



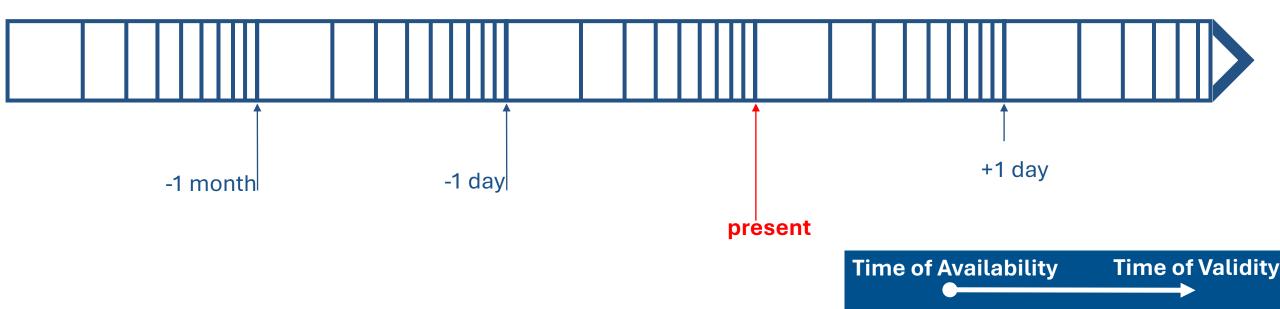
## IPWG-11 Training Course : Data assimilation within cloudy and precipitating areas

Philippe Chambon, with the contribution of many colleagues

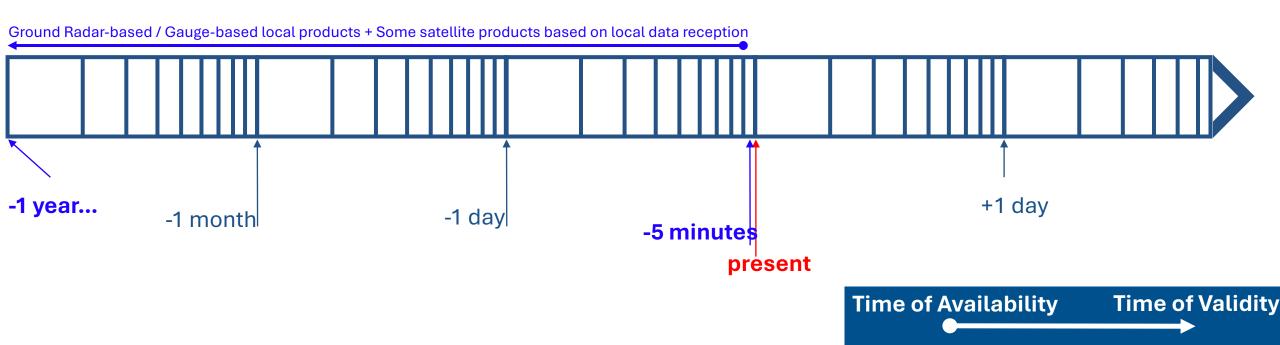
Météo-France, Toulouse, France

11th Workshop of International Precipitation Working Group (IPWG-11), 15-18 July 2024







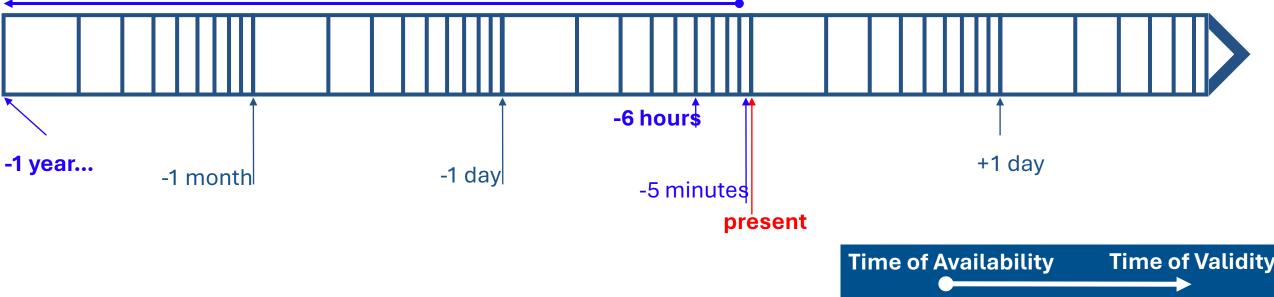




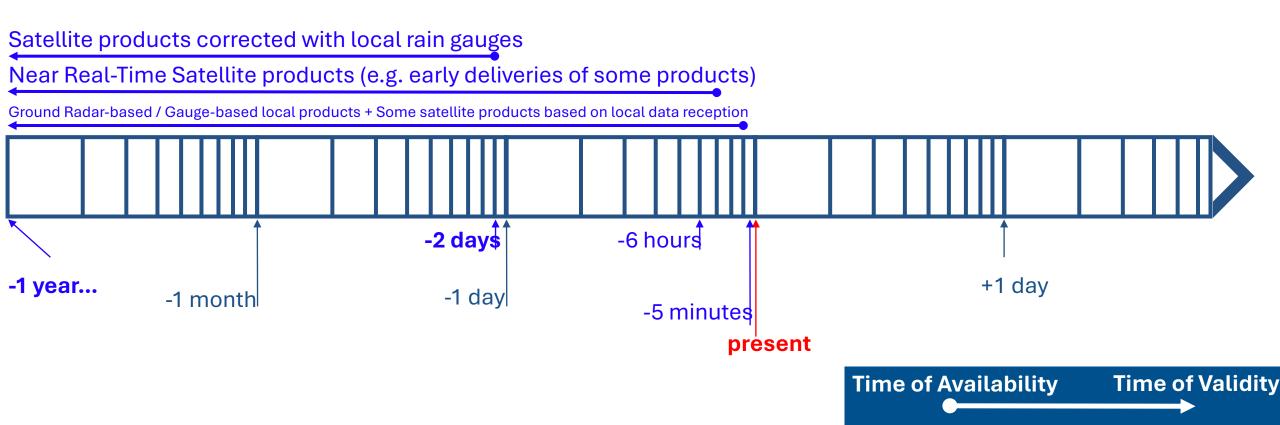
Rainfall products are needed for various applications : two requirements that will provide guidance on what product could be used are the Times of Validity/Availability of the needed rainfall estimates :

### Near Real-Time Satellite products (e.g. early deliveries of some products)

Ground Radar-based / Gauge-based local products + Some satellite products based on local data reception









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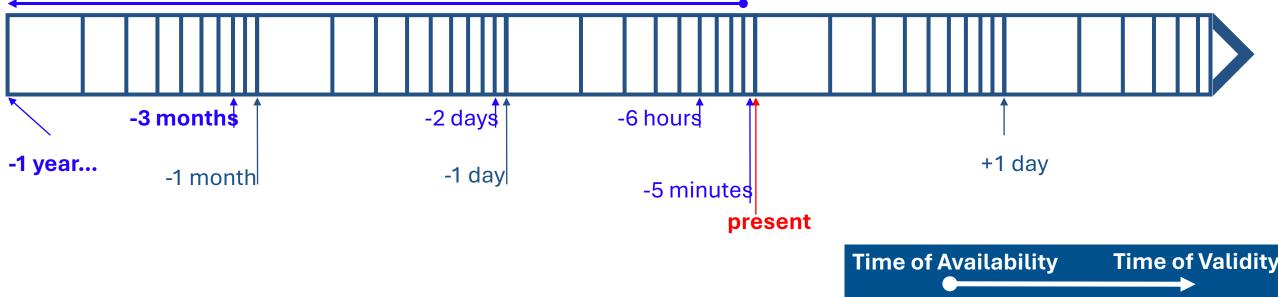
Satellite products corrected with rain gauges globally

Gauge-based global products

Satellite products corrected with local rain gauges

Near Real-Time Satellite products (e.g. early deliveries of some products)

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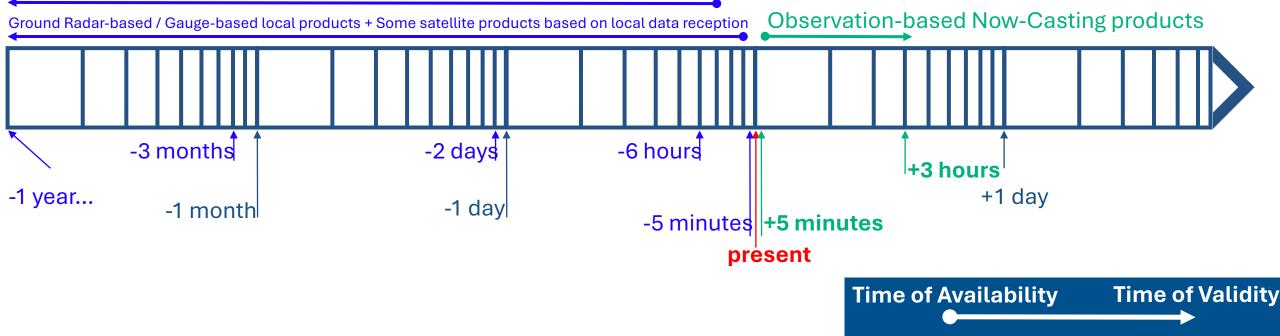
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Satellite products corrected with rain gauges globally

