



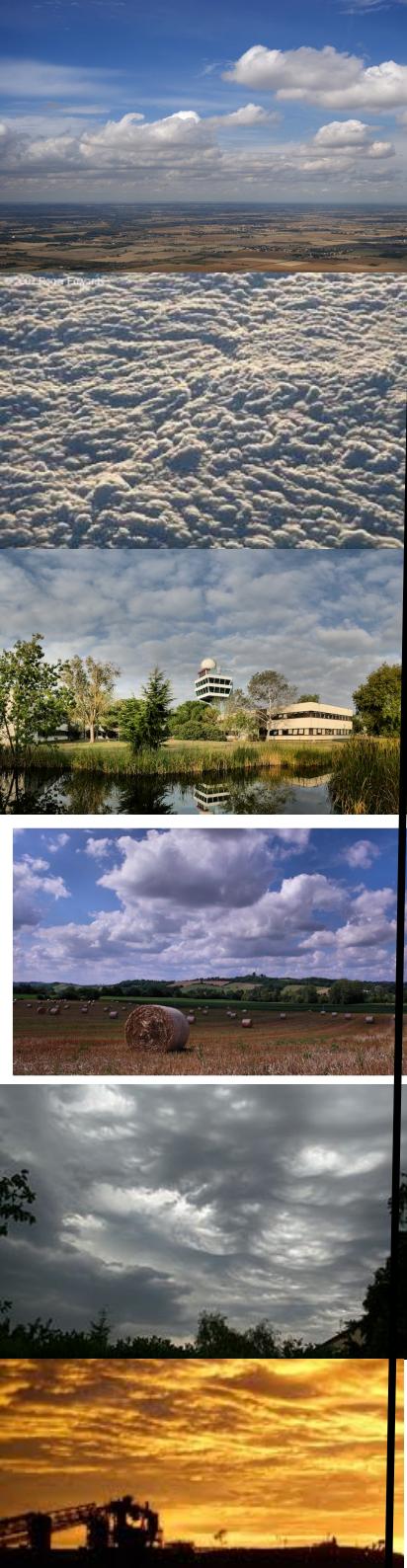
HIGH-TUNE, High-resolution simulations to improve and Tune boundary-layer cloud parameterizations

F Couvreux*, R Honnert, and the HIGH-TUNE team
(CNRM, LMD, University of Exeter)

Using state-of-the art statistical tools applied to the comparison
SCM/LES to tune the boundary-layer cloud parameterizations



More info :
Fleur.couvreux@meteo.fr



The HIGH-TUNE project

Context : important biases in NWP and climate models for the reproduction of low-level clouds

Objective : improve the parameterizations involved in the representation of low-level clouds

Deadlock : many free parameters in any parameterization

Proposition : use state-of-the-art statistical tools to propose values for those parameters based on a comparison SCM/LES on an ensemble cases

Advantages : a new way to tune parameterization one by one with 1D simulations

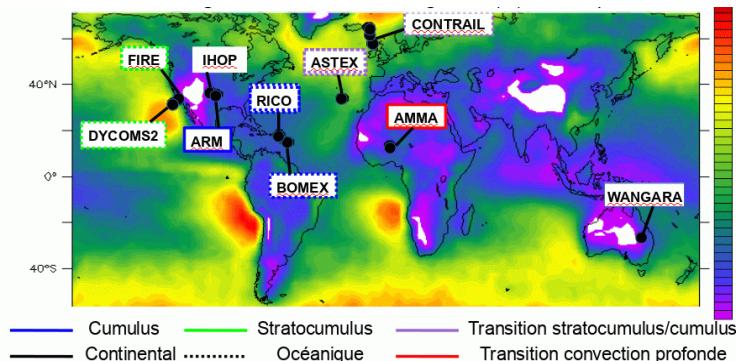
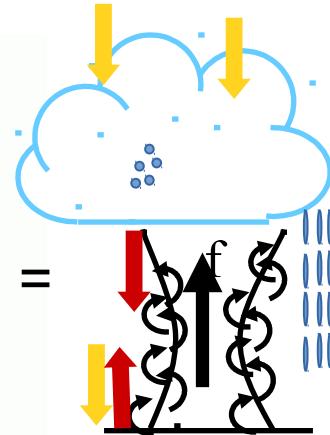
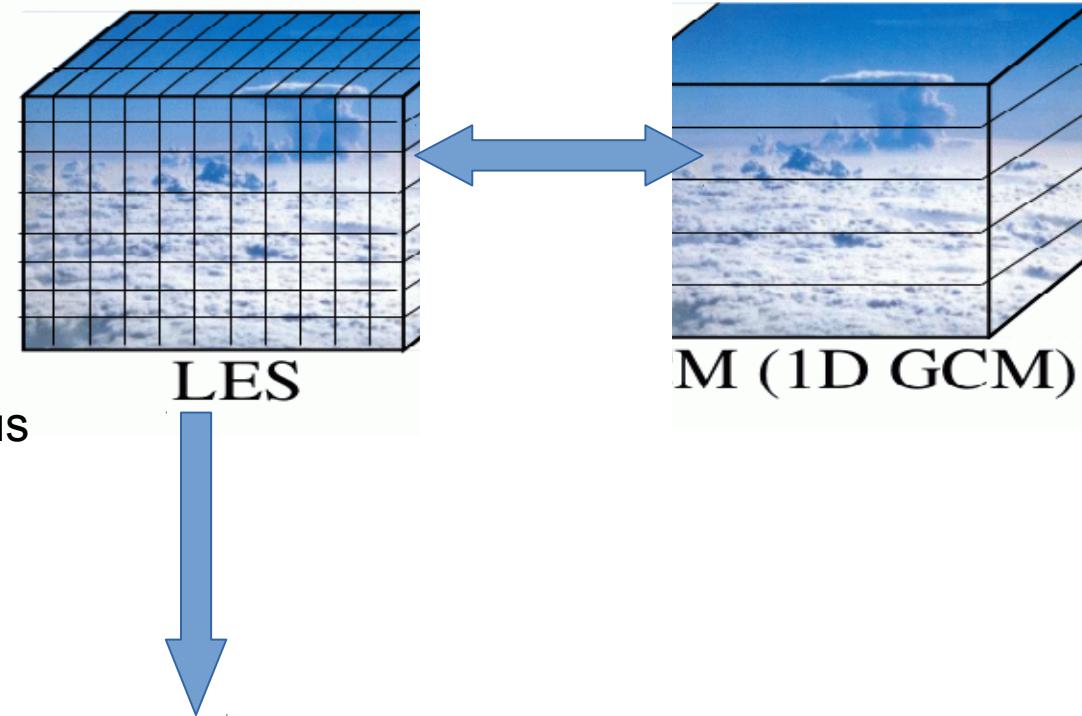
Bonus : a better understanding of the behaviour of the parameterizations ; a tool to disentangle structural errors from tuning issues

