



# Seasonal Forecasting Workshop on agro-hydro-climatic characteristics of the main rainfall season in the Gulf of Guinea countries / PRESAGG -11

Accra, GHANA

February 26 to March 01, 2024



## Principles of quantiles and their use in seasonal forecast verification

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An initiative of the Organisation of African, Caribbean and Pacific States funded by the European Union





# Outline

What and why LRF verification

Types of LRF verification methods

Sample products and interpretation

How to generate the verification products

Demonstration and exercises



# What is LRF verification

- *Seasonal Forecast are usually made in term of “probabilistic forecast”*
- *So the aim here is to provide you with the tools to measure the attribute of a probabilistic forecast*
- **An accurate probability forecast system has:**
  - Reliability - agreement between forecast probability and mean observed frequency*
  - Sharpness - tendency to forecast probabilities near 0 or 1, as opposed to values clustered around the mean*
  - Resolution - ability of the forecast to resolve the set of sample events into subsets with characteristically different outcomes*



# Why LRF verification (1/3)

Why we do seasonal forecast verification?

- Testing if the forecast is “consistent”
- Testing if the forecast has “quality” if it corresponds to what happened.
- Testing if the forecast has “value” if it can be used to help realize some benefit, whether economic, social, or otherwise.



# Why LRF verification (2/3)

**Without information about the quality of the forecasts  
how is anybody to know whether to believe them?**

*It's on the forecaster to demonstrate that his products are worth taking note of.*

**Provide fairly detailed information about the quality of forecasts**

*Go beyond asking "Can these forecasts be believed?" to address questions such as "How can these forecasts best be used?" and "How can these forecasts be improved?"*



# Why LRF verification (3/3) - Benefits

**A forecast is like an experiment** -- *given a set of conditions, you make a hypothesis that a certain outcome will occur.*

The process shouldn't consider to be complete until you find out whether the forecast was successful.

The three most important reasons to verify forecasts are:

- to **monitor forecast quality** - how accurate are the forecasts and are they improving over time?
- to **improve forecast quality** - the first step toward getting better is discovering what you're doing wrong.
- to **compare the quality of different forecast systems** - to what extent does one forecast system give better forecasts than another, and in what ways is that system better?



# Types of forecast performance evaluation (verification)

## 1. Visual / qualitative verification

- A forecast is “consistent”

## 2. Quantitative verification

- A forecast has “quality” if it corresponds to what happened.

## 3. Impact Assessment

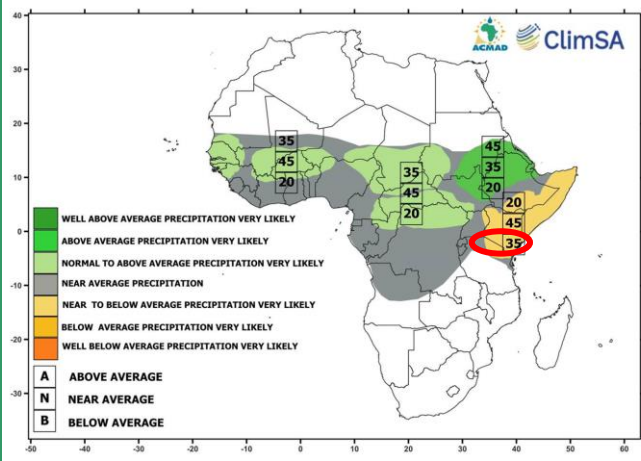
- A forecast has “value” if it can be used to help realize some benefit, whether economic, social, or otherwise.



