A flexible three-dimensional stratocumulus, cumulus and cirrus cloud generator (3DCLOUD) based on drastically simplified atmospheric equations and the Fourier transform framework.

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Abstract

The 3DCLOUD algorithm for generating stochastic three-dimensional (3D) cloud fields is described in this paper. The generated outputs are 3D optical depth (τ) for stratocumulus and cumulus fields and 3D ice water content (IWC) for cirrus clouds. This model is designed to generate cloud fields that share some statistical properties observed in real clouds such as the inhomogeneity parameter ρ (standard deviation normalized by the mean of the studied quantity), the Fourier spectral slope β close to -5/3 between the smallest scale of the simulation to the outer L_{out} (where the spectrum becomes flat). Firstly, 3DCLOUD assimilates meteorological profiles (humidity, pressure, temperature and wind velocity). The cloud coverage C, defined by the user, can also be assimilated, but only for stratocumulus and cumulus regime. 3DCLOUD solves drastically simplified basic atmospheric equations, in order to simulate 3D cloud structures of liquid or ice water content. Secondly, the Fourier

1   filtering method is used to constrain the intensity of ρ, β, L_{out} and the mean of τ or IWC

2   of these 3D cloud structures. 3DCLOUD model was developed to run on a personal computer

3   under Matlab environment with the Matlab statistics toolbox. It is used to study 3D

4   interactions between cloudy atmosphere and radiation.

5

6   **1   Introduction**

7

8   Clouds have a significant effect on the Earth radiation budget. They reflect the solar radiation

9   and reduce the warming of the Earth (albedo effect). They also create a greenhouse effect by

10 trapping the thermal radiation emitted from the Earth's surface, reducing the radiative cooling

11 of the Earth (Collins and Satoh, 2009). Cloud feedback has remained, however, the largest

12 uncertainty in the study of climate sensitivity for almost twenty years (Bony et al., 2006). In

13 almost all climate models, clouds are assumed plane-parallel with homogeneous optical

14 properties (PPH), and radiation codes use a one-dimensional (1D) scheme. Therefore,

15 improving parameterisations of clouds in large-scale model, especially their interaction with

16 radiation, is a challenge in order to reduce uncertainty in model projections of the future

17 climate (Illingworth and Bony, 2009). Improved global characterization of the three

18 dimensional (3D) spatial distribution of clouds is, thus, necessary (Clothiaux et al., 2004).

19 Moreover, satellite passive sensors, such as multi-spectral and multi-angular radiometers, and

20 satellite active sensor, such as LIDAR and RADAR in the A-train mission, allow the retrieval

21 of cloud horizontal and vertical optical properties with an adequate spatial and temporal

22 coverage. For practical and computational cost purposes, interpretation of such measurements

23 generally also assumes 1D radiative algorithm and PPH cloud. This assumption can be far

24 from being realistic and leads to biases on the retrieved properties from passive sensors

25 (Barker and Liu, 1995; Várnai and Marshak, 2002; Lafont and Guillemet, 2004; Cornet et al.,

26 2005; Várnai and Marshak, 2007; Cornet et al., 2013) and active sensors (Battaglia and

27 Tanelli, 2011). These biases depend at least, on the cloud coverage and on the variability of

28 cloud optical depth or water content. This variability is quantified by an inhomogeneity

29 parameter, often defined as the standard deviation normalized by the mean of the studied

30 quantity (Szczap et al., 2000; Carlin et al., 2002; Oreopoulos and Cahalan, 2005; Sassen et al.,

31 2007; Hill et al., 2012).

Determining the significance of the 3D inhomogeneity of clouds for climate and remote sensing applications requires the measurement and the simulation of the full range of actual cloud structure. Apart from the computational time, accurate 3D cloudy radiative transfer problem is not an issue, per se (Evans and Wiscombe, 2004). Monte Carlo transfer models can indeed accurately and efficiently compute radiative properties for arbitrary cloud fields (Battaglia and Mantovani, 2005; Pincus and Evans, 2009; Mayer, 2009; Cornet et al., 2010; Battaglia and Tanelli, 2011; Cornet et al., 2013; Fauchez et al., 2013). The difficulty is to generate cloud property fields that are statistically representative of cloud fields in nature.

Cloud fields generated by dynamic cloud models, such as the cloud resolving model (CRM) or the large-eddy simulation model (LES), are very attractive, as they contain the state of the art of physical processes (resolution of atmospheric equations, detailed microphysics, radiation, .etc.). The goal of the LES approach is to simulate the three-dimensional atmospheric turbulent flows. There are different scales of turbulent eddies; large eddies (from 100 to 1000 m and more) that are produced directly by the instability of the mean flow and small eddies (from a few centimeters to 100 m) as well as by the energy-cascade process from the larger eddies (Moeng, 1984). LES seeks to capture accurately the larger eddies, while only modeling the smaller ones. Instead of reproducing all the scales of turbulence flow, they can integrate a flow in which small scale details are removed from the solution. The spatial filtered equations can, therefore, be integrated with available resources (Bryan et al., 2003). Nevertheless, they are still very expensive to run in a 3D domain.

Stochastic models have the capability to simulate quickly realistic 2D and 3D cloud structures with just a few parameters. Examples of these types of cloud models are: the bounded cascade model (Cahalan et al., 1994 ; Marshak et al., 1998), the IAAFT algorithm (Venema et al., 2006), the SITCOM model (Di Guiseppe and Thompkins, 2003), the tdMAP model (Benassi et al., 2004), the model developed by Evans and Wiscombe (2004) for low liquid clouds (stratocumulus and cumulus) or by Alexandrov et al. (2010) and the Cloudgen model (Hogan and Kew, 2005) for high ice clouds (cirrus). These stochastic models are based on fractal or Fourier framework. The scale invariant properties observed in real clouds can be controlled. The power spectra of the logarithm of their optical properties (optical depth, liquid water content or liquid water path for low clouds and ice water content for high clouds) typically exhibits a spectral slope of around -5/3 (Davis et al., 1994 ; Cahalan et al., 1994 ; Davis et al., 1996 ; Davis et al., 1999 ; Davis et al., 1997 ; Benassi et al., 2004 ; Hogan and Kew, 2005 ;

1 Hill et al., 2012, Fauchez et al., 2014) from small scale (a few meters) to the "integral scale"

2 or the outer scale (few tenth to hundred kilometres), where the spectrum becomes flat (i.e.

3 decorrelation occurs). The disadvantage of such models arises from the fact that effects of

4 meteorological processes are not always considered and dominant scales of organization

5 related to turbulent eddy due, for example, to wind shear, convection, and entrainment are not

6 directly modeled. At the same time, it should be noted that Cloudgen does consider the effect

7 of wind shear on cirrus cloud.