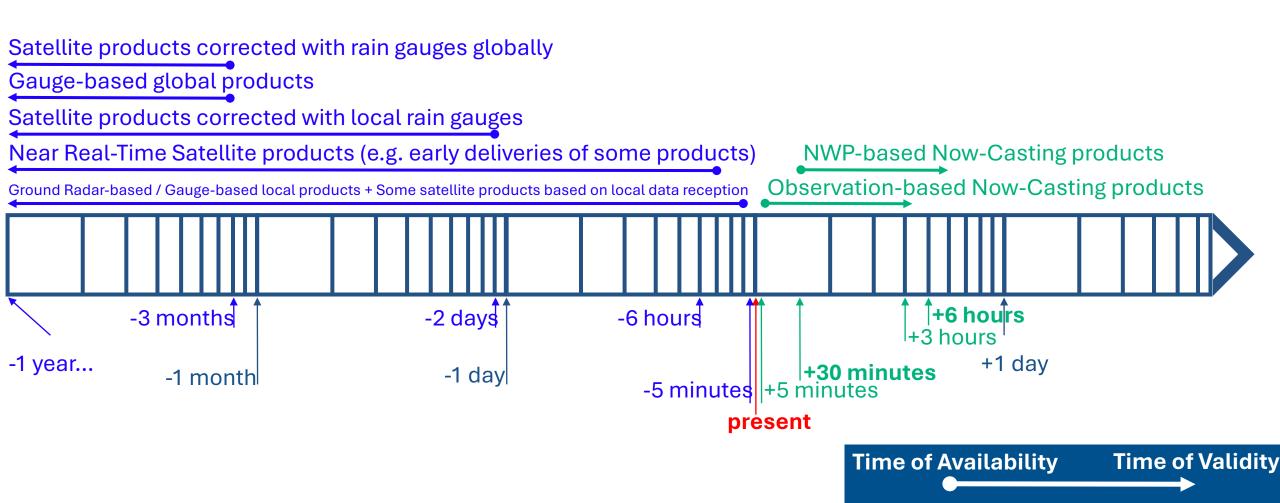
Gauge-based global products

Satellite products corrected with local rain gauges

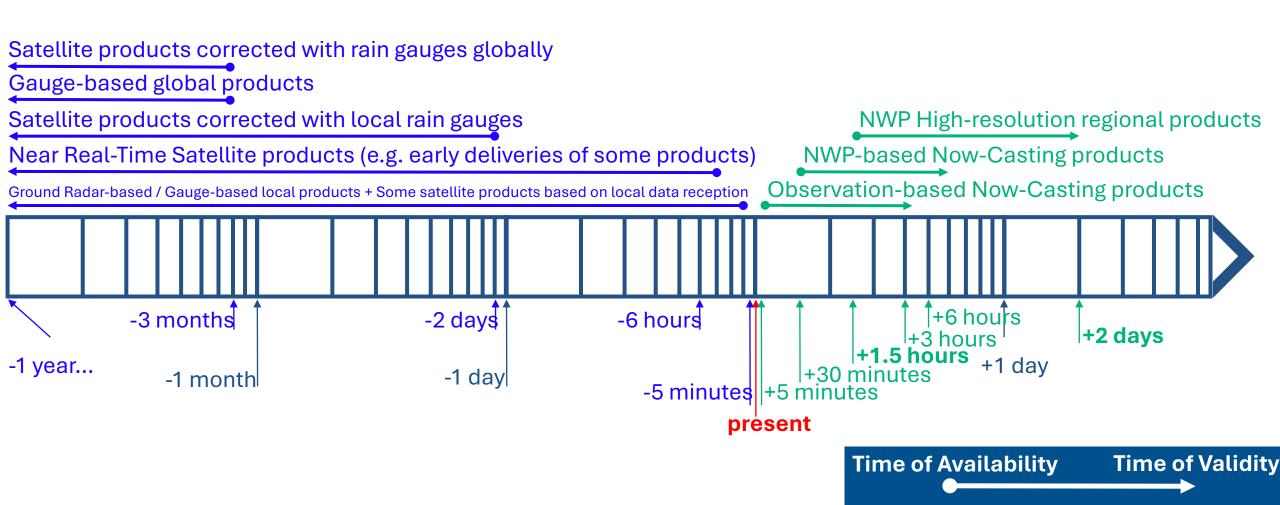
Near Real-Time Satellite products (e.g. early deliveries of some products)



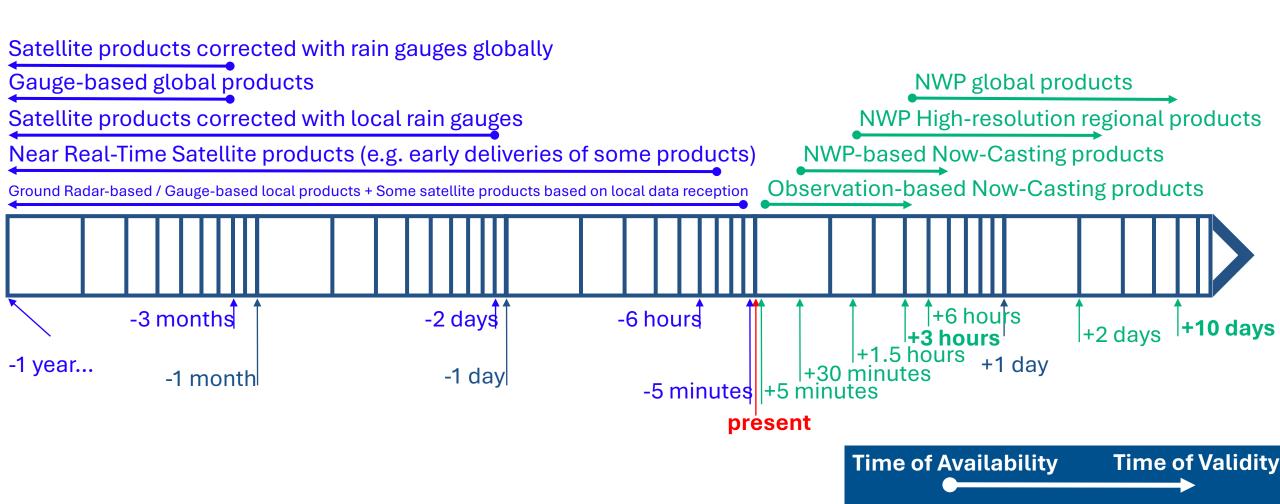




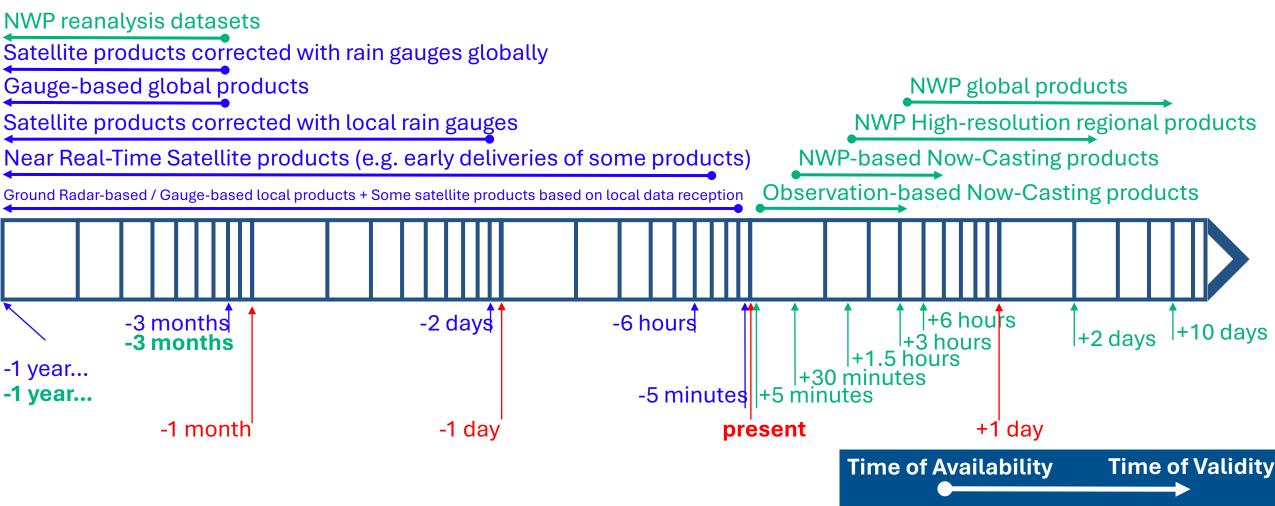




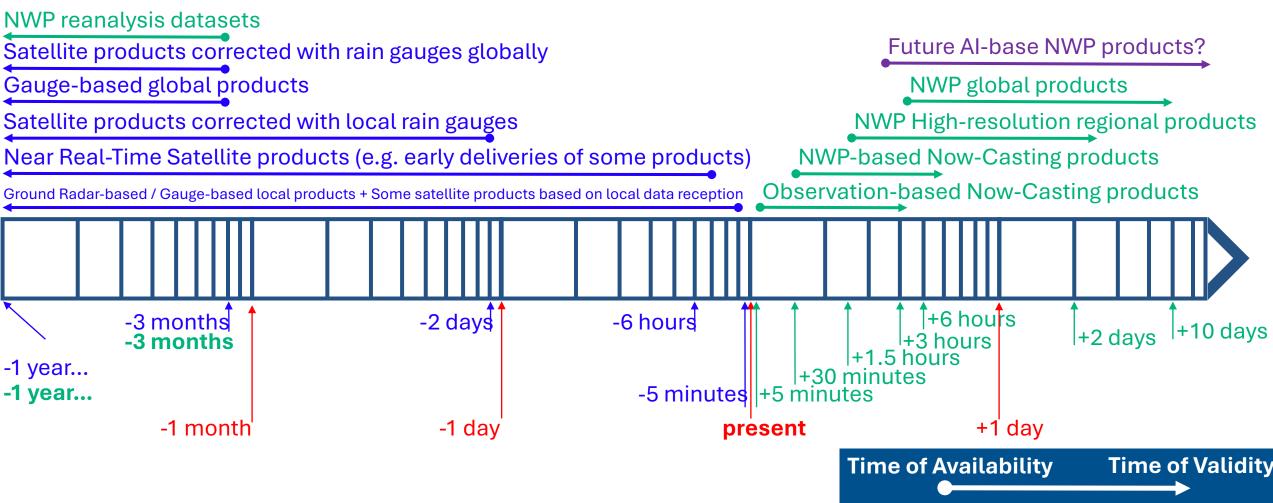














# **Outline of the presentation**

### **1. Current and Future NWP models: what are their strengths and**

weaknesses regarding rainfall forecasts ?