1D cases :

- BOMEX }
- RICO } Oceanic cumulus
- ARM }
- SCMS } Continental cumulus
- FIRE }
- DYCOMS } Stratocumulus

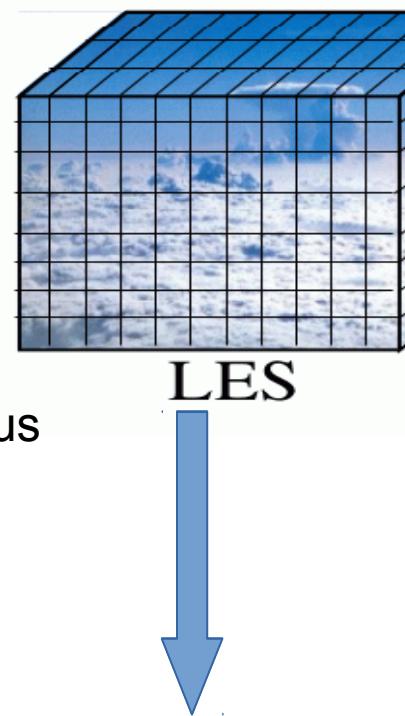
- ASTEX } Transition Cu → Stocu
- IHOP }
- WANGARA }
- AYOTTE (x6) }

- AMMA Deep convection
- GABLS4 Stable boundary layer

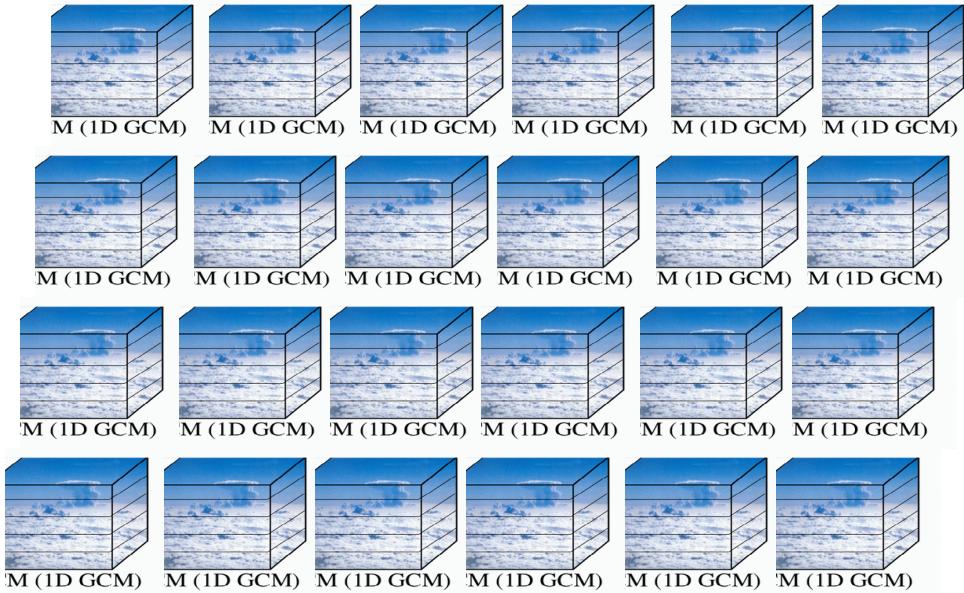


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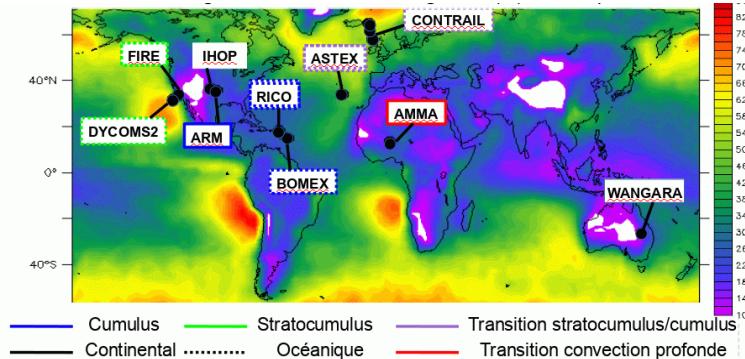


