



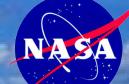
Cloud Microphysics Across Scales for Seamless Prediction

A. Gettelman (NCAR), R. Forbes, M. Fielding (ECMWF) R. Marchand (UW)

+ a huge cast







Outline

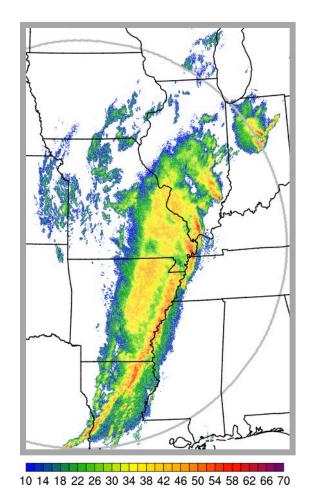
- Motivation/Statement of the 'Cloud Problem'
 - Weather Extremes, Cloud Feedbacks, Aerosol Forcing
- What is cloud microphysics
- Some major issues for Weather and Climate
- Experiments and Progress towards seamless prediction
- Next Steps

Weather: Cloud Microphysics Kills!

- Clouds are responsible for most severe weather
 - Tornadoes, Thunderstorms, Hail, Tropical Cyclones
- Critical processes depend on cloud microphysics (Thunderstorms, Hail, even Tornadoes)

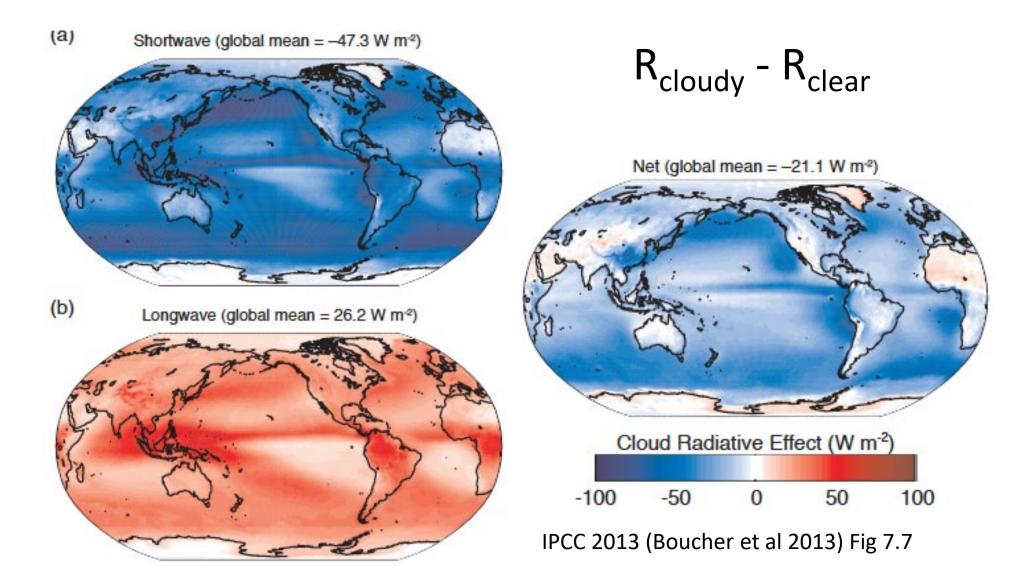


Radar Reflectivity, 27 April 2017



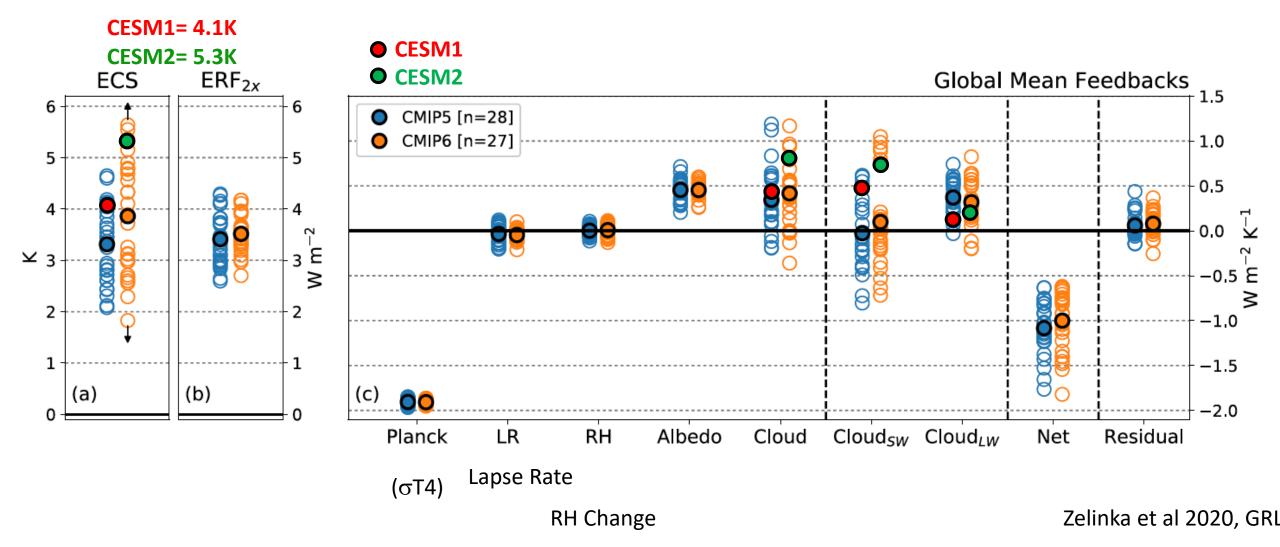
Clouds and Climate

Climate: Cloud Radiative Effects are Large



Climate Feedbacks

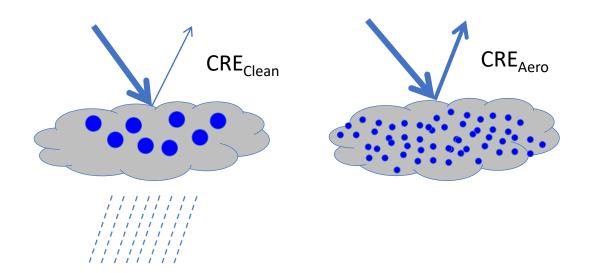
Equilibrium Climate Sensitivity (ECS) Uncertainty: *It's all about cloud feedback*