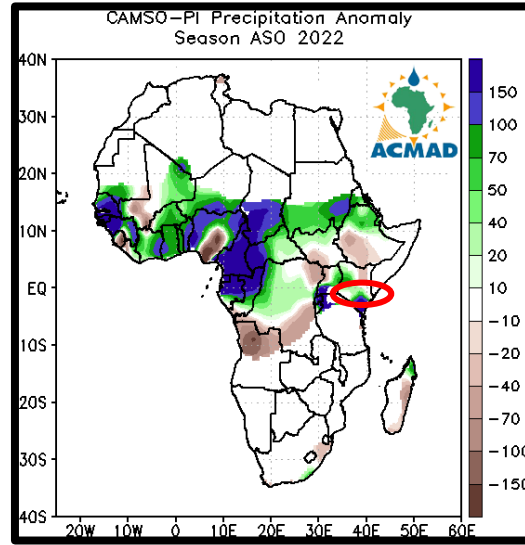
# Types of verification (1/3)

## Visual evaluation of a forecast

SEASONAL PRECIPITATION FORECAST  
FOR AUGUST-SEPTEMBER-OCTOBER 2022  
ISSUED ON JULY 25, 2022

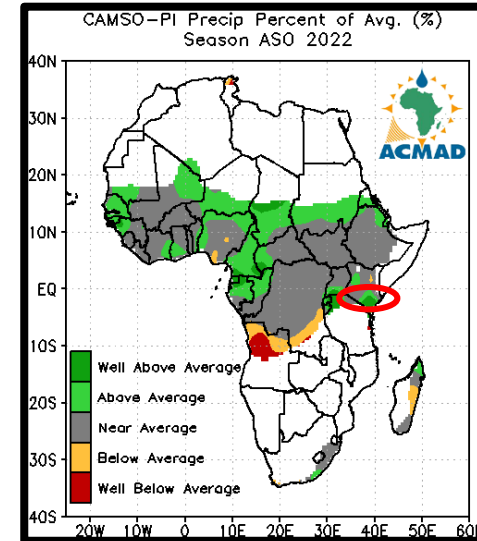


CAMS0-PI Precipitation Anomaly  
Season ASO 2022



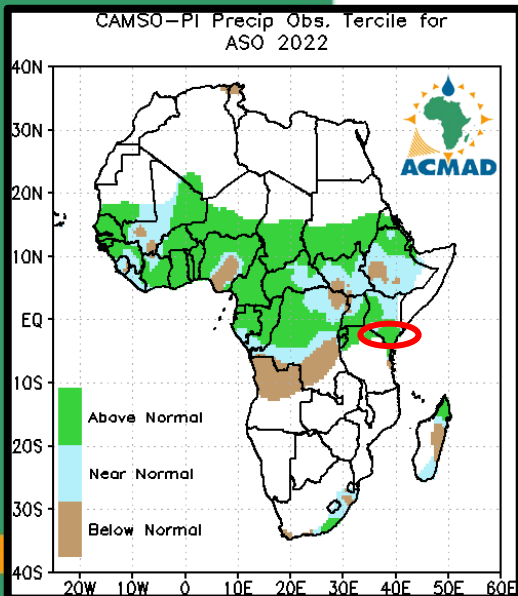
**Anomaly** – used to visually detect the shift (from the mean) of observed precip

CAMS0-PI Precip Percent of Avg. (%)  
Season ASO 2022



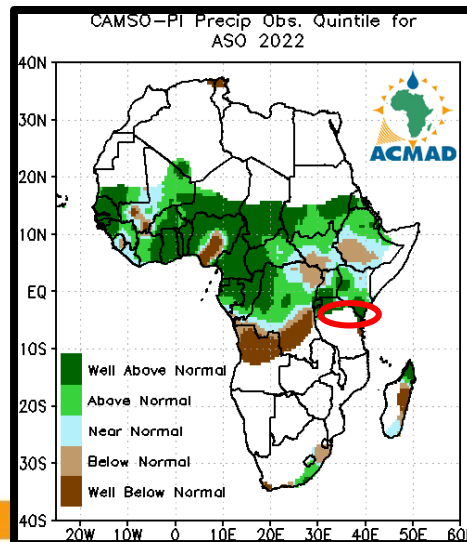
**Percent of Average** – Used to detect magnitude of observed shift from the mean