Different free parameters



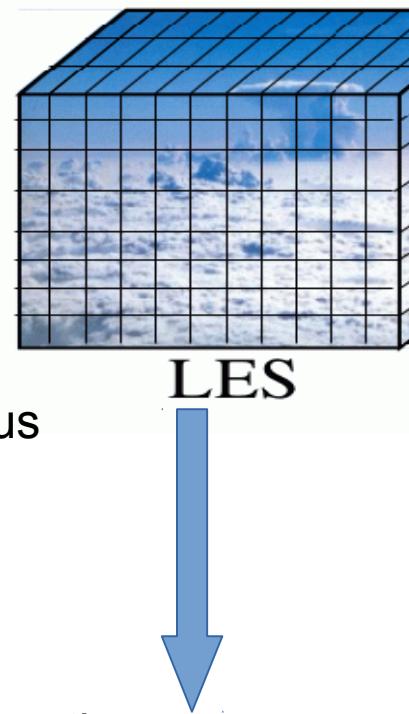
Metrics

- AMMA Deep convection
- GABLS4 Stable boundary layer

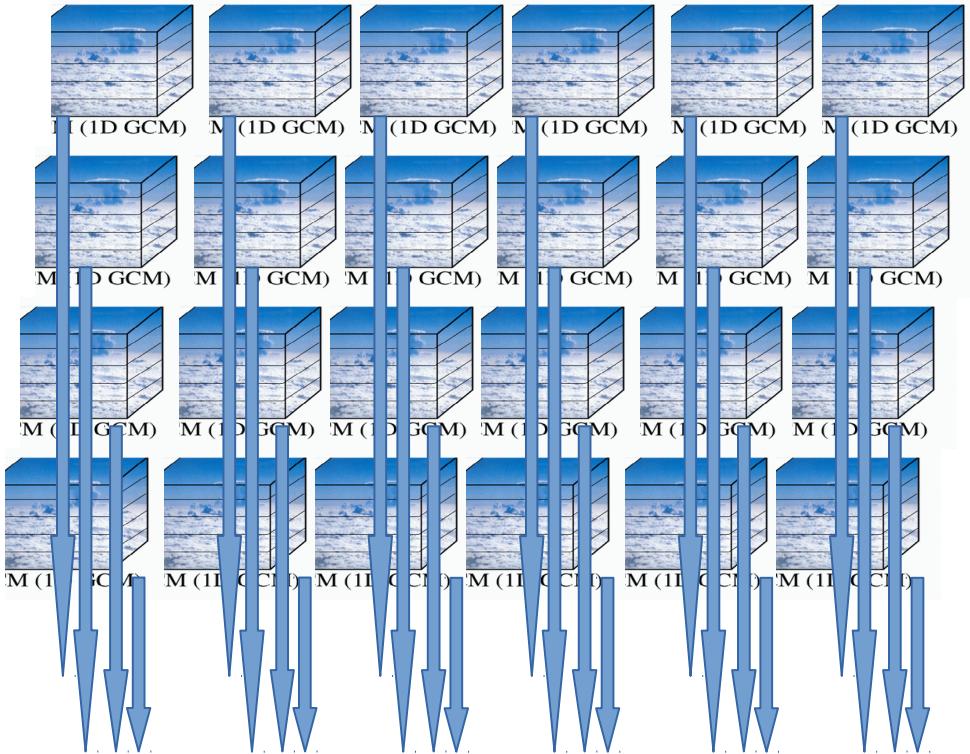


1D cases :

- BOMEX }
- RICO } Oceanic cumulus
- ARM }
- SCMS } Continental cumulus
- FIRE }
- DYCOMS } Stratocumulus
- ASTEX }
- Transition Cu → Stocu
- IHOP }
- WANGARA }
- AYOTTE (x6) } Clear convective boundary layer
- AMMA Deep convection
- GABLS4 Stable boundary layer



