

Advancing the Use of ARM Observations for Large-Scale Earth System Model Development

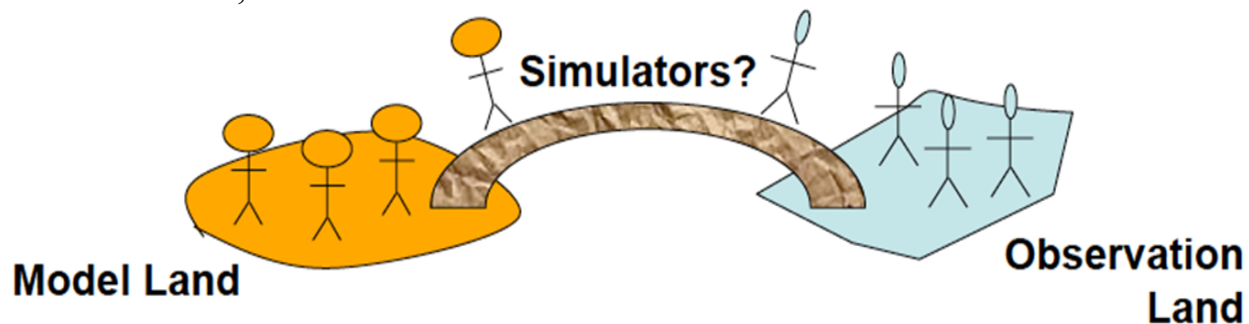
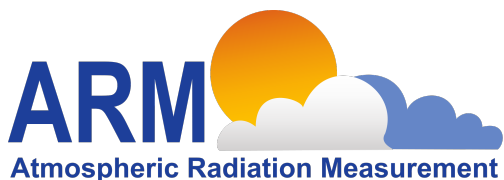
Breakout session

Co-organized by Katia Lamer, Daniel Feldman, Jiwen Fan and Shaocheng Xie

2019 ARM/ASR PI meeting

Rockville, Maryland

June 10, 2019



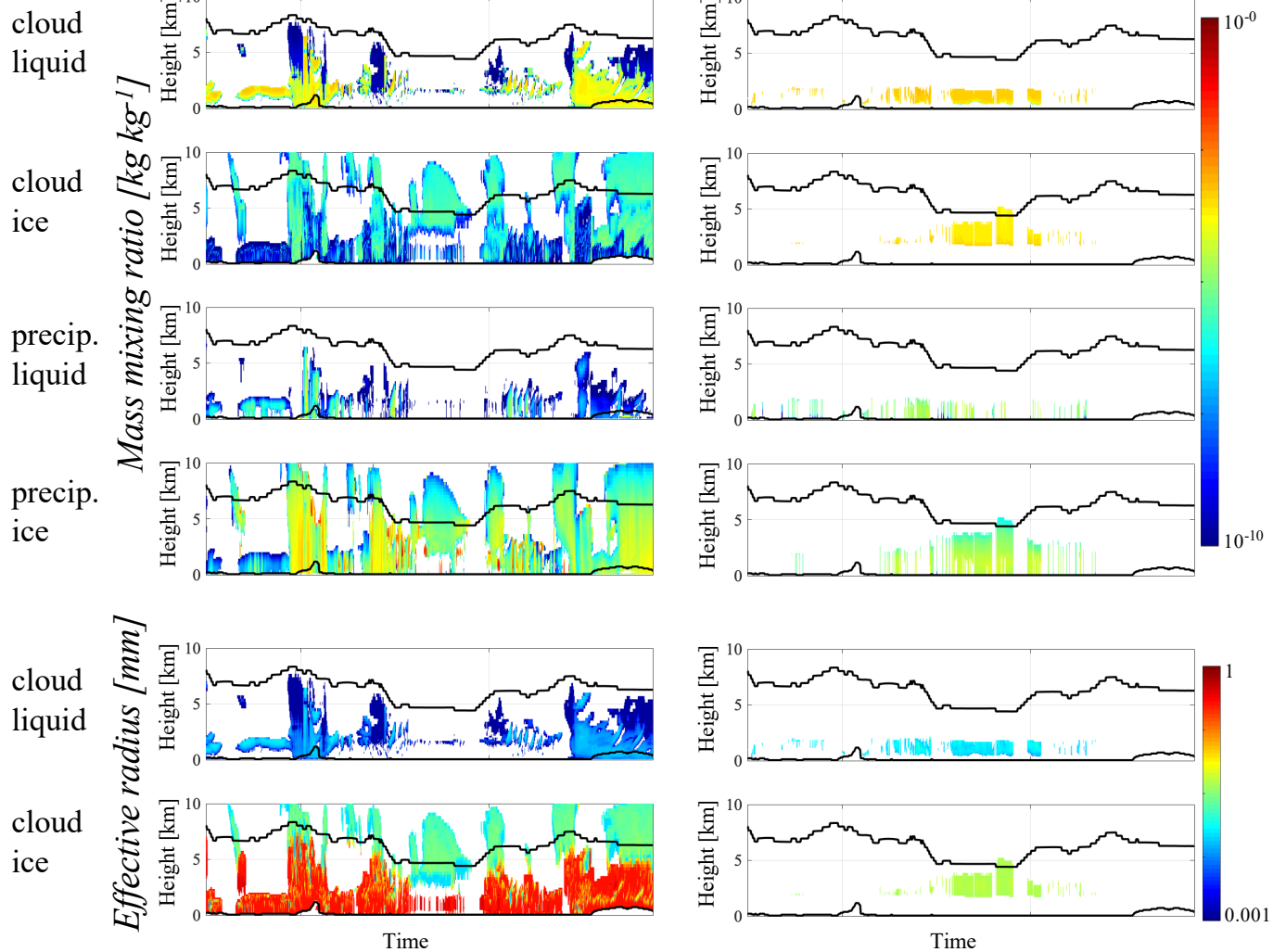
Challenges in the intercomparison of ground-based observations and large-scale models

	Large-scale models	Ground-based observations
Definition	Fractional area coverage Mass-weighted hydrometeor properties	Size-related hydrometeor properties
Detection	All hydrometeor are simulated Presence of numerical noise	Only a fraction of all hydrometeor is detectable Detection limitations are instrument specific Presence of instrument noise
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Statistical significance	Perturbations Sensitivity to parameters	Partial uncertainty quantification In retrieval space In forward-simulation space

