

Physics Parameterizations in Global Atmospheric Models

Intro to CAM-CLUBB Tutorial 2019

NCAR

Monday June 10th

Outline

- What are “AGCM Physics”?
- Resolution, subgrid variability, nonlinearity
- Basic design of parameterizations
- Future directions

Equations of Motion – explicitly resolved dynamics

Where do the “physics” appear?

$$d\bar{\mathbf{V}}/dt + f\mathbf{k} \times \bar{\mathbf{V}} + \nabla\bar{\phi} = \mathbf{F},$$

(horizontal momentum)

$$d\bar{T}/dt - \kappa\bar{T}\omega/p = Q/c_p,$$

(thermodynamic energy)

$$\nabla \cdot \bar{\mathbf{V}} + \partial\bar{\omega}/\partial p = 0,$$

(mass continuity)

$$\partial\bar{\phi}/\partial p + R\bar{T}/p = 0,$$

(hydrostatic equilibrium)

$$d\bar{q}/dt = S_q.$$

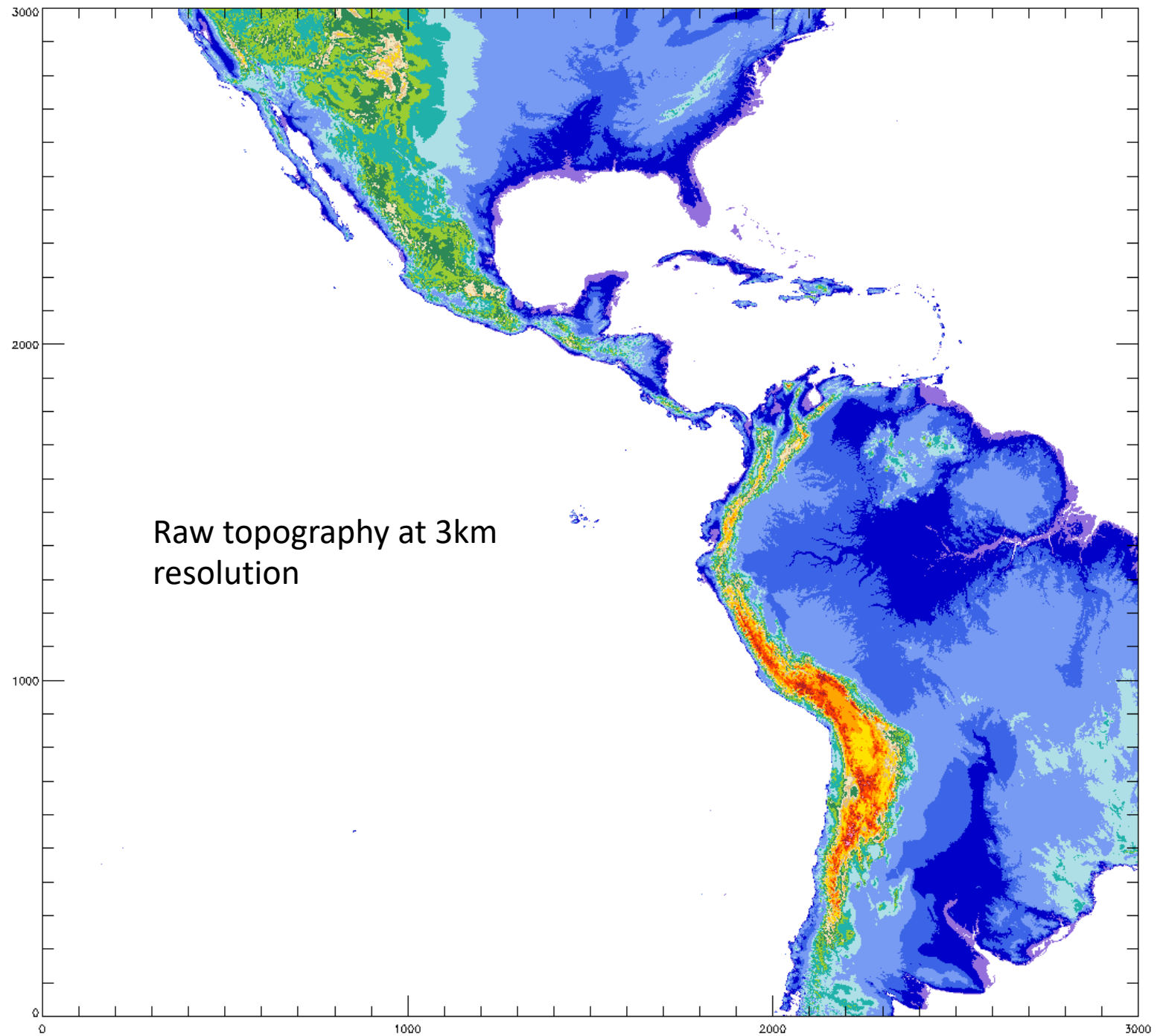
(water vapor mass continuity)