Aerosol Effects on Clouds

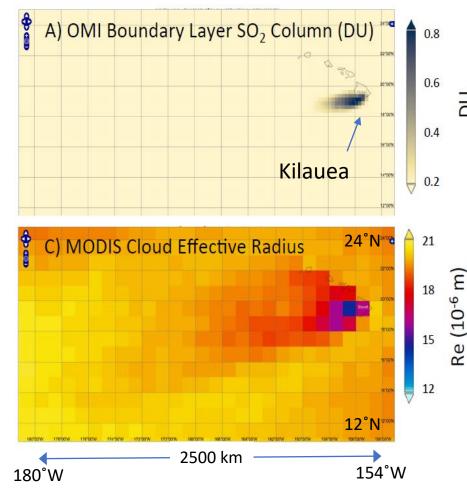
- Scattering & Absorption = Direct effects
- Aerosol Cloud Interactions (ACI)

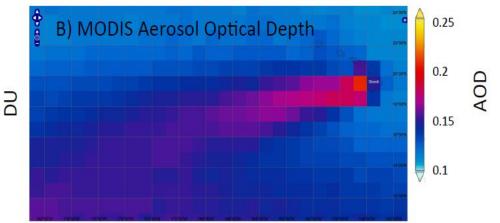
+Aerosols \rightarrow +CCN \rightarrow +N_c $\rightarrow \Delta$ CRE



Brighter clouds (albedo effect) with smaller drops (S. Twomey 1977) Also: delay in precipitation (B. Albrecht, 1989). Longer lived Clouds?

'Volcano Tracks': Satellite Climatology

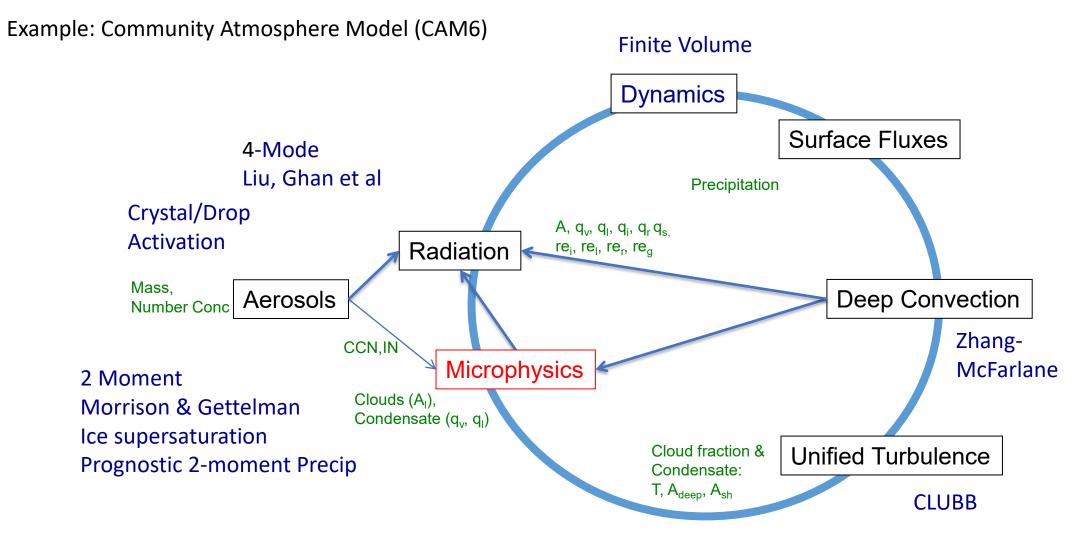




Satellite data for C. Pacific, 10 year climatology around the Hawaiian Islands

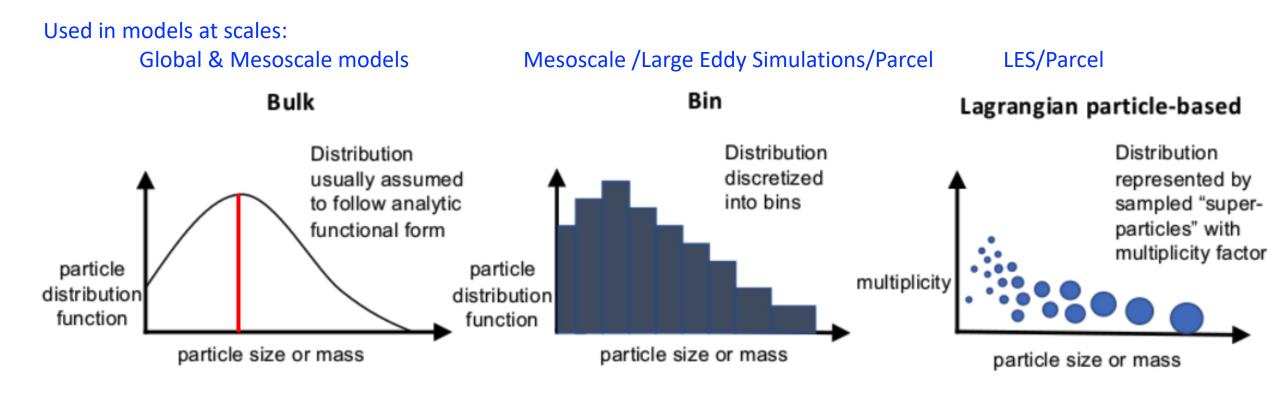
SO₂ is from Kilauea Volcano on Hawaii (Small-Griswold & Gettelman, in Prep)

What is Cloud Microphysics?