Different free parameters

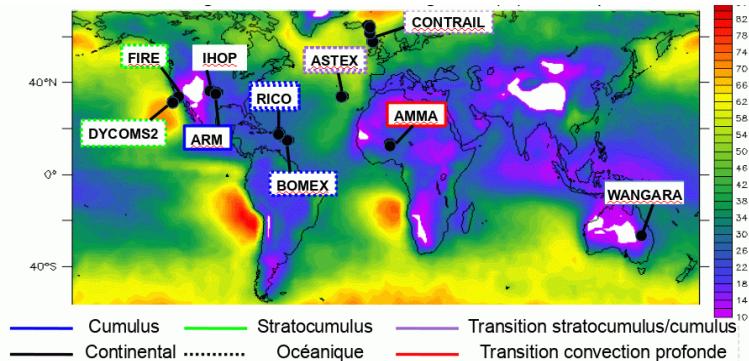


Emulator

Metrics = $f(\text{param}_1, \text{param}_2, \dots, \text{param}_n)$

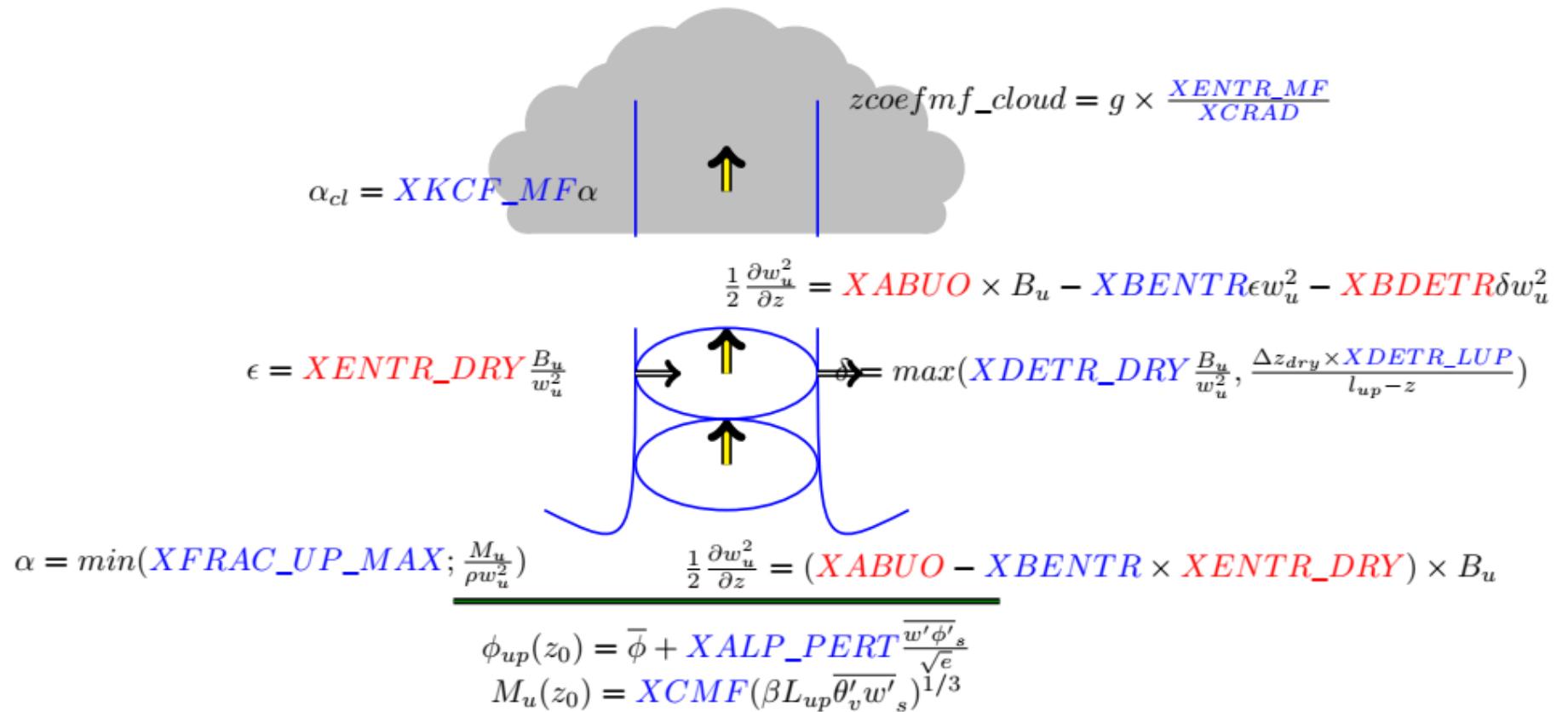
History Matching (Williamson et al 2013)

Ensemble of possible values of
($\text{param}_1, \text{param}_2, \dots, \text{param}_n$)



One example with AROME model

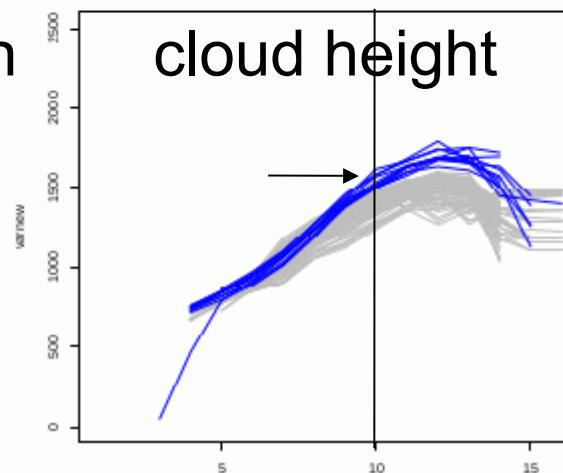
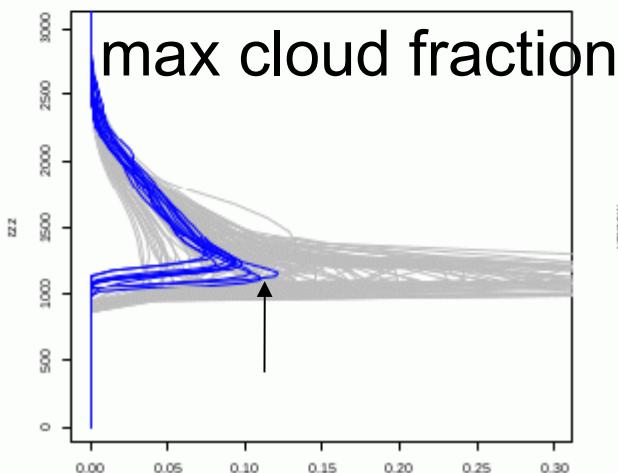
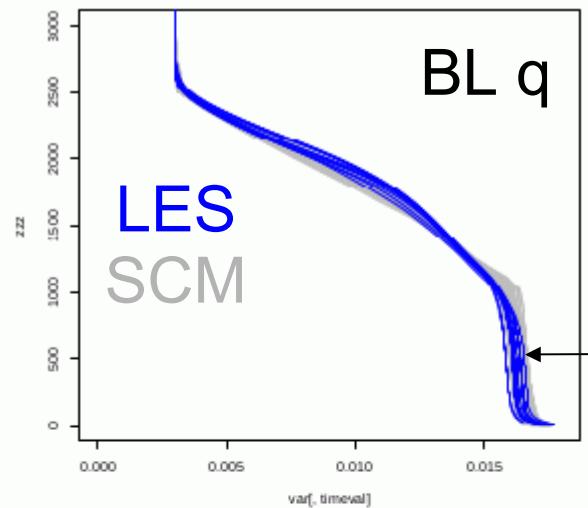
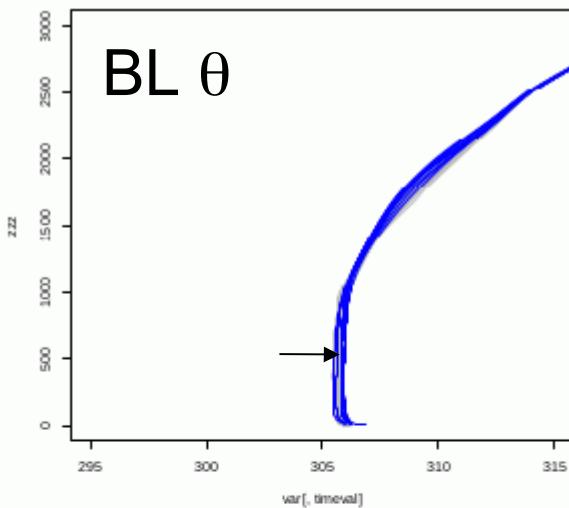
Parameters: 3 parameters from the mass-flux scheme of Pergaud et al (2009) involved in entrainment/detrainment formulation