Large-scale models

Stratiform hydrometeors

Convective hydrometeors

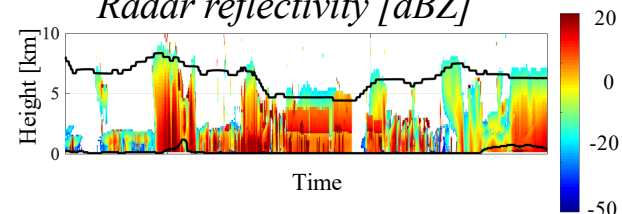


Ground-based observations

Radar Observations

cloud liquid + cloud ice
+ precip liquid + precip ice

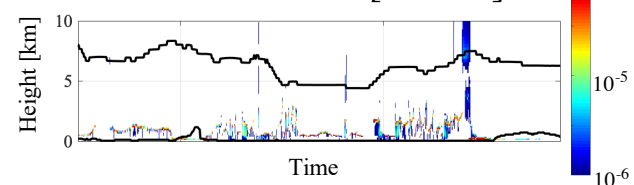
Radar reflectivity [dBZ]



Lidar Observations

cloud liquid + cloud ice

Lidar backscatter [$\text{m}^{-1} \text{sr}^{-1}$]

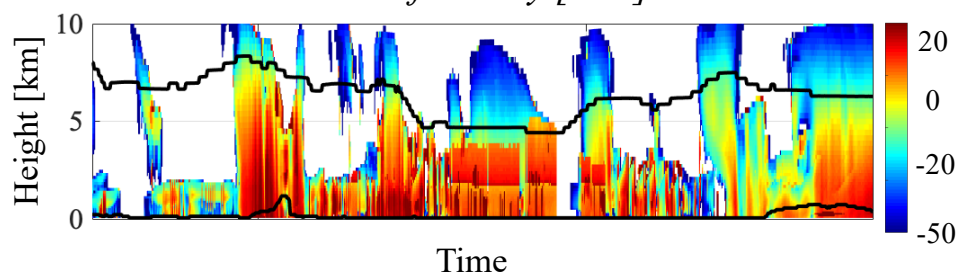


Large-scale models Forward-simulation

Radar forward-simulation

cloud liquid + cloud ice
+ precip liquid + precip ice

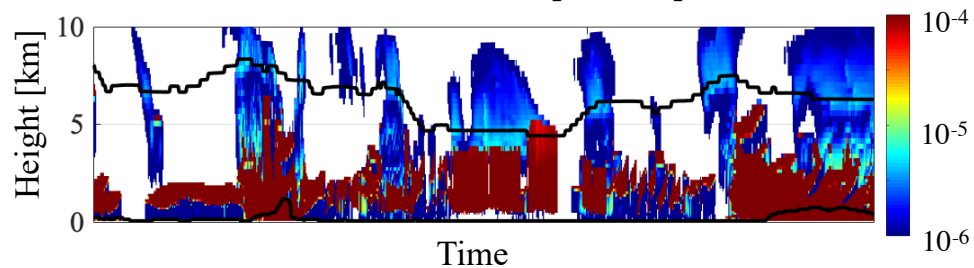
Radar reflectivity [dBZ]



Lidar forward-simulation

cloud liquid + cloud ice

Lidar backscatter [$m^{-1} sr^{-1}$]

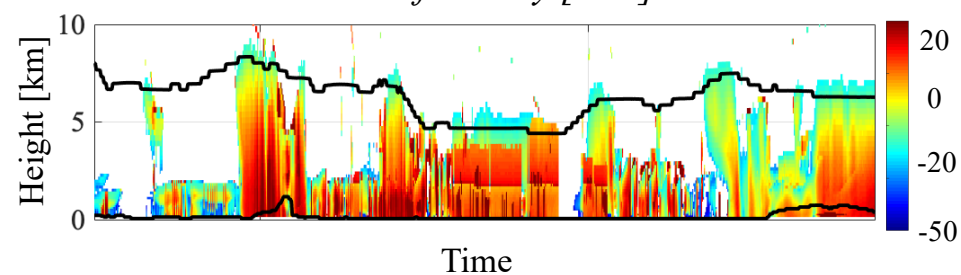


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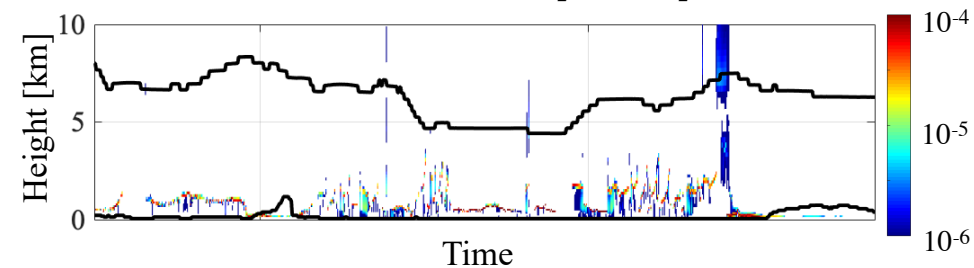
Radar reflectivity [dBZ]



Lidar observations

cloud liquid + cloud ice

Lidar backscatter [$m^{-1} sr^{-1}$]



Challenges in the intercomparison of ground-based observations and large-scale models

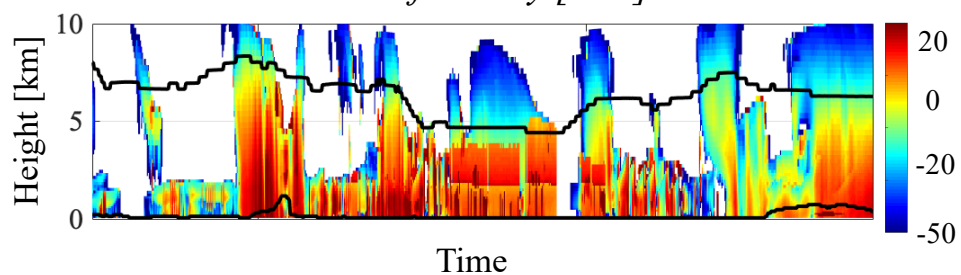
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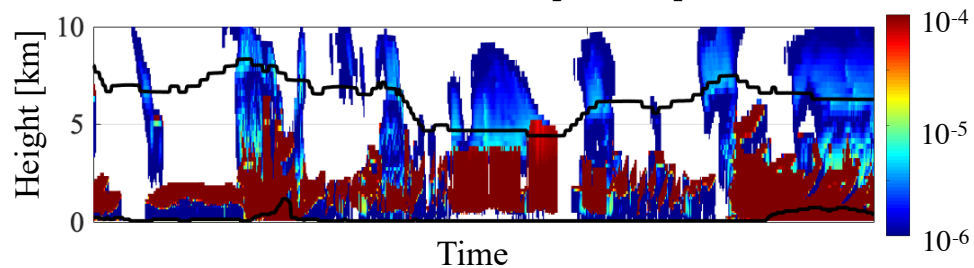
Radar reflectivity [dBZ]