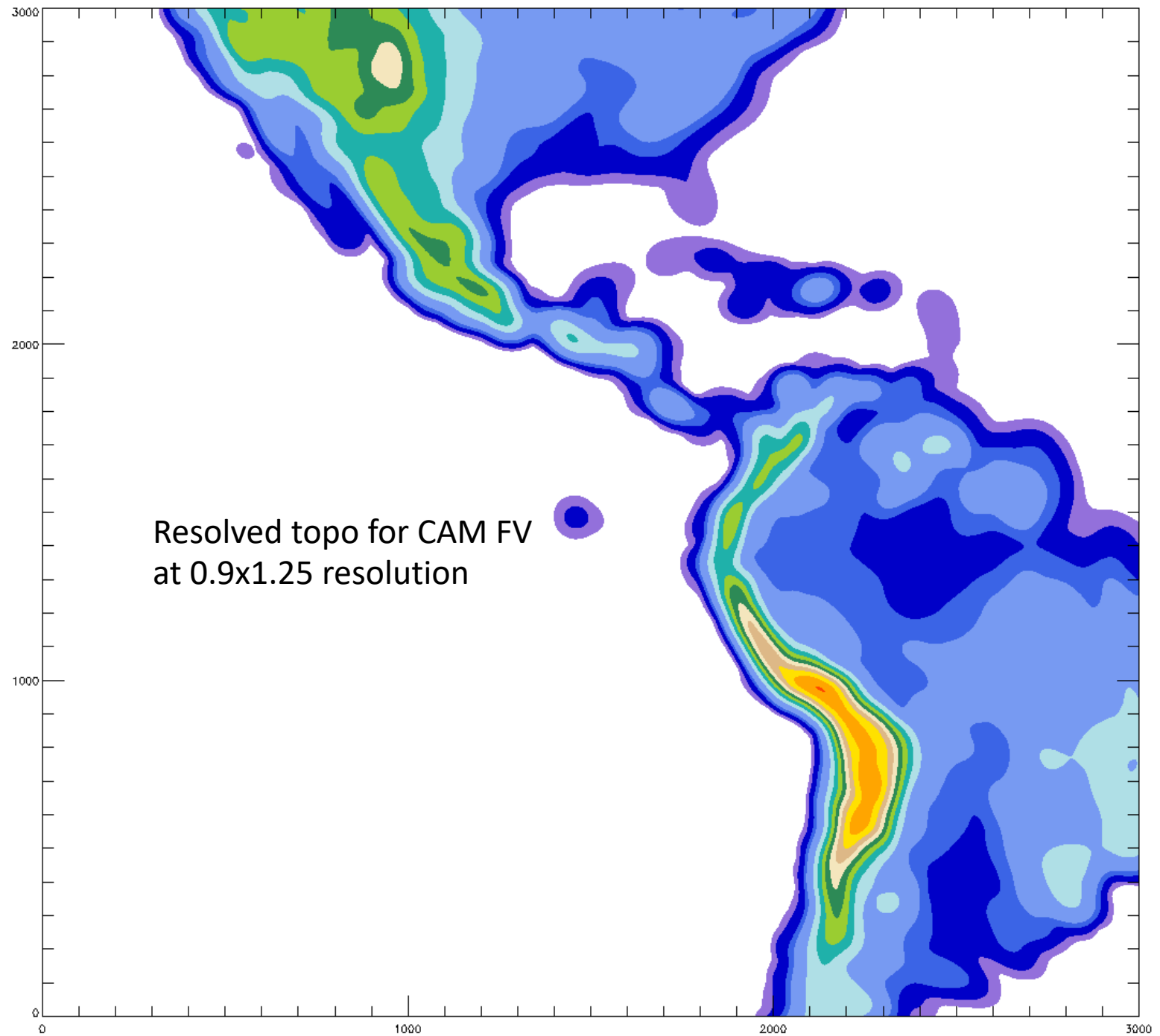
$$dq_{\{l,i,r,\dots\}}/dt =$$

$F_{QV}, F_{QL}, F_{QI} \dots?$ *(water substance evolution equations, chemistry ...)*