CAMS0-PI Precip Obs. Tercile for  
ASO 2022



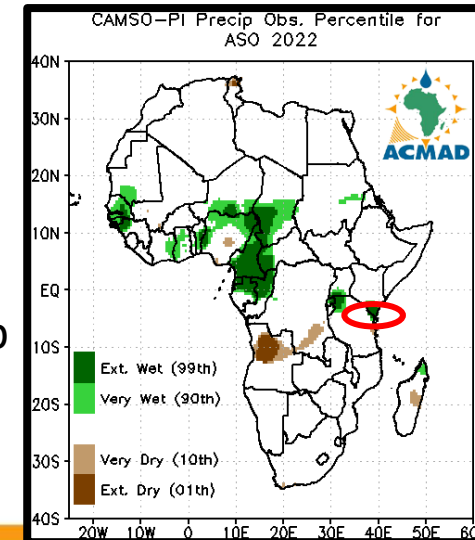
**Tercile** – used to visually detect spatial shape of observed precip

CAMS0-PI Precip Obs. Quintile for  
ASO 2022



**Quintile** – used to visually detect the magnitude of observed precip

CAMS0-PI Precip Obs. Percentile for  
ASO 2022



**Percentile** – used to visually detect magnitude of extreme observed precip (extremely wet and dry areas)



# Types of verification (2/3) - Quantitative verification

## What are we trying to detect by verification of LRF

### False alarm:

A warning or forecast issued for an event that did not actually occur.

### **False-alarm rate (FAR):**

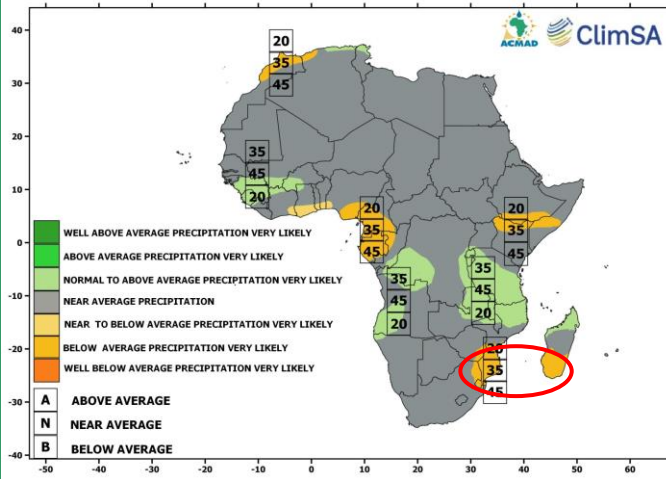
A measure of the quality of deterministic forecasts; specifically, the number of false alarms divided by the number of non-events. The FAR measures the proportion of non-events that were incorrectly forewarned, and should be distinguished from the false-alarm ratio, which measures the proportion of incorrect warnings.

### Hit:

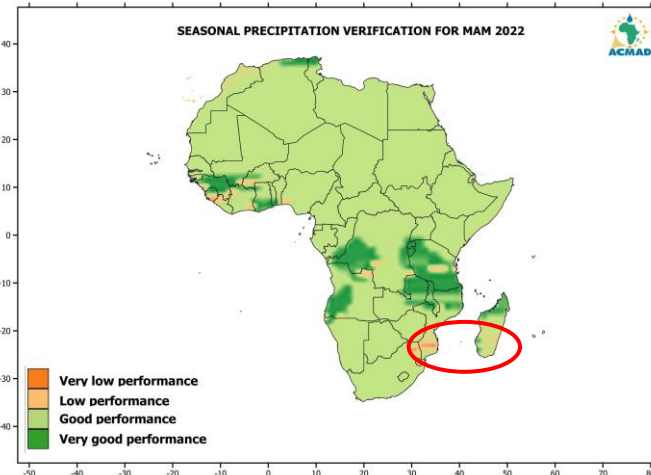
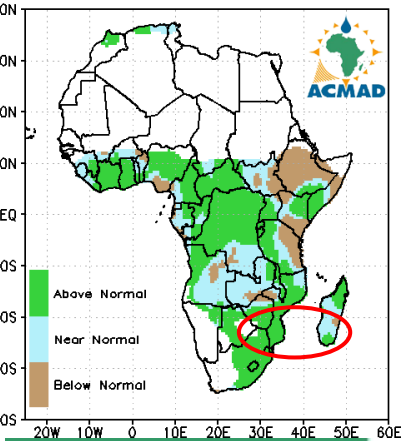
A warning or forecast issued for an event that occurs.