A = cloud fraction, $q=H_2O$, re=effective radius (size), T=temperature (i)ce, (l)iquid, (v)apor, (r)ain, (s)now

Types of Microphysical Schemes

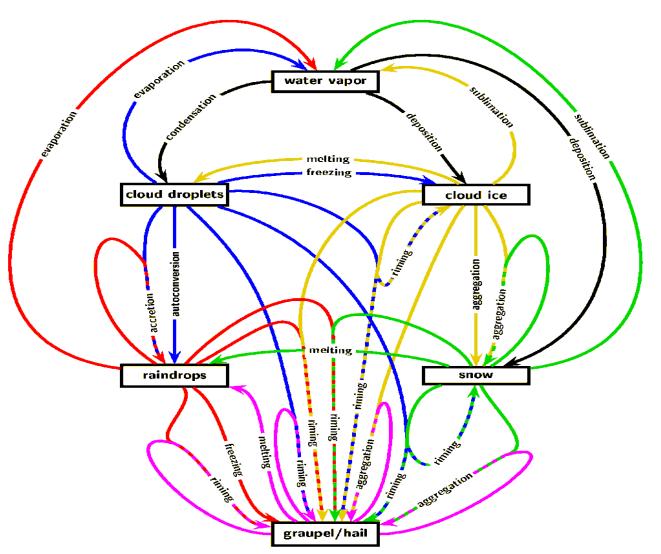


Two Moment = Prognostic Mass and Number One Moment = Prognostic Mass, Diagnostic Number/Size

Figure: Morrison et al 2020, JAMES

Cloud Microphysical Processes

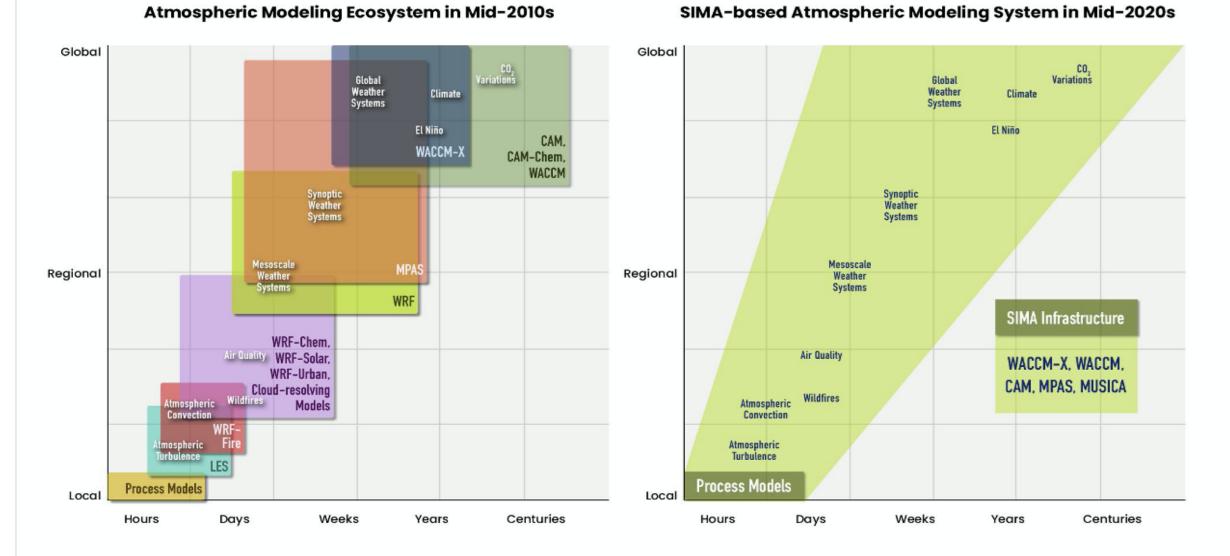
A. Seifert, Personal Communication



Key processes:

- **Mixed phase** processes important for weather and climate (supercooled liquid meta-stable)
- Conversion processes (vapor deposition onto ice, riming snow to graupel)
- **Aggregation** processes grow particles to where they fall fast (rain formation)
- Ice particle density and fall speeds vary strongly and are important for precipitation

A (not so unique) vision System for Integrated Modeling of the Atmosphere (SIMA)



Major Issues for Clouds, Precipitation & Aerosols

- Aerosol activation (cloud-aerosol interactions)
 - Vertical velocity critical
 - Size distributions govern process rates
 - Ice Nucleation can be highly uncertain
- Cloud Phase
 - Radiative effects at high latitudes AND for cloud feedbacks
 - Also weather impacts: extreme precipitation, riming
- Size Distributions
 - Consistent treatment of process rates and radiation
- Parameter Uncertainty
 - Can we narrow the range of uncertainty in cloud physics / microphysics
- Precipitation Formation: Frequency & Intensity
 - The chronic drizzle problem is a cloud physics problem
- Seamless prediction
 - Can we push across scales: examples from 100km \rightarrow 3km

Cloud Phase

SOCRATES in-situ flights over the S. Ocean used to understand & improve models

CAM6: Too little ice. This contributes to high climate sensitivity.

Similar work on importance of cloud phase by Forbes (2014, 2015)