- **2.** How rainfall forecasts can be validated ?
- 3. How rainfall forecasts can be improved ?



# **Outline of the presentation**

# 1. Current and Future NWP models: what are their strengths and weaknesses regarding rainfall forecasts ?

**2.** How rainfall forecasts can be validated ?

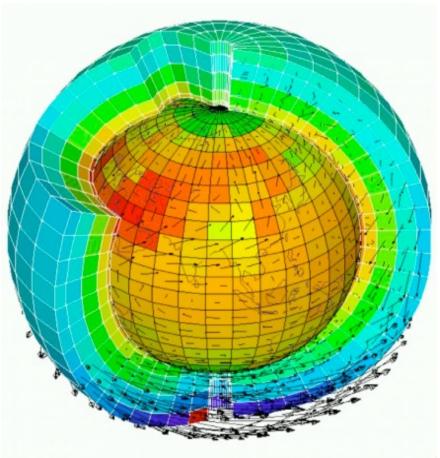
**3. How rainfall forecasts can be improved ?** 

11th Workshop of International Precipitation Working Group (IPWG-11), 15-18 July 2024



# Current and Future NWP models: what are their strengths and weaknesses regarding rainfall forecasts ?

- Numerical Weather Prediction uses mathematical models of the atmosphere and oceans, in which time and the physical fields are discretized, to predict the weather based on current weather conditions.
- It is impossible to solve exactly the partial differential equations that govern the atmosphere and errors grow with the forecast range during the model integration in time



(Source: http://pedagotech.inp-toulouse.fr)

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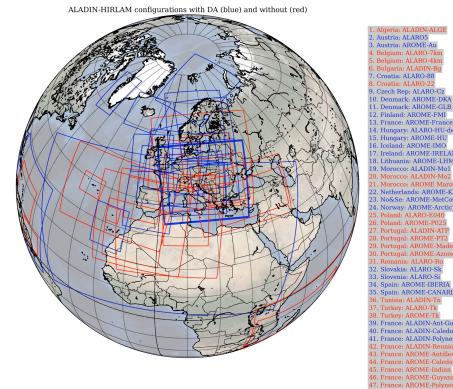
### **Current and Future NWP models: what are their** strengths and weaknesses regarding rainfall forecasts?

The current generation of <u>operational</u> global **models** use parametrization for various processes like convection and provide forecasts at **O**(10km) resolution



Artist view of the ECMWF 'cubic-octahedral grid' (Source: http://www.ecmwf.int)

The current generation of operational regional models are non-hydrostatic are resolve convection explicitly. They provide forecasts at **O**(1km) resolution



#### Limited Area Domains of the ALADIN/HIRLAM consortia with DA and without **DA** (Source http://www.cnrm-game-meteo.fr/aladin/)

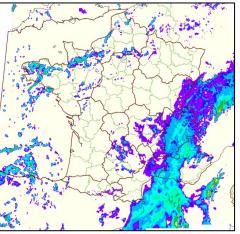


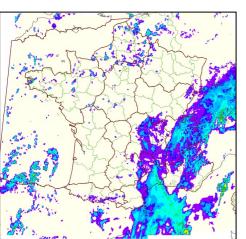
# Current and Future NWP models: what are their strengths and weaknesses regarding rainfall forecasts ?

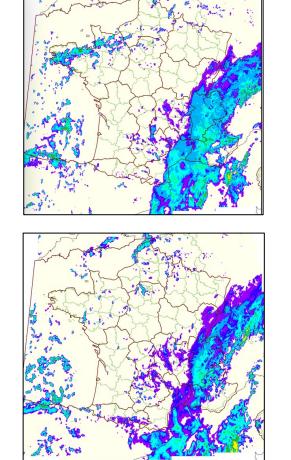
 Numerical Weather Prediction systems are now often based on ensembles, providing different scenario. The spread between these scenario can give an idea of the uncertainty

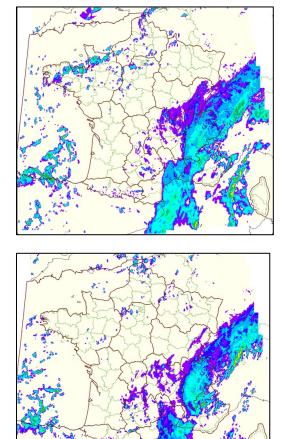
level of the forecast

Example of several members from the AROME ensemble of regional forecasts of Météo-France for Oct 28, 2018



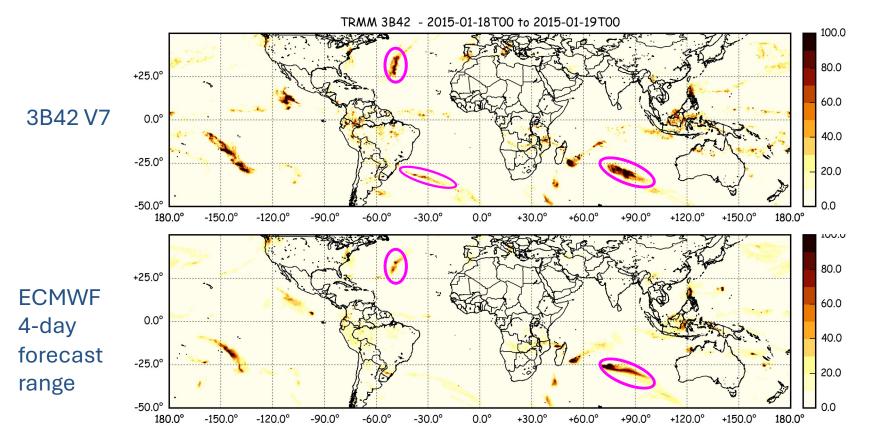






### Current and Future NWP models: what are their Strengths and weaknesses regarding rainfall forecasts ?

• One example of forecast error growth: 24h precipitation forecast for Jan 18<sup>th</sup>, 2015 from the ECMWF model



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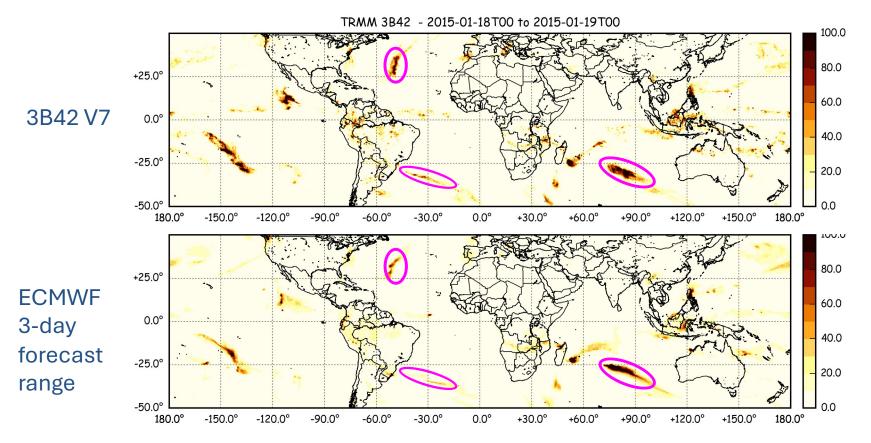
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At long range, precipitation forecasts suffer from mis-locating clouds and precipitation features due to either initial conditions problems, model physics deficiencies or

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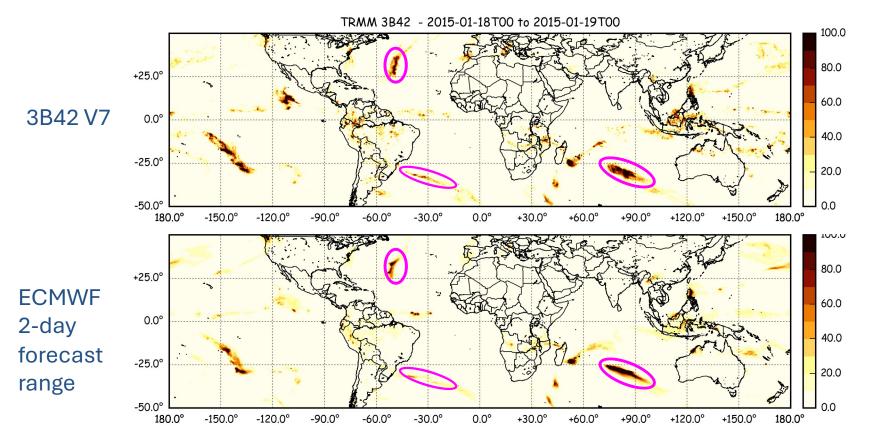
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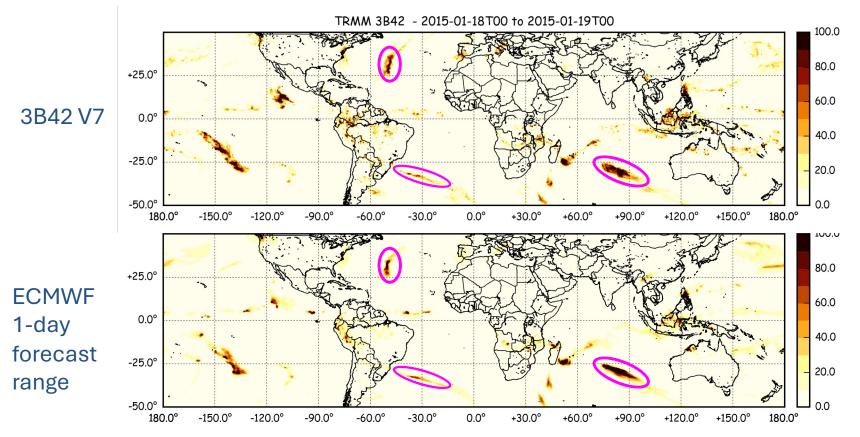
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With shorter forecast range, precipitation forecasts skills improve ! Depending on the application and the accuracy needed, precipitation forecasts can be highly relevant products.