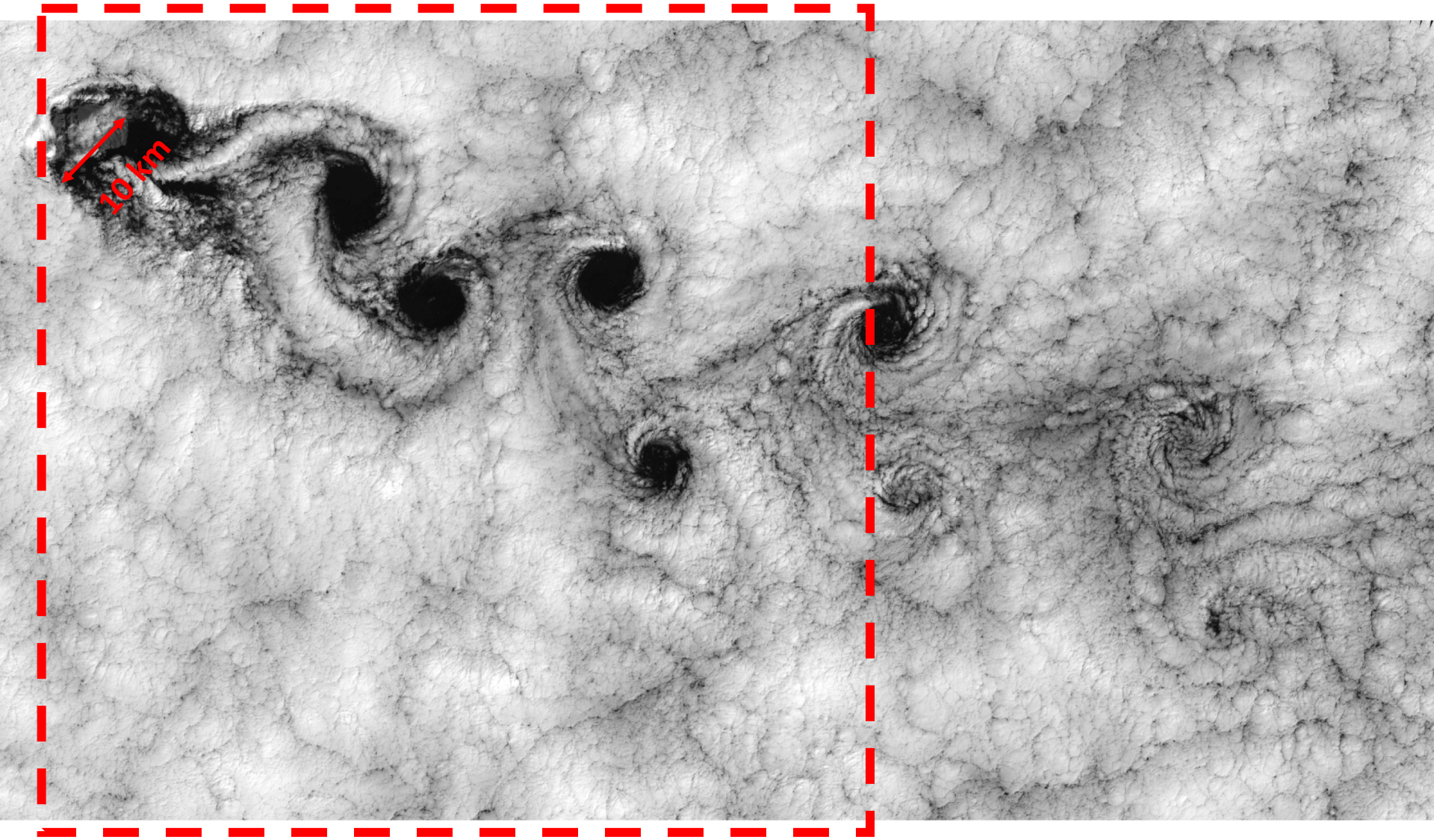
What are the “physics” trying to represent?

