

Karhunen-Loève Expansion Analysis of Uncertainties in Cloud Microphysical Property Retrievals

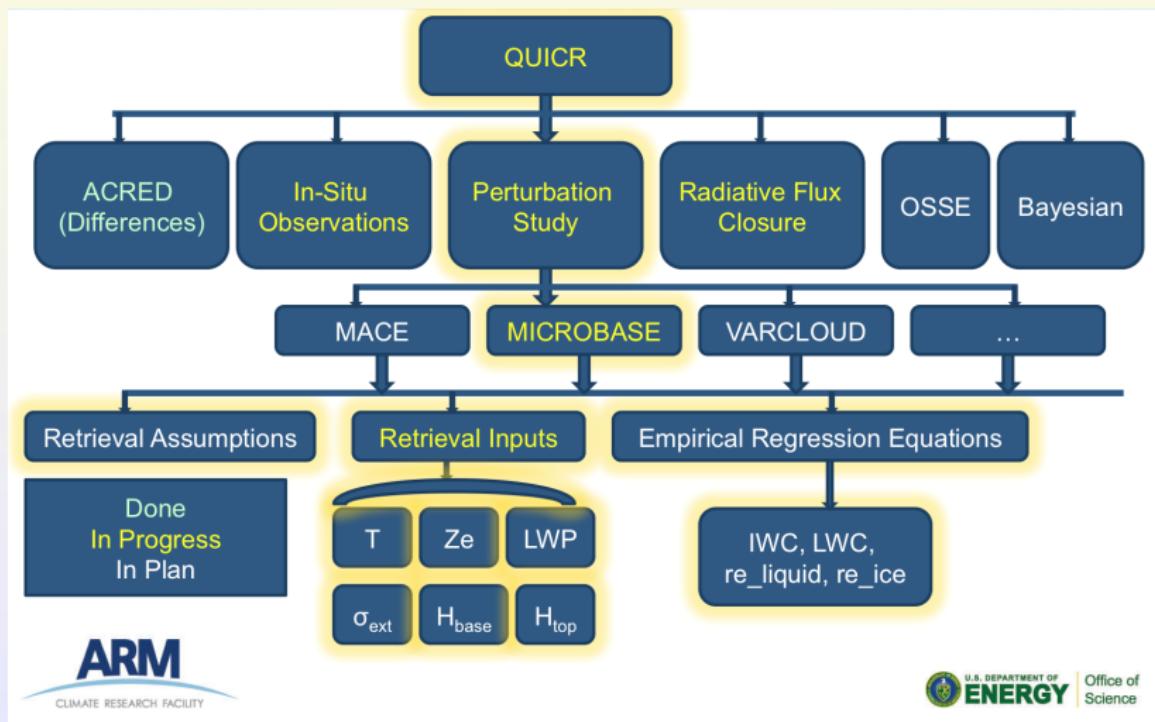
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Chitra Sivaraman, Timothy Shippert, and Charles Tong*