# Current and Future NWP models: what are their strengths and weaknesses regarding rainfall forecasts ?

- From its beginnings, Numerical Weather Prediction has been confronted with having to solve an initial-value problem without the right initial data.
- Model state space:
  - MASS (temperature, pressure...)

Radiosondes, surface observations, satellite sounders, aircraft

**MOISTURE** (humidity, clouds, precipitation...)

Radiosondes, surface obs., satellite sounders and imagers, aircraft, radar, lidar

**DYNAMICS** (wind, vorticity, convergence...)

Radiosondes, surface obs., satellite imagers, satellite scatt./ radar/lidar, aircraft **COMPOSITION** (ozone, aerosol...)

Ozone sondes, surface observations, satellite sounders



Source: http://pedagotech.inp-toulouse.fr

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# Current and Future NWP models: what are their strengths and weaknesses regarding rainfall forecasts ?

- The number of observations of the present observing system for analysis is orders of magnitude smaller than the number of values required to specify the model state. The initial state of the numerical model cannot therefore be determined from the available observations alone.
- An analysis of the present weather combines information from both the model and observations, thanks to data assimilation.

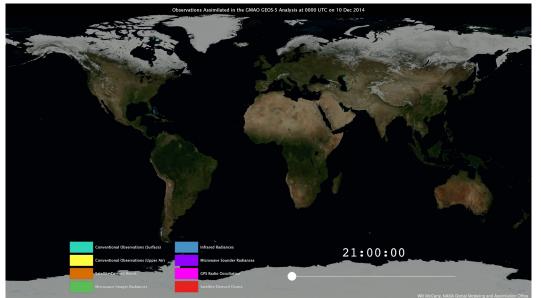






# Current and Future NWP models: what are their strengths and weaknesses regarding rainfall forecasts ?

- Performing an analysis is a complex mathematical problem which has many common points with the inverse problem (retrievals).
- One of the challenges it needs to face is to combine millions of observations of different kinds in an optimal way with a model prior information and compute the best estimate of the atmospheric state



NASA's Global Modeling and Assimilation Office GEOS-5 global model

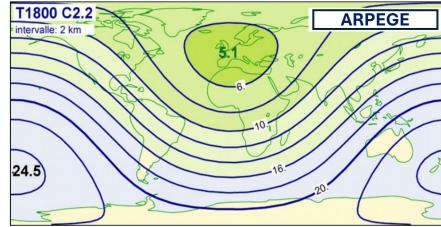
5 millions of observations during each 6-hour assimilation period

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### Current and Future NWP models: what are their strengths and weaknesses regarding rainfall forecasts?

• The current generation of NWP systems already suffer from a lack of constrain on its initial conditions with the current observing system.



 $\sim 8.10^6$  observations assimilated every 6 hours for ~200.10<sup>6</sup> variables to initialize



Radiosondes

Microwave sounding

from space



Surface stations

Infrared sounding

from space



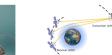
Aircraft data



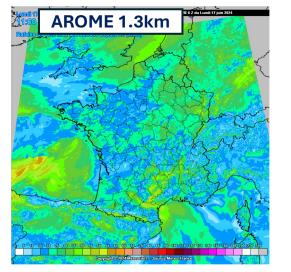
Winds derived

from imagery





Scatterometer GNSS Radio-occultation



~100 000 observations assimilated every hours for  $\sim 800.10^6$  variables to initialize





# Current and Future NWP models: what are their strengths and weaknesses regarding rainfall forecasts ?

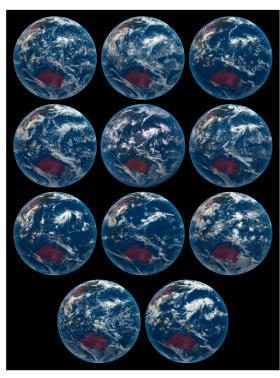
- Model forecast skills vary from one region to another over the globe (e.g. over Europe, ECMWF rain forecasts often have more skills than any satellite retrievals, not the case in the Tropics)
- These precipitation products are forecasts, they thus fulfill different needs than satellite retrievals => different times of validity/availability
- There is still a lot of room for improving model rain forecasts, in particular with better initial conditions constraining clouds and precipitation
- $\Rightarrow$  Satellite clouds and precipitation observations can be used to this endeavor, See Section 3
- ⇒ Model errors in clouds and precipitation need to be documented first with different kind of observations, See Section 2



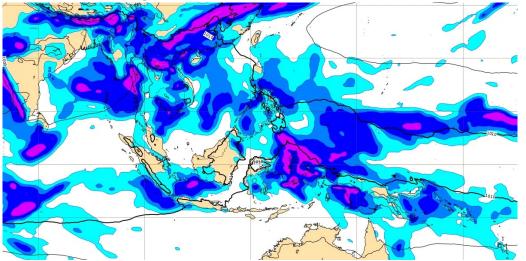
# Current and Future NWP models: what are their strengths and weaknesses regarding rainfall forecasts ?

• Research on two major evolutions is ongoing with very different upcoming challenges for data assimilation :

*Kilometric scale modeling of the Earth system* 



Numerical forecasts with AI-based models



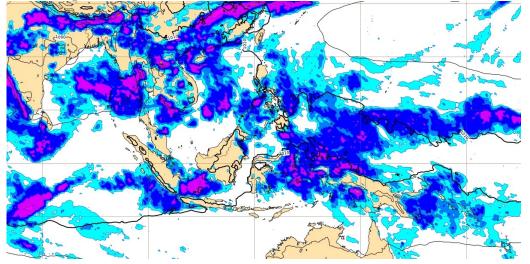
(Source : ECMWF, AIFS ML model)

(Source : Stevens et al. 2019)

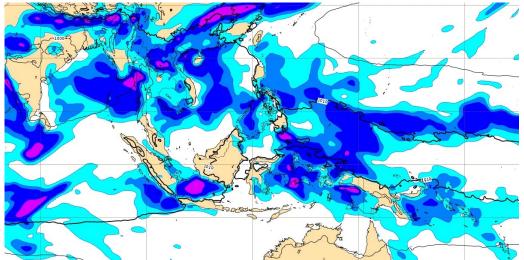


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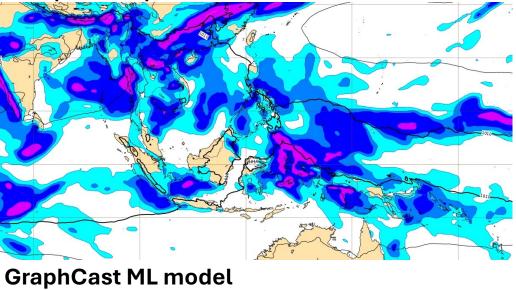
**IFS High resolution forecast** 

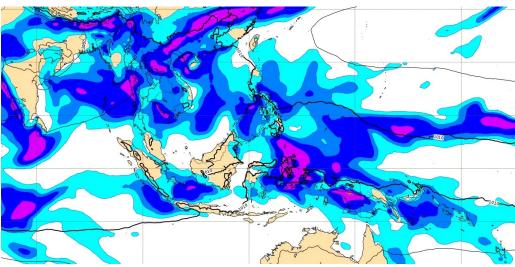


### FuXi ML model



#### AIFS (ECMWF) ML model



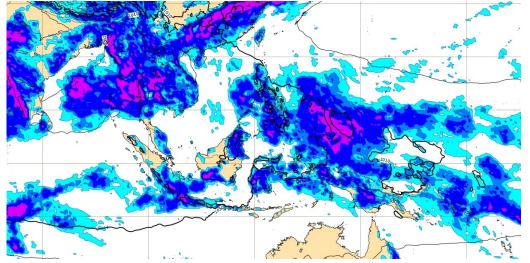


Runs of different Al-based models performed at ECMWF, initialized with the same analysis 

