

# Parametrization of moisture condensation using stochastic Lagrangian models

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## Condensation of water vapour

specific humidity of an air parcel:

$$Q = \frac{\text{mass of water vapour}}{\text{total air mass}}$$

• saturation specific humidity,  $q_s(T)$ •  $Q = \begin{cases} q_s & \text{if } Q > q_s & (\text{excessive moisture condensed}) \\ Q & \text{otherwise} \end{cases}$ 

 ${\scriptstyle { \bullet } \ } q_s(T)$  decreases with temperature T, hence generally decreases with height



## Water vapour transport in the atmosphere

- specific humidity described by a continuous field q(x, y, t)
- evolution of q(x, y, t) governed by a partial differential equation (advection-diffusion-condensation):

$$egin{aligned} &rac{\partial q}{\partial t} + ec{u} \cdot 
abla q = \kappa 
abla^2 q + S - C \ & C = rac{1}{ au_c} (q - q_s) \, \mathcal{H}(q - q_s) \end{aligned}$$

 $\kappa:$ eddy diffusivity representing small-scale turbulence

#### Water vapour transport in the atmosphere

- specific humidity described by a continuous field q(x, y, t)
- evolution of q(x, y, t) governed by a partial differential equation (advection-diffusion-condensation, fast condensation limit):

$$\frac{\partial q_*}{\partial t} + \vec{u} \cdot \nabla q_* = \kappa \nabla^2 q_* + S$$
$$C: q \to \min[q_*, q_s]$$

 $\kappa:$ eddy diffusivity representing small-scale turbulence

a set of equations for cloudy planetary boundary layer (Bougeault, JAS, 1981)

$$\frac{\partial u_i}{\partial t} = -\frac{\partial}{\partial x_\alpha} (u_\alpha u_i) - \frac{1}{\rho_0} \frac{\partial p}{\partial x_i} + \beta \delta_{3l} \theta_v + \nu \nabla^2 u_i + \epsilon_{ijk} \Omega_j u_k \frac{\partial \theta_l}{\partial t} = -\frac{\partial}{\partial x_\alpha} (u_\alpha \theta_l) + \nu_\theta \nabla^2 \theta \frac{\partial q_w}{\partial t} = -\frac{\partial}{\partial x_\alpha} (u_\alpha q_w) + \nu_q \nabla^2 q \theta = \theta_l + \left[\frac{p_0}{p_r(z)}\right]^k \frac{L}{c_p} q_l q_l = \{q_w - q_s[\theta, p + p_r(z)]\} \times H\{q_w - q_s[\theta, p + p_r(z)]\}$$

## Subgrid-scale fluctuations



**9** governing PDEs describe the atmospheric state at all scales

- weather/climate models: numerical solutions of a discrete version of these equations on a coarse grid (horizontal resolution ~ kilometers)
- $\blacksquare$  humidity inside a grid box is represented by a single value of q
- missing subgrid-scale moisture variability has significant effects on the resolved scales—system tends to saturate!

## Advection-condensation model

- bounded domain:  $[0, \pi] \times [0, \pi]$ , reflective boundaries
- large-scale cellular flow:  $\psi = \sin x \sin y$ ;  $(u, v) = (-\psi_y, \psi_x)$
- small-scale turbulence
- $q_s(y) = q_{\max} \exp(-\alpha y)$ :  $q_s(0) = q_{\max}$  and  $q_s(\pi) = q_{\min}$
- **9** resetting source:  $Q = q_{\text{max}}$  if parcel hits y = 0



### **Eulerian field formulation: PDE**

$$\frac{\partial q_*}{\partial t} + \vec{u} \cdot \nabla q_* = \kappa \nabla^2 q_*$$
$$C: q(x, y, t) \to \min[q_*(x, y, t), q_s(y)]$$

• boundary source:  $q(x, y = 0, t) = q_{\max}$ 

• relative humidity:  $r(x, y, t) = \frac{q(x, y, t)}{q_s(y)}$ 



## Stochastic Lagrangian model: SDE

$$dX(t) = u(X, Y) dt + \sqrt{2\kappa} dW_1(t) \qquad \psi = -\sin x \sin y$$
  

$$dY(t) = v(X, Y) dt + \sqrt{2\kappa} dW_2(t) \qquad u = -\psi_y$$
  

$$dQ(t) = [S(Y) - C(Q, Y)] dt \qquad \psi = \psi_x$$
  

$$\kappa = 10^{-2}$$

fast condensation  $C: Q \to \min[Q, q_s(Y)]$ 



Tsang & Vanneste, Proc. R. Soc. A 473, 20170196 (2017)

## **Eulerian versus Lagrangian formulation**

• to facilitate comparison, divide the domain into small bins and average over parcels in each bin to produce a field  $r_{\text{bin}}(x, y)$ : making a coarse-resolution "observation"





 $\frac{1}{x}$ 

"observed"  $r_{\rm bin}(x,y)$ 

1.0 0.9 0.8 0.7 0.6

 $0.5 \\ 0.4 \\ 0.3 \\ 0.2 \\ 0.1 \\ 0.0$ 

#### **Eulerian versus Lagrangian formulation**



 $1.0 \\ 0.9 \\ 0.8 \\ 0.7 \\ 0.6$ 

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## Why are PDE models so wet?

