

Measuring dynamical properties of atmospheric convection using C²OMODO a tandem of microwave radiometers

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1 Scientific context: Deep Convection

- **Extreme cloud systems**, occurs in Intertropical zone
- air, aerosols & water updrafts up to the tropopause
- Strong impacts on : Water & Energy Cycles
- Issues around deep convection :
 - Lack of observations of convective updrafts
 - High spatial and temporal sampling
 - In-situ measurement difficulties over oceans.
 - Limited representation in **meteo/climate models**

2 Space mission : C²OMODO

C²OMODO (Convective Core Observations through MicrOWave Derivatives in the trOPics) :

- Contribution to Atmosphere Observing System (AOS) program (NASA-led with CSA, JAXA, CNES)
- Observation of **clouds and storms dynamics**
- Twin multispectral **passive microwave radiometers**
- Measurement expressed in brightness temperature (Tbs)
- **Tbs** related to **ice scattering attenuation**
- $\frac{dTbs}{dt}$ related to growth of convection [1][2]

General Description	Frequency [GHz]	IFOV [km]
Orbits inclined ±55° latitudes	183.31	5
Launch : [2028 – 2030]	325.15	3
Mean altitude : 427 km	89	10
Swath : > 750 km		

Table 1 : Description of C²OMODO and instrumental configuration used. IFOV at nadir.

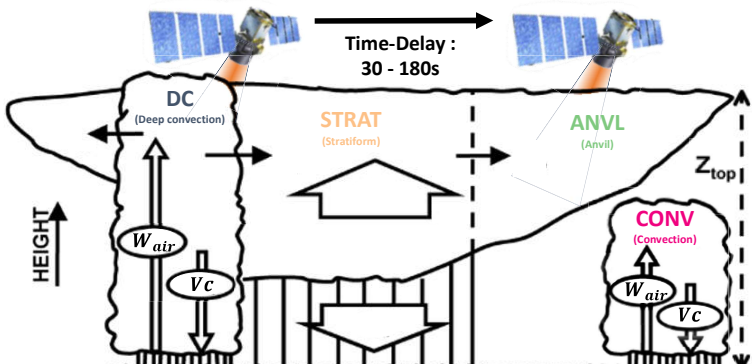


Fig.1: C²OMODO principle. Cumulonimbus (liquid and frozen water) observed with twin radiometers
 W_{air} : In-cloud vertical wind velocity ; V_c : Terminal velocities ; In red : Cloud classes

3 Numerical tools

MESO-NH [5] (Cloud Resolving Model) :

- Laboratoire d'A erologie et M et eo-France
- Non-hydrostatic (Resolved convection)
- Hector simulation [6]:
 - 6h ; 256*200 km² ; up to 30km
 - $\Delta x = 1\text{km}$; $\Delta z = 200\text{m}$; $\Delta t = 30\text{s}$

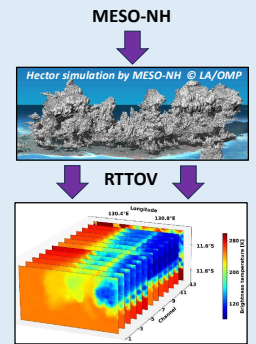


Fig.3: Tbs samples across channels

RTTOV.13 [7] (Radiative Transfer Model) :

- Simulation of satellite measurements
- No spatial observation geometry yet
- Default particle size/shape distribution [8]
 - To be explored in another study

4 Machine learning method

Method : GBDT - LightGBM [9] (Article in preparation) :

- Convective cell structure (Based on [4]) :
 - Single-radiometer : Cloud structure (Fig.4)
 - Tandem-based : Ice variation phases
- High performance retrieval of IWP and $\frac{dIWP}{dt}$

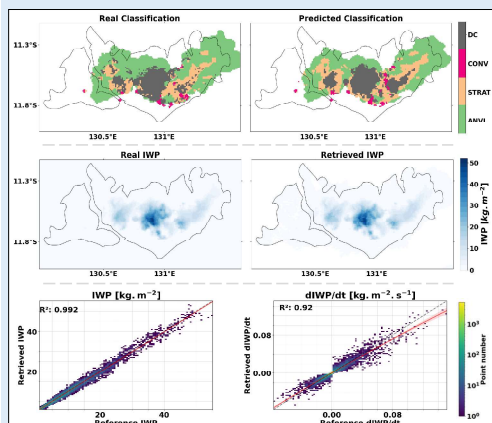


Fig.4: Results of the Machine learning methods.

Top line : map of the real and predicted classes.
 Middle line : map of the real and retrieved IWP (Ice Water Path [kg.m⁻²] : total atmospheric ice content).
 Bottom line : Diagram of dispersion of IWP (left) and dIWP/dt (right).

5 Variational approach method (1D-VAR)

Method :

- Iterative principle - Based on Bayesian theory [5]
- Retrieval of ice water content
- Retrieval of vertical wind velocities profile, $W(z)$
 - Gaussian model based (Fig.5)
 - Advection scheme similar to MESO-NH (Fig.6)

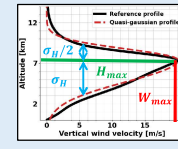


Fig.5: Wind vertical velocity profile based on gaussian model

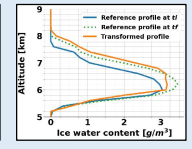


Fig.6: Advection scheme compute vertical motion between initial and final time measurement

First results :

- Fig.7: Spatial coherence of dynamical properties
- Fig.8: Consistent ice content and vertical velocity profile
- Fig.9: W_{max} and H_{max} well estimated

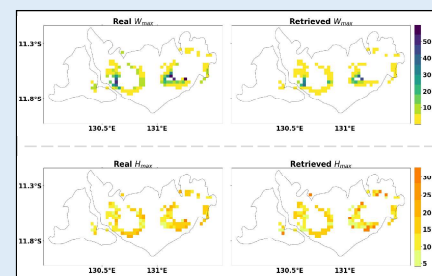


Fig.7: 1D-VAR performance retrieval map.

Top line : map of the real and retrieved Wmax. Bottom line : map of the real end retrieved Hmax.

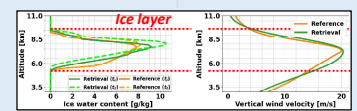


Fig.8: Real and retrieved ice profile and vertical wind velocities

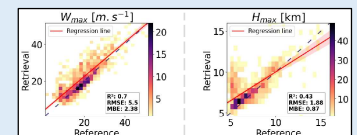


Fig.9: Scatter plot of 1D-VAR retrieval in function of reality-like values. Red line : regression line.

5 Conclusions and perspectives

Among other possible products, C²OMODO enable the retrieval of:

- Using **Machine Learning** : Cells structure and Integrated geophysical variables
- Using **1D-VAR** : Profile of : **Ice Water Content & Vertical wind velocities**

PHD perspectives : Investigation of 1D-VAR biases, Application on other cloud structure, Sensitivity analysis to tandem parameters (Tandem simulator), Publication on the 1D-VAR.



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