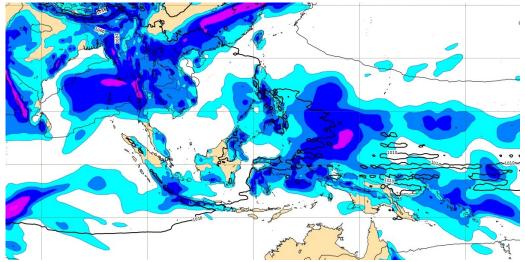


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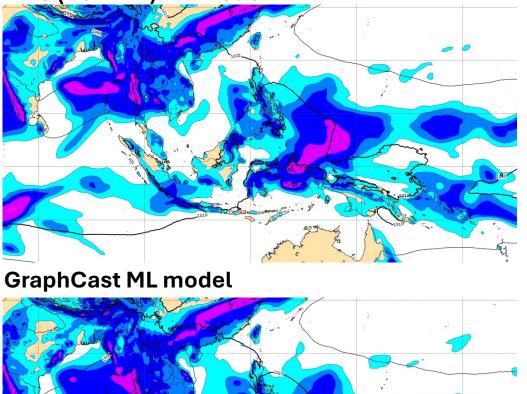
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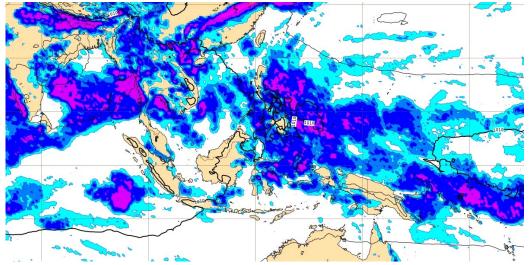


Runs of different Al-based models performed at ECMWF, initialized with the same analysis **C**ECMWF

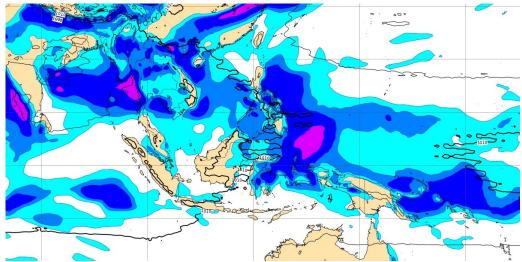


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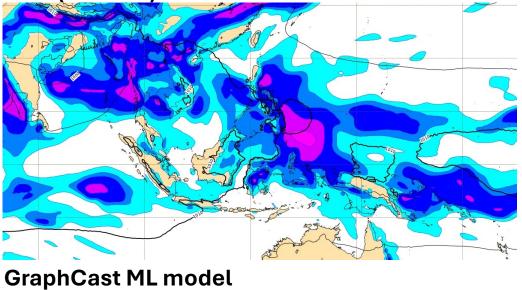
**IFS High resolution forecast** 

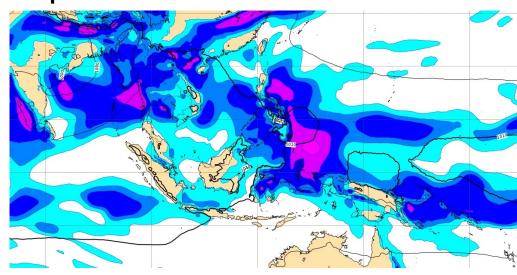


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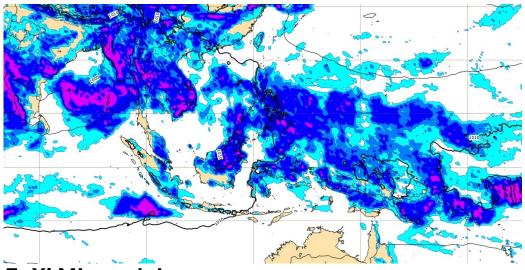


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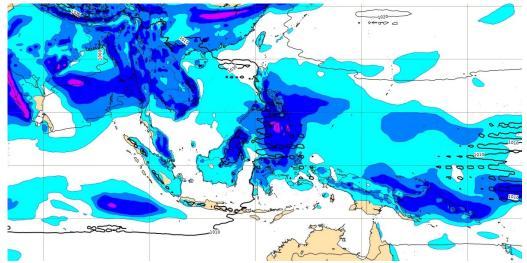


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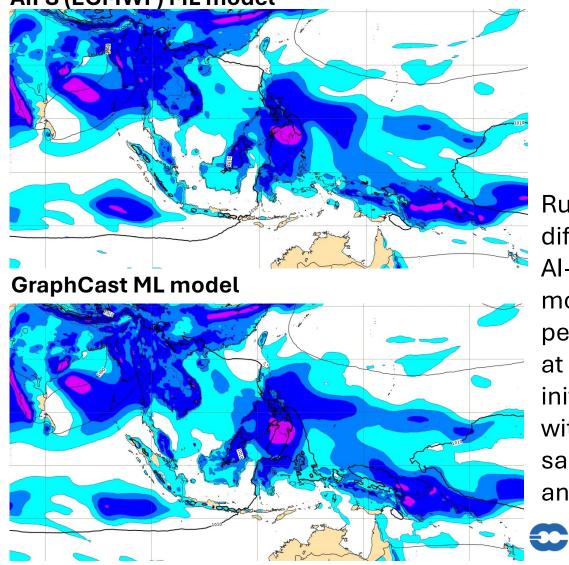
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Runs of different Al-based models performed at ECMWF, initialized with the same analysis 



# **Outline of the presentation**

### **1. Current and Future NWP models: what are their strengths and**

weaknesses regarding rainfall forecasts ?

### 2. How rainfall forecasts can be validated ?

3. How rainfall forecasts can be improved ?



Rainfall forecasts need to be compared to observations for documenting their strengths and weaknesses:

- Over land and instrumented areas, rain gauges or radar data can be good references
- Over oceans or other less instrumented areas, satellite rainfall products can be good alternatives

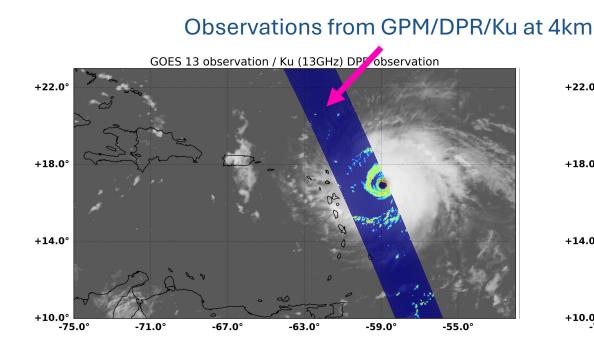


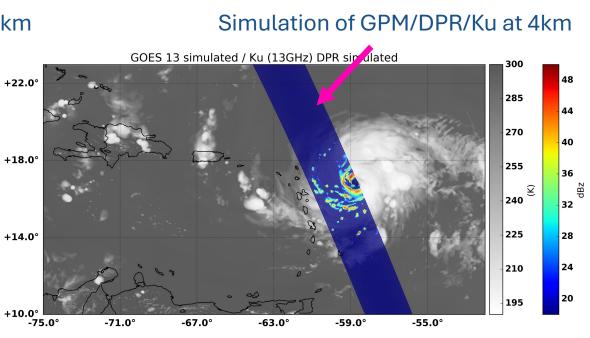
Example of the AROME-OM domains operated at Météo-France:



An important aspect is the spatial and temporal scale at which you want to validate your forecasts:

 At fine scale (e.g. kilometric, instantaneous), interesting references to use are space radar data: the comparison can be performed in both observation space (reflectivity) or model space (rain rate retrieval)

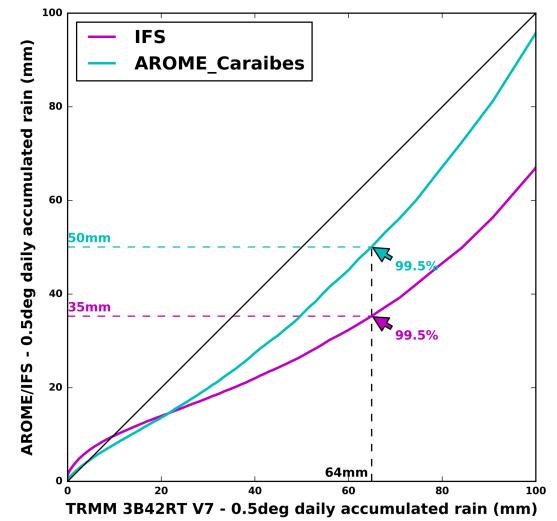




### RÉPUBLIQUE FRANÇAISE Image: Comparison of the comparison

At larger scale (e.g. daily accumulation, 0.5° x 0.5°) combined IR+MW products are highly relevant references

Example of comparisons between AROME over the Caribbean's, the ECMWF model and TRMM 3B42 rain estimations at 1-day/0.5°



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At larger scale (e.g. daily accumulation, 0.5° x 0.5°) combined IR+MW products are highly relevant references

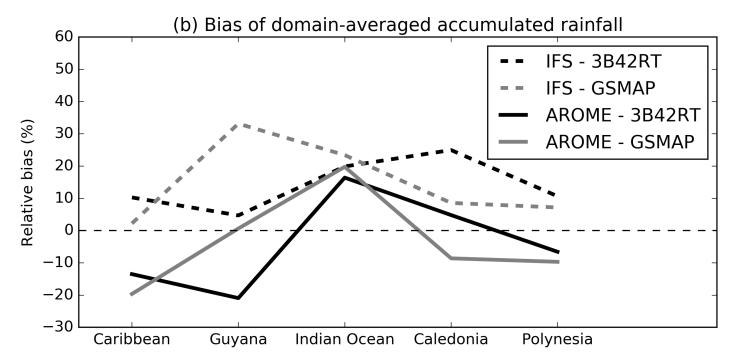
For assessing biases, using several references are important, In particular when part of the model domain is over land (more uncertainty of the satellite rainfall products)

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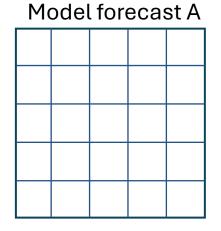
=> When a bias is identified, then the tricky part is to characterize the causes of it !