8 The aim of the 3DCLOUD algorithm is to reconcile these two approaches. In Sect. 2, we

9 describe the 3DCLOUD generator. In Sect 3, 3DCLOUD outputs are compared to LES

10 outputs to check the validity of the chosen basic atmospheric equations. In Sect 4,

11 stratocumulus, cumulus and cirrus examples provided by 3DCLOUD are presented.

12

13 2   The 3DCloud generator

14

15 3DCLOUD generates, in two distinct steps (see Fig. 1), a 3D optical depth field for

16 stratocumulus and cumulus or a 3D ice water content field for cirrus clouds. These cloud

17 fields were chosen as most of the papers dealing with scale invariant properties focus on

18 liquid water path and optical depth for stratocumulus and cumulus and on ice water content

19 for cirrus. During the first step, meteorological vertical profiles (temperature, pressure, wind,

20 humidity), defined by the user, are assimilated and basic atmospheric equations are resolved.

21 During the second step, cloud scale invariant properties are constrained in a Fourier

22 framework. At the same time, a gamma distribution of local optical depth or IWC is mapped

23 onto the LWC/IWC generated during the first step. This gamma distribution is iteratively

24 computed in such way that the mean optical depth or IWC and the inhomogeneity parameter

25 satisfy the values imposed by the user. Details of these two steps are presented below.

26 2.1   Step 1: the 3D LWC/IWC generator

27 The essential basic quantities to generate cloud fields are the condensed water mixing ratio

28 $q_c = q_l + q_i$ where q_l is the liquid water mixing ratio and q_i is the ice water mixing ratio, the

29 wind velocity vector $\boldsymbol{u}$, air pressure p, temperature T, and vapour water mixing ratio q_v.

4

1 Mixing ratios are the mass of vapour or condensed water per unit of dry air mass. We describe

2 in this section, the equations used to generate clouds with the associated simplifications used.

2.1.1 The simplification of basic atmospheric equations

4 The continuity and momentum equations of the atmosphere can be written (Houze, 1993):

$$5 \quad \begin{cases} \dfrac{D\rho}{Dt} = -\rho \nabla . \boldsymbol{u} \\[2mm] \dfrac{D\boldsymbol{u}}{Dt} = -\dfrac{1}{\rho}\nabla p - f\boldsymbol{k} \wedge \boldsymbol{u} - g\boldsymbol{k} + \boldsymbol{F} \end{cases} \tag{1}$$

6 where t is time, ρ the air density, f the Coriolis parameter, g the acceleration due to

7 gravity and $\boldsymbol{F}$ the acceleration due to other forces (frictional acceleration for example).

8 $D/Dt = \partial/\partial t + \boldsymbol{u} \cdot \nabla$ is the Lagrangian derivative operator following a parcel of air, $\partial/\partial t$ is the

9 Eulerian derivative operator and ∇ is the three-dimensional gradient operator. $\boldsymbol{u} = u\boldsymbol{i} + v\boldsymbol{j} +$

10 $w\boldsymbol{k}$ is the wind velocity vector with horizontal components u, v and vertical component w

11 projected in the Cartesian geometry system, where $\boldsymbol{i}$, $\boldsymbol{j}$ and $\boldsymbol{k}$ are the unit vectors in the x, y

12 and z directions. The continuity and momentum equations of the atmosphere under the

13 anelastic and Boussinesq approximation, assuming shallow motion, neglecting Coriolis

14 parameter, neglecting frictional forces, and neglecting the molecular viscosity can be written

15 (Holton, 2004, p. 117 ; Houze, 1993, p. 35 ; Emanuel, 1994, p. 11)

$$16 \quad \begin{cases} \nabla . \boldsymbol{u} = 0 \\[2mm] \dfrac{D\boldsymbol{u}}{Dt} = -\dfrac{1}{\rho_0}\nabla p^* + B\boldsymbol{k} \end{cases} \tag{2}$$

17 where B is the buoyancy acceleration, ρ_0 is the constant mean value of air density and p^* is

18 the pressure perturbations. The above differential operators are valid only in the limit when

19 $\delta t,\ \delta x,\ \delta y$ and δz approach 0 (Pielke, 2002, p. 41). Nevertheless, turbulent motions (shear

20 induced eddies, convection eddies) have spatial and temporal variations at scales much

21 smaller than those resolved by LES and 3DCLOUD. If we assume field variables can be

22 separated in slowly varying mean field and rapidly varying turbulent component, and if we

23 apply the Reynolds decomposition, we can rewrite the above equation set as:

$$24 \quad \begin{cases} \nabla . \boldsymbol{u} = 0 \\[2mm] \dfrac{D\boldsymbol{u}}{Dt} = -\dfrac{1}{\rho_0}\nabla p^* + B\boldsymbol{k} + \boldsymbol{\Phi} \end{cases} \tag{3}$$

1   where Φ is the three dimensional convergence of the eddy flux of moment (Houze, 1993, p.

2   42), the turbulent flux (Holton, 2002, p 119) or the sub-grid correlation term (Pielke, 2002, p.

3   44). The Reynolds decomposition is not used in LES. The atmospheric equations are derived

4   by spatial filtering, where a special function is applied. Thus, the filtering operation acts on

5   atmospheric quantities and separates them in two categories: the resolved one (large eddy)

6   and unresolved one (subgrid-scale). An unknown term remains in the filtered equations of

7   LES, often called the subgrid-scale stress, which needs to be parameterized or estimated with

8   the help of subgrid-scale modeling. This subgrid-scale stress for LES equations is analogous

9   to the Φ term for Reynolds decomposition. In 3DCLOUD, the Φ term is voluntarily

10   neglected. Indeed, the guiding idea of 3DCLOUD is to simulate, in the fastest way, 3D

11   fluctuations of LWC/IWC of a cloud showing turbulent properties (or invariant scale

12   properties).

13   When water phase changes are only associated with condensation and evaporation (or

14   sublimation), the first law of thermodynamics can be written (Houze, 93):

15
$$\frac{D\theta}{Dt} = -\frac{L}{C_p \Pi} \frac{Dq_v}{Dt} \tag{4}$$

16   where $L = 2500$ kJkg^{-1} and $L = 2800$ kJkg^{-1} are the usual latent heat of vaporization of water

17   and ice, respectively. $C_p = 1.004$ kJkg^{-1}K^{-1} is the usual specific heat of dry air at constant

18   pressure, θ is the potential temperature and $\Pi = (p/p_0)^\kappa = T/\theta$ is the Exner function where

19   $p_0 = 1000$ hPa and $\kappa = 0.286$. In addition to the equation of motion and the first law of

20   thermodynamics, air parcels follow the water continuity equation:

21
$$\frac{Dq_i}{Dt} = S_i, \quad i = 1, ..., n \tag{5}$$

22   where S_i are the sum of the sources and sinks for a particular category (among n categories)

23   of water indicated by i (vapor, solid, liquid water category for example).