Lidar forward-simulation

cloud liquid + cloud ice

Lidar backscatter [$m^{-1} sr^{-1}$]

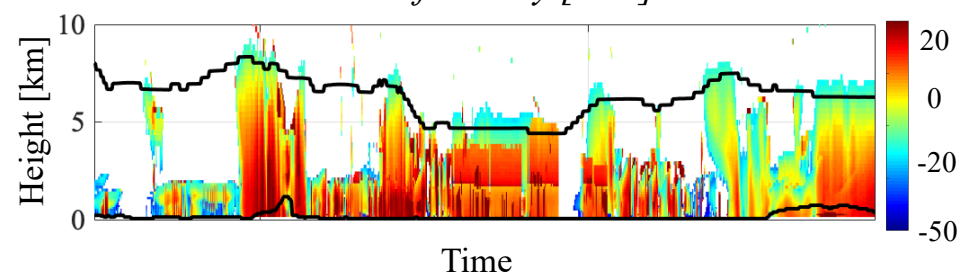


Ground-based observations

Radar observations

SOME OF THE cloud liquid + cloud ice
+ precip liquid + precip ice

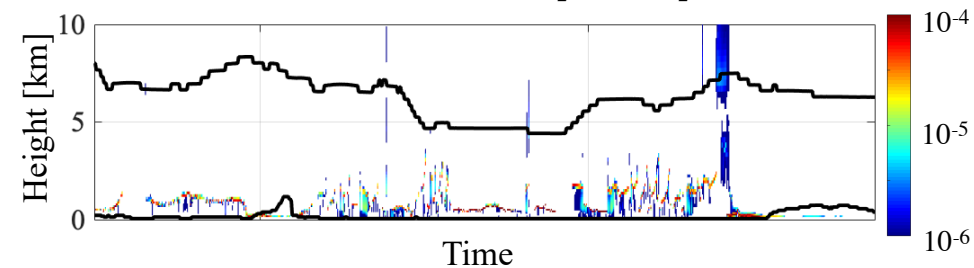
Radar reflectivity [dBZ]



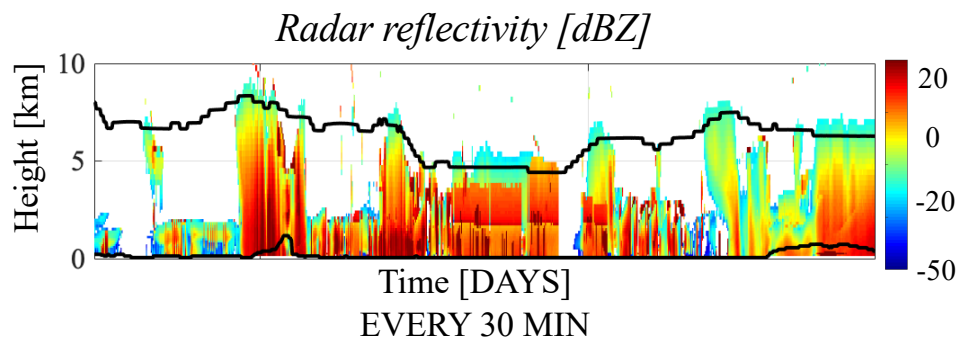
Lidar observations

SOME OF THE cloud liquid + cloud ice

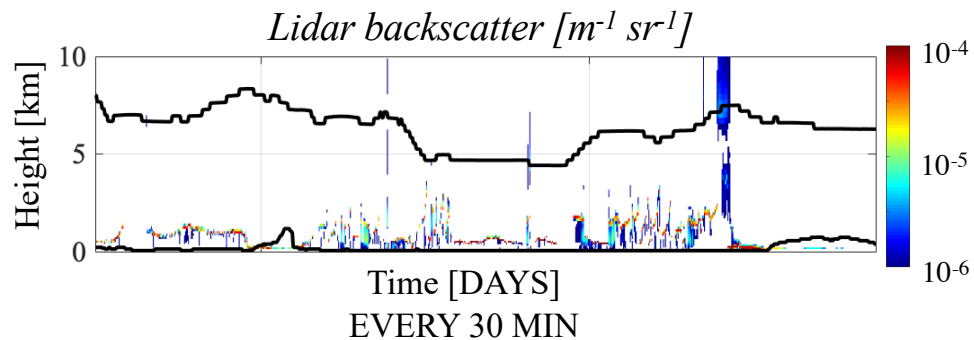
Lidar backscatter [$m^{-1} sr^{-1}$]



**Large-scale models Forward-simulation
with Instrument model**
Radar forward-simulation

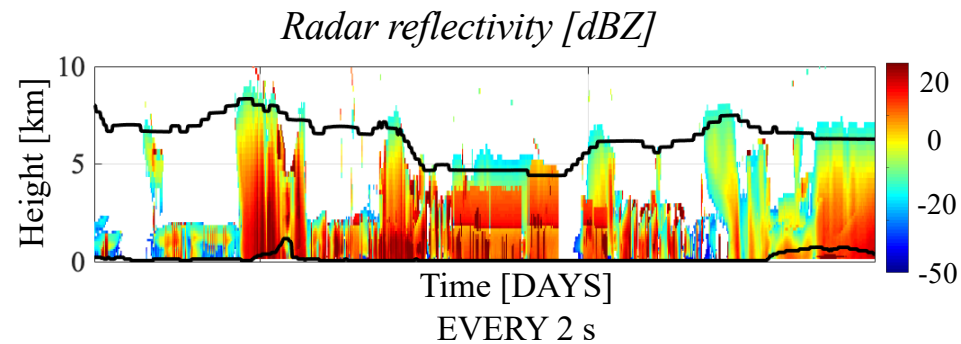


Lidar forward-simulation

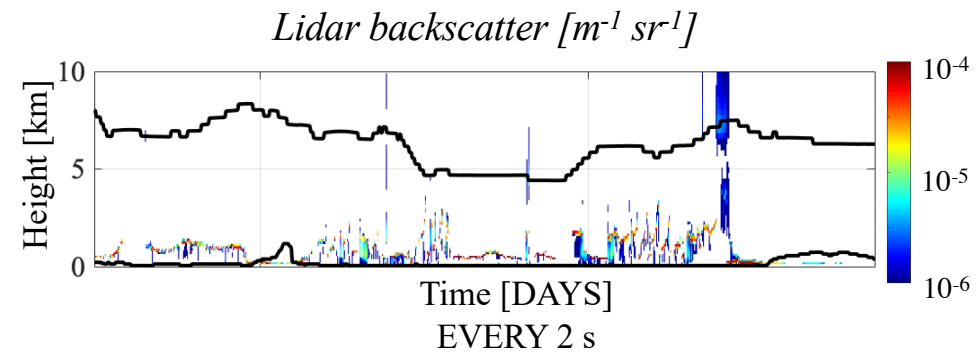


Ground-based observations

Radar observations
SOME OF THE cloud liquid + cloud ice
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Lidar observations
SOME OF THE cloud liquid + cloud ice