### **Hit rate (HR):**

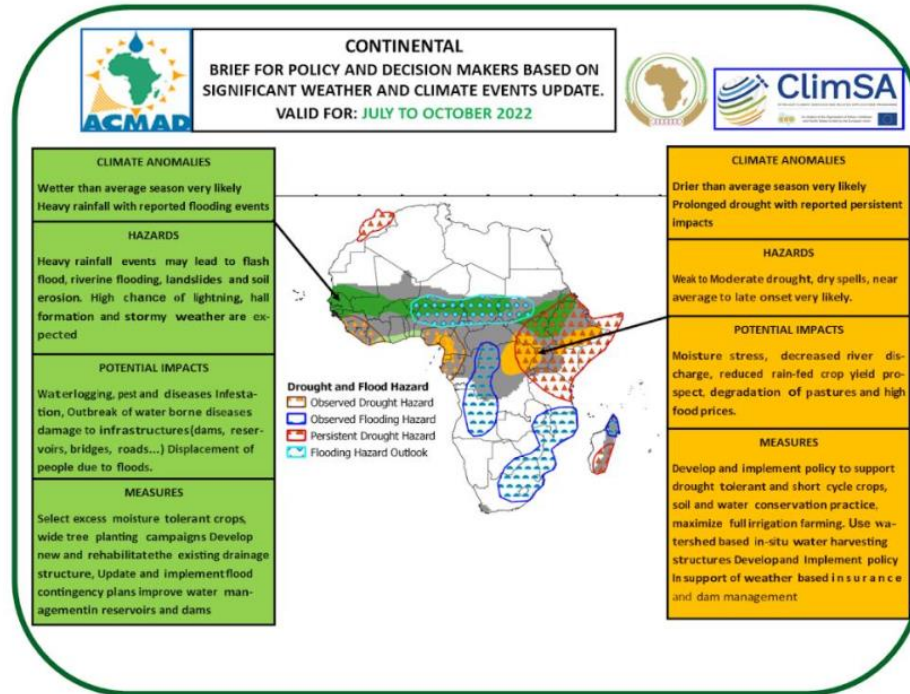
A measure of the quality of deterministic forecasts; specifically, the number of hits divided by the number of events. The HR measures the proportion of events that were forewarned, and should be distinguished from the hit score, which measures the proportion of correct warnings.