24   As the horizontal extension of the simulated cloud fields is around a few km, horizontal

25   pressure is assumed to be constant. Therefore, the current version of 3DCLOUD does not

26   have a large enough domain to contain power in the mesoscale. All these considerations lead

27   to a dramatic simplification of the dynamic equations. The simplified equations of

28   3DCLOUD governing the formation of 3D cloud structures are:

$$1 \quad \begin{cases} \dfrac{D\boldsymbol{u}}{Dt} = g\left(\dfrac{\theta_v^*}{\theta_{v0}} - q_c\right)\boldsymbol{k} - \dfrac{1}{\rho_0}\dfrac{\partial p^*}{\partial z}\boldsymbol{k} \\[2mm] \nabla \cdot \boldsymbol{u} = 0 \\[2mm] \dfrac{D\theta}{Dt} = \dfrac{L}{C_p \Pi}\xi \\[2mm] \dfrac{Dq_v}{Dt} = -\xi \\[2mm] \dfrac{Dq_c}{Dt} = \xi \end{cases} \qquad (6)$$

2   where the reference state is denoted by subscript $_0$ and the deviation from the reference state

3   by an asterisk, $\theta_v = \theta(1 + 0.61 q_v)$ is the virtual potential temperature. For stratocumulus and

4   cumulus fields ξ is estimated as:

5   $\xi = \min(q_{vs} - q_v, q_c)\Delta t$ \hfill (7)

6   where Δt is the simulation time step and $q_{vs}(T, p)$ is the saturation mixing ratio derived

7   from Thetens and Magnus formula: $q_{vs}(T, p) = \dfrac{0.622 P_{sat}}{(p/100 - 0.378 P_{sat})}$, where

8   $P_{sat} = 6.107 \exp\left[\dfrac{4028(T - 273.15)}{234.82(T - 38.33)}\right]$ for water, $P_{sat} = 6.107 \times 10^{\left[\frac{9.5(T - 273.15)}{265.5 + (T - 273.15)}\right]}$ for ice.

9   Computation of ξ at each simulation step is based on the work of Asai (1965). For cirrus

10   clouds, condensation, evaporation and ice crystals sedimentation processes are very complex

11   and still not well understood (Karcher and Spichtinger, 2009). In order to take into account

12   super-saturation and sub-saturation regions in cirrus clouds in a very simple way, we used the

13   parameterisation of Starr and Cox (1985) to compute the values of ξ every 2.5 min.

14   Sedimentation processes are taken into account in Eq. (5.5) by adding ice fall speed v_{fall}

15   taken from Starr and Cox (1985):

16   $v_{fall} = \dfrac{1.5}{6}\log 10\left[\max\left(\text{IWC}, 1 \times 10^{-6}\right)\right] + 1.5$ \hfill (8)

17   where v_{fall} is in m/s and the ice water content IWC in g/m^3.

18

1 ## 2.1.2 Assimilation of meteorological profiles and cloud coverage

2 In order to control the structure and nature of clouds and especially vertical position and

3 extension, it is necessary to impose a large-scale environment. Practically, forcing terms are

4 added to the 3DCLOUD equations to nudge the solutions towards observations. Our state

5 observations are the initial meteorological profiles (provided by the user for example) and do

6 not change during the simulation. The technique used is based on the initialization integration

7 method (Pielke, 2002). Consequently, 3DCLOUD equations become:

8

9
$$
\begin{cases}
\dfrac{Du}{Dt} = G_u(z)\left[u_{ini}(z) - \bar{u}(z)\right] \\[2mm]
\dfrac{Dv}{Dt} = G_v(z)\left[v_{ini}(z) - \bar{v}(z)\right] \\[2mm]
\dfrac{Dw}{Dt} = g\left(\dfrac{\theta_v^*}{\theta_{v0}} - q_c\right) - \dfrac{1}{\rho_0}\dfrac{\partial p^*}{\partial z} \\[2mm]
\nabla \cdot \boldsymbol{u} = 0 \\[2mm]
\dfrac{D\theta}{Dt} = \dfrac{L}{C_p \Pi}\xi + G_\theta(z)\left[\theta_{ini}(z) - \theta\right] \\[2mm]
\dfrac{Dq_v}{Dt} = -\xi + G_{q_v}(z)\left[q_{v_{ini}}(z) - q_v\right] \\[2mm]
\dfrac{Dq_c}{Dt} = \xi
\end{cases}
\tag{9}
$$

10 where for variables X, $\bar{X}(z)$ is the mean of X at height z and quantities $G_X(z)$ are

11 adjusted during the simulation in such a way that X or $\bar{X}(z)$ do not diverge far from the

12 initial conditions $X_{ini}(z)$. In a general way, G is the inverse of timescale but, because the

13 contribution of G is artificial, it must not be a dominant term in the governing equations and

14 should be scaled by the slowest physical adjustment processes in the model (Cheng et al.,

15 2001). This timescale was first set to 1h but this value was found to be too large and must be

16 adjusted as a function of altitude, especially at height where large vertical gradients of X

17 appear (e.g., at the top of a stratocumulus cloud, for example). Therefore, we developed a

18 very fast and simple numerical method to adjust the values of $G_X(z)$ during the simulation.

19 At each level, we compute the relative difference $\alpha_X(z) = \dfrac{X_{ini}(z) - \bar{X}(z)}{\bar{X}(z)} \cdot 100$. $G_X(z)$ is

1   assumed to be proportional to $\alpha_X(z)$ and is estimated as

2     $G_X(z) = \min\left[G_{\min} + \dfrac{G_{\max} - G_{\min}}{\alpha_{X,\max}}\alpha_X(z), G_{\max}\right]$ where $G_{\min} = \dfrac{1}{3600}\,\text{s}^{-1}$ and $G_{\max} = \dfrac{1}{2\Delta t}$.

3   Values of $\alpha_{X,\max}$ were estimated during our numerous numerical experiments. For

4   stratocumulus and cumulus, $\alpha_{X,\max}$ values for horizontal wind, temperature and humidity are

5   20%, 2% and 20% respectively and for cirrus, $\alpha_{X,\max}$ values are 20%, 10% and 10%,

6   respectively.

