Cloud condensate control variable transform and background errors

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Tangent linear cloud variables

Additional 3D variables δq_l and δq_i are advected by the tangent linear model using updated linear cloud physics. Background error is in terms of cloud condensate $\delta q_c = \delta q_l + \delta q_i$. Transform back from cloud condensate control variable uses temperature dependent splitting, $\delta q_l = \alpha(T^b) \delta q_c$, $\delta q_i = (1 - \alpha(T^b)) \delta q_c$.

Balanced and symmetric control variable

The control variable includes grid point balance operator between temperature δT , specific humidity δq_v and cloud condensate δq_c increments (mainly reflecting cloud condensation effects). Balance coefficients Q_{cq} and Q_{qT} are determined from model level and relative humidity background rh^b dependent regression,

$$\begin{pmatrix} (\delta \hat{q}_c)_u \\ (\delta \hat{q}_v)_u \\ \delta \hat{T} \end{pmatrix} = \begin{pmatrix} 1 - Q_{cq}(rh^b) & 0 \\ 0 & 1 & -Q_{qT}(rh^b) \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} \delta \hat{q}_c \\ \delta \hat{q}_v \\ \delta \hat{T} \end{pmatrix}$$

where $\delta \hat{T} = \delta T r h^b L_v / [R_v (T^b)^2]$, $\delta \hat{q}_v = \delta q_v / q_s (T^b)$ and $\delta \hat{q}_c = \delta q_c / \sigma_{cc}(q_s(T^b) - q_s(T_l))$, using liquid temperature T_l . This is followed by normalisation by background error standard deviations,

$$egin{pmatrix} \delta ilde q_c \ \delta ilde q_v \ \delta ilde T \end{pmatrix} = egin{pmatrix} \delta \hat q_c / \sigma_{cr}(rh) \ \delta \hat q_v / \sigma_q(rh) \ \delta \hat T / \sigma_T \end{pmatrix}$$

Figure 1: Symmetrizing of cloud condensate background error pdf by fitting σ_{cc} against the frequency A statistical regression on a large sample of ensemdistribution along the mean of the background and ble forecast differences determines $\sigma_{cc}(q_s(T^b) - q_s(T_l))$ analysis cloud condensate $q_s(T^b) - q_s(T_l^b + \frac{1}{2}\delta T_l)$ (right) $\sigma_{cr}(rh)$, and $\sigma_{a}(rh)$. To obtain symmetric error disinstead of just the background $q_s(T^b) - q_s(T^b_l)$ (left). tributions, the regressions are made on forecast differences stratified against $rh^b + \frac{1}{2}\delta rh$ and $q_s(T^b) - \frac{1}{2}\delta rh$ $q_s(T_l^b + \frac{1}{2}\delta T_l)$ instead of the background values. In **3** Wavelet B 4D-VAR using multiple outer loops, this nonlinear trans-Cloud condensate has been added to a global wavelet form is performed at outer loop level.



Figure 5: Single cloud liquid observation (bold red/black circle) at start of 6H 4D-VAR window in nearly saturated area (left), very dry area with no background condensate (middle), and nearly saturated boundary layer (right): cloud condensate (black isolines) and specific humidity (white isolines) analysis increments [units 1E-6 kg/kg] and background relative humidity (colour).



sample is obtained which fills the troposphere. Stratospheric correlations are added, but have no effect due to cloud condensate error variances set to very low values above the tropopause. There is considerable geographical variation of the correlations.



Figure 2: Vertical background error correlation geographical variation in the wavelet B (averages over seasons).

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seasons).

Flow-dependent variances 4

Flow-dependent variance $\sigma_{cc}\sigma_{cr}$ calculated from statistical regression agree reasonably with ensemble spread. Current work is introducing ensemble generated variances of the day for cloud condensate and further study performance in precipitating conditions.



Figure 3: Horizontal background error correlation geographical variation in the wavelet B (averages over

Figure 4: Cloud condensate background error standard deviation $\sigma_{cc}(q_s(T^b) - q_s(T^b_l + \frac{1}{2}\delta T_l))\sigma_{cr}(rh)$ (at ca. 670 hPa) from a flow-dependent statistical regression model compared with the ensemble spread from ten ensemble forecasts valid at the same time: background (upper left), statistically estimated standard deviation (upper right), ensemble mean (lower left), ensemble spread (lower right) [units 1E-3 kg/kg].

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