Challenges in the intercomparison of ground-based observations and large-scale models

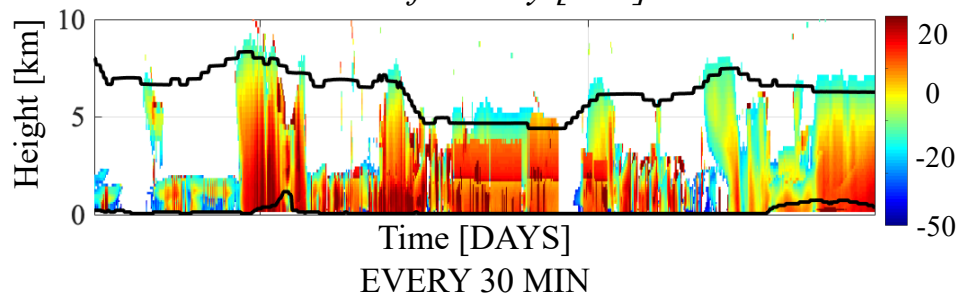
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Large-scale models Forward-simulation with Instrument model

Radar forward-simulation

FOR A 2° x 2.5° LAT x LON GRID BOX
ROUGHLY 100 KM x 100 KM

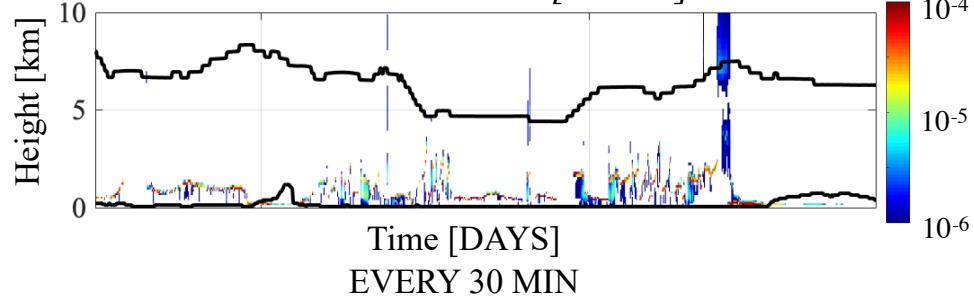
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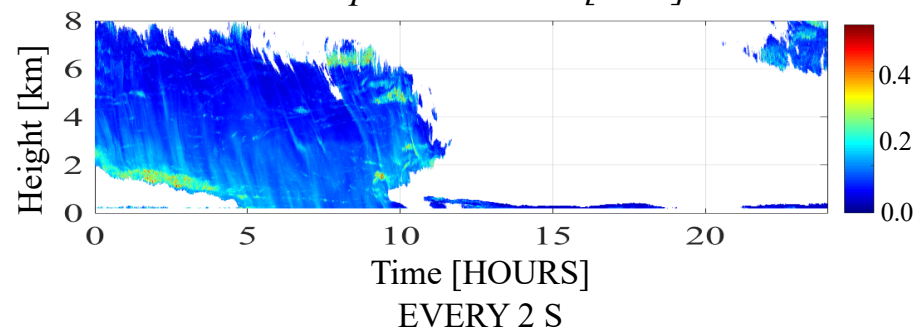
Lidar backscatter [$m^{-1} sr^{-1}$]



Ground-based observations

Radar observations FOR A PENCIL BEAM

Radar spectrum width [$m s^{-1}$]

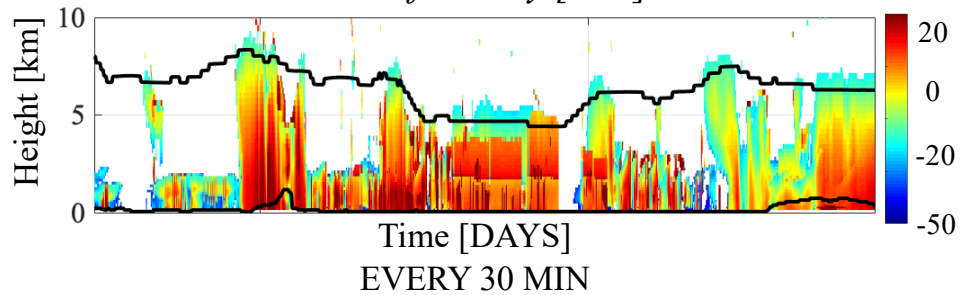


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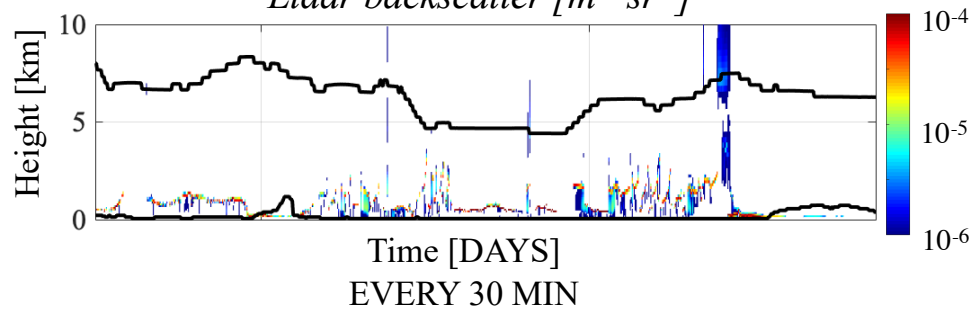
Radar reflectivity [dBZ]