Unresolved motions, sub-grid
variability, photons ...
chemistry/microphysics



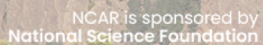


Boundary layer clouds



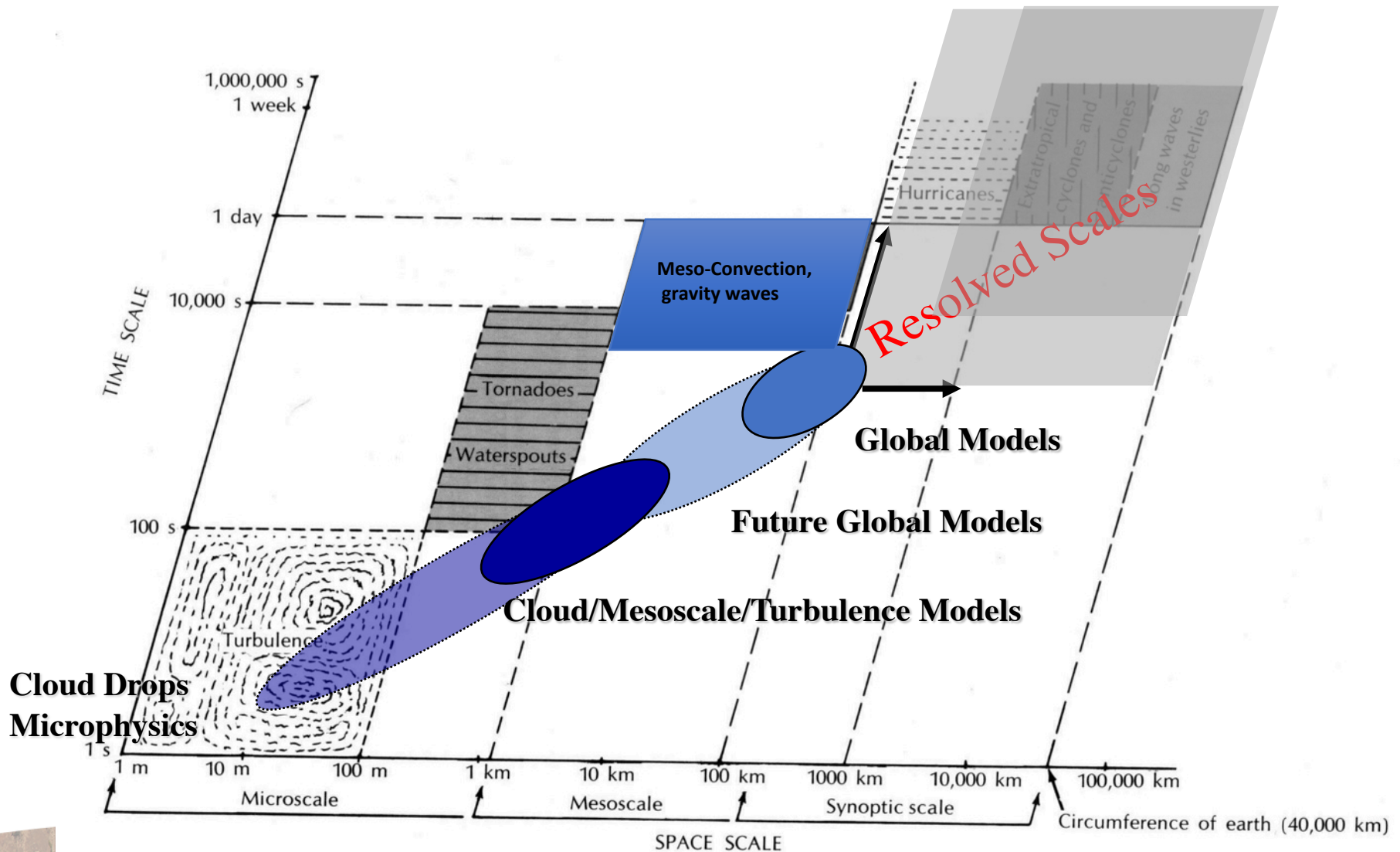
Alejandro Selkirk Island (33S 80W)

July 15, 2015



Scales of Atmospheric Processes

Resolved processes vs parameterized



Atmospheric models with large grid
boxes miss a lot of interesting stuff

....

