

1. Background

The goal of **data assimilation** (DA) is to update a forecast with observations of the modelled system in a dynamically consistent and optimal way, given **estimates** of their respective **errors** [1]. DA techniques need to evolve in order to keep up with advances in **high-resolution** Numerical Weather Prediction (NWP) as it approaches the scale at which convection is explicitly resolved. To investigate the potential of high-resolution DA schemes, **idealised models** can be employed that capture some fundamental features of **convective-scale dynamics** while remaining computationally inexpensive.

3. Model: SWEs with ‘rain’

We outline an idealised NWP model (after [2]) which is designed to represent an **idealised atmosphere with moist convection**. Moisture is incorporated into the traditional SWEs by the introduction of a ‘**rain mass fraction**’ variable, r , which acts on the momentum equation (1b) via the geopotential, and is allowed to precipitate via an additional conservation equation (1d).

$$\partial_t h + \partial_x(hu) = 0, \quad (1a)$$

$$\partial_t(hu) + \partial_x(hu^2 + p(h)) + gh\partial_x b + hc_0^2 \partial_x r - fhv = 0, \quad (1b)$$

$$\partial_t(hv) + \partial_x(huv) + fhu = 0, \quad (1c)$$

$$\partial_t(hr) + \partial_x(hur) + h\tilde{\beta}\partial_x u + \alpha hr = 0, \quad (1d)$$

$$\partial_t b = 0, \quad (1e)$$

where $p = p(h)$ is ‘effective’ pressure defined by:

$$p(h) = \begin{cases} \frac{1}{2}gH_c^2, & \text{for } h + b > H_c, \\ \frac{1}{2}gh^2, & \text{otherwise,} \end{cases}$$

and:

$$\tilde{\beta}(h, \partial_x u) = \begin{cases} \beta, & \text{for } h + b > H_r \text{ and } \partial_x u < 0, \\ 0, & \text{otherwise.} \end{cases}$$

(Black - standard SWEs; red - modifications.)

- h = fluid depth, (u, v) = velocities, b = bottom topography, r = rain mass fraction; all as a function of (x, t) .
- H_c, H_r = threshold heights, above which convection and ‘rain’ processes occur.
- $c_0^2 = gH_r$, α = rate of ‘rain’ removal, β controls ‘rain’ formation, f = Coriolis frequency, g = gravity.

4. Why SWEs?

The **traditional** SWEs:

- admit **gravity waves** and have **scale interactions** (fast/slow modes).
- admit discontinuous solutions, akin to the **propagation of atmospheric fronts**.

The **modified** SWEs:

- artificially mimic **conditional instability** (positive buoyancy) and include idealised transport of **moisture**.
- contain **switches** for the onset of convection and precipitation - realistic features of operational NWP models.

