



Bridging observations and LES towards reducing uncertainties in convectionpermitting models

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1 Land-surface, and Aeros

Transitions in Convective Cloud Populations TAerosol Populat

errelationships

Land-Atmosphere-Cloud Interactions

ICLASS -



Motivation and Objective

- Convection-permitting models (CPM) are the future of Earth System Models
- But CPMs have various dynamical and microphysical biases
- Deep convection initiation and growth under realistic environments are poorly understood, near-cloud environmental factors and key cloud structures are difficult to observe
- **Goal:** Better understand processes controlling deep convective cloud growth under a variety of realistic environmental conditions during CACTI through developing an observation-model integration framework, and ultimately jointly improve model and observation capabilities







Science questions:

- 1. Are deep convection populations and their associated precipitation realistic in DYAMOND GCPMs?
- 2. How well are mesoscale convective systems (MCSs) represented?

Key findings:

- Diverse range in simulating tropical DCC and MCS
- Most models overestimate **DCC/MCS** in Maritime Continents (MC), but underestimate tropical MCSs over continents, Indian/Atlantic Oceans, SPCZ
- **All models** overestimate MCS precipitation in MC, but most underestimate those in other tropical regions

Feng et al. (2022), in prep.



 Δ Number of MCS (T_b+PF)



Examples of Simulated DCC and MCS



SCREAM



- SCREAM simulated a lot more unorganized DCC ("pop corn" convection) and less MCSs than OBS •
- DYAMOND models results are quite diverse in the morphology of organized convection •





Model MCS Properties and Interpretations

- **Most models** capture MCS lifetime, cloud-shield area and total volume rain quite well
- Widespread max cloud-top height (min T_b) but generally deeper than **obs**., indicate updraft intensity may be too strong over ocean
- All models underestimate PF area (stratiform bias is common), overestimate rain intensity (too much convective rain)



Feng et al. (2022), in prep.



Pacific

MCSs in WRF Show Similar Biases



- Simulated rainfall biases differ between short-lived and long-lived MCSs \rightarrow biases depends environmental conditions
- Model biases cannot be solely explained by satellite-retrieved rainfall • uncertainties, as rain gauge comparisons show similar bias with smaller magnitude

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Zhang et al. (2021) MWR

Approach: Convective Cell Tracking Database





Continental Convective Upscale Growth Biases in WRF during CACTI Northwest



• Applied cell tracking to 6-month WRF simulations (3 km)

Pacific

- CPM tend to over produce convective cells in low MUCAPE environments
- Model fails to simulate observed dependence of cell area growth on MUCAPE when cell formations are abundant

Zhang et al. (2022), in prep.





Opportunities from LASSO-CACTI Simulations

- LASSO offers opportunities to understand resolution dependence on convective upscale growth in realistic environments
- We adapted PyFLEXTRKR to track convective cells in LASSO simulations at CPM and LES grid spacings
- Tag environmental conditions for each tracked cell at CI locations in LASSO
- Multiple LASSO simulations days are coming online, large CPM ensembles for 20 events (~660 simulations) are available:

https://www.arm.gov/news/blog/post/778 33





Updraft Width Dependence on Relative Humidity

- Wide updrafts in 2.5 km runs are associated with drier mid-level RH than in 500 m runs, which may suggest updrafts in CPM are less sensitive to environmental humidity (i.e., weaker turbulent entrainment effects)
- More work is needed to disentangle other processes contributing to these differences



*Narrow cells: ≤ 1/3 updraft area distribution; Wide cells: ≥ 2/3 updraft area distribution



PyFLEXTRKR Software Package for Community Use

- PyFLEXTRKR (Python-based atmospheric feature) tracking software package)
- Current capabilities:
 - Tracking convective cells using radar reflectivity data [Feng et al. (2022) MWR]
 - Tracking MCSs using satellite (T_b) data, or model outgoing longwave radiation (OLR) data, with optional collocated precipitation data to identify robust MCSs [Feng et al. (2021) JGR]
 - Generic 2D objects defined by simple thresholds
- Works on observations and model outputs, optimized to run on large datasets, scalable parallelization
- Provides visualization scripts, Jupyter notebooks for statistical analysis
- Now available: • https://github.com/FlexTRKR/PyFLEXTRKR



Feng et al. (2022), submitted

Precipitation (mm h⁻¹)



Path Towards Reducing Uncertainties in CPMs

- Develop an observation-model integration framework centered around Lagrangian convective cloud lifecycle
- Build a large, comprehensive convective feature tracking database from ARM OBS and available CPM/LES in multiple regimes, use observationallyconstrained simulations to understand and improve processes contributing to CPM biases
- Example from **CACTI-LASSO** efforts:
 - Radar cell tracking database released: https://doi.org/10.5439/1844991
 - OBS: will add synoptic forcing, aerosols, and cloud vertical structures
 - LASSO: matched updraft/downdraft statistics, nearcloud environments, updraft entrainments
- We welcome collaborations and contributions



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