Multi-layer cloud conditions in trade wind shallow cumulus

Confronting models with airborne observations

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The treatment of shallow clouds over the vast, sub-tropical oceans remains a large source of uncertainty in climate models. Therefore, cloud-resolving kilometer-scale resolution models are applied in climate studies as a means of improving the cloud representation. But...



4 Overview: Cloud boundaries – The influence of different sensors

- Observation of cloud tops in two layers. Lower layer is mostly visible to lidar only.
- Both models reproduce lower layer, but only LEM clearly develops upper layer.



Fig. 1: Shallow cumulus clouds. Small size – high impact.

- ... how do those models represent marine shallow cumuli compared to observations?
- What is the best way to asses the clouds?
- And how does the liquid water path help to interpret differences between observed and simulated cloud structures?

The research aircraft HALO offers us the opportunity to answer this question with respect to two cloud-resolving models.

2 Airborne observations and atmospheric models

Nadir pointing instruments on-board the <u>High</u> <u>Altitude</u> and <u>LO</u>ng range research aircraft (HALO):

- Aerosol backscatter <u>lidar</u>: Backscatter ratio (BSR) detects cloud top height of small cloud droplets.
- Cloud and precipitation radar: Radar reflectivity is scattered back by large droplets and precipitation from cloud top to base.
- Microwave <u>radiometer</u>: Retrieval of integrated liquid water path.



Fig. 6: Cloud boundaries in all observations and forward simulated radar and lidar signals. Same thresholds for cloud detection are used for the observed and simulated radar and lidar signals. Height is in relation to the lifted condensation level (Icl). Shadings depict western (bright edge) and eastern (dark edge) half of each dataset. Observations are from RF 1 to 8. SRM data are sub-sampled (0.5°, hourly) for 24 days. LEM data are taken from 10 grid points at high temporal resolution (every 36 s) for 4 days. All data is during daytime (\sim 8 AM to 5 PM local time).

5 Details: Liquid water path enriches cloud analysis



ICOsahedral Nonhydrostatic storm resolving model (<u>SRM</u>)

- Forced with ECMWF data
- 1.25 km grid, 75 levels
- One-moment microphysics
- Resolves deep convection

ICON large eddy model (<u>LEM</u>)

- Nested in SRM
- 300 m grid, 150 levels
- Two-moment microphysics
- Resolves cloud circulation





Fig. 4: Example scene from ICON LEM. Forward simulated radar signal and lidar cloud top height from meteogram output. Height Heigh Heigh Lifted condensation leve Lifted condensation leve Lifted condensation leve HALO (94.1%) —— HALO (74.8%) --- HALO ICON SRM 1.25 km (98.7%) ICON SRM 1.25 km (2.0%) --- ICON SRM 1.25 km -1000-1000-1000ICON LEM 300 m (26.2%) --- ICON LEM 300 m ICON LEM 300 m (99.7%) 0.6 0.8 0.6 0.8 0.6 0.2 0.4 0.2 0.4 0.2 0.4 Normalized frequency (km⁻¹) Normalized frequency (km⁻¹) Normalized frequency (km⁻¹) 3000 3000 3000 g/m² lcl (m) lcl (m) /e lcl (m) LWP 2000 2000 2000 Deepening of 1000 clouds with LWP. 300 V 1000 1000 ab 100 Height Heigh Heigh Lifted condensation leve Lifted condensation leve ifted condensation level HALO (97.1%) — HALO (91.1%) HALO ICON SRM 1.25 km (98.2%) ICON SRM 1.25 km (21.9%) --- ICON SRM 1.25 km -1000-1000-1000ICON LEM 300 m (99.8%) ICON LEM 300 m (46.6%) --- ICON LEM 300 m 0.4 0.6 0.8 0.2 0.4 0.6 0.8 0.4 0.6 0.2 0.2 Normalized frequency (km⁻¹) Normalized frequency (km⁻¹) Normalized frequency (km⁻¹) 3000 3000 3000 g/m² lcl (m) LWP lcl (m) (m 2000 2000 Models produce 2000 -Ω high-LWP clouds 1000 without rain. 1000 1000 1000 ab 300 Height Lifted condensation leve Lifted condensation leve — HALO (96.3%) HALO (96.3%) HALO ICON SRM 1.25 km (96.1%) — ICON SRM 1.25 km (78.9%) ICON SRM 1.25 km -1000-1000-1000--- ICON LEM 300 m ICON LEM 300 m (99.4%) ICON LEM 300 m (76.6%)

1000

3 Benefit of forward simulations

The observable signals are forward simulated from drop size distributions of cloud and rain water given by the models. The lidar signal is sensitive to the number of droplets and therefore depends only on the high number of small cloud droplets. The radar signal is more sensitive to large droplets and thus detects rain or thick clouds.



Fig. 7: Cloud boundaries classified by liquid water path (LWP) in observations and forward simulated radar and lidar signals.

0.2

6 Conclusions and outlook

0.2

0.4

Normalized frequency (km⁻¹)

0.6

0.8

• Lidar and radar forward simulates allow to impose instrumental thresholds to model data.

1000

V

- Connection with retrieved LWP helps to understand differences between models and observations.
- Comparison reveals lack in clear layer separation in SRM.

0.6

0.4

Normalized frequency (km⁻¹)

0.8

- Both models are unable to represent larger but non-raining droplets (drizzle).
- Methods are ready to be applied on even more coordinated model and observation activities during the upcoming EUREC⁴A campaign Jan/Feb 2020.



often

precipitate.

0.8

0.8

0.8

0.6

1000

Fig. 8: EUREC⁴A: Elucidating the Role of Cloud-Circulation Coupling in Climate.

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0.2

0.4

Normalized frequency (km^{-1})

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