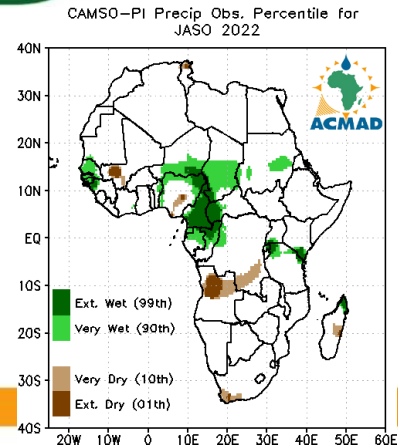
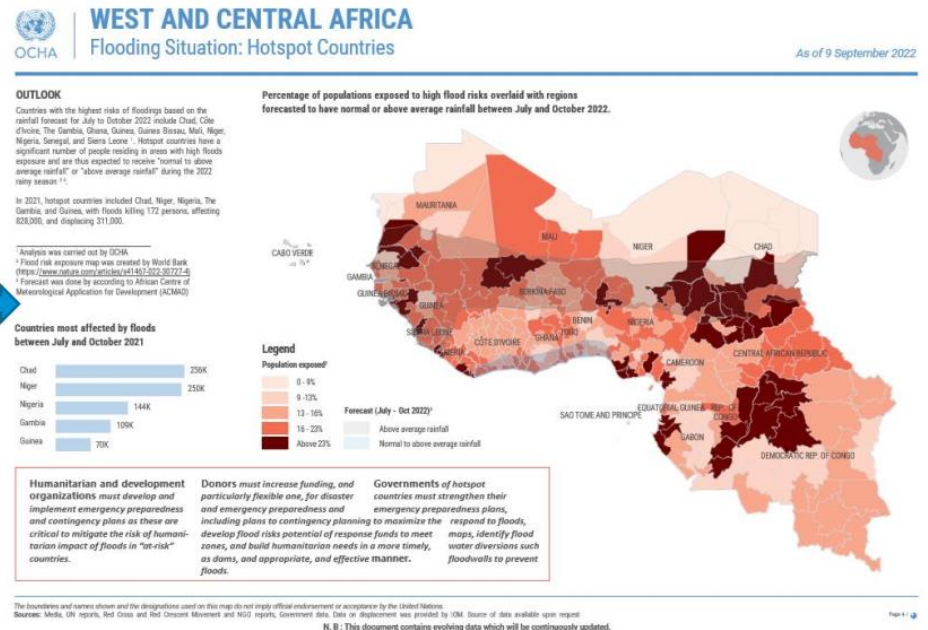
CAMS0-PI Precip. Obs. Tercile for MAM 2022



## Impact Assessment



## Seasonal Forecast overlaid on population density

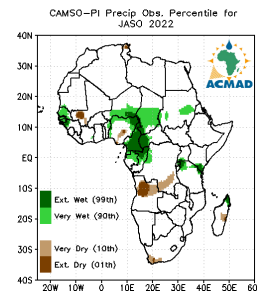
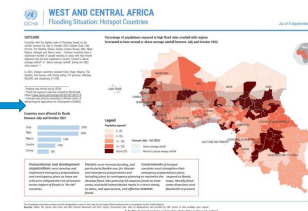
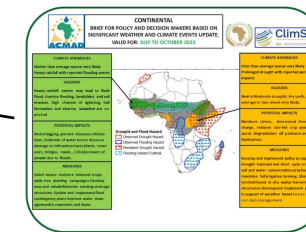
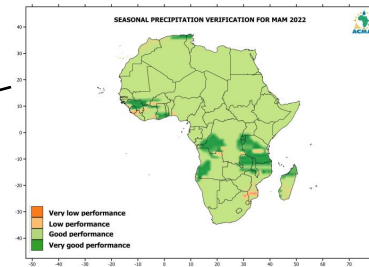
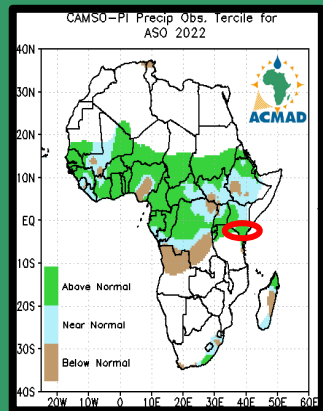
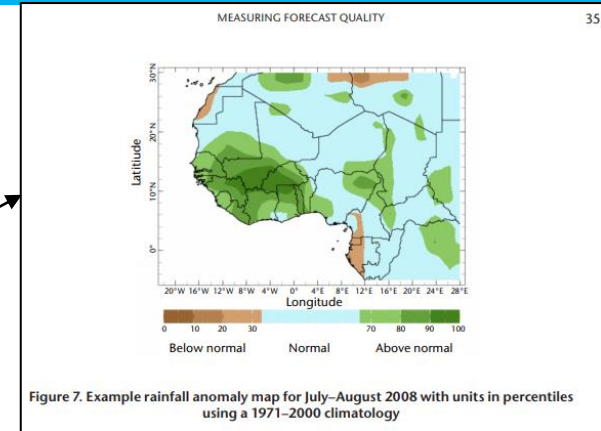


# How are the verification and forecast maps produced?



## LRF Verification:

- Verification scores:-
  - WMO Guidelines on verification of LRF: [library.wmo.int/doc\\_num.php?explnum\\_id=4886](http://library.wmo.int/doc_num.php?explnum_id=4886)
- Commonly used is the Ranked probability skill score (RPSS) and Heidke Skill Score (HSS)
- **Eyeballing**
- Socio-economic impact assessment





# Demonstration and exercises



# Quantile q-quantile computation

## Computing Quantiles or q-quantile

Count the number of observations in the dataset ( $n$ ).