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Overview



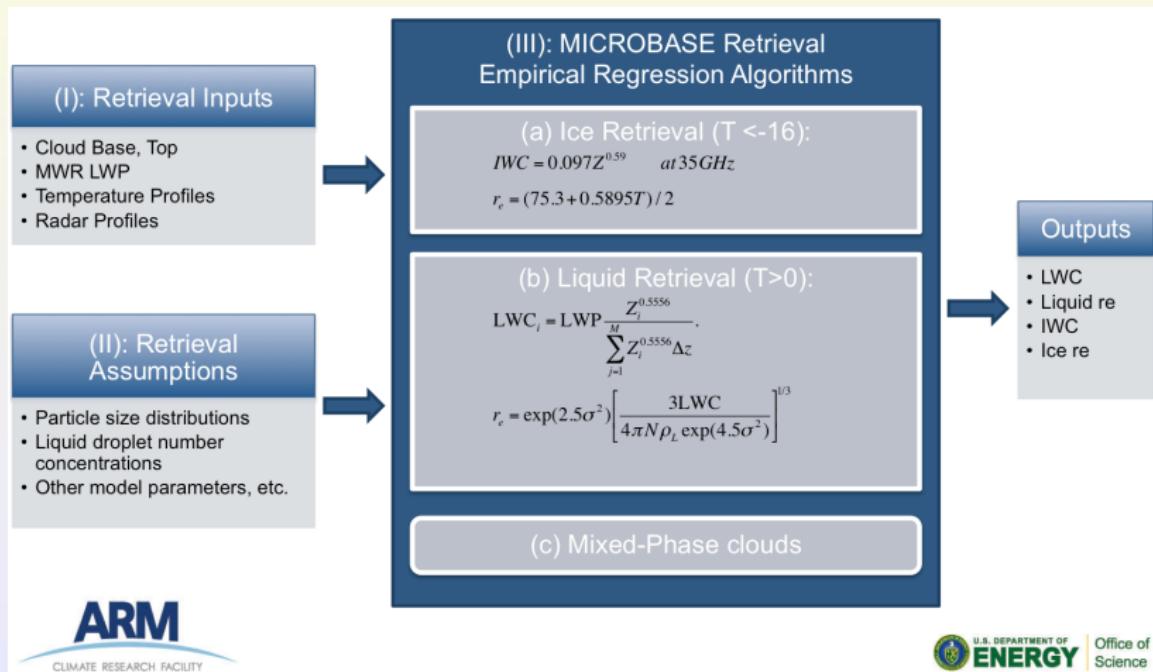
QUICR mission statement

*“to develop a **methodology** for ... quantifying uncertainties in
... ARM cloud retrievals ... in support of both **retrieval**
algorithm improvement and **cloud modeling** study”*

Questions

- What are the vertically-resolved cloud retrieval uncertainties on model temporal resolution (~ 30 min)?
- How large are the uncertainties from individual sources (e.g., instrument noise, parameters, sampling, etc)?

MICROBASE algorithm



Perturb MICROBASE (w/ PSUADE). HOW?

Problem Solving environment for Uncertainty Analysis and Design Exploration

- “Curse of dimensionality” in the stochastic input space
 - $315(T) + 512(Ze) + 1(LWP) = 828$ input layers prohibit large random sampling. Correlated?
- Parameterization of inputs by random variables
- Sampling input random variables (i.e., PDF)
 - Assumption of probability distributions (e.g., uniform) disagrees with observations.
- Current MICROBASE frame work **CANNOT** do parallel runs.
Large number runs are infeasible.
 - 50,000 times of perturbations per day require
50,000 min \approx **35 days & 11 TB** disk space

Karhunen-Loève (KL) + Central limit theorem (CLT)

- KLE reduces dimensions (828 to 14 modes) and extracts uncorrelated, independent input random variables.
- CLT shows that sample mean of random input variables approximately follows normal distribution ($n \rightarrow \infty$).

The central mean subtracted KLE w/ observation noise:

$$\overline{\mathcal{X}} = \sum_{i=1}^r U_i \frac{S_i}{\sqrt{m}} \sqrt{1 + \left(\frac{\sigma_0}{\frac{S_i}{\sqrt{m}}} \right)^2} \frac{z_i}{\sqrt{m}}$$

$z_i \sim N(0, 1)$, $\text{noise}_i \sim N(0, \sigma_0^2)$, where z_i and noise_i are i.i.d.

Uniform distribution is used for parameter uncertainties.

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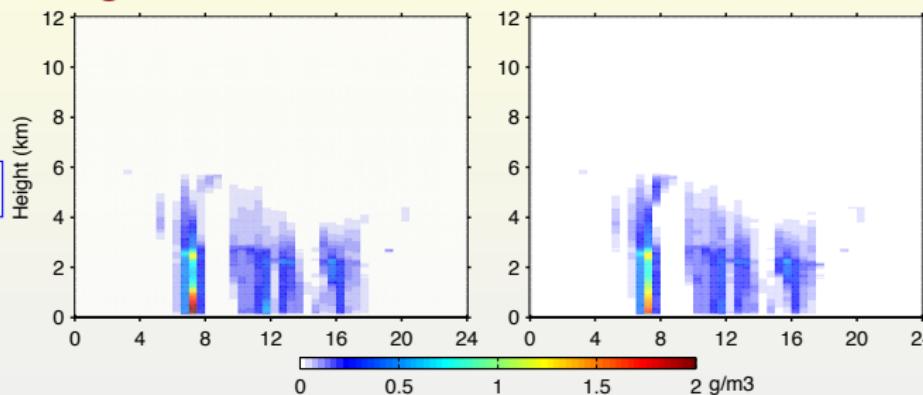
The perturbation range follows Zhao et al. 2013, submitted to JGR.

Input	T (K)	Ze (dbZ)	LWP (%)
Range	0.5	0.5	15

Pars	a (g/m^3)	d ($\mu\text{m}^\circ\text{C}$)	g	σ	N (cm^{-3})
Range	0.03–0.22	0.2311–0.8211	0.5–0.6	0.2–0.6	10–350

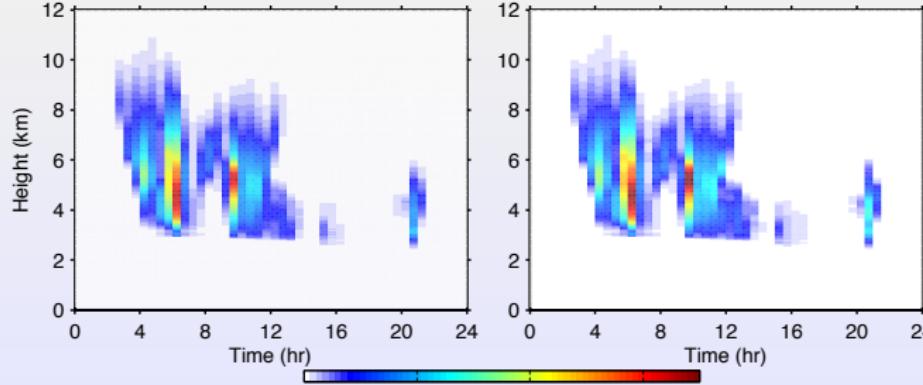
0.5-hr averages of LWC & IWC at SGP on 2006/05/10