One example with AROME model

Parameters: 3 parameters from the mass-flux scheme of Pergaud et al (2009) involved in entrainment/detrainment formulation

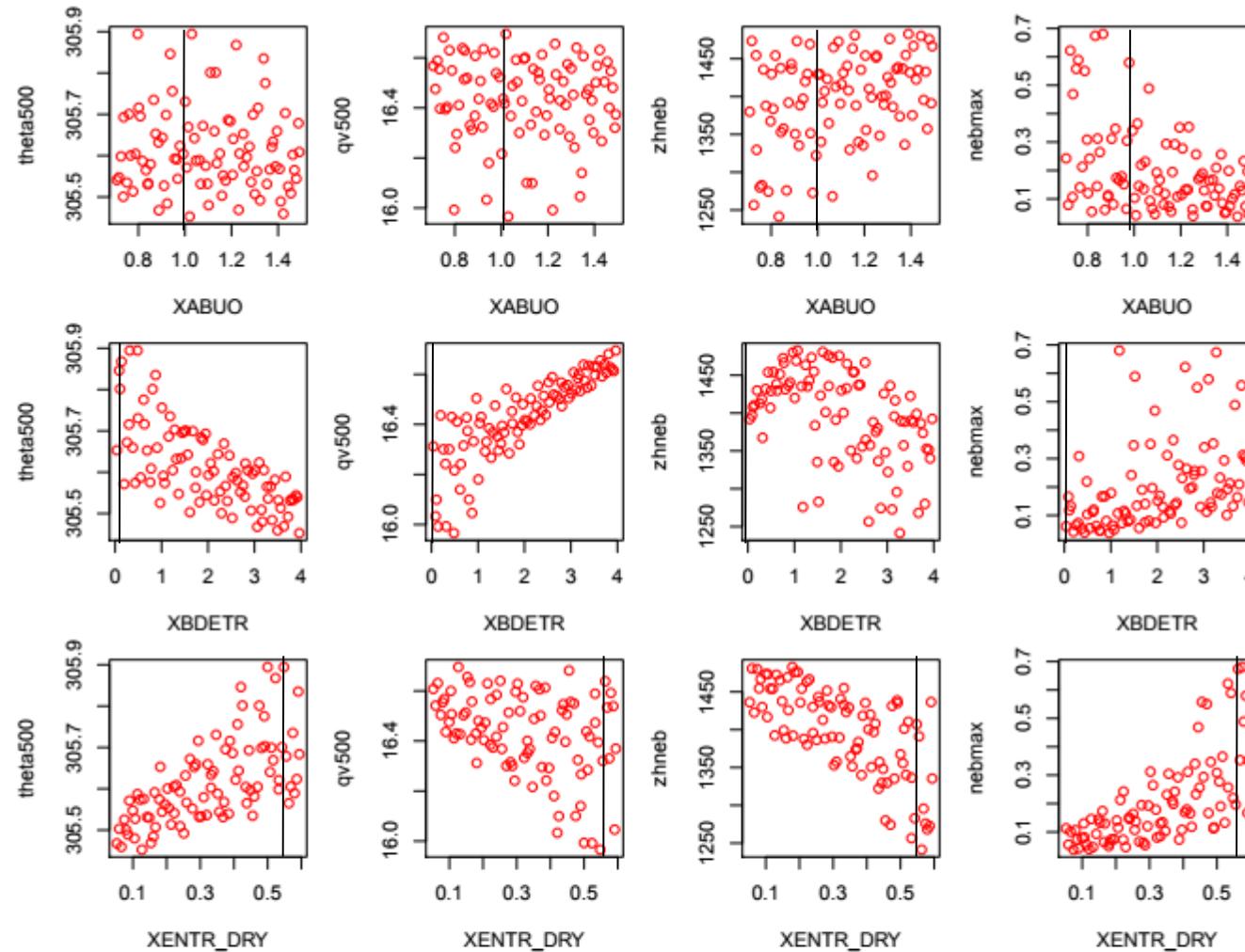
Case and metrics: ARM (Brown et al 2002) ; BL θ , BL q, max of cloud fraction, cloud equivalent height



One example with AROME model

Parameters: 3 parameters from the mass-flux scheme of Pergaud et al (2009) involved in entrainment/detrainment formulation