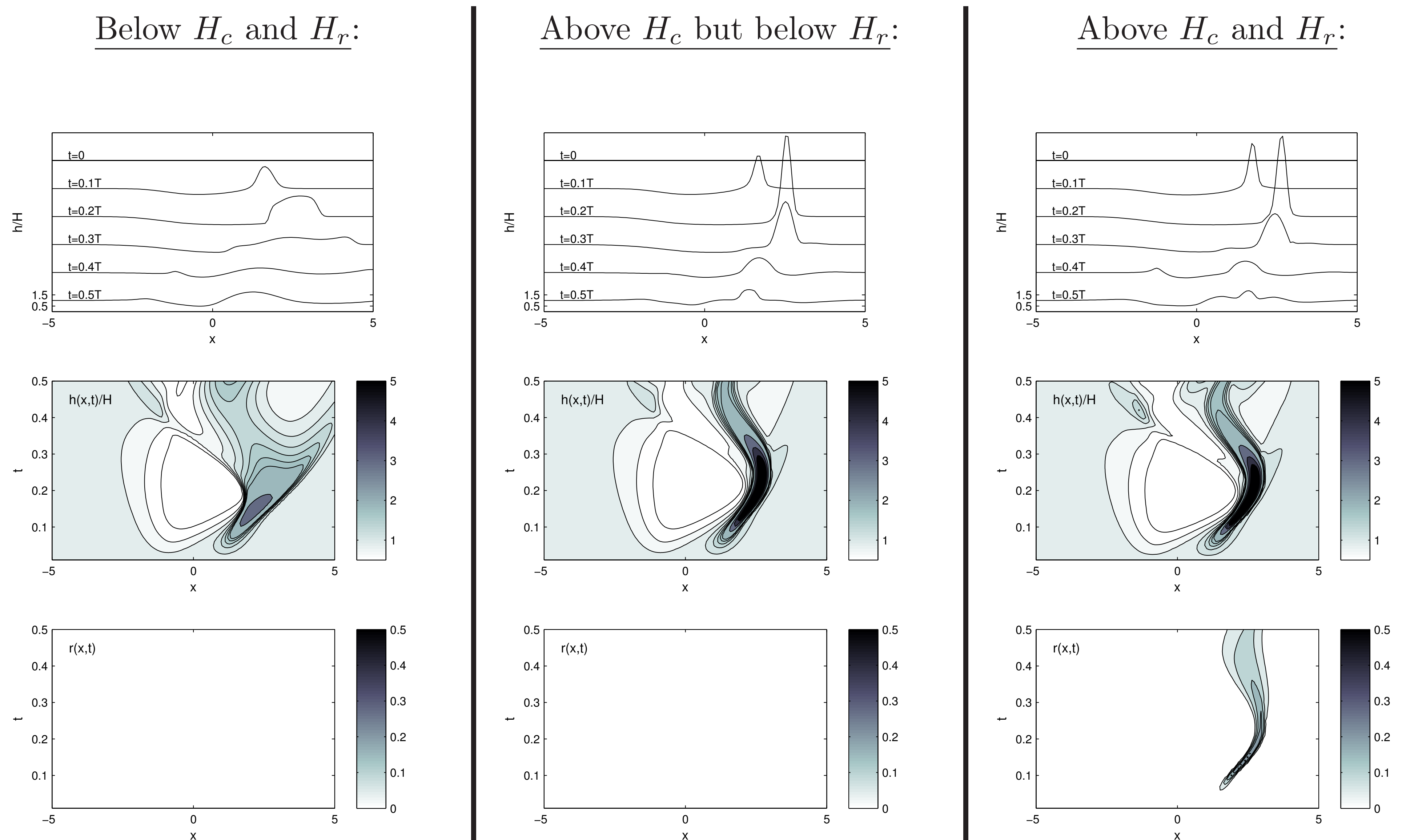
2. Approach

1. Introduce a **physically plausible idealised model** and implement numerically.
 - based on the shallow water equations (SWEs), often used in the meteorological community for modelling atmospheric circulation.
 - compare dynamics of the modified model to those of the classical shallow water theory, e.g., the Rossby geostrophic adjustment and flow over topography.
2. Idealised **convective-scale data assimilation** experiments: Ensemble Kalman Filter (EnKF) with perturbed observations.
 - ‘truth’ trajectory is determined by very high resolution numerical solution and pseudo-observations are generated by randomly perturbing this ‘truth’.
 - ensembles are generated at a coarser ‘forecast’ resolution in which small-scale features are not fully resolved.

5. Dynamics: classical and modified

Rossby geostrophic adjustment in a periodic domain

- describes the evolution of the free surface height h when disturbed from its rest state by a transverse jet.
- to adjust to this initial momentum imbalance, the height field evolves rapidly, emitting inertia gravity waves which propagate from the jet and eventually occupy the whole domain.
- periodic boundary conditions mean that waves travelling out from the jet core interact at a later time (not shown here).
- solved numerically using a shock-capturing finite volume/element method [3] which deals robustly with the high nonlinearity of the switches and dynamics.



The model reduces to the classical rotating shallow water model when the fluid does not exceed the threshold heights H_c and H_r .

Exceedence of H_c triggers positive buoyancy leading to a convective updraft, but no ‘rain’ is produced as H_r is not exceeded.

Given H_r exceedence and convergence ($\partial_x u < 0$), ‘rain’ is produced and then slowly precipitates, providing a downdraft to suppress convection.

6. Current and future steps

1. Set up meaningful experiments for the EnKF, relevant for convective-scale NWP:
 - choose initial ensemble to have sufficiently **fast perturbation growth**, providing a **representation of forecast error**.
 - “tune” the observing system by varying observation density, frequency, and noise.
2. Implement a **variational** DA algorithm and compare methods.

References

- [1] Kalnay, E., 2003: *Atmospheric Modeling, Data Assimilation, and Predictability*. Cambridge University Press.
- [2] Würsch, M., and Craig, G.C., 2014: A simple dynamical model of cumulus convection for data assimilation research. *Meteorologische Zeitschrift*. 23(5), 483-490.
- [3] Rhebergen, S., Bokhove, O., and Van der Vegt, J., 2008: Discontinuous Galerkin finite element methods for hyperbolic nonconservative partial differential equations. *J. Comp. Phys.*, 227(3), 1887-1922.

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