Lidar forward-simulation

FOR A 2° x 2.5° LAT x LON GRID BOX
ROUGHLY 100 KM x 100 KM

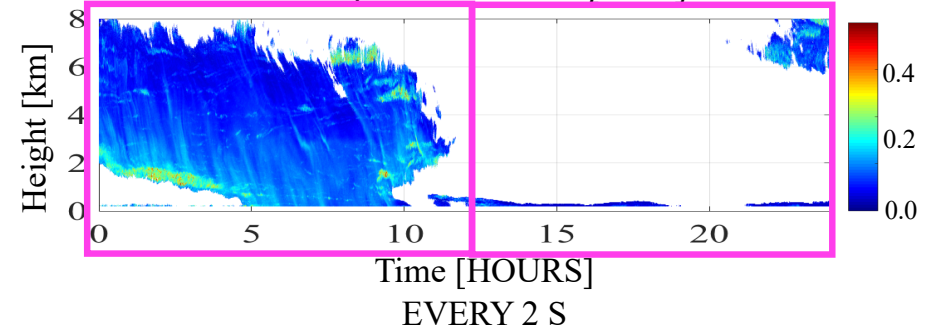
Lidar backscatter [$m^{-1} sr^{-1}$]



Ground-based observations

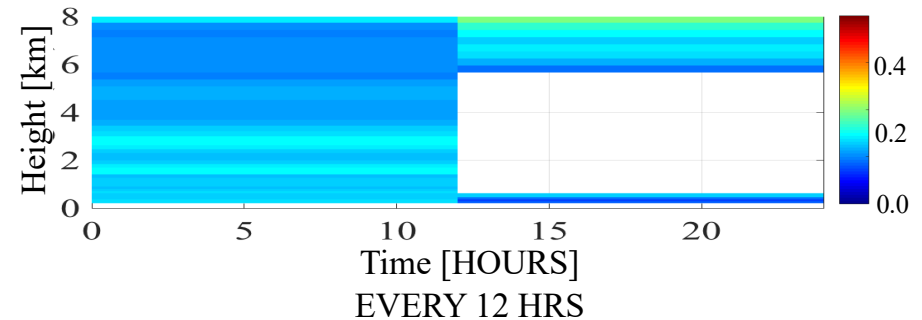
Radar observations FOR A PENCIL BEAM

Radar spectrum width [$m s^{-1}$]



Roughly the wind should take 12 hrs to advect
hydrometeor forming within 100 km from the site

?DEGRADED?



Challenges in the intercomparison of ground-based observations and large-scale models

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Agenda

- 1:40-1:50** Comparison of vertically pointing and scanning observations for the retrieval of domain average light precipitation rate – Katia Lamer
- 1:50-2:05** Using CPOL data as an observational target for E3SM – Robert Jackson
- 2:05-2:20** Evaluation of simulated convective cloud system and precipitation using ARM data in combination with NEXRAD and GPM data – Jiwen Fan
- 2:20-2:35** Evaluation of parameterizations of mesoscale convective organization in Earth System Models – Daehyun Kim
- 2:35-2:50** ENA observations of boundary layer clouds and atmospheric condition relationships: application to CAM6 evaluation – Catherine Naud
- 2:50-3:05** Experiences with LASSO and CMDV-MCS for Bridging the Observation–Model Divide – William Gustafson
- 3:05-3:30** Discussion

Characterization of Shallow Oceanic Precipitation using Profiling and Scanning Radar Observations at the Eastern North Atlantic ARM Observatory

Katia Lamer

City University of New York, The City College

In collaboration with

Bernat Puigdomènech Treserras, Zeen Zhu,
Bradley Isom, Nitin Bharadwaj, and Pavlos Kollias



2019 ARM/ASR PI meeting
Rockville, Maryland
June 10, 2019



Observations/Retrieval of Light Precipitation Rate

KAZR2

Vertically pointing observations

Rain rates retrieved using the *O'Connor et al. (2005)* radar-lidar technique on calibrated backscatter observations

Now available as a PI product in the ARM archive

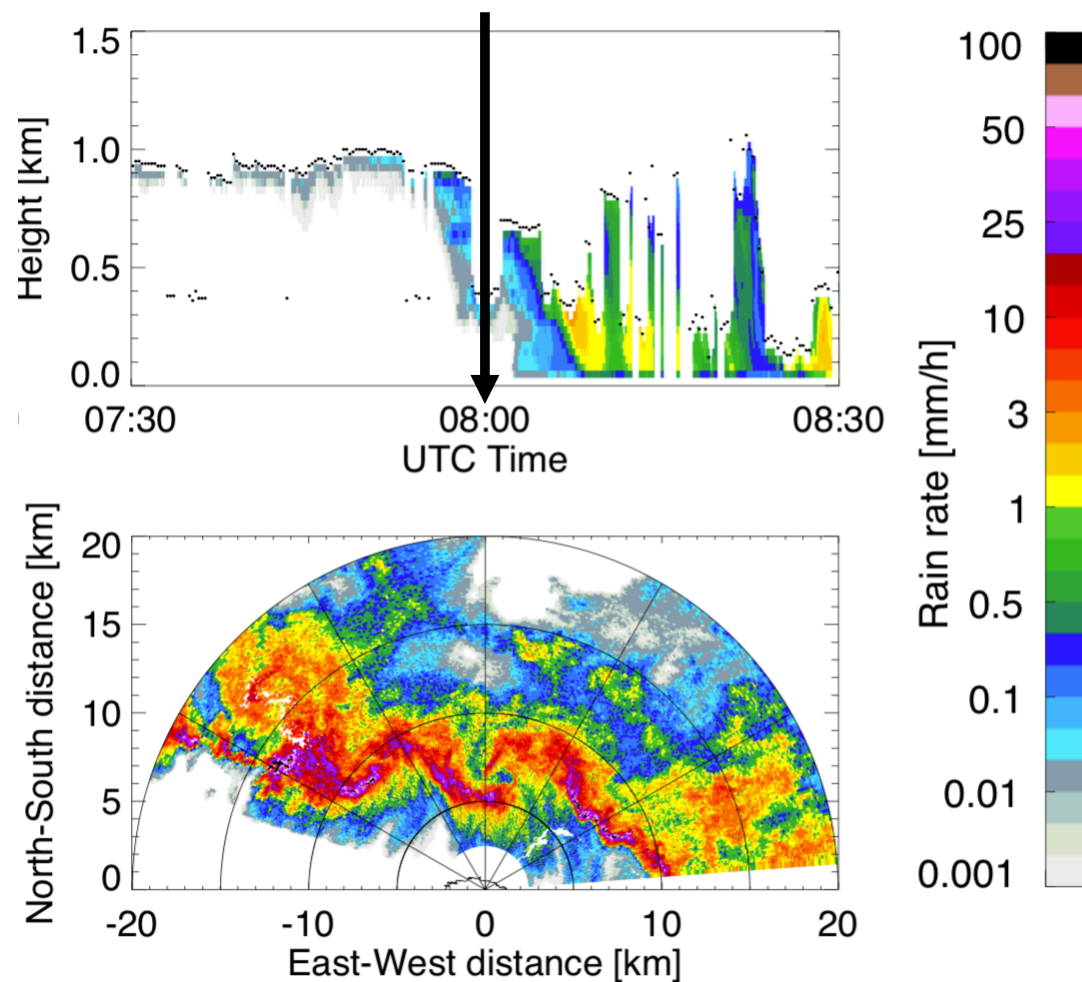
XSAPR2

1° elevation PPI scan covering ~ 2,500 km²

Rain rates retrieved using the *Lamer et al. (2019)* adaptive Z-R technique on calibrated radar reflectivity observations