7   The cloud coverage C is defined as the fraction of the number of cloudy pixels to the total

8   number of pixels in the 2D horizontal plan. The value of C is chosen with the assimilation of

9   initial meteorological profiles. At each time step, the initial profile of vapour mixing ratio

10    $q_{v_{ini}}(z)$ is modified between cloud base and top height if $C \geq 50\%$ or between ground and

11   cloud top height if $C < 50\%$ until C agrees with the desired value within few percents. The

12   new "initial" profile of vapour mixing ratio $q_{v_{ini}}^{new}(z)$ is computed from the currently simulated

13   (old) profile of vapour mixing ratio $q_{v_{ini}}^{old}(z)$ as $q_{v_{ini}}^{new}(z) = q_{v_{ini}}^{old}(z) \pm \dfrac{n_z - n_{base}}{n_{top} - n_{base}}q_{v_{ini}}^{old}(z) \times \dfrac{0.1}{100}$

14   where z is height, and n_z, n_{top} and n_{base} are the levels indexes (in z direction) corresponding

15   to cloud top height and to cloud base height (or ground), respectively.

16   This method gives satisfactory results for stratocumulus and cumulus cloud fields (see Sect.

17   4.1.2), but not for cirrus fields. This is because condensation/evaporation and dynamic

18   processes are different for stratocumulus/cumulus and cirrus regimes. Indeed, for liquid and

19   warm stratocumulus/cumulus regime, liquid super or sub-saturation regions are not allowed in

20   3DCLOUD. Therefore, the distinction between cloudy and free cloud voxels is sharp.

21   Moreover, as stratocumulus/cumulus fields are often driven by convection processes in a

22   well-mixed planetary boundary layer, vertical correlation occurs between cloudy voxels (free

23   cloud voxels) and updrafts (downdrafts). Thus, the fractional cloud coverage is easily

24   controlled by adjusting the vertical profile of vapor mixing ratio during the simulation. By

25   contrast, in ice cirrus regimes, (large) ice crystals can survive even if ice relative humidity is

26   less than 100%. Ice super or sub-saturation regions are often observed in cirrus and are taken

27   into account in the Starr and Cox parameterization used in 3DCLOUD. Therefore, many

28   cloudy voxels still exist in our cirrus simulations, even if the ice water content is very small.

1   The distinction between cloudy and free cloud voxels is, thus, very tenuous. Moreover, cirrus

2   dynamics is often driven by wind shear: small fractional cloud coverage can exist at the top of

3   the cirrus field due to convection or radiative cooling coexisting with large fractional cloud

4   coverage and can also exist at the bottom of cirrus field due to wind shear. Finally, the total

5   cloud coverage could be large. If we adjust the vertical profile of vapor mixing ratio during

6   the simulation in the same way as for the stratocumulus/cumulus field, the total cloud

7   coverage will be difficult to control. Further investigations are thus needed to perfectly

8   control the cloud coverage of cirrus simulated by 3DCLOUD.

9

10 2.1.3 Implementation of 3DCLOUD algorithm

11   To implement the previously described equations, space is divided in

12    $(N_x + 2) \times (N_y + 2) \times (N_z + 2)$ cells or voxels where N_x, N_y and N_z are the voxel numbers in

13   each direction. A voxel is characterized by its spatial resolution with $\Delta x = \Delta y \neq \Delta z$.

14   Horizontal extensions are $L_x = L_y$ and can be different from the vertical extension L_z. In

15   order to take into account the boundary conditions, one layer of voxel is added around the

16   simulation domain.

17   A semi-Lagrangian scheme was chosen to solve the equation:

18    $$\frac{DX}{Dt} = \frac{\partial X}{\partial t} + \boldsymbol{u} \cdot \nabla X = 0 \qquad\qquad (10)$$

19   where X is a scalar advected by the wind velocity $\boldsymbol{u}$. X can be the potential temperature θ,

20   the condensed water q_c or the vapor mixing ratios q_v, and also the three components of wind

21   velocity u, v and w. Two steps are needed in order to compute the value of $X(x, t + \Delta t)$ at

22   a fixed position x and at time $t + \Delta t$. $X(x,t)$ and $\boldsymbol{u}(x,t)$ are known values and Δt is the

23   time step. First, we compute the previous position $p(X, t - \Delta t) = x - \boldsymbol{u}(x,t)\Delta t$ of X at time

24    $t - \Delta t$. In a second step, we compute the value of $X(p,t)$ at the position p and at the time t

25   by an interpolation scheme. This interpolated value $X(p,t)$ is the desired value $X(x, t + \Delta t)$.

26   The main advantage of this approach is that the time step is not restricted by the Courant-

27   Friedrichs-Lewy stability limit, but by the less restrictive condition that parcels do not

1   overtake each other during the time step (Riddaway, 2001). Therefore, at each iteration, the

2   maximum value of time step Δt_{max} can be roughly estimated as

3    $\Delta t_{max} = 1/\left(\max|\Delta u/\Delta x| + \max|\Delta v/\Delta y| + \max|\Delta w/\Delta z| \right)$. The accuracy of this approach depends

4   on the accuracy of the interpolation scheme. Due to CPU time, we chose a linear

5   interpolation, which, unfortunately, provides numerical dissipations. However, this drawback

6   is overcome using the Fourier transform performed in the second step of the 3DCLOUD

7   algorithm (see Section 2.2.2).

8   As the Fourier transform is easy to implement, this method was chosen to solve the equation

9    $\nabla \cdot \mathbf{u} = 0$. In the Fourier domain, the gradient operator ∇ is equivalent to the multiplication

10   by $i\mathbf{k}$, where $i \equiv \sqrt{-1}$ and $\mathbf{k}$ is the wave number vector. Thus, the following equation

11   $i\mathbf{k}.\hat{\mathbf{u}}(\mathbf{k}) = 0$, where $\hat{\mathbf{u}}$ is the transform of wind velocity $\mathbf{u}$ in the Fourier domain, has to be

12   solved. This implies that the Fourier transform of the velocity of a divergent free field is

13   always perpendicular to its wave numbers. Therefore, the quantity $1/k^2 \left(\mathbf{k}.\hat{\mathbf{u}}(\mathbf{k}) \right) \mathbf{k}$ is

14   removed from $\hat{\mathbf{u}}$. Keeping the real part of inverse transform of $\hat{\mathbf{u}}$ provides the new wind

15   velocity $\mathbf{u}$ with the desired free divergent property.