The coarse-graining process and the condensation process do not commute and local fluctuation is lost:



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The coarse-graining process and the condensation process do not commute and local fluctuation is lost:



- Lagrangian formulation is generally computational expensive
- However, it is possible for the Eulerian model to mimic the Lagrangian model by parametrizing subgrid-scale condensation

#### An effective subgrid condensation

● In the stochastic Lagrangian model, (Q, X, Y) are governed by SDEs ⇒ the joint PDF P(q, x, y; t) satisfies the Fokker–Planck equation

$$\frac{\partial P}{\partial t} + \vec{u} \cdot \nabla P - \frac{\partial}{\partial q} (CP) = \kappa \nabla^2 P \qquad C = \frac{1}{\tau_c} (q - q_s) \mathcal{H}(q - q_s)$$

•  $q_{\text{bin}}(x,y,t) \approx \overline{q}(x,y,t) = \int_{q_{\min}}^{q_{\max}} q' \, \hat{P}(q'|x,y;t) \, \mathrm{d}q'$ 

$$\frac{\partial \overline{q}}{\partial t} + \vec{u} \cdot \nabla \overline{q} = \kappa \nabla^2 \overline{q} - \int_{q_{\min}}^{q_{\max}} C(q', q_s) \hat{P}(q'|x, y; t) \, \mathrm{d}q'$$

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**9** In the Eulerian model, the coarse-grained field q(x, y, t) satisfies

$$\frac{\partial q}{\partial t} + \vec{u} \cdot \nabla q = \kappa \nabla^2 q - C(q, q_s)$$

Effective condensation to put back some local variability:

$$C(q,q_s) \to C_{\text{eff}}(q,q_s) = \int_{q_{\min}}^{q_{\max}} C(q',q_s) \Phi_*(q'|x,y;t) \,\mathrm{d}q'$$

 $\Phi_*(q'|x,y;t)$  is an approximation to  $\hat{P}(q'|x,y;t)$ . How?

## Assumed PDF and matching moments

1. method of assumed PDF with small number of parameters:  $(\beta, a, \sigma)$ 

$$\Phi_*(q'|x, y, t) = \beta(x, y, t)\delta(q - q_{\min}) + \tilde{\Phi}_*(q'|x, y, t)$$

2. (a) from the Fokker–Planck equation, derive the equations governing the moments in the stochastic Lagrangian model (m = 1, 2, ..., M)

$$\mu_m(x, y, t) \equiv \int q'^m \hat{P}(q'|x, y, t_n) \mathrm{d}q',$$

(b) to determine  $(\beta, a, \sigma)$ , solve for  $\mu_m$  and set

$$\int q'^m \Phi_*(q'|x,y,t_n) \mathrm{d}q' = \mu_m$$



### **Results: relative humidity**



1.00.9 0.80.70.6

0.50.40.30.20.10.0

> 1.00.9 0.8 0.7 0.6

0.50.40.30.20.10.0

## Condensation parametrization in atmospheric models

- the idea of representing subgrid-scale moisture variability by a probability distribution has been employed in weather and climate models since Sommeria & Deardorff (1977),
  - such distributions are introduced in an ad hoc manner
  - the parameters in the PDF are determined using various forms of turbulent closures
- $\checkmark$  Here, using an advection–condensation model, we demonstrate:
  - mathematically how the Lagrangian formulation gives more realistic results than Eulerian coarse-resolution models by accounting for local variability.
  - an Eulerian model for a coarse-grained moisture field  $q(\vec{x}, t)$  can mimic its Lagrangian counterpart when an effective condensation is implemented

Based on the present results, we propose a strategy to build a condensation parametrization in atmospheric GCMs using stochastic Lagrangian models.

## Condensation parametrization in atmospheric models

- 1. choose a suitable stochastic Lagrangian model to describe parcel trajectories in subgrid-scale atmospheric transport (Lagrangian Modeling of the Atmosphere, 2013)
- 2. fix the assumed PDF  $\Phi_*(q'|x, y; t)$  in the effective condensation:

$$C_{ ext{eff}} = \int_{q_{ ext{min}}}^{q_{ ext{max}}} C(q',q_s) \Phi_*(q'|x,y;t) \, \mathrm{d} q' \quad C = rac{1}{ au_c}(q-q_s) \, \mathcal{H}(q-q_s)$$

using information from the stochastic Lagrangian model, e.g. by matching moments

- 3. implement  $C_{\text{eff}}$  in the Eulerian PDE governing the moisture field in the atmospheric GCM
- a consistent unified approach: the stochastic Lagrangian model
   (i) provides the theoretical basis for the probabilistic effective condensation as well as (ii) fixes the assumed PDF

Tsang & Vallis, J. Atmo. Sci., accepted (2018)