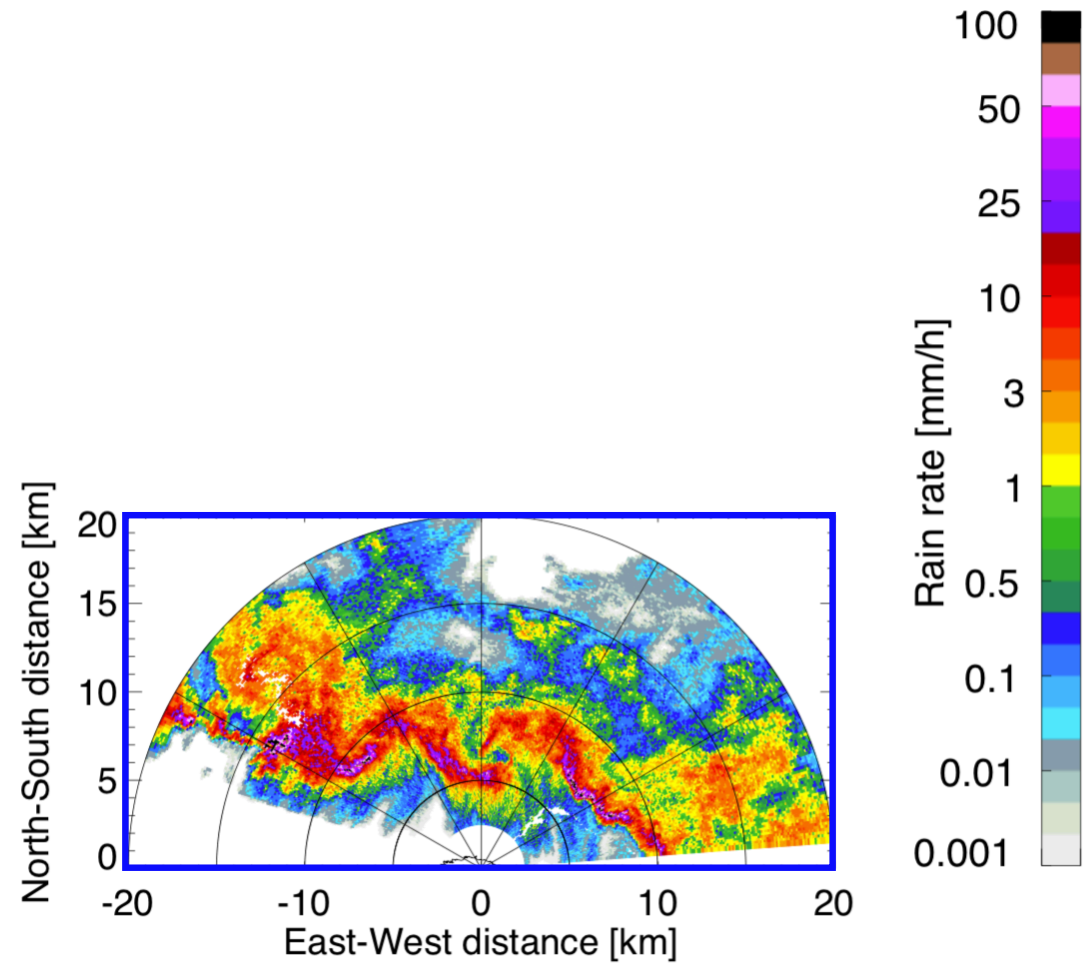
Data accessible upon request

Observations collected at the ENA site
Period 01/2018 to 04/2018 ~ 4 months



Estimation of Domain Average Light Precipitation Rate

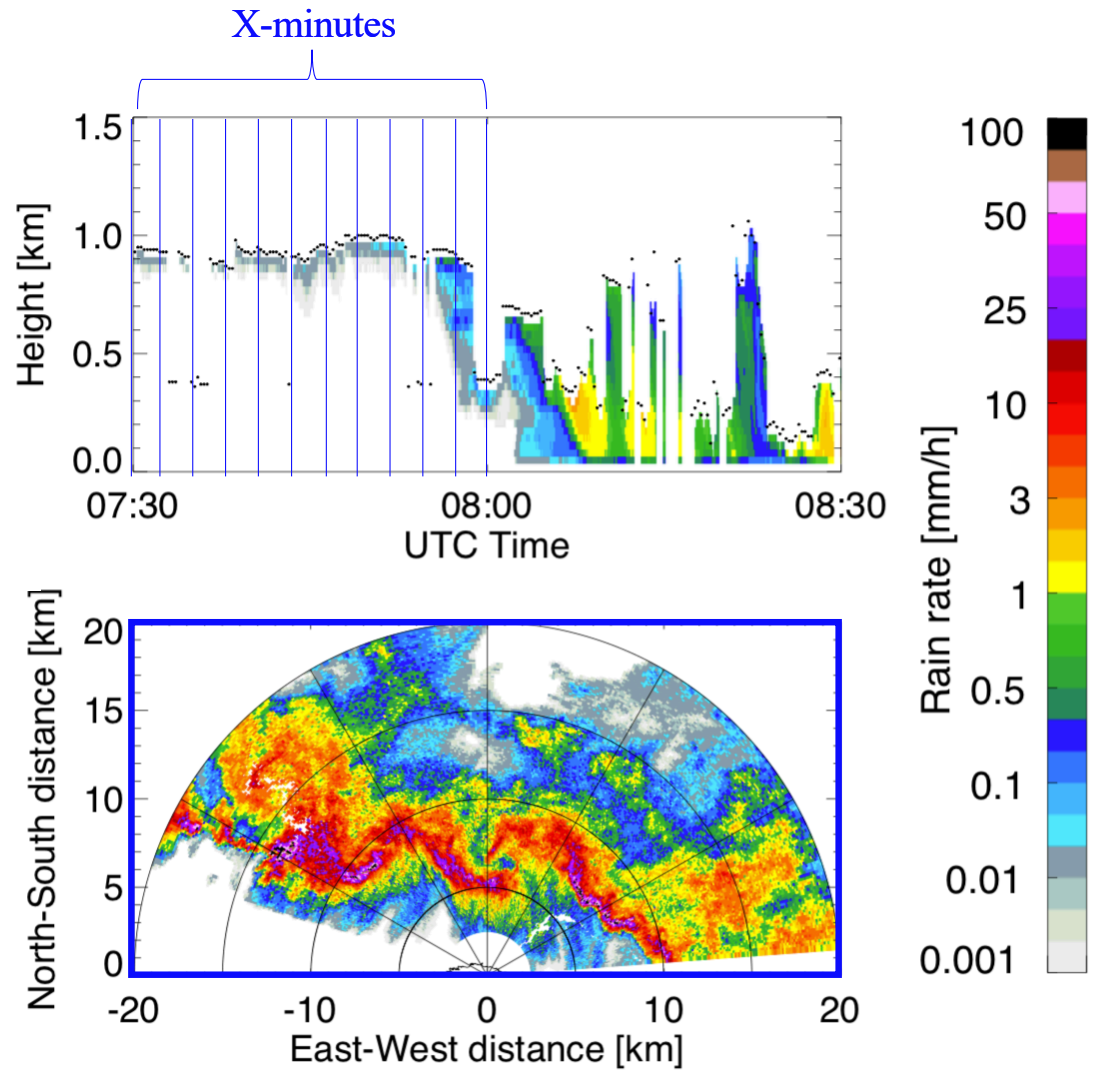
1. Estimate a domain average rain rate for each $\sim 2,500 \text{ km}^2$ domain snapshot. Domain “snapshots” are collected every 5-minutes.