Sort the observations from smallest to largest.

Find the first q-quantile:

- Calculate  $n * (1 / q)$ .
- If  $n * (1 / q)$  is an integer, then the first q-quantile is the [mean](#) of the numbers at positions  $n * (1 / q)$  and  $n * (1 / q) + 1$ .
- If  $n * (1 / q)$  is not an integer, then round it up. The number at this position is the first q-quantile.



# Quantile $q$ -quantile computation

Find the second  $q$ -quantile:

- Calculate  $n * (2 / q)$ .
- If  $n * (2 / q)$  is an integer, the second  $q$ -quantile is the mean of the numbers at positions  $n * (2 / q)$  and  $n * (2 / q) + 1$ .
- If  $n * (2 / q)$  is not an integer, then round it up. The number at this position is the second  $q$ -quantile.

Find the  $(q-1)^{\text{th}}$   $q$ - quantile:

- Calculate  $n * ((q-1) / q)$ .
- If  $n * ((q-1) / q)$  is an integer, then the  $(q-1)^{\text{th}}$   $q$ - quantile is the mean of the numbers at positions  $n * ((q-1) / q)$  and  $n * ((q-1) / q) + 1$ .
- If  $n * ((q-1) / q)$  is not an integer, then round it up. The number at this position is the  $(q-1)^{\text{th}}$   $q$ - quantile.



# Quartiles from q-quantile

## Computing Quartiles (4-quantile)

Count the number of observations in the dataset ( $n$ ).

Sort the observations from smallest to largest.

Find the first quartile:

- Calculate  $n * (1 / 4)$ .
- If  $n * (1 / 4)$  is an integer, then the first quartile is the mean of the numbers at positions  $n * (1 / 4)$  and  $n * (1 / 4) + 1$ .
- If  $n * (1 / 4)$  is not an integer, then round it up. The number at this position is the first quartile.



# Quartiles from Quantile $q$ =quantile

Find the second quartile:

- Calculate  $n * (2 / 4)$ .
- If  $n * (2 / 4)$  is an integer, the second quartile is the mean of the numbers at positions  $n * (2 / 4)$  and  $n * (2 / 4) + 1$ .
- If  $n * (2 / 4)$  is not an integer, then round it up. The number at this position is the second quartile.

Find the third quartile:

- Calculate  $n * (3 / 4)$ .
- If  $n * (3 / 4)$  is an integer, then the third quartile is the mean of the numbers at positions  $n * (3 / 4)$  and  $n * (3 / 4) + 1$ .
- If  $n * (3 / 4)$  is not an integer, then round it up. The number at this position is the third quartile.



# Exercises



**Exercise 1:** Compute quartiles of the following data

Luanda CHIRPS station:

1. Compute seasonal totals for OND totals
2. Rank the OND totals
3. Compute the quartiles ( $q = 4$ ) and  $n$ =total number of years

Identify the values of the 3 positions of ( $q$ )

Q1:  $n \cdot (1/q)$  for quartile 1 position in your data

Q2:  $n \cdot (2/q)$  for quartile 2 position in your data

Q3:  $n \cdot (3/q)$  for quartile 3 position in your data

**Exercise 2:** Compute the terciles for the following datasets

Use the steps above with  $q=3$

Tercile 1:  $n \cdot (1/q)$  for Quartile 1 position in your data

Tercile 2:  $n \cdot (2/q)$  for Quartile 1 position in your data



*THANK YOU*

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