Does it matter??

Nonlinearity

5 Values of “x” = 1,1,1,2,10

$$M(p) = \left[\frac{1}{5} \sum x^p \right]^{\frac{1}{p}}$$

p=1

$$M(p) = 3.00000$$

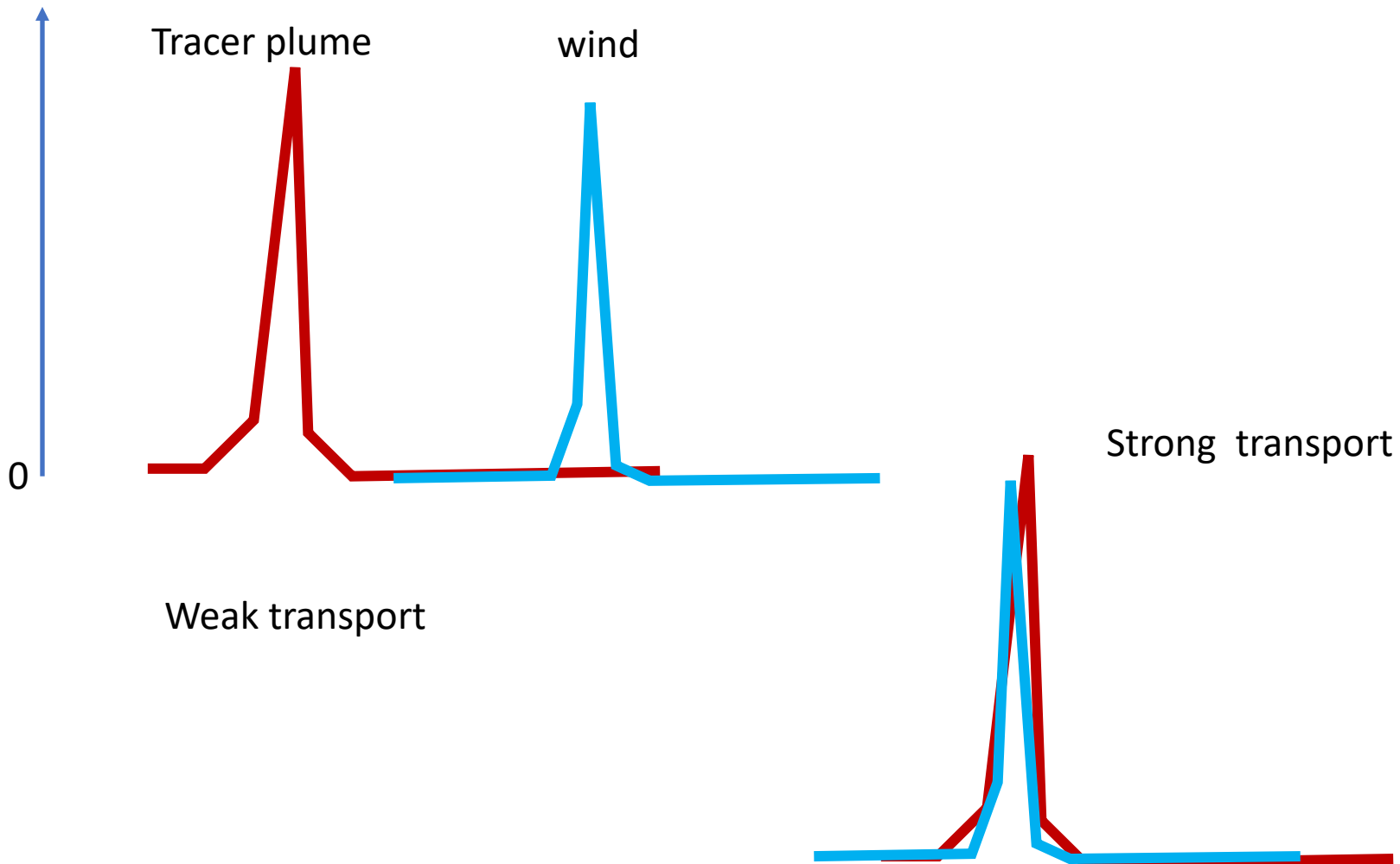
p=2

$$M(p) = 4.62601$$

p=10

$$M(p) = 8.51340$$

Nonlinearity



How do nonlinearities arise?

e.g., Cloud microphysics:

Autoconversion of cloud
water to rain

$$P_{l \rightarrow r} = k q_l^a N_l^b$$

a ranges from 2 to 4;

b from around -1 to -2

Condensation at RH=1

How do nonlinearities arise?