Statistical tools: Able to predict the metrics for any values of parameters



One example with AROME model

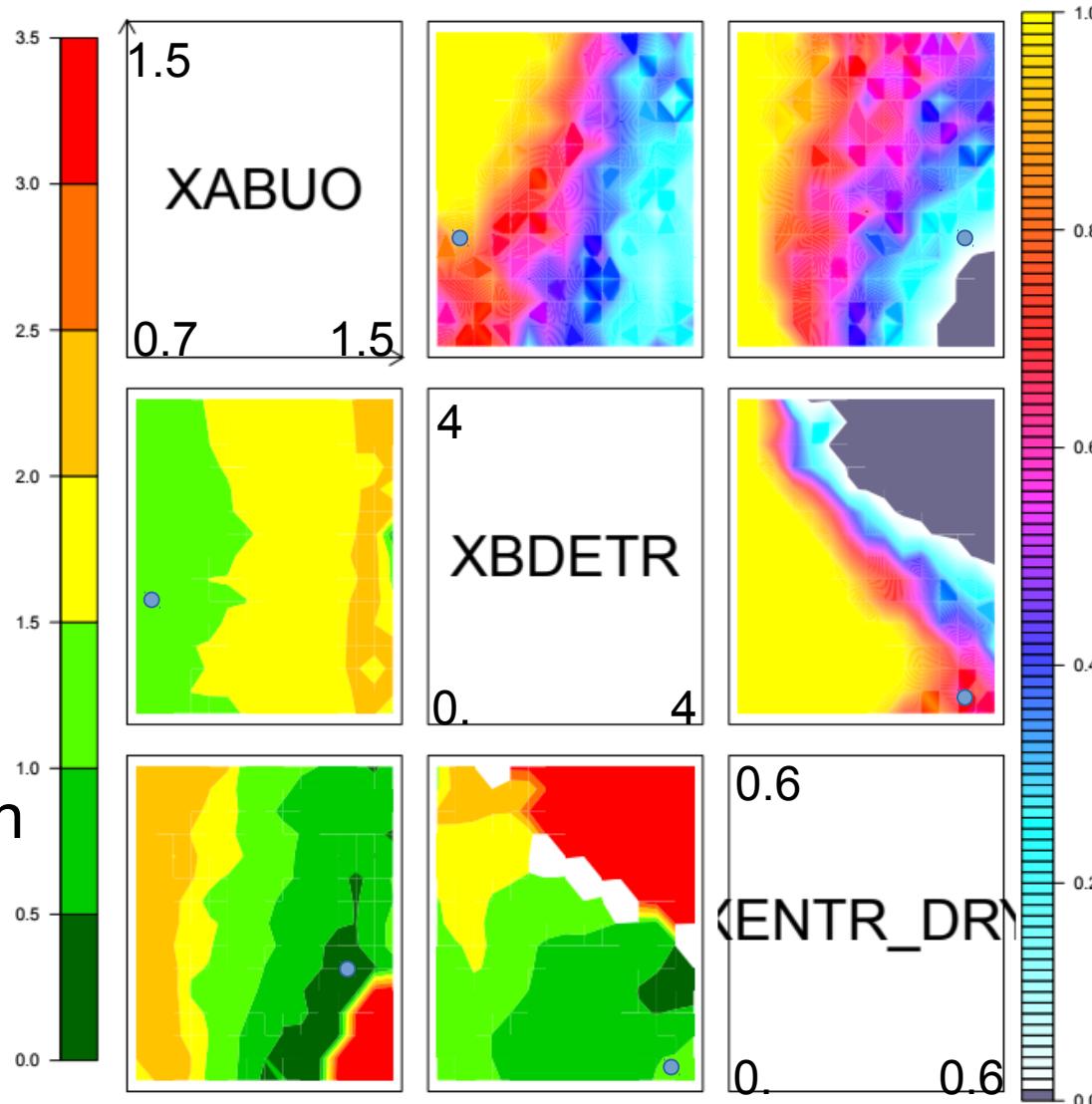
Parameters: 3 parameters from the mass-flux scheme of Pergaud et al (2009) involved in entrainment/detrainment formulation

Results: Determine the Non-Ruled Out Yet space of parameter values

Values in
• the default
operational
version

Minimum
Implausibility
in the 3D domain

$$I = \frac{(M_{LES} - M_{Emulator=f(p_1, \dots, p_n)})^2}{\varepsilon_{LES}^2 \varepsilon_{Structural}^2 + \varepsilon_{Emulator}^2}$$