4000

3500

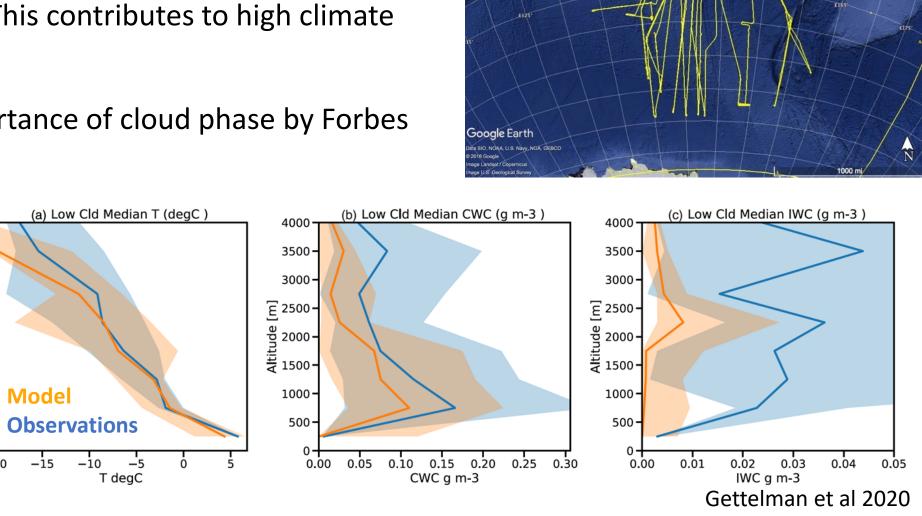
3000

- 2500 - 4 Hitnde 2000 - 500 - 15000 - 15000 - 1500 - 1500 - 1500 - 1500

1000-

500-

-20



SOCRATES All Flights

Legend Track, last hour

So Track, older

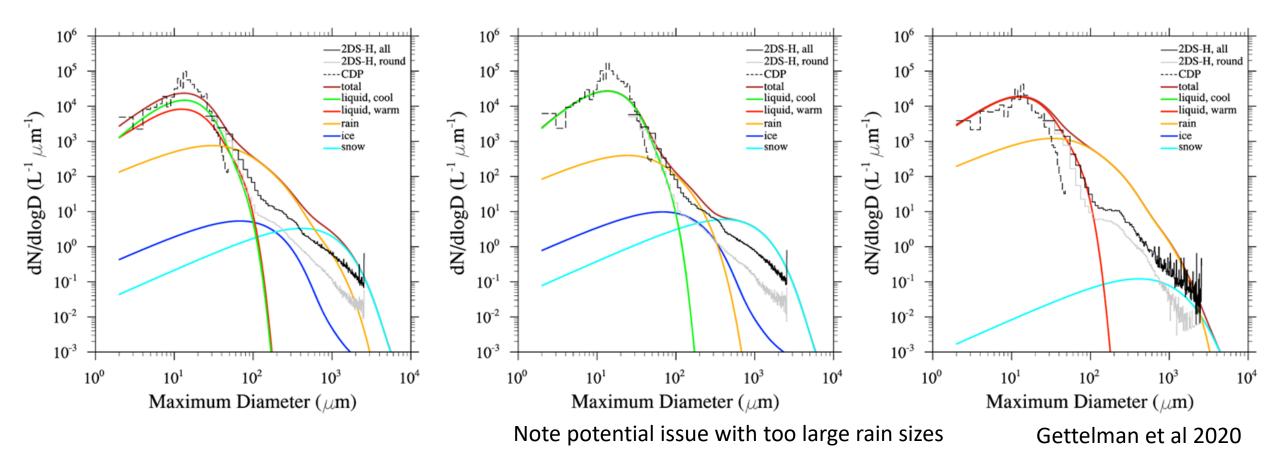
Microphysics, Size distributions Advanced GCMs/GSRMs can be compared directly to cloud microphysical size distributions (here from SOCRATES)

Comparison is GCM cloud microphysics along aircraft flight tracks with in-situ data

(a) All Clouds

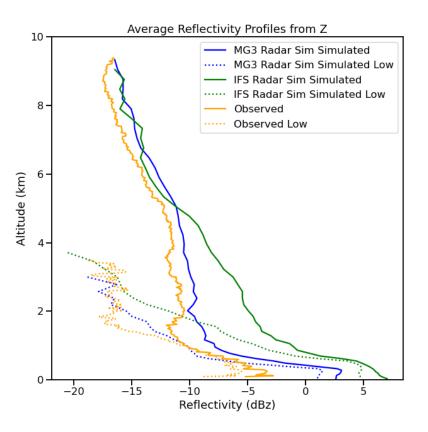
(b) Cold (T<0°C) Clouds

(c) Warm (T>0°C) Clouds

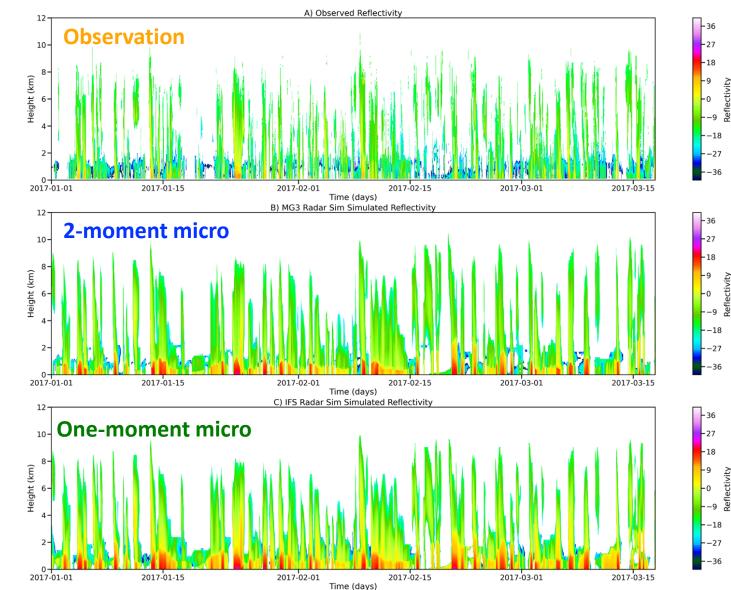


Microphysics: Comparing to Reflectivity

Comparisons over Macquarie Island in S. Ocean between a precipitation radar and single column simulations with **one-moment** and **2-moment** microphysics in the ECMWF-IFS SCM.