16   Lateral periodic conditions and continuity conditions to bottom and top are applied. For wind

17   velocity, free slip boundary conditions are applied at the bottom and top of the domain, which

18   are assumed to be a solid wall (i.e. $w = 0$). But, as the Fourier transform (which is needed to

19   solve the equation $\nabla \cdot \mathbf{u} = 0$) requires periodic conditions, it provides spurious oscillations

20   during the simulations. In order to limit this effect, extra levels with wind velocity set to zero

21   are added under and above the model domain.

22   The 3DCLOUD algorithm to simulate 3D structures of LWC or IWC is, in summary:

23   1) Definition of initial meteorological profiles u_{ini}, v_{ini}, θ_{ini}, $q_{v_{ini}}$ from idealized cloud

24   conceptual models or from the user. The vertical pressure profile is generally computed from

25   the hydrostatic law, but can be provided by the user.

26   2) Initial perturbations $\mathbf{u}'$ are added to wind velocity $\mathbf{u}$. $\mathbf{u}'$ is free-divergent and turbulent

27   with a spectral slope of -5/3 (see more explanations in Sect. 2.2.2) .

28   3) Assimilation of initial meteorological profiles (optional, see Eq. 9 and Sect. 2.1.2).

29   4) Constrain of divergent-free velocity $\mathbf{u}$ (see Eq. 9).

11

1   5) Computation of ice fall speed v_{fall} (only for cirrus cloud, see Eq. 8).

2   6) Advection of u, v, w, θ, q_v and q_c by wind velocity $\boldsymbol{u}$ (see Eq. 10).

3   7) Modification of θ, q_v and q_c due to evaporation/condensation processes (see Eq. 7).

4   8) Modification of the vertical velocity due to buoyancy (see Eq. 9).

5   9) Modification of $q_{v_{ini}}$ in order to assimilate cloud coverage C (optional, only for

6   stratocumulus and cumulus, see Sect. 2.1.2).

7   10) Return to 3) until maximum iteration number is reached.

8   11) Computation of LWC or IWC.

9

2.2   Step 2 : statistical adjustment

11   Hereafter, we present the second step of the 3DCLOUD algorithm that is the methodology to

12   adjust, according to user requirements, the mean optical depth $\overline{\tau}$ (or the mean ice water

13   content $\overline{\text{IWC}}$) and the inhomogeneity parameter of the optical depth ρ_τ (or the ice water

14   content ρ_{IWC}) from the LWC (or from the IWC) simulated at the step 1. The distribution of τ

15   or IWC is assumed to follow a gamma distribution. Indeed, distribution of τ and IWC are

16   usually well represented by a lognormal or gamma distribution (Cahalan et al., 1994; Barker

17   et al., 1996; Carlin et al., 2002; Hogan and Illingworth, 2003; Hogan and Kew, 2005). The

18   scale invariant cloud properties, controlled at each level, are characterised by the spectral

19   exponent β_{1D} close to -5/3 (slope β of the one dimension wave number spectrum in log–log

20   axes of the Fourier space). This spectral slope is computed from the outer scale L_{out} (defined

21   by the user) to the smaller scale (voxel horizontal size).

22

2.2.1   Control of the mean and of the inhomogeneity parameter

24   The relation between *local* optical depth $\tau(x,y,z)$, liquid water path LWP and density of

25   water ρ_l in each voxel is given by (Liou, 2002):

1     $$\tau(x,y,z) = \frac{3}{2\rho_l}\frac{\mathrm{LWP}(x,y,z)}{R_{eff}} \text{ with } \mathrm{LWP}(x,y,z) = \rho_{air}q_c(x,y,z)\Delta z \qquad (11)$$

2   where x, y, z are the spatial positions inside the simulation domain, $\mathrm{LWP}(x,y,z)$ is the

3   *local* liquid water path and Δz is the vertical resolution. *Local* quantity means that the

4   quantity is estimated inside a voxel. The optical depth $\tau(x,y)$ for each pixel is the sum of

5   *local* optical depths along the z-axis:

6     $$\tau(x,y) = \sum_{z=1}^{N_z}(x,y,z). \qquad (12)$$

7   The mean optical depth $\overline{\tau}$ is then defined as:

8     $$\overline{\tau} = \frac{1}{N_x N_y}\sum_{x=1}^{N_x}\sum_{y=1}^{N_y}\tau(x,y). \qquad (13)$$

9   In the same way, for ice cloud, the mean IWC is obtained with

10     $$\overline{\mathrm{IWC}} = \frac{1}{N_x N_y N_z^*}\sum_{x=1}^{N_x}\sum_{y=1}^{N_y}\sum_{z=1}^{N_z^*}\mathrm{IWC}(x,y,z) \qquad (14)$$

11   where N_z^* is the number of layers between cloud top and cloud bottom.

12   To describe the amplitude of the optical depth for 1D and 2D overcast cloud, Szczap et al.

13   (2000) defined the inhomogeneity parameter of optical depth ρ_τ. For 3D broken fields, this

14   parameter is defined according to

15     $$\rho_\tau = \frac{\sigma\left[\tau^{>0}(x,y)\right]}{\overline{\tau}^{>0}(x,y)} \qquad (15)$$

16   where $\sigma\left[\tau^{>0}(x,y)\right]$ and $\overline{\tau}^{>0}(x,y)$ are the standard deviation and the mean of the strictly

17   positive optical depth.

18   Due to the flexibility of the mathematical formulation of the gamma distribution and to its

19   ability to mimic the attributes of other positive-value distributions, such as lognormal and

20   exponential distributions, we choose to control $\overline{\tau}$ or $\overline{\mathrm{IWC}}$ and ρ_τ or ρ_{IWC} by mapping

21   theoretical gamma-distributed properties onto the simulated properties. This mapping

22   technique is analog to the "amplitude adaptation" technique explained in Venema et al.

1 (2006), where amplitudes are adjusted based on their ranking. The gamma distribution is a

2 two-parameter family of continuous probability distribution. It has a shape parameter a and

3 scale parameter b. The equation defining the probability density as a function of a gamma-

4 distributed random variable Y is:

5 $$Y = f\left(\mu; a, b\right) = \frac{1}{b^a \Gamma\left(a\right)} \mu^{a-1} e^{-\mu/b} \qquad (16)$$

6 where $\Gamma\left(.\right)$ is the gamma function. We develop a simple iterative algorithm, where values of

7 a and b are adjusted until mean and inhomogeneity parameters reach the required user

8 values within few percents.

