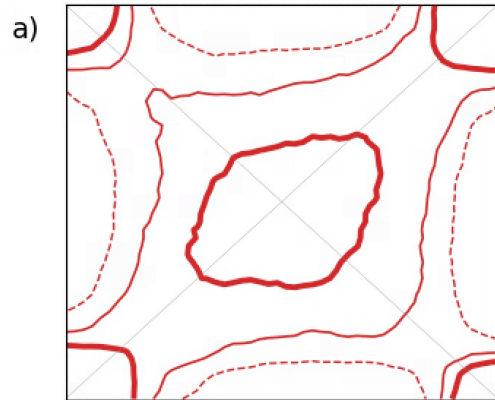
**Blue:** observation

## The 2D ensemble rank histogram

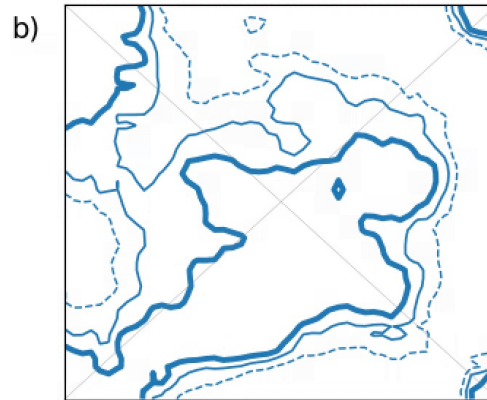
[stand-point = 2D ensemble]

### 3. U and V at 200hPa (N. Hem.)

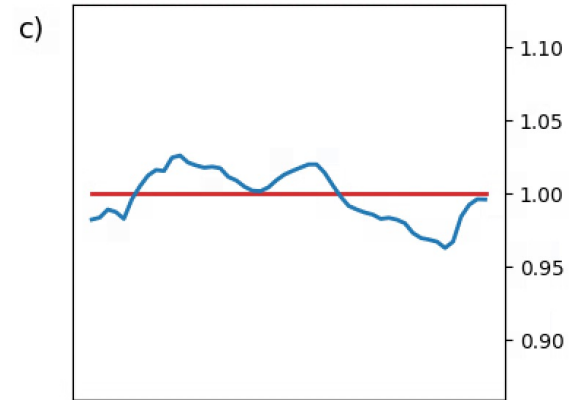
Ensemble copula



2D rank histogram



1D rank histogram



**Red:** reference

**Blue:** observation

## SUMMARY / Outlook

Rank histograms are **versatile tools** for exploring and comparing forecast and observation **statistical structures**, also in a **multivariate space**.

- **The observation rank histogram** is a complementary tool to the Q-Q plot. It can be used to assess **forecast activity** and forecast ability to capture **observed extremes**.
- **The 2D ensemble rank histogram** is a **generalisation** of the ensemble rank histogram to assess **dependences** in bivariate ensemble forecasts, using the **ensemble copula** as a reference.
- **Outlook:**
  - Statistical **significance test** for 2D rank histograms?
  - **2D observation rank histogram** to compare observation and forecast “climatological” copulas