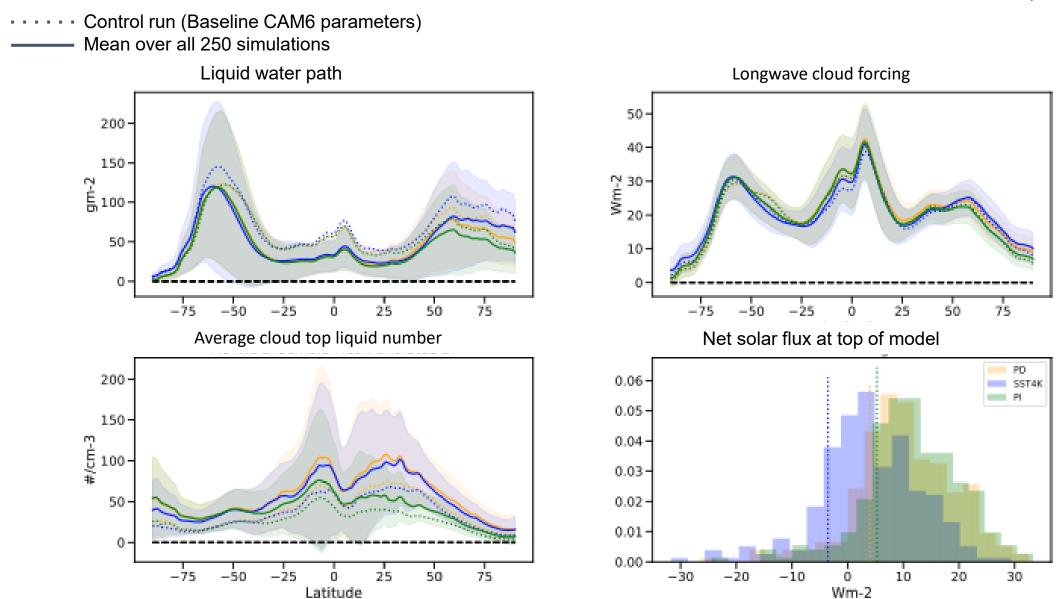


Gettelman, Forbes, Fielding, Marchand, in Prep



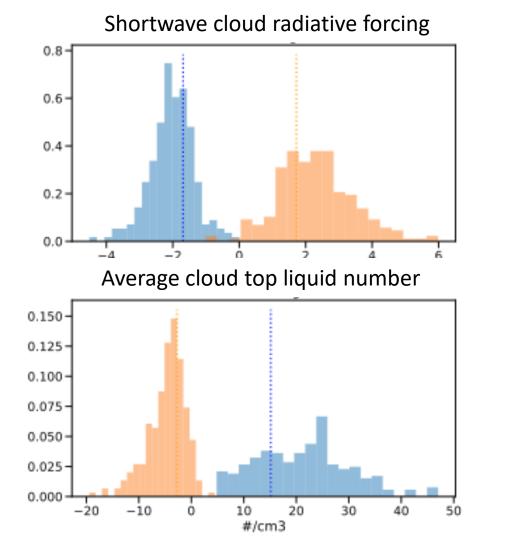
Robustness: Parameter Uncertainty Perturbed Parameter Ensemble (PPE)

Eidhammer, Gettelman, Thayer-Calder, Duffy

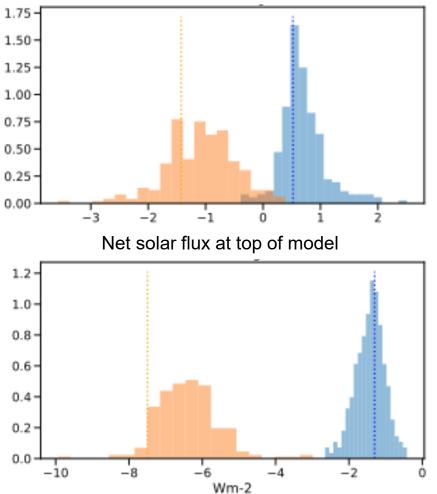


Forcing and Feedback

Forcing:Present day (PD) - Pre-industrial (PI) aerosolFeedback:4K SST - Present day (PD)

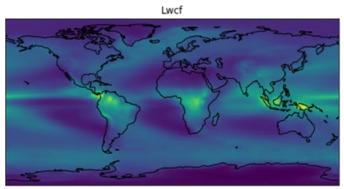


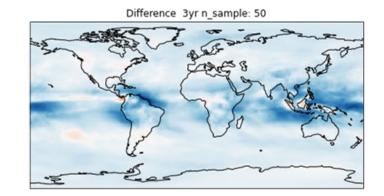
Longwave cloud radiative forcing

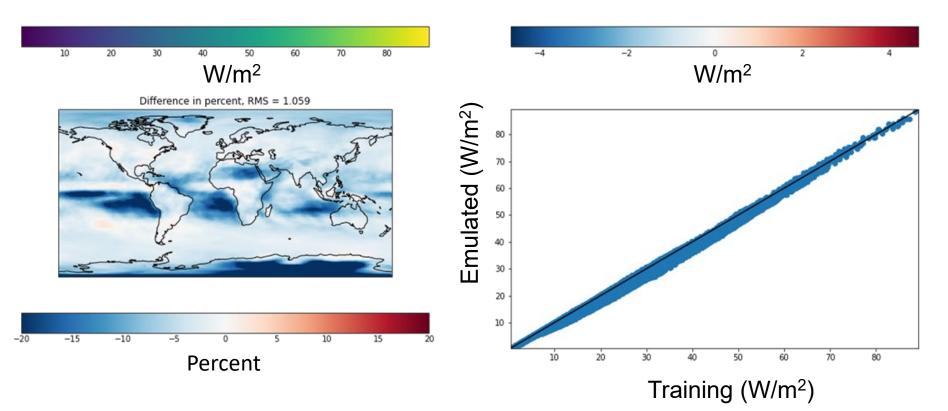


Emulating Longwave Cloud Radiative Forcing Gaussian Process Emulator (University of Oxford)

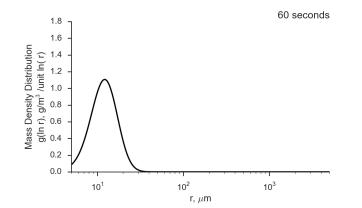
Emulate LW Cloud Radiative Effect (LWCRE)