9

2.2.2 Control of invariant scale properties by adjustment of spectral exponent
in Fourier space

12 The spectral slope value β_{1D} of the horizontally 2D field is adjusted according to the

13 following methodology. As proposed by Hogan and Kew (2005), we choose to manipulate 2D

14 plan of Fourier amplitudes of *local* optical depth τ_{2D} (or IWC_{2D}) with a 2D Fourier

15 transform performed at each height of cloudy layer. Suppose a 2D isotropic field $g\left(x, y\right)$

16 characterized by a Gaussian probability density function (PDF) and a 1D power spectrum

17 $E_1\left(k\right)$ with a spectral slope β_{1D} at all scales defined as:

18 $$E_1\left(k\right) = \widehat{E_1} k^{-\beta_{1D}} \qquad (17)$$

19 where k is the wave number in any direction and $\widehat{E_1}$ is the spectral energy density at $k = 1$

20 m^{-1}. Following Hogan and Kew (2005), for the idealized case where $g\left(x, y\right)$ is continuous at

21 small scales and infinite in extent, its 2D spectral density matrix $E_2\left(k_x, k_y\right)$ can be written as

22 $$E_2\left(k\right) = \kappa \widehat{E_1} k^{-\beta_{1D}-1} \qquad (18)$$

23 where $k = \sqrt{k_x^2 + k_y^2}$ and κ a constant. In general, a 2D cloud layer of τ_{2D} (or IWC_{2D}) is

24 anisotropic and, in our case, the optical depth (or IWC) is gamma-distributed. Therefore, the

1   1D power spectrum $E_1(k)$ seldom has the required spectral slope β_{1D}. In this context, a

2   numerical method has to be developed to perform our objectives.

3   Let set $\widehat{Y_{2D}}$ be the 2D Fourier transform of Y_{2D}, where Y_{2D} can be τ_{2D} (or IWC_{2D}) at a given

4   cloudy layer. This quantity, estimated with the help of a direct 2D Fast Fourier Transform

5   algorithm can be written as:

6    $$\widehat{Y_{2D}}(k) = E_{2D}(k)\exp\left(i\varphi_{2D}(k)\right) \tag{19}$$

7   where $E_{2D} = \left|\widehat{Y_{2D}}\right|$ is the magnitude or spectral energy, $\varphi_{2D}(k)$ are the phase angles and

8    $k = \sqrt{k_x^2 + k_y^2}$ is the absolute wave number. The cloud field domain is defined to measure L_x

9   and L_y and they have spatial resolutions of Δx and Δy. The resulting wave number for E_{2D}

10   ranges from $-K_x$ to $+K_x$ with a resolution of $\Delta k_x = 1/L_x$, where $K_x = 1/2\Delta x$. It is similar for

11   k_y direction.

12   Our objective is to modify spectral energy $E_{2D}(k)$ in such a way that the 1D spectral slope

13   value μ_{1D} estimated in one dimension from Y_{2D} for $k \geq k_{out}$ ($k_{out} = 1/L_{out}$) satisfies the

14   desired value β_{1D} required by the user. Practically, μ_{1D} is defined as:

15   $$\mu_{1D} = (\beta_x + \beta_y)/2 \tag{20}$$

16   where β_x and β_y are the 1D spectral slope values of $\widehat{Y_{2D}}$ estimated in the x and y directions

17   respectively.

18   In order to conserve the spatial repartition of Y_{2D}, we keep $\varphi_{2D}(k)$ phase angles unchanged

19   for all values of k. We also keep unchanged $E_{2D}(k)$ for $k < k_{out}$. For $k \geq k_{out}$, two 2D matrix

20   $E_{2D}^*(k)$ and $E_{2D}^{**}(k)$ can be computed. $E_{2D}^*(k)$ is based on Eq. (18) and is defined as:

21   $$E_{2D}^*(k) = \frac{k^{(-\beta_{1D}-1)}}{k_{out}^{(-\beta_{1D}-1)}} \overline{E_{2D}}(k = k_{out}) \tag{21}$$

22   whereas $E_{2D}^{**}(k)$ is defined as:

1 $$E_{2D}^{**}(k) = \frac{E_{2D}(k)}{\overline{E_{2D}}(k)} \frac{k^{(-\beta_{1D}-1)}}{k_{out}^{(-\beta_{1D}-1)}} \overline{E_{2D}}(k = k_{out}) \qquad (22)$$

2 where $\overline{X}$ is the mean of variable X. If the degree of anisotropy of Y_{2D} is small, such as for

3 stratocumulus and cumulus, we use $E_{2D}^{*}(k)$ and if not, such as for cirrus clouds, $E_{2D}^{**}(k)$.

4 Nonetheless, the user can also choose one of these methods.

5 Finally, the new 2D *local* optical depth (or the 2D new ice water content Y_{2D}^{new}) at the given

6 cloud layer is computed by keeping the real part of the inverse 2D Fast Fourier Transform of

7 the new quantity:

8 $$\widehat{Y_{2D}^{new}}(k) = E_{2D}^{*}(k)\exp(i\varphi_{2D}(k)) \text{ or } \widehat{Y_{2D}^{new}}(k) = E_{2D}^{**}(k)\exp(i\varphi_{2D}(k)) \qquad (23)$$

9 But as a result, the distribution of Y_{2D}^{new} is not the same Y_{2D} at the given cloud layer, and the

10 equality between the estimated spectral slope μ_{1D} of Y_{2D}^{new} and the required values β_{1D} is not

11 always guaranteed. Therefore, we have to redo an "amplitude adaptation", as explained in

12 Venema et al. (2006), and to iterate the process explained in this section by changing the

13 value of β_{1D} in Eq. (21) or Eq. (22), until the estimated value μ_{1D} reaches the required value

14 within a few percent.

15

16 **2.2.3 Implementation**

17 We describe here the part of the 3DCLOUD algorithm that establishes the cloud field mean

18 optical depth $\overline{\tau}$ ($\overline{\text{IWC}}$), the inhomogeneity parameter ρ_{τ} (or ρ_{IWC}) and the spectral exponent

19 β_{1D}.

20 For stratocumulus and cumulus clouds, the algorithm is the following:

21 1) Transformation of $\text{LWC}(x, y, z)$ to $\tau'_{3D}(x, y, z)$ with Eq. (10). Effective radius can be set

22 to 10 µm for example.

23 2) Application of the algorithm explained in Sect. 2.2.2 in order to constrain β_{1D} of each

24 cloudy layer of $\tau'_{3D}(x, y, z)$. We obtain $\tau''_{3D}(x, y, z)$.

3) Computation of optical depth $\tau'(x, y)$ from $\tau''_{3D}(x, y, z)$ (Eq. (11)).