Fluxes:

$$\partial_t(\rho s) + \partial_x(\rho u s) + \partial_y(\rho v s) + \partial_z(\rho w s) = P_s$$

$$\partial_t(\rho \bar{s}) + \partial_x(\rho \bar{u} \bar{s}) + \partial_y(\rho \bar{v} \bar{s}) + \partial_z(\rho \bar{w} \bar{s}) = \\ \bar{P}_s - \partial_x \overline{\rho u' s'} - \partial_y \overline{\rho v' s'} - \partial_z \overline{\rho w' s'}$$

$\overline{(\)}$ large-scale horiz. average; $(\)'$ deviation from avg.

What do parameterizations do?

- Physics schemes - “*parameterizations*” - need to return tendencies as functions of model grid mean variables
- Tendency calculations may include representation of subgrid variability

$$d\bar{\mathbf{V}}/dt + f\mathbf{k} \times \bar{\mathbf{V}} + \nabla\bar{\phi} = \mathbf{F},$$

(horizontal momentum)

$$d\bar{T}/dt - \kappa\bar{T}\omega/p = Q/c_p,$$

(thermodynamic energy)

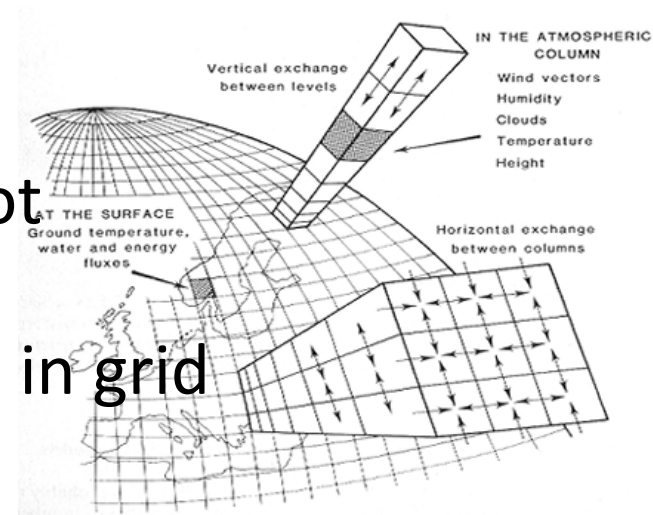
$$d\bar{q}/dt = S_q.$$

(water vapor mass continuity)

Parameterizations

High level design

1. Inputs and effects totally contained within single columns
 - Single grid point structures are believed
 2. Most (many common) schemes do not possess a “memory”
 3. Assume sufficient space-time volume in grid means for “good” statistics
 4. For climate should be mass, momentum and energy conserving (limiters and fixers)
- 1,2 and 3 begin to cause trouble as resolution increases and time-steps decrease*



“Column physics”

Subgrid horizontal fluxes are typically ignored in atmospheric models

$$\partial_t(\rho\bar{s}) + \partial_x(\rho\bar{u}\bar{s}) + \partial_y(\rho\bar{v}\bar{s}) + \partial_z(\rho w s) =$$
$$\bar{P}_s - \cancel{\partial_x \rho \overline{u' s'}} - \cancel{\partial_y \rho \overline{v' s'}} - \partial_z \rho \overline{w' s'}$$

$$\partial_t(\rho\bar{s}) + \partial_x(\rho\bar{u}\bar{s}) + \partial_y(\rho\bar{v}\bar{s}) + \partial_z(\rho w s) =$$
$$\bar{P}_s - \partial_z \rho \overline{w' s'}$$

Column physics don't need to communicate with neighboring grid columns → “embarrassingly parallel”

Physics Parameterizations needed by an AGCM

- Radiation
 - Clear sky (typically no subgrid variability used)
 - Cloudy
- Surface exchanges
- Boundary Layer Turbulence
- Shallow convection
- Cloud “macrophysics”
- Deep Convection
- Cloud microphysics
- PBL form drag
- Gravity wave drag