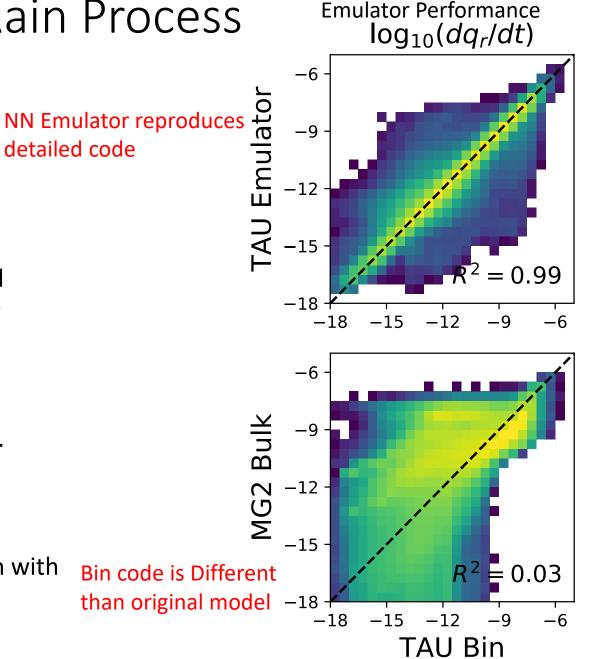
Machine Learning the Warm Rain Process

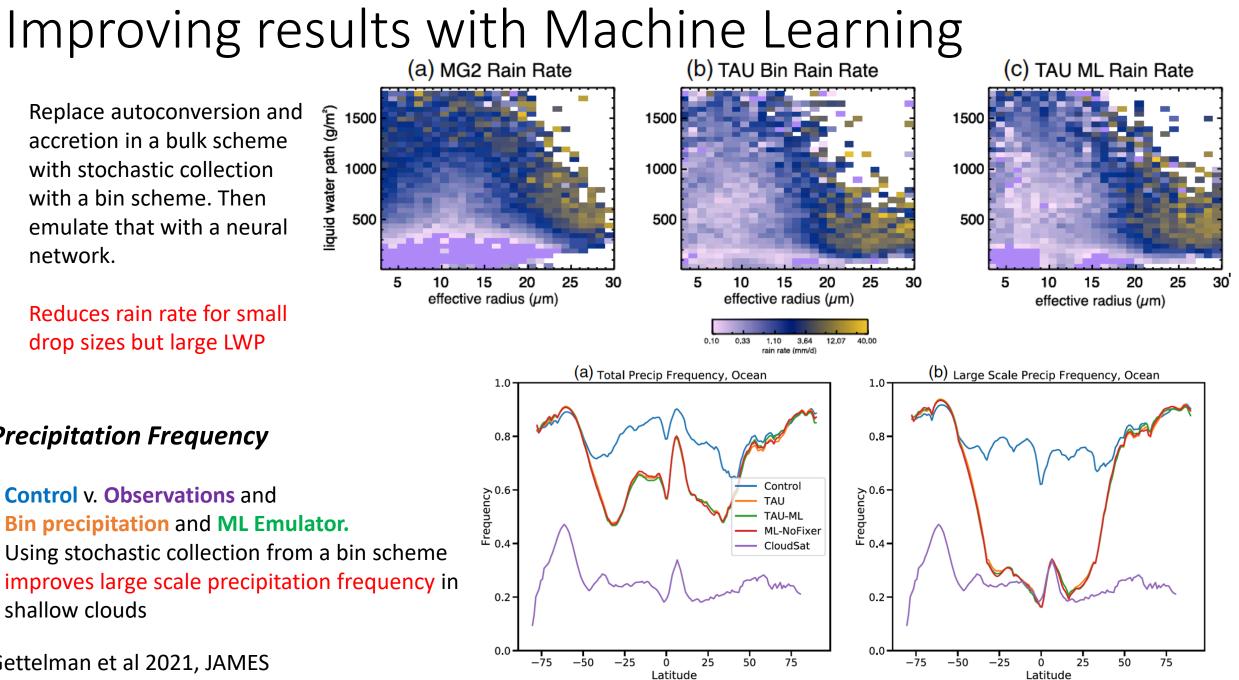


Replace traditional GCM bulk rain formation with a bin model formulation for stochastic collection. This is too expensive for climate use. So emulate it with a neural network.

Results:

- We can change the answer in the model with the bin code.
- Very slow when using full treatment
- Recover speed and recover results with a neural network emulator (it works)
- Embedded NN in the microphysics: maintains conservation with series of checks





Precipitation Frequency

Bin precipitation and **ML Emulator.** Using stochastic collection from a bin scheme shallow clouds

Gettelman et al 2021, JAMES

Climate Extremes: Variable-Resolution ($60 \rightarrow 3$ km)

- Global Model: CESM-MPAS: 3km regional, nonhydrostatic dynamics.
- Regional climate model: WRF (CONUS) 4km (Rasmussen et al., 2021)

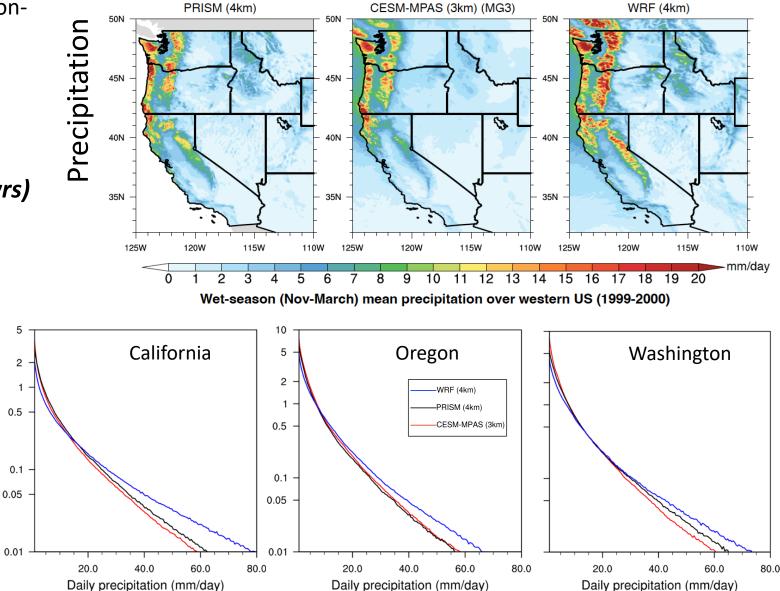
W. USA Wet-season (Nov-Mar) precip (5yrs)

- CESM-MPAS results compare well to obs
- Smaller biases than WRF mesoscale model

Daily precipitation Intensity PDF

4km Mesoscale Model (WRF) 3km Global Model (CESM) 4km Observations

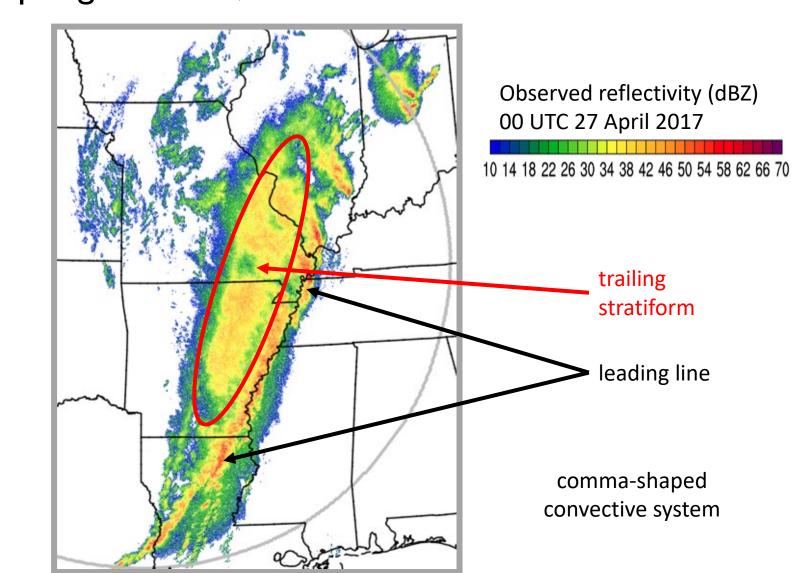
CESM captures observed PDF better than WRF, especially for extreme precipitation X. Huang, NCAR



Atmosphere Testing MPAS - CAM6 physics at convection-permitting scales Central US spring test case, 24 hour forecast