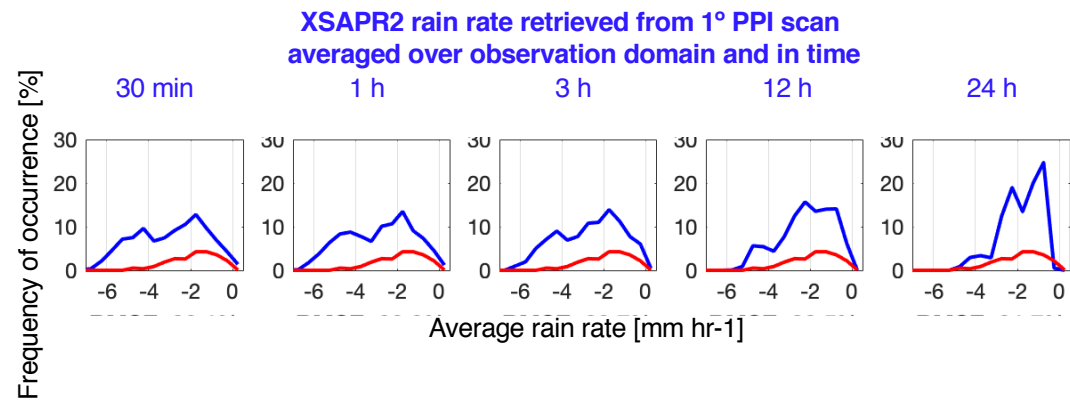
Characterizing variations in Domain Average Precipitation Rate Across Various Timescales

2.
To characterize domain-average precipitation rate changes across different timescales

Average X-minutes worth of domain average rain rates

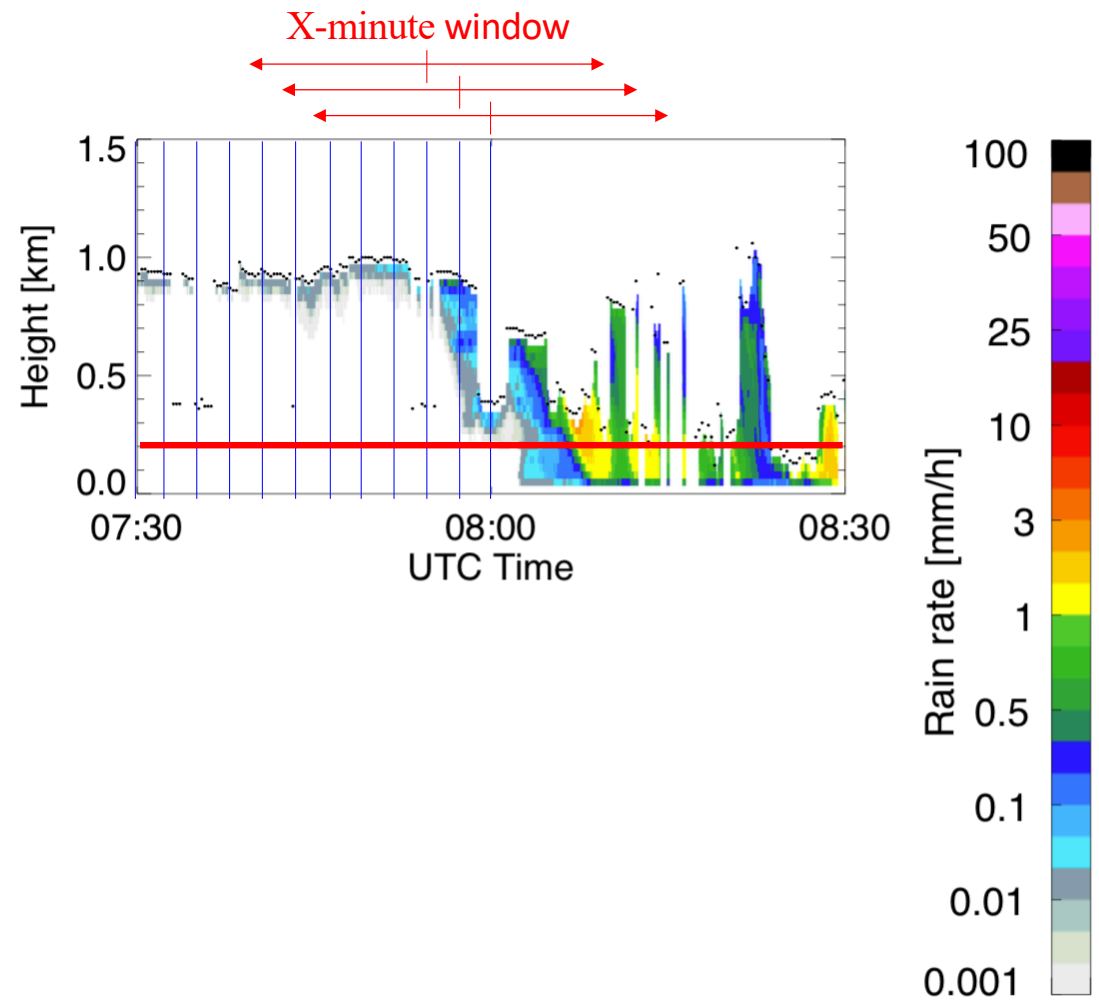


Characterizing variations in Domain Average Precipitation Rate Across Various Timescales