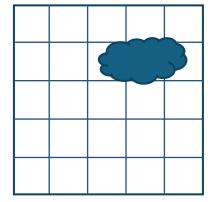
At larger scale (e.g. daily accumulation, 0.5° x 0.5°) combined IR+MW products are highly relevant references

For assessing the representation of the model variability, Fuzzy verification is often used (e.g. Ebert, 2008) => Take into neighborhood for the verification

Observation



Model forecast B

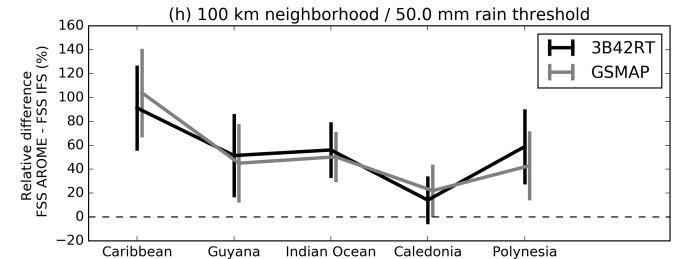


Is model-forecast B more useful than forecast A for your application? For forecaster, the answer is often yes!

## RÉPUBLIQUE FRANÇAISE Image: Comparison of the comparison

At larger scale (e.g. daily accumulation, 0.5° x 0.5°) combined IR+MW products are highly relevant references

Example of Brier Skill Score differences between AROME in the Tropics, the ECMWF model and TRMM 3B42/GSMAP rain estimations at 1-day/0.5 Over a one year period



=> The representation of strong rainfall events is often better forecasted by regional non hydrostatic model with no parametrized convection



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Several pathways can be used to improve model precipitation forecasts:

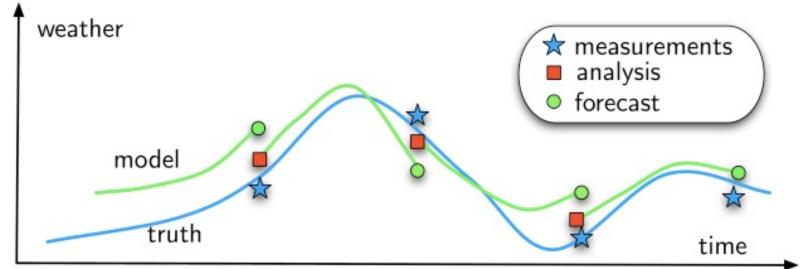
- Model physics can be improved through different developments like parametrization improvements, microphysics scheme improvements etc
- Initial conditions can be improved through the assimilation of observations of clouds and precipitation => same data that are used in satellite precipitation product
  - > Improvements can arise from new data assimilation algorithmic developments
  - > Or new observational developments (e.g. usage of new observation types)



• The analysis minimizes a combination of distances :

$$J(\blacksquare) = \frac{1}{2} \parallel \blacksquare - \bigcirc \parallel_B^2 + \frac{1}{2} \parallel \bigstar - \mathcal{G}(\blacksquare) \parallel_R^2$$

The model provides a forecast from which the observation operator extracts quantities which are compared to actual data observation. The analysis is a field that combines these observations with the model forecast. It is then used as an initial condition for the next forecast.



Source: http://pedagotech.inp-toulouse.fr

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$$\mathbf{x}_{a} = \operatorname{Arg min} J$$
  

$$J(\mathbf{x}) = (\mathbf{x} - \mathbf{x}_{b})^{\mathrm{T}} \mathbf{B}^{-1} (\mathbf{x} - \mathbf{x}_{b}) \cdot (\mathbf{y} - H[\mathbf{x}])^{\mathrm{T}} \mathbf{R}^{-1} (\mathbf{y} - H[\mathbf{x}])$$
  

$$= J_{b}(\mathbf{x}) + J_{o}(\mathbf{x})$$

In the classical data assimilation formalism:

- x is the model state vector (temperature, humidity, hydrometeors,...)
- y is the observation vector (conventional, satellite, etc...)
- H refers to the observation operator

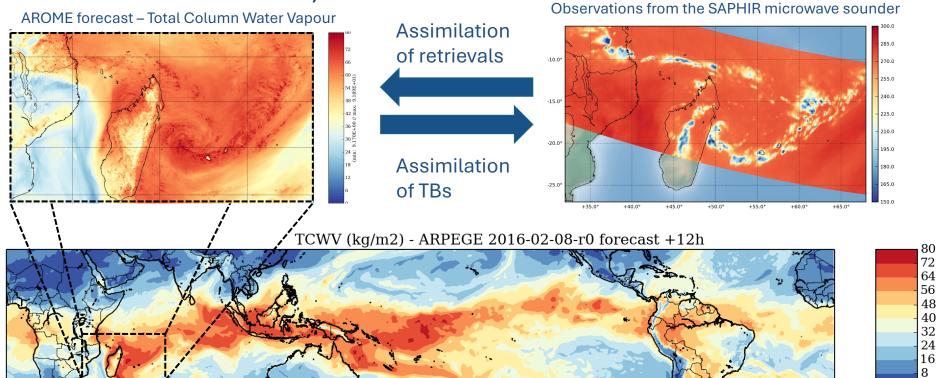
=> operator which transform the model state variables into observable quantities (e.g. for satellite, radiances or brightness temperatures)

R refers to the covariance matrix of observation errors

=> R describes the errors of the difference (y-H[x]), including radiometric uncertainties and H errors



- Observations and model first guess need to be in the same space to be compared within a DAS framework => two possibilities
- The current trend in NWP centers is to assimilate TBs (mainly because observation errors) tend to be easier to characterize)





$$\mathbf{x}_{a} = \operatorname{Arg min} J$$
  

$$J(\mathbf{x}) = (\mathbf{x} - \mathbf{x}_{b})^{\mathrm{T}} \mathbf{B}^{-1} (\mathbf{x} - \mathbf{x}_{b}) \cdot (\mathbf{y} - H[\mathbf{x}])^{\mathrm{T}} \mathbf{R}^{-1} (\mathbf{y} - H[\mathbf{x}])$$
  

$$= J_{b}(\mathbf{x}) + J_{o}(\mathbf{x})$$

- B refers to the covariance matrix of background errors (a priori information from former forecast). B describes the uncertainties of the background (previous model forecast in a cycled assimilation process), as well as the error correlations between model variables.
- Accurate Background Error Covariances are required to update observed and unobserved model variables in a **balanced** way.



- Depending on the "control variable" selected for clouds and precipitation, different background error statistics will be needed for the assimilation framework.
- Specific Humidity? Relative humidity? (See Geer et al., 2017 for a summary of the control variables currently used in different NWP centers)

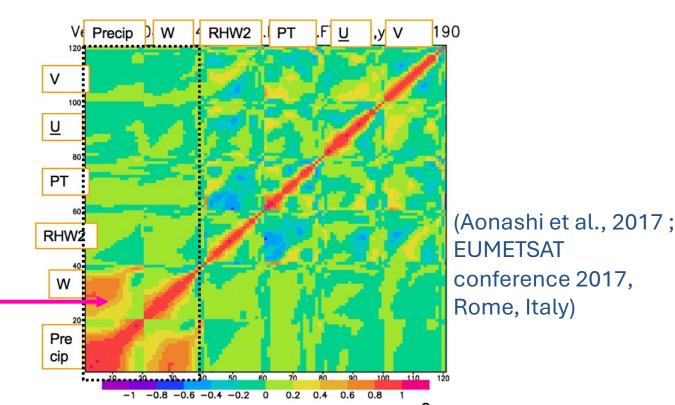
≻Hydrometeors?

(e.g. Michel et al., 2011; Zhang et al., 2013)

➤Vertical wind speed?

(e.g. Aonashi et al., 2017; Lee, et al., 2018)

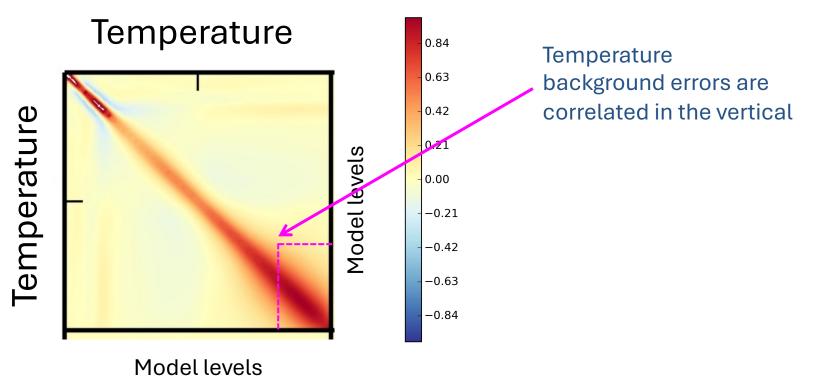
Strong correlation between vertical wind speed and precipitation





Example of multivariate background errors vertical covariances modeled in cloudy and rainy sky over the Indian Ocean of the AROME model

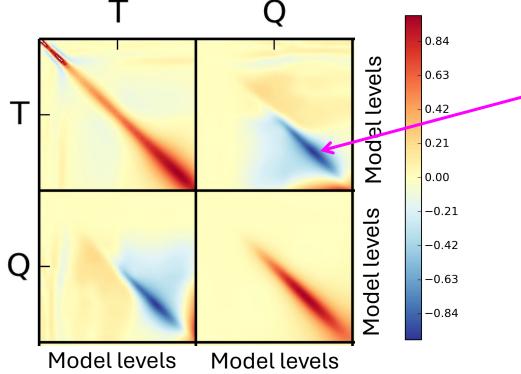
(Montmerle and Berre, 2011)





Example of multivariate background errors vertical covariances modeled in cloudy and rainy sky over the Indian Ocean of the AROME model