Physics Parameterizations in CAM6

- Radiation **RRTMG**
 - Clear sky (typically no subgrid variability used)
 - Cloudy
- Surface exchanges **Similarity theory (Monin-Obukhov ...)**
- Boundary Layer Turbulence
- Shallow convection **CLUBB prognostic moments**
- Cloud “macrophysics”
- Deep Convection **Zhang & McFarlane mass flux scheme**
- Cloud microphysics **Morrison Gettelman 2-moment**
- PBL form drag **Beljaars et al neutral shear flow over obstacles**
- Gravity wave drag **Lindzen-type schemes for various sources**
- **Complex prognostic aerosol model**

How are parameterizations built?

- Basic physics
- Empirical formulations from observations or high-resolution calculations (e.g. LES, CRMs)
- Some simple conceptual model – “cartoon”

Parameterizations

How do they couple to other parts of model?

Process splitting versus ***time splitting***

Process splitting:

- All parameterizations work on same state.
Provide tendencies for unified update

Time splitting

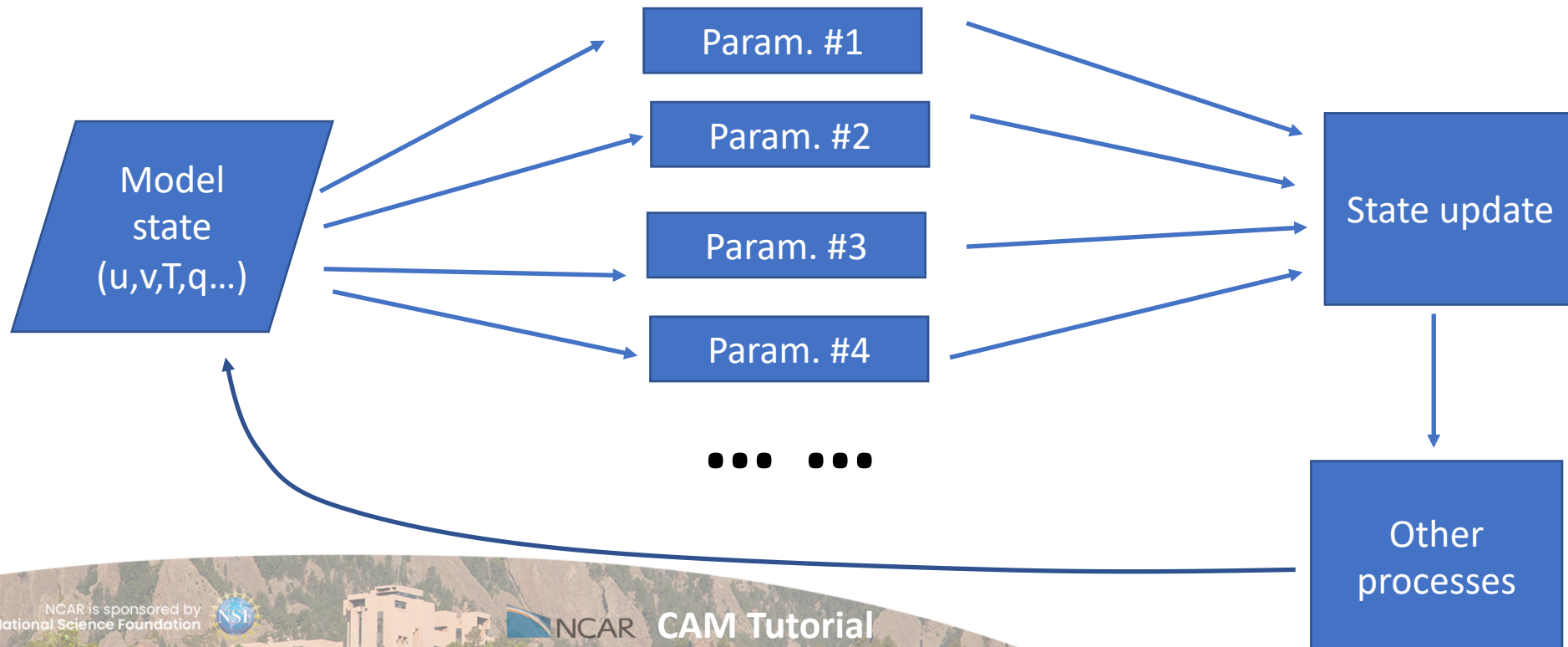
- Parameterizations update state as they work
and pass updated state to next param.