4) Transformation of $\tau'(x, y)$ to $\tau''(x, y)$ with the help of the algorithm explained in Sect. 2.2.1 in order to constrain $\bar{\tau}$ and ρ_{τ} values.

5) Transformation of $\tau''(x, y)$ to $\tau'''(x, y)$ with the algorithm explained in Sect 2.2.2 in order to control β_{1D} value.

6) Normalization of $\tau'''(x, y)$ to the required $\bar{\tau}$ value, in order to obtain $\tau_{3D}(x, y, z)$.

For cirrus clouds, the algorithm is as follows:

1) Transformation of $\mathrm{IWC}(x, y, z)$ to $\mathrm{IWC}'(x, y, z)$ with the algorithm explained in Sect. 2.2.1 in order to constrain $\overline{\mathrm{IWC}}$ and ρ_{IWC} values.

2) Application of the algorithm explained in Sect. 2.2.2 in order to constrain β_{1D} of each cloud layer of $\mathrm{IWC}'(x, y, z)$. We obtain $\mathrm{IWC}''(x, y, z)$.

3) Transformation of $\mathrm{IWC}'''(x, y, z)$ to $\mathrm{IWC}_{3D}(x, y, z)$ with the algorithm explained in Sect. 2.2.1 in order to constrain $\overline{\mathrm{IWC}}$ and ρ_{IWC} values.

2.3 Differences between 3DCLOUD, IAFFT method and Cloudgen models

Both IAAFT (Venema et al., 2006) and Cloudgen (Hogan and Kew, 2005) models are purely stochastic Fourier based approaches that are able to generate synthetic or surrogate cloud. On the contrary, 3DCLOUD solves, in a first step, basic atmospheric equations, in order to generate an intermediate cloud field. In its second step, as for both IAFFT and Cloudgen models, it uses Fourier tools (manipulation of energy and phase in frequency space) and amplitude adaptation (manipulation of distributions) in order to generate the final cloud field. IAAFT and Cloudgen are designed to simulate stratocumulus/cumulus fields for the first and cirrus fields for the second, when 3DCLOUD is able to simulate stratocumulus, cumulus and cirrus field within the same framework.

More specifically, the IAAFT method is designed to generate surrogate clouds having both the amplitude distribution and power of the original cloud (2D LWC from 1D LWP measurement, 3D LWC from 2D LWC fields or 3D LWC from 3D fields generated by LES). It needs LES inputs or measurements. As explained in Venema et al. (2006), stratocumulus often display beautiful cell structures, similar to Bénard convection, and LES clouds show such features. But their 3D IAFFT surrogates show these much less and do not show fallstreak or a filamentous structure. Due to the specific manipulations of Fourier coefficients presented in the paper, we show that 3DCLOUD is able to simulate the cell structure of stratocumulus (see Fig. 7c and Fig. 8), the filamentous structure of cirrus (see Fig. 13) and the cirrus fallstreaks (see Fig. 14 and Fig. 15) relatively well. Moreover, the objective of 3DCLOUD is not to provide many surrogate clouds with the same amplitude distribution and power spectrum from an LES original cloud, but to provide 3D LWC (or optical depth) with the required cloud coverage, the -5/3 spectral slope (often observed in real clouds), the mean value of the gamma distribution of the optical depth and the inhomogeneity parameter, all these parameters being very pertinent for radiative transfer.

Cloudgen is designed to simulate surrogate cirrus with the cirrus specific structural properties: fallstreak geometry and shear-induced mixing. It first generates a 3D fractal field by performing an inverse 3D Fourier transform on a matrix of simulated Fourier coefficients with amplitude consistent with observed 1D spectra. Then random phases are generated for the coefficient allowing multiple cloud realizations with the same statistical properties. Horizontal slices from the domain are manipulated in turn to simulate horizontal displacement and to change the spectra with height. The final field is scaled to produce the observed mean and fractional standard deviation of ice water content. 3DCLOUD does not use a 3D fractal field, but a 3D IWC field simulated by the simplified atmospheric equation set. Therefore, cloud structures due to wind shear are physically obtained by taking into account the advection (a nonlinear term in momentum equation) rather than by a linear horizontal displacement of phase. Afterwards, in the current version of 3DCLOUD, for each level, 2D horizontal slices of this 3D IWC are manipulated in 2D Fourier domain in such a way that the Fourier coefficient amplitude is consistent with the 2D spectra of the simulated IWC, with the constraint that the 1D spectral slope is equal to -5/3 (this value can be change easily in future version of 3DCLOUD). At each level, the 2D phase for the coefficient is kept unchanged. Finally, the mean value of the 3D IWC and the inhomogeneity parameter are adjusted. As explained by Hogan and Kew (2005), it is difficult with Cloudgen to generated anisotropic

1 cirrus structure such as roll-like structure near cloud top. 3DCLOUD, using physically based

2 equations, allows simulating such kinds of anisotropy, as for example, 3DCLOUD Kelvin-

3 Helmholtz wave breaking (see Fig. 14).

4

3 Comparison between 3DCLOUD and large-eddy simulation (LES) outputs

6 The objective of this section is to check that the basic atmospheric equations used in

7 3DCLOUD (see Sect. 2.1.1) are solved correctly. We compare 3DCLOUD and LES outputs

8 found in the scientific literature for marine stratocumulus, cumulus and cirrus regimes. Note

9 that assimilation techniques of meteorological profiles and of cloud coverage (see Sect. 2.1.2)

10 are not used here, except in Sect 3.4.

11 The test cases come from the output LES numerical files provided by the Global Water and

12 Energy Experiment (GEWEX) Cloud System Studies (GCSS) Working Group 1 (WG1) and

13 Working Group 2 (WG2), easily downloadable from the web. They are often used as a

14 benchmark. We choose the DYCOMS2-RF01 case (the first Research Flight of the second

15 Dynamics and Chemistry of Marine Stratocumulus) for the marine stratocumulus regime

16 (Stevens et al., 2005), the BOMEX case (Barbados Oceanographic and Meteorological

17 Experiment) for the shallow cumulus regime (Siebesma et al., 2003), and the ICMCP case

18 (Idealized Cirrus Model Comparison Project) for cirrus regimes (Starr, 2000).