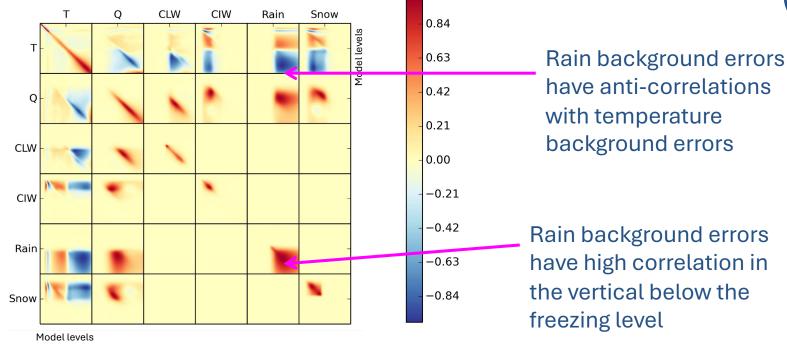
(Montmerle and Berre, 2011)



Humidity background errors are anti correlated in the vertical to temperature background errors



Example of multivariate background errors vertical covariances modeled in cloudy and rainy sky over the Indian Ocean of the AROME model

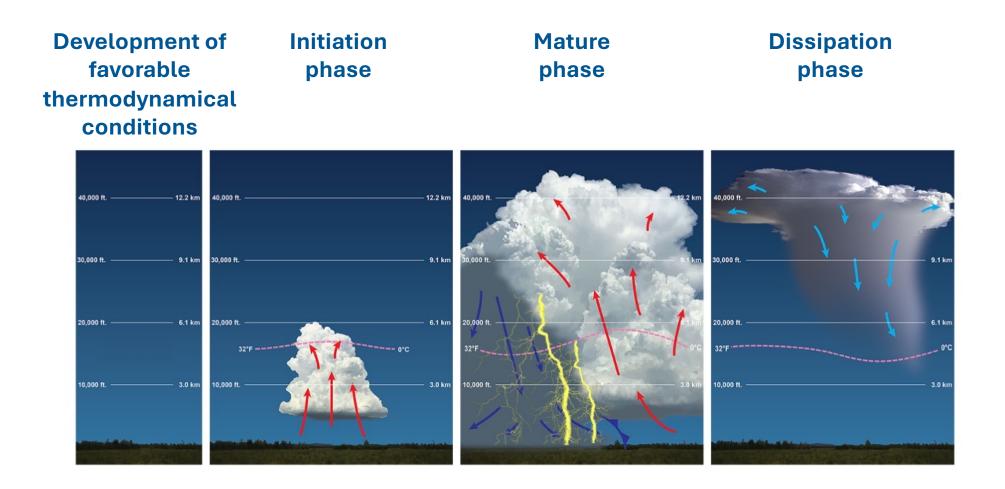


(Montmerle and Berre, 2011)

These balances need to be preserved (in 3D) if one want the increments done during the analysis process to be sustained during a forecast

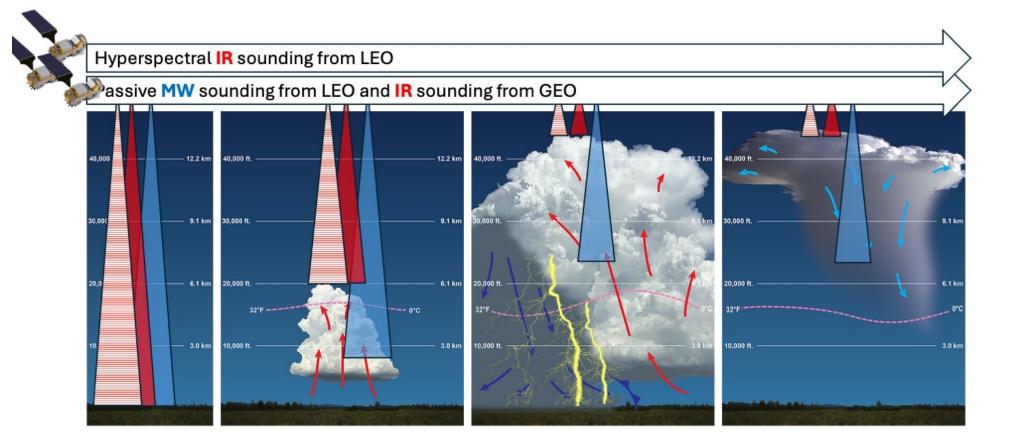


### How rainfall forecasts can be improved ?



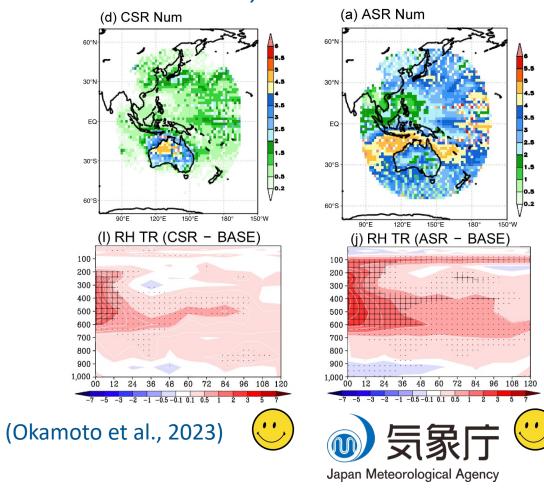
(Figure adapted from https://www.weather.gov/spotterguide/life)







#### *Transition from Clear Sky Radiances to Himawari AllSky Radiances*



## How rainfall forecasts can be improved ?

#### Transition from Clear Sky Radiances to Himawari AllSky Radiances

- A.

Liberté Égalité

Fraternite

RÉPUBLIQUE

FRANCAIŠE

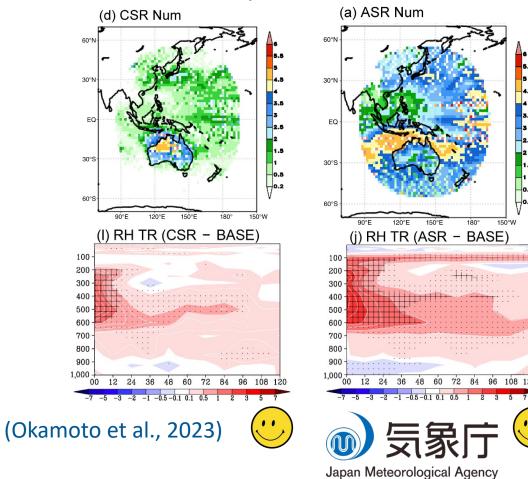
Ø

**METEO** 

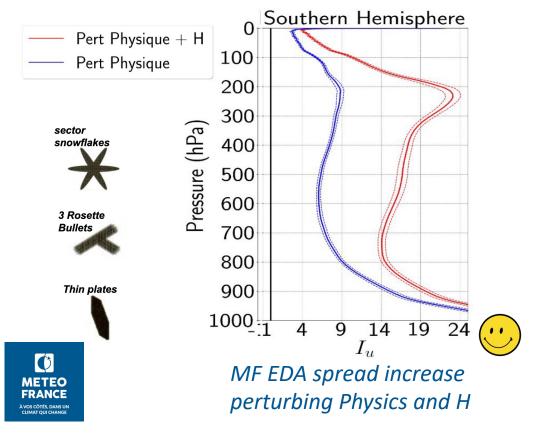
FRANCE

À VOS CÔTÉS, DANS UN

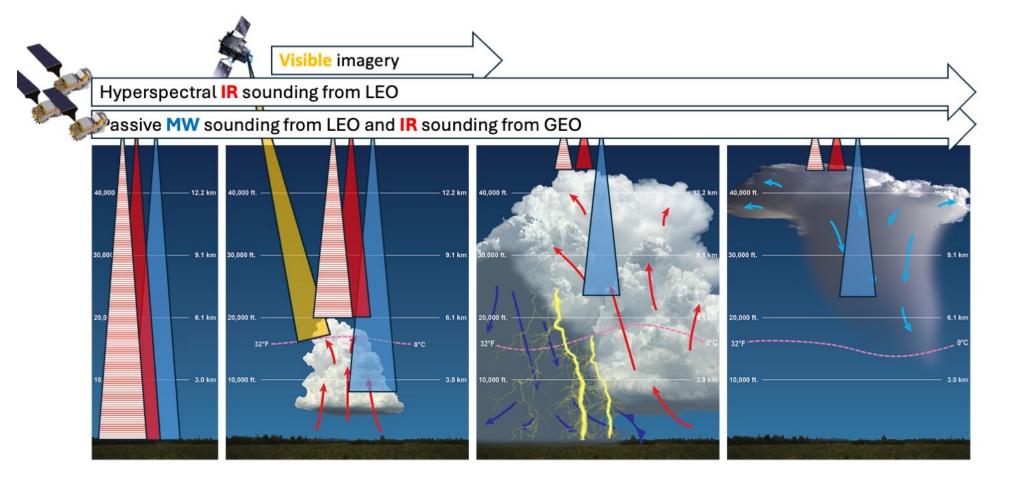
CLIMAT OUI CHANGE



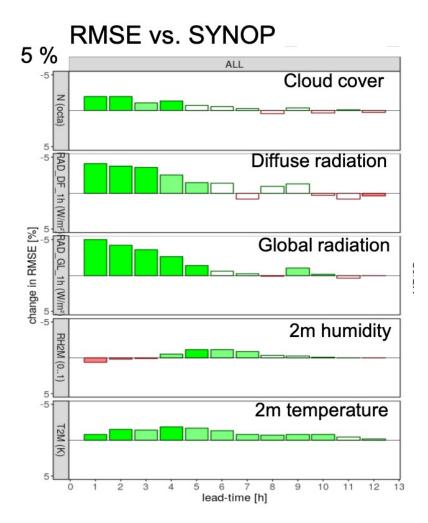
#### Taking into account uncertainties of the forward model within DA => perturbing the observation oeprator