0 UTC 26 April 2017 initialization 60-3 km variable-resolution mesh

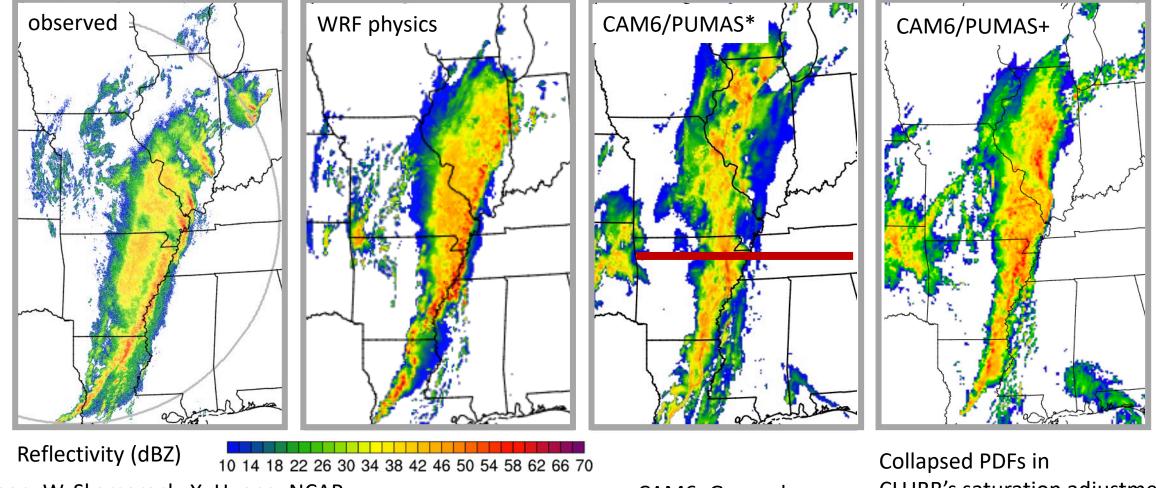




Central US: initial results

- Central US Summertime squall line. 24 hour forecast valid 0 UTC 27 April 2017
- Mesoscale model v. Climate Model

(* Computed using single-moment reflectivity diagnostic)

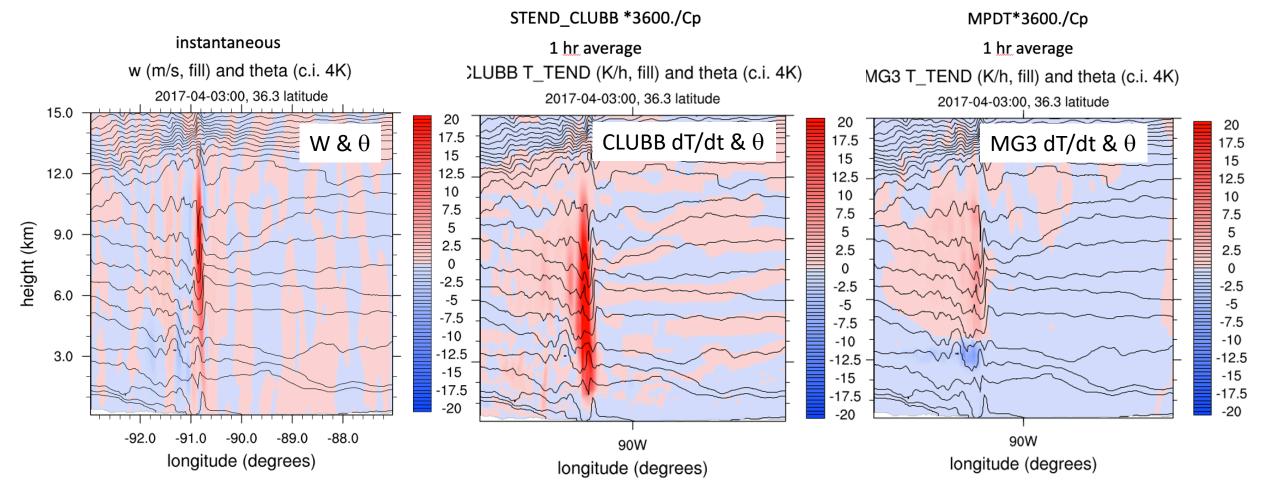


M. Chen, W. Skamarock, X. Huang, NCAR

CAM6+Graupel

CLUBB's saturation adjustment

Instantaneous outputs across squall line

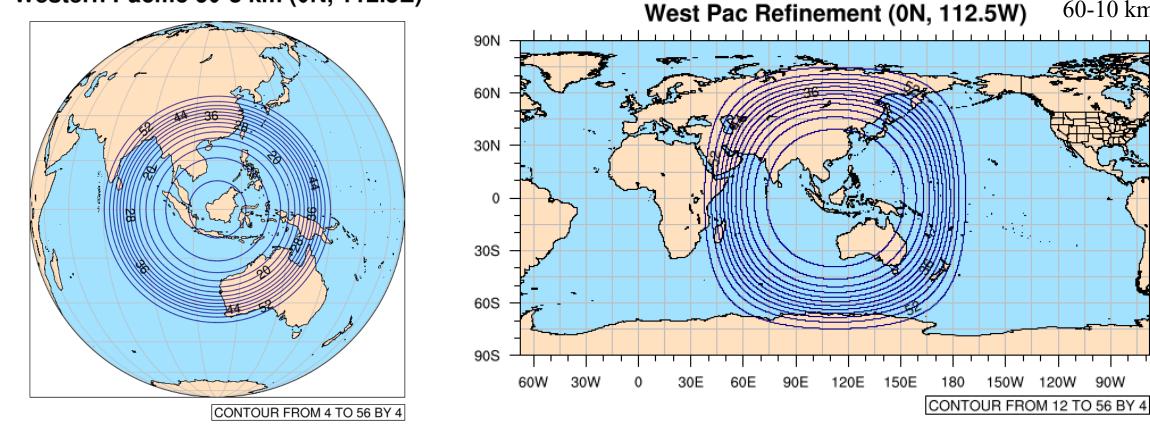


Issue: CLUBB seems to be doing a lot here.

After some discussion: remove 'partial cloudiness' (collapse PDF) in CLUBB saturation adjustment calculation

Western Pacific Cases

Western Pacific 60-3 km (0N, 112.5E)



60-10 km

90W

120W

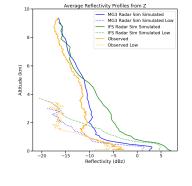
Western Pacific Movies...