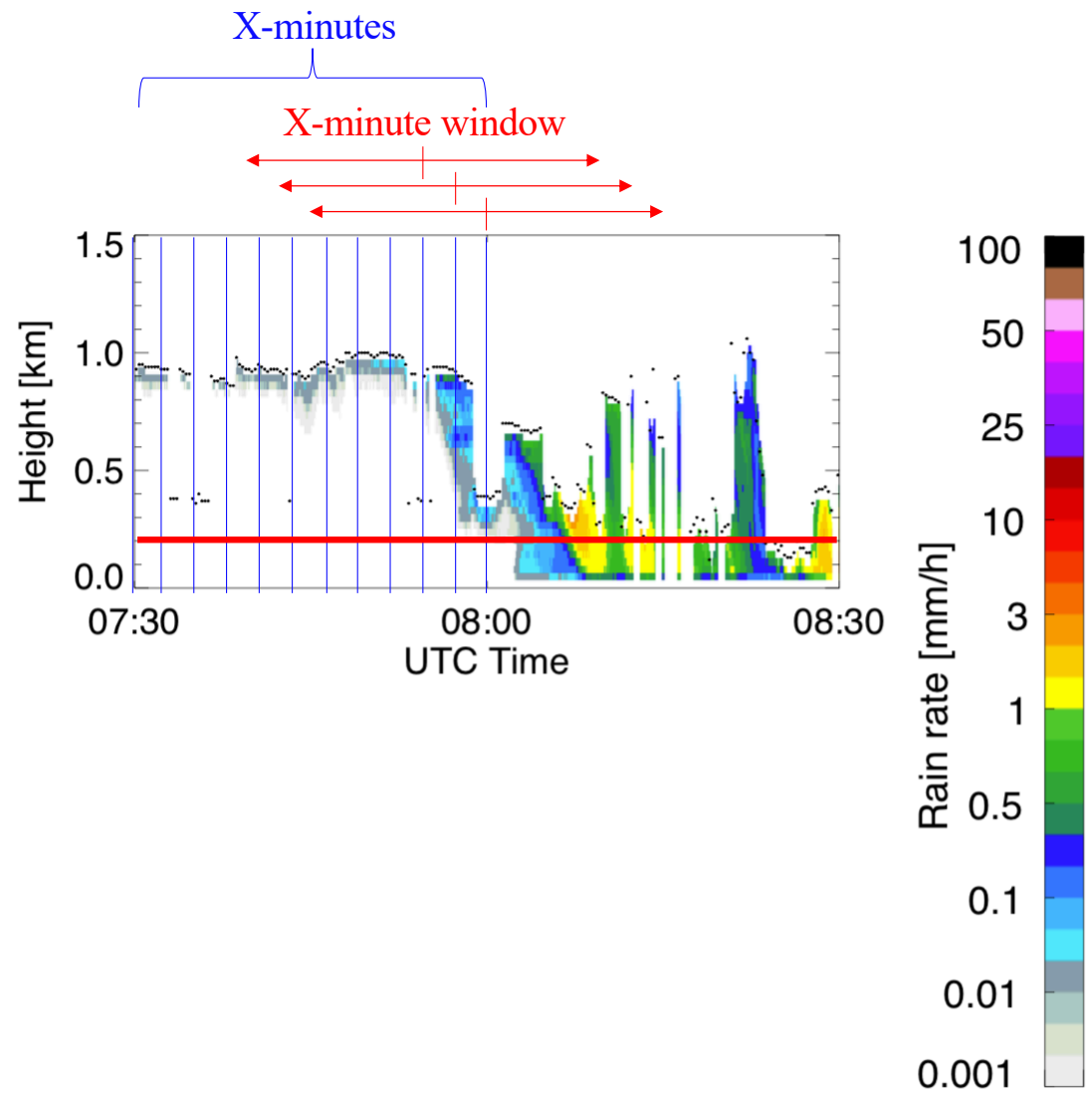


Estimating a Proxy for Domain Average Light Precipitation Rate Using Zenith observations

1. Centered on each X-SAPR2 domain snapshot estimate the average rain rate observed by KAZR2 at 200m in a X-minute time window

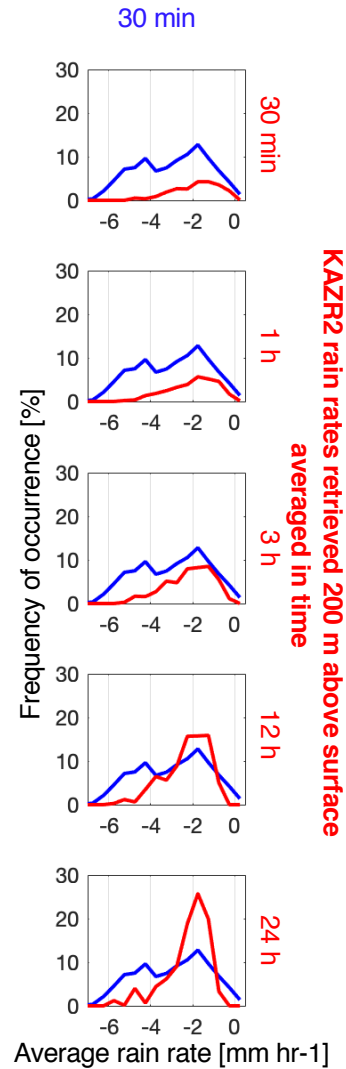


2. Average the same X-minutes worth of “domain” average rain rates as used when treating the XSAPR2 observations



Characterizing variations in Domain Average Precipitation Rate Across Various Timescales

XRAPR2 rain rate retrieved from 1° PPI scan averaged over observation domain and in time



Can point observations be used to characterize the temporal variability of near surface rain rate over a $\sim 2,500 \text{ km}^2$ domain?

Based on 4-months of ENA observations

- ✓ Low-level precipitation rate statistics only converge when characterizing long-term (12-hourly) variability
KAZR2 observations collected in 12-h windows around each XSAPR2 scan are needed to capture the $\sim 40 \text{ km}$ radius domain
- ✓ Scanning sensors are better suited to document sporadic and horizontal inhomogeneous precipitation
- ✓ Zenith-pointing radars more suited for precipitation that varies rapidly with height for instance owing to an active evaporation process

More details in Lamer et al. (2019) available in AMTD