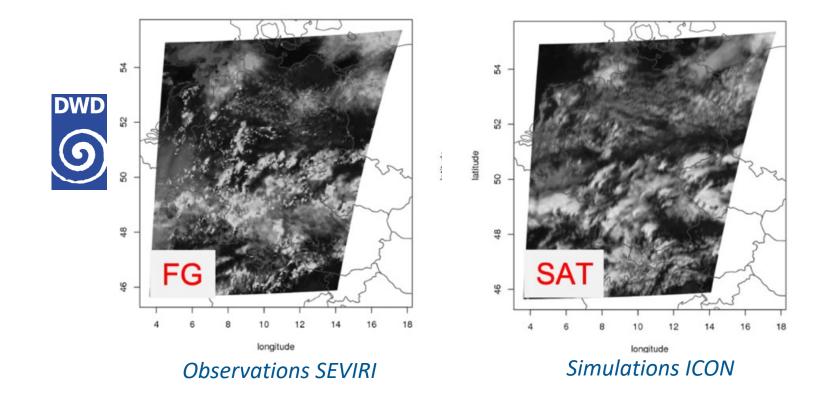






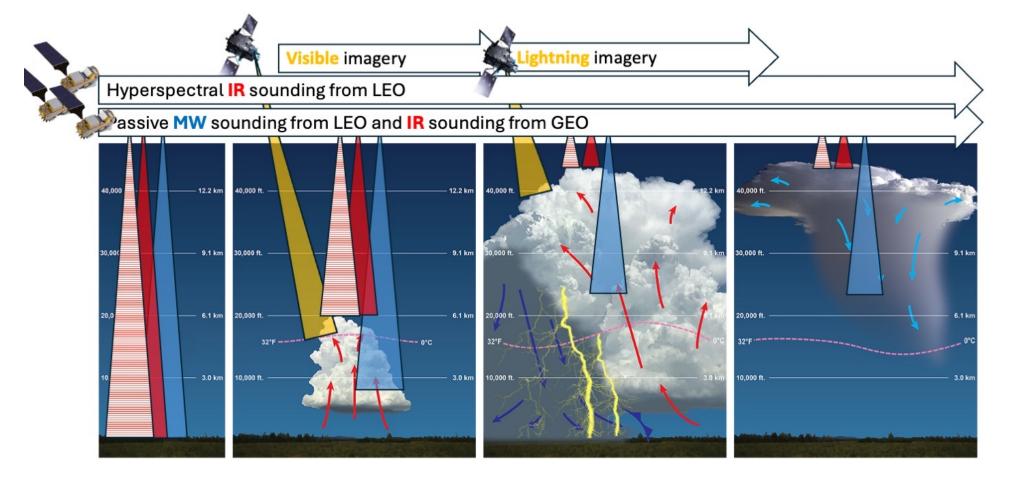




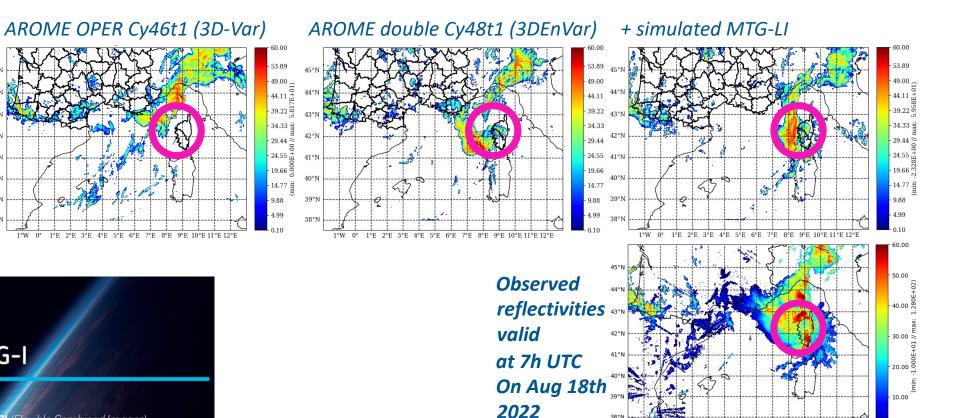


Impact of Visible observations in ICON









1°E 2°E 3°E 4°E 5°E 6°E 7°E 8°E 9°E 10°E 11°E 12°E

FCI (Flexible Combined Imager) LI (Lightning Imager) DCS (Data Collection System) GEOSAR (Search and Rescue)

**Forecasted** 

reflectivities

at range +4h

Valid at 7h

on Aug 18th,

2022

40°N

3001

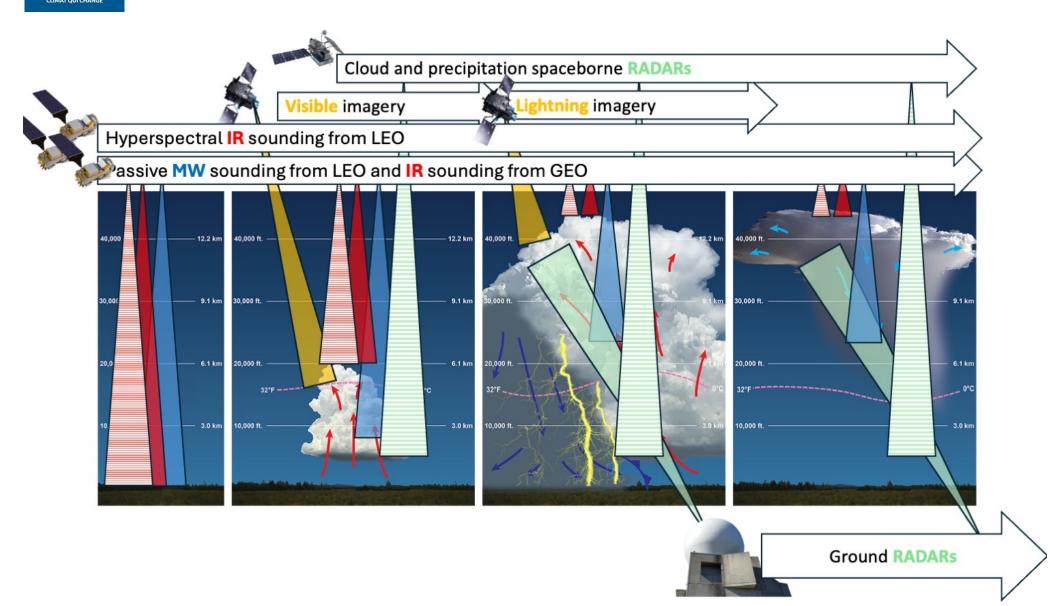
38°1

MTG-I

1°W

#### Ø RÉPUBLIQUE METEO FRANCE How rainfall forecasts can be improved ? FRANÇAIŠE À VOS CÔTÉS, DANS UN CLIMAT QUI CHANGE

Liberté Égalité Fraternité



## How rainfall forecasts can be improved ?

Preparing EarthCARE assimilation with Cloudsat/Calipso data

Ø

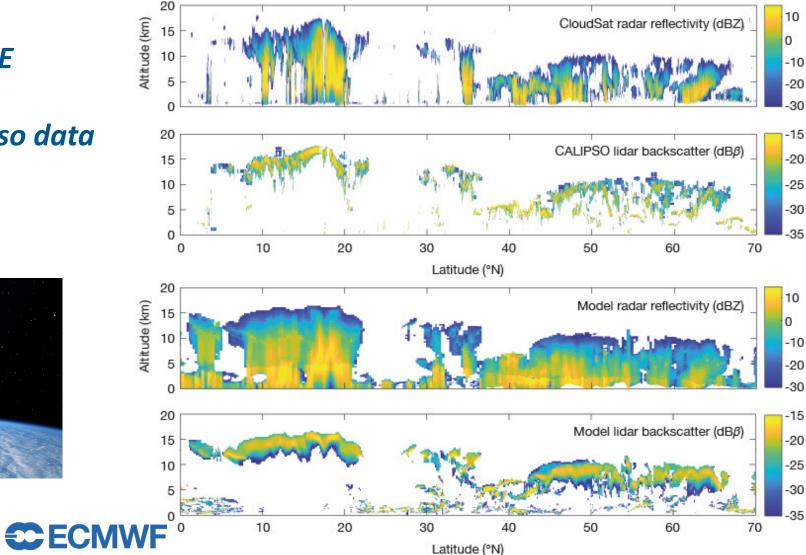
METEO FRANCE

À VOS CÔTÉS, DANS UN CLIMAT QUI CHANGE

Liberté Égalité Fraternité

RÉPUBLIQUE

FRANCAISE

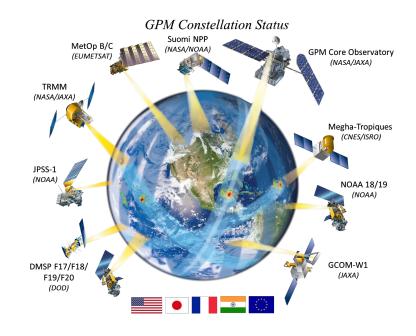






# Assimilating Satellite Observations of Clouds and Precipitation is a challenging field of research !!

- Non-linear observation operator
- Non-Gaussian error distributions
- Under-determined problem
- Complex, flow-dependent balance
- Significant model errors
- Variety of spatio-temporal scales



(Source http://pmm.nasa.gov)