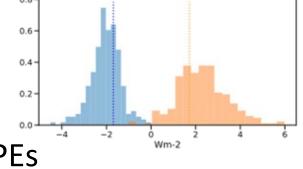
- Looking at scales of variability
- Eastward propagating cells
- Diurnal cycles over Indonesia
- Tropical Cyclones
- Larger scale organization

Next Steps: Paths forward

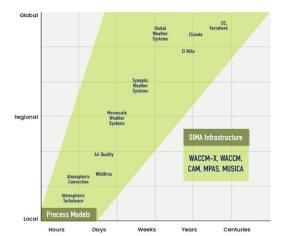
- For cloud microphysics: we can do 'seamless' (100km \rightarrow 3km)
 - Requires some complexity (rimed ice)
 - Working on modifications to physics
 - Exploring regime by regime in refined (regional) simulations
- Rain formation is critical, we need to do it better
 - Frequency, Intensity of precipitation
 - Aerosol forcing/response
 - May want to 'emulate' detailed models: microphysics is a good candidate
- 'Simplify' schemes
 - Unified ice/snow desirable
 - Think about also learning model structure (size distributions, process rates)
 - Emulators to help constrain process rates (objective adjustments)
- Cloud Phase still important
 - Requires improved knowledge of aerosols
- Couple the physics and dynamics
 - May be necessary at large scales. Where is microphysics gray zone?
 - Gray zone for activation, evaporation, coupling with turbulence

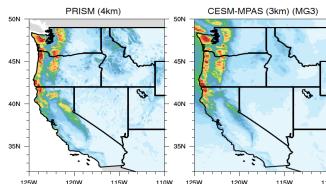
- Future Work: Where does it all go?
 - Better uncertainty quantification (parameters, processes): PPEs
 - Model-data fusion, using observations
 - Crossing scales (LES--> Global)
 - New modeling capabilities with refined ESMs and with global km scale models
 - Machine learning/emulators:
 New tools, but still need understanding



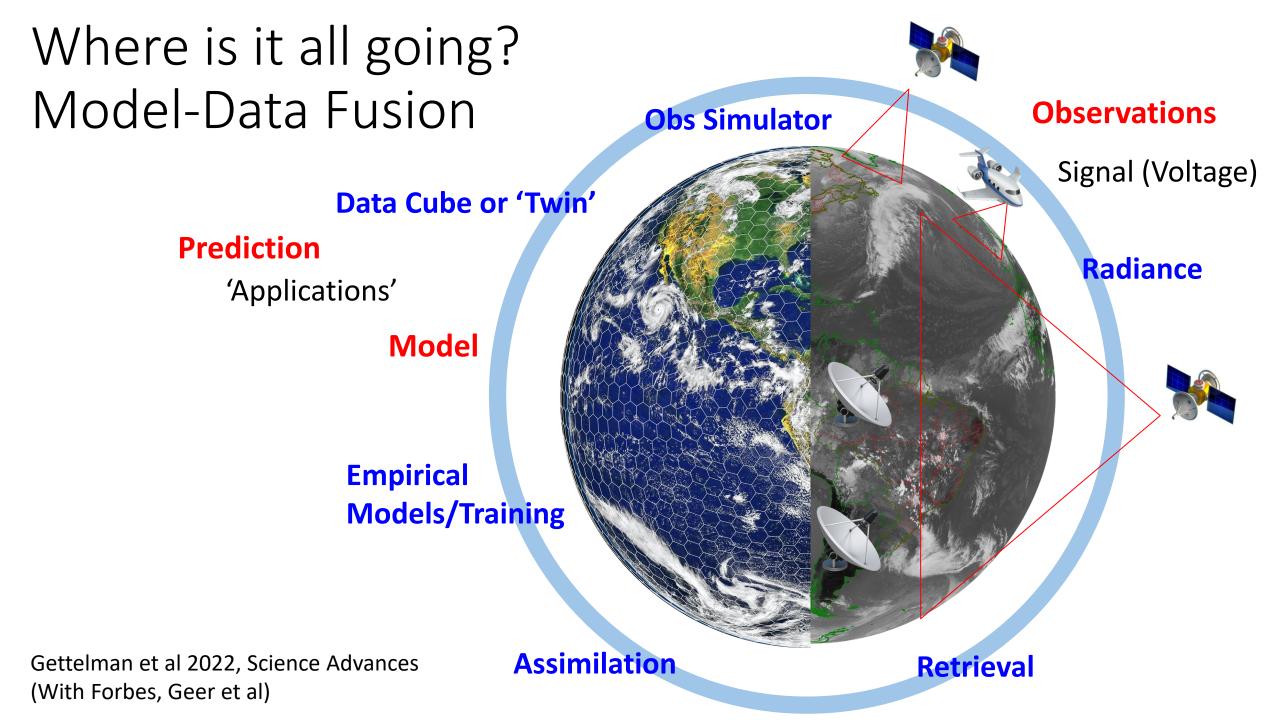


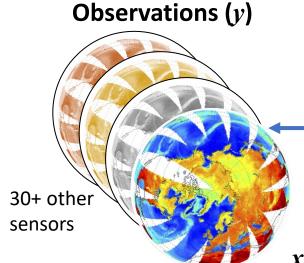
∆SWCF Histogram

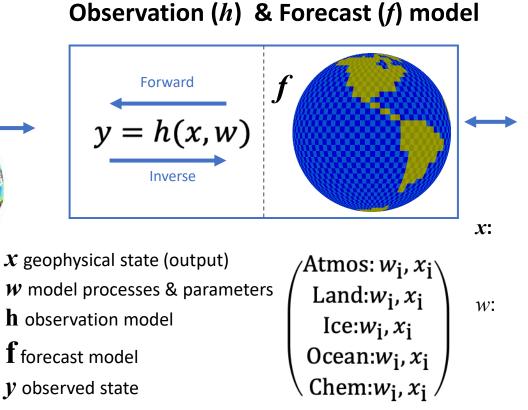




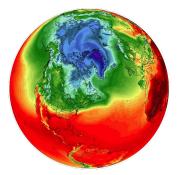








Predictions (x, w)



Atmospheric NWP analysis (e.g. 2m temp) Ocean and Sea Ice analysis Land and Hydrology Analysis Process Modelling Deep Learning Parameter Estimation