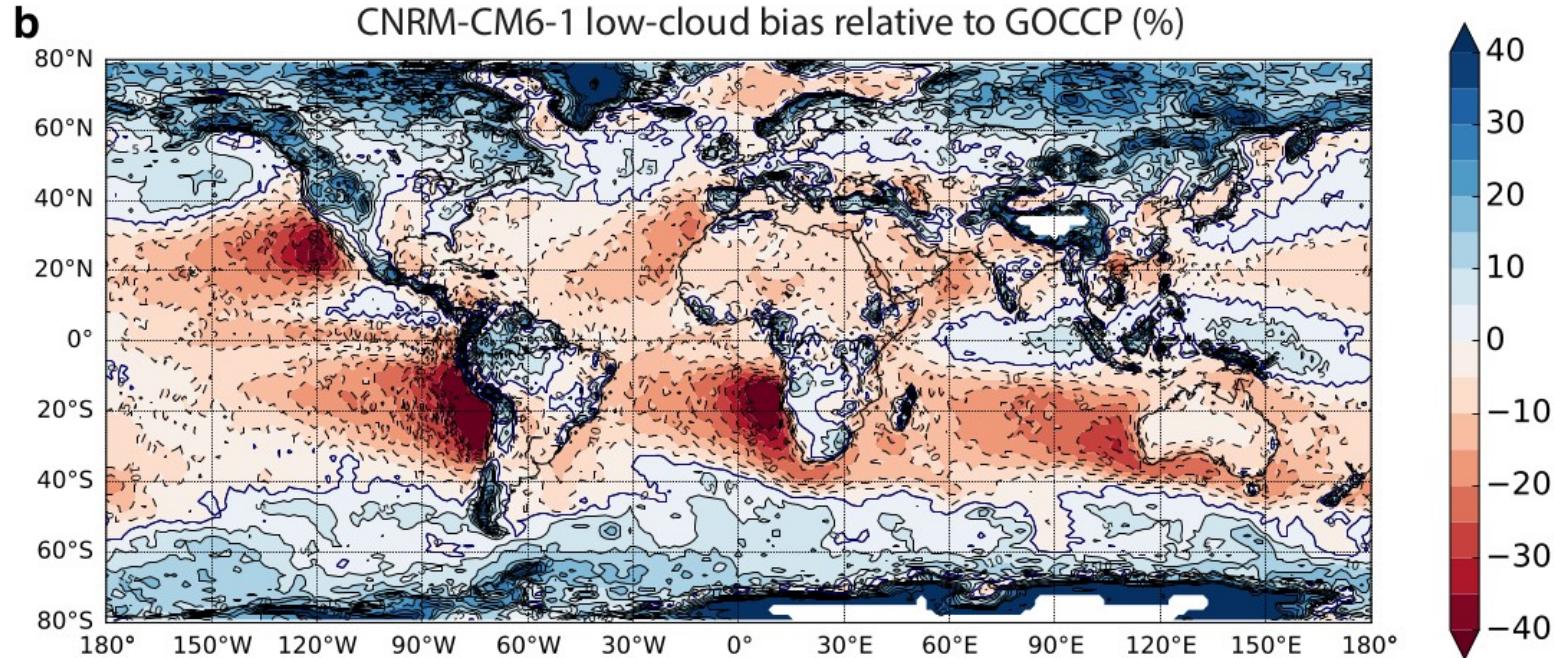
% of the 3D
domain of value
of parameters
kept



Evaluating marine stratocumulus clouds in the CNRM-CM6-1 model using short- term hindcasts

Florent Brient, Romain Roehrig, Aurore Volodire
In revision to JAMES

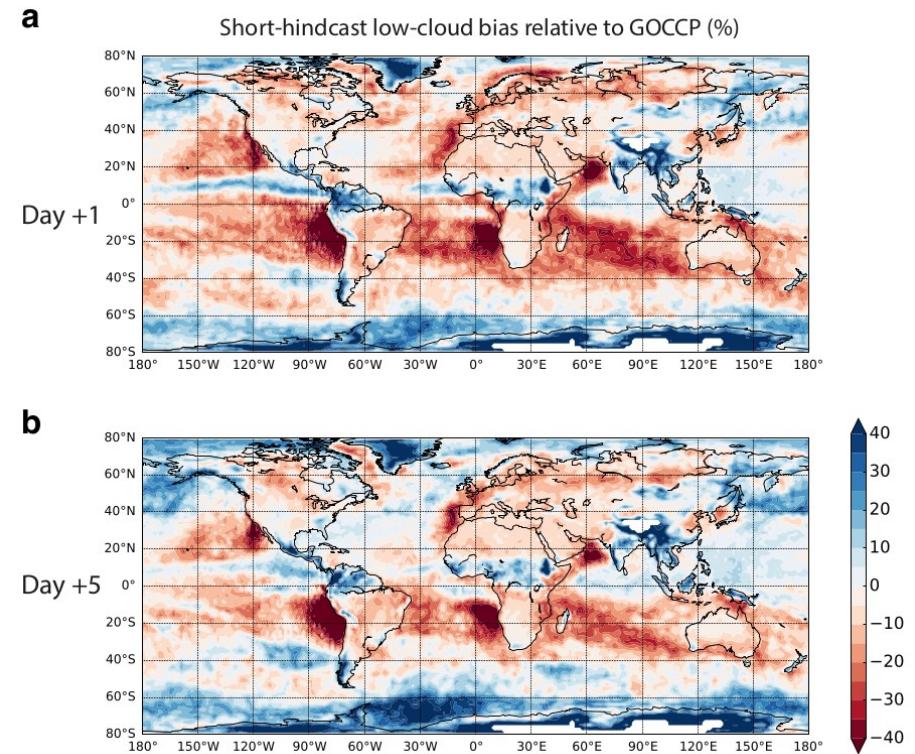
Low-cloud biases in the CNRM-CM6 model



- Strong low-cloud underestimation in the eastern parts of tropical oceans (up to -40 %)