Parameterizations

How do they couple to other parts of model?

Process splitting:

- All parameterizations work on same state.
Provide tendencies for unified update

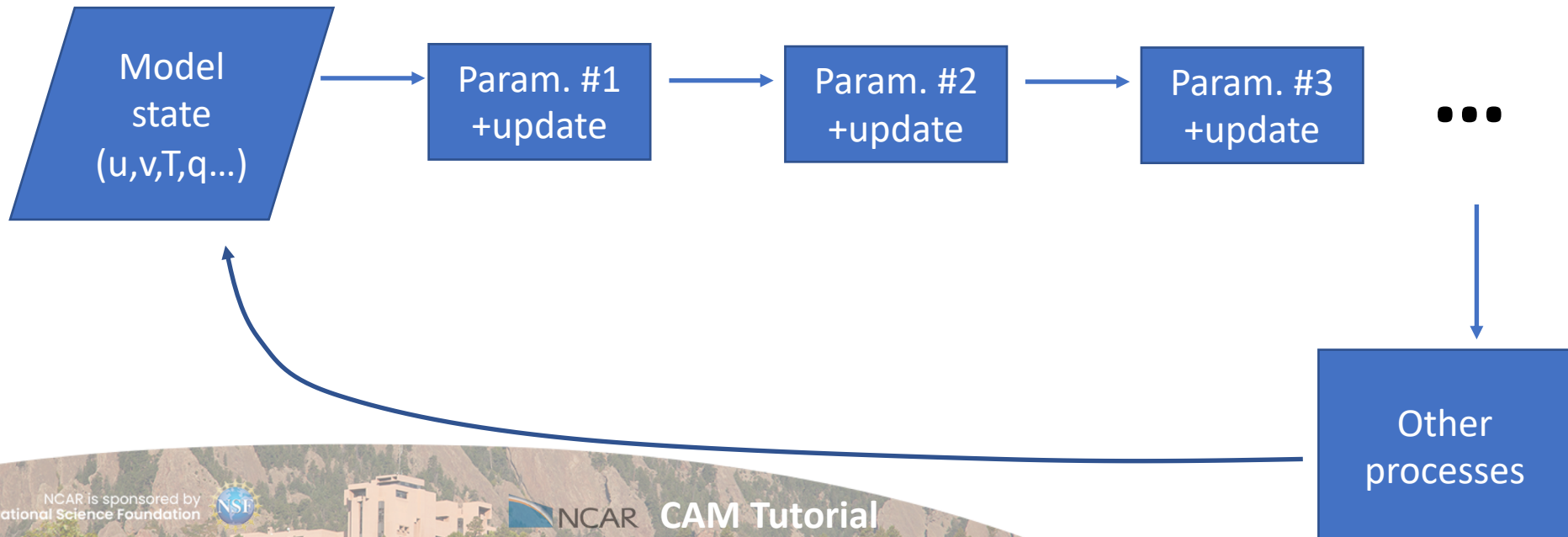


Parameterizations

How do they couple to other parts of model?

Time splitting (currently used in CAM)

- Parameterizations update state as they work and pass updated state to next param.



Future Directions for Physics in Models?

What do we need to consider?

As grid-sizes and time steps decrease, parameterizations may need to communicate across space and time

As grid-sizes and time steps decrease, resolved scales may not contain enough information to close parameterizations

- Stochastic elements?
- Life-cycles of processes?

At any resolution, better sub-grid representations are needed

- Subcolumns?

Future Directions for Physics in Models?

Science insurgents plot a climate model driven by artificial intelligence

By [Paul Voosen](#) Jul. 26, 2018 , 2:00 PM

<http://www.sciencemag.org/news/2018/07/science-insurgents-plot-climate-model-driven-artificial-intelligence>

Learning the climate

A new data-driven climate model will use satellite observations and high-resolution simulations to learn how best to render its clouds. Similar methods will also be applied to other, small-scale phenomena, such as sea ice and ocean eddies.