19

3.1 DYCOMS2-RF01 (GCSS-WG1) case

21 We remember briefly the conditions of simulations and configurations, explained in detail in

22 Stevens et al. (2005). A 4-h simulation on a horizontal grid of 96 by 96 points with 35-m

23 spacing between grid nodes was required. Vertical spacing was required to be 5 m or less. In

24 3DCLOUD, we thus set $N_x = N_y = 96$ and $N_z = 240$, $L_x = L_y = 3.5$ km and $L_z = 1200$ m, so

25 that $\Delta x = \Delta y \approx 36.5$ m and $\Delta z = 5$ m. Initial profiles of the liquid water potential temperature

26 θ_l and of the total water mixing ratio q_t are $\theta_l = 289.0$ K and $q_t = 9.0$ g/kg if $z \leq z_i$ and

27 $\theta_l = 297.5 + (z - z_i)^{1/3}$ K and $q_t = 1.5$ g/kg if $z > z_i$. Other required forcings include

28 geostrophic winds ($U_g = 7$ m/s and $V_g = -5.5$ m/s), divergence of the large-scale winds (

29 $D = 3.75 \times 10^{-6}$ s^{-1}), surface sensible heat flux (15 W/m^2) and surface latent heat flux (115

1   W/m^2). The momentum surface fluxes where the total momentum is specified by setting

2    $u^* = 0.25$ m/s and the radiation schemes is based on a simple model of the net longwave

3   radiative flux (see Eq. (3) and Eq. (4) in Stevens et al., 2005).

4   Figure 2 shows the evolution of the mean cloud top height, the mean cloud base height, the

5   cloud coverage and the liquid water path from the master ensemble and for 3DCLOUD during

6   the 4-h simulations. Even though we can notice slight discrepancies between 3DCLOUD and

7   master ensemble results in the first 2 hours ("spinup" period), 3DCLOUD results are quite

8   consistent with master ensemble results, especially at the end of the simulation. Nevertheless,

9   3DCLOUD tends to generate a lower cloud height than the mean results with a higher liquid

10   water path.

11   Figure 3 shows the mean profiles averaged over the fourth hour of the longwave net flux, the

12   liquid water potential temperature, the total water mixing ratio, the liquid water mixing ratio,

13   the horizontal velocity components, and the air density. Even though the 3DCLOUD

14   longwave net flux is smaller compared to master ensemble, again 3DCLOUD results are quite

15   consistent with other results for all the variables.

16

3.2 BOMEX (GCSS-WG1) case

18   For the BOMEX case (Siebesma et al., 2003), a 6-h simulation on a horizontal grid of 64 by

19   64 points with 100-m spacing between grid node was required. Vertical spacing was required

20   to be 40 m. In 3DCLOUD, we set thus $N_x = N_y = 64$ and $N_z = 76$ with $L_x = L_y = 6.4$ km and

21   $L_z = 2980$ m, so $\Delta x = \Delta y = 100$ m and $\Delta z \approx 39.2$ m. Initial profiles of the liquid water potential

22   temperature θ_l and the total water mixing ratio q_t and the other requirement including

23   geostrophic winds, divergence due to the subsidence, surface sensible heat flux, surface latent

24   heat flux, momentum surface fluxes, moisture large scale horizontal advection term and

25   longwave radiative cooling (radiative effects due to the presence of clouds are neglected) are

26   presented in appendix B in Siebesma et al., 2003).

27   Figure 4 shows the evolution of the cloud coverage and the liquid water path from the master

28   ensemble and for 3DCLOUD during the 6-h simulation. We can notice the small value of the

29   cloud coverage (less than 10%). 3DCLOUD results are quite consistent with the master

ensemble results, even if the simulated 3DCLOUD liquid water path (LWP) may be too low at the end of the simulation.

Figure 5 shows mean profiles, averaged over the fifth hour of the cloud coverage, potential temperature, water vapour mixing ratio, liquid water mixing ratio, horizontal velocity components, and air density. The 3DCLOUD results are again quite consistent with the master ensemble results. We note, however, that the 3DCLOUD cloud coverage and liquid water mixing ratio are smaller at all heights. We also see small differences (less than 1 m/s) for the wind velocity below 500 m, and for the potential temperature (less than 1K) and water vapour mixing ratio (less than 1g/kg) for altitudes 1,800 m.

3.3   ICMCP (GCSS-WG2) case

For the cirrus case detailed in Starr et al. (2000), the baseline simulations include night-time "warm" cirrus and "cold" cirrus cases where cloud top initially occurs at -47°C and -66°C, respectively. The cloud is generated in an ice super-saturated layer with a geometric thickness around 1 km (120% in 0.5 km layer) and with a neutral ice pseudo-adiabatic thermal stratification. Cloud formation is forced via an imposed diabatic cooling over a 4-h time span followed by a 2-h dissipation stage without cooling. All models simulate radiative transfer, contrary to 3DCLOUD. In 3DCLOUD, we set $N_x = N_y = 60$ and $N_z = 140$ with $L_x = L_y = 6.3\,\mathrm{km}$ and $L_z = 14\,\mathrm{km}$, so that $\Delta x = \Delta y = 105\,\mathrm{m}$ and $\Delta z \approx 100\,\mathrm{m}$.

Figure 6 shows the evolution of the ice water path (IWP) from the master ensemble and for 3DCLOUD during the 6-h simulation. In a general way, most of the tested models and 3DCLOUD have similar behaviour: indeed, the IWP increases during the first 4-h simulation (cirrus formation due to imposed cooling) and decreases after (cirrus dissipation due to non-imposed cooling). The IWP range of the tested models is very large (factor of 10), but we can notice that 3DCLOUD behaviour is closer to bulk microphysics models behaviours, especially for "warm cirrus".

For "cold" cirrus, the 3DCLOUD IWP is smaller than most participating models during all the simulation duration. It is probably because 3DCLOUD does not account for the radiative transfer, as opposed to the participating models. Indeed, neglecting cirrus top cooling due to

1   radiative processes restricts the formation of thin "cold" cirrus. This radiative diabatic effect is

2   probably less important for the "warm" cirrus because the latent heat diabatic effect is larger.

3

4   3.4   Comparison between 3DCLOUD and BRAMS for the DYCOM2-RF01 case

5   In order to underscore differences between 3DCLOUD and LES for comparable scenes, we

6   choose again the well documented DYCOMS2-RF01 case. Snapshots can be found, for

7   example, in Stevens et al. (2005) and in Yamaguchi and Feingold (2012). We performed the

8   4h simulations of the DYCOMS2-RF01 case with 3DCLOUD and with the Brazilian

9   Regional Atmospheric Modelling System (BRAMS v4) model (Pielke at al., 1992 ; Cotton et

10 al., 2003). BRAMS simulations were provided by G. Penide (Penide et al., 2010). The

11 BRAMS model is constructed around the full set of nonhydrostatic, compressible equations.

12 The cloud microphysics parameterization is based on a two-moment scheme (Meyers et al.