Usefulness of short-term hindcasts

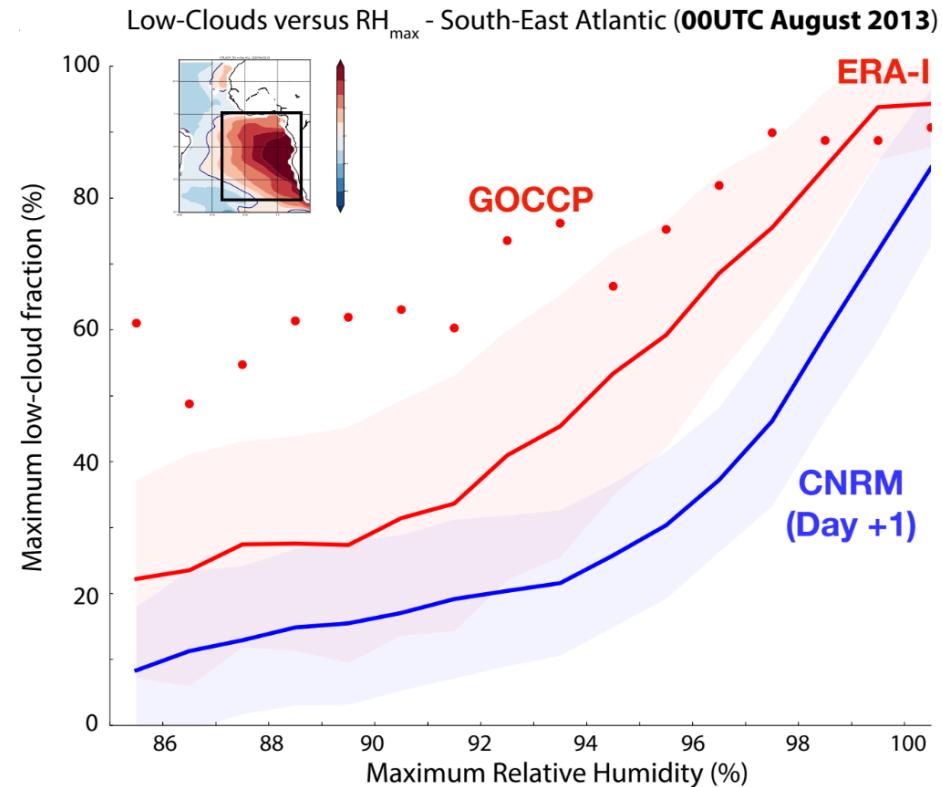
- Ensemble of short-term hindcasts starting each day of August 2013 from ERA-Interim (u, v, T, qv and ps)



- Low-cloud biases appear within only a few hours, thus independently of errors in the large-scale circulation
- Key processes underlying the low-cloud errors are thus mainly local
- Transpose-AMIP is a useful framework to evaluate and improve parameterizations

Process-oriented analysis of low-cloud errors

- RHmax vs CFmax (< 3km) in the South Atlantic
- CNRM model highly biased in all RH regimes

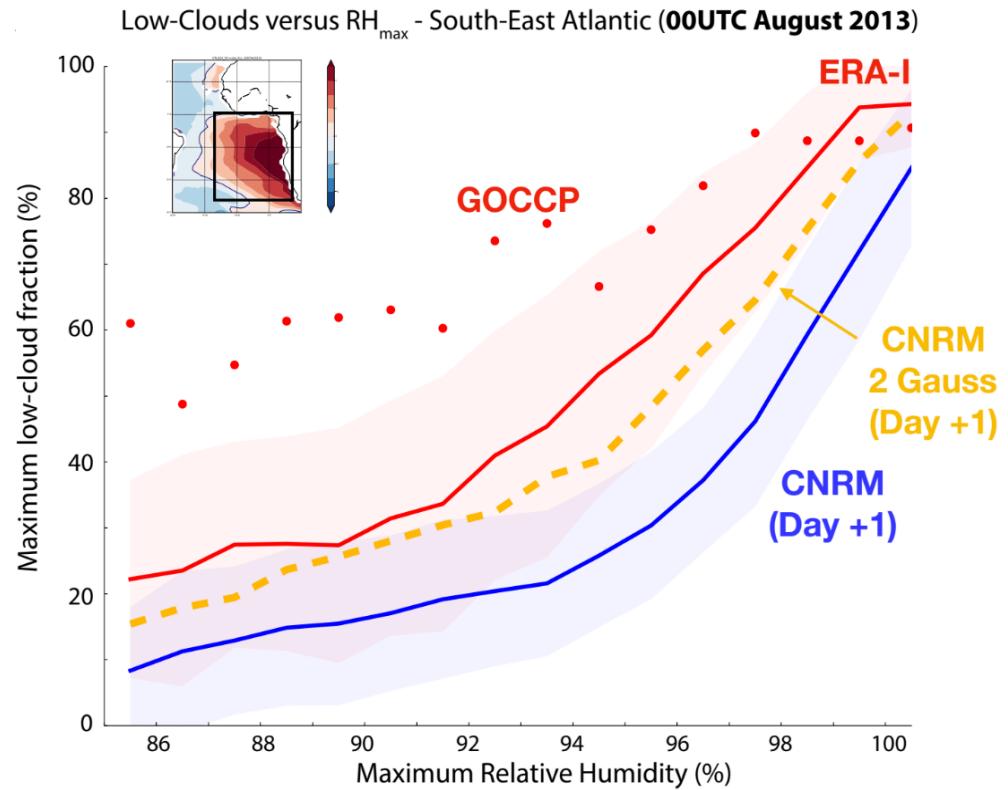


Subgrid-scale distributions of T/qv mostly explain cloud errors

- RHmax vs CFmax (< 3km) in the South Atlantic
- CNRM model highly biased in all RH regimes

$$\overline{CF} = \int_{-Q_1}^{+\infty} G(t) dt$$

$$\overline{Q_1} = a \left[\frac{\overline{q_t} - q_{sat}(\overline{T_l})}{2\sigma_s} \right]$$



- Sensitivity test to the subgrid distribution of T/qv (**G**) improve the relationship between RH and CF
- This calls a revisit of the cloud parameterization subgrid-scale distribution, e.g. by accounting for other sources of moisture subgrid-scale variability.