Raw LWC



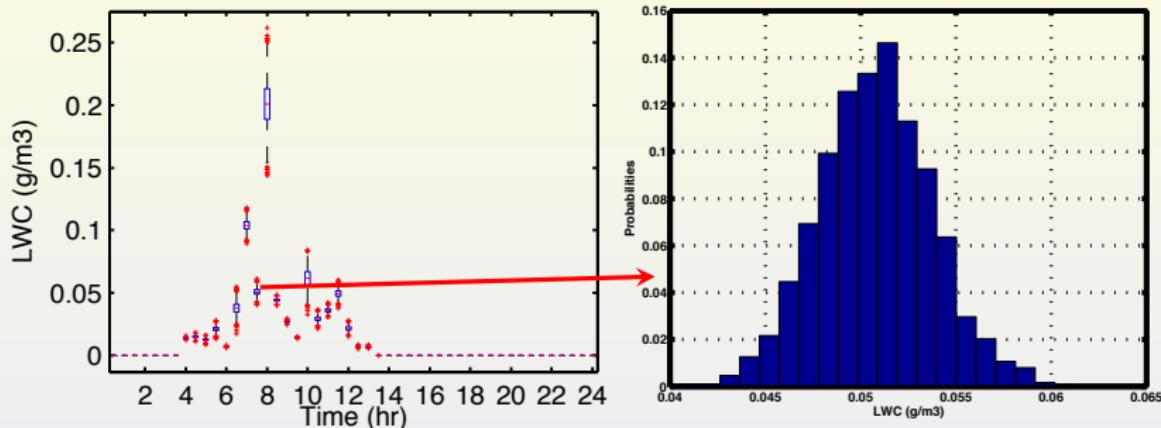
UQ LWC

Raw IWC



UQ IWC

PDF of LWC at 5 km on 2006/05/10

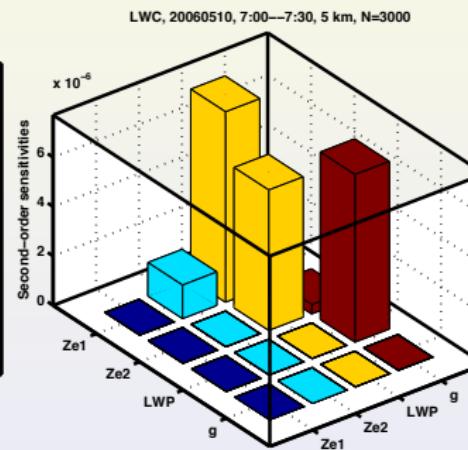
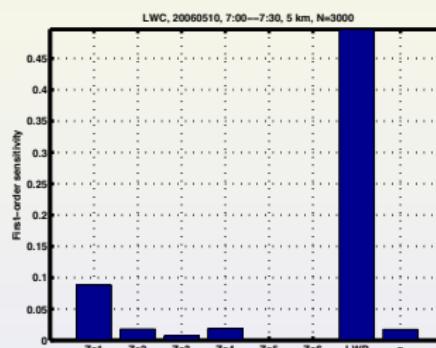


- Perturbing MICROBASE inputs by adding instrument noise and parameter uncertainties yields height-time probability distribution functions (PDFs) for retrieval products.
- These observation-based PDFs are better a priori for other UQ studies (e.g., MCMC).

Sensitivities of LWC to input random variables

$$LWC = LWP \frac{Z_e^g}{\sum_{j=1}^M Z_e^g \Delta z}$$

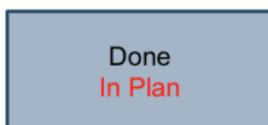
PSUADE



- Allows attributing uncertainties to individual input variables.
- Provides directions to improve instruments & obs strategies.

Summary of MICROBASE UQ

Source	Type	Approach implemented by PSUADE	Mathematical Theorem
Retrieval Inputs	Inputs Data Variability and Noise	Perturb principal modes	Principal Component Analysis or Karhunen-Loeve expansion
		Sampling distribution of sample mean	Central Limit Theorem
Parameters in Retrieval Assumptions and Algorithms	Parametric Uncertainties	Alertory variability	Uniform/Triangular Distribution Assumption
		Epistemic uncertainty quantification	Evidence Theory
Equations of Retrieval Algorithms	Model-Form Uncertainties	Model error/discrepancy analysis	Validation, Verification and Uncertainty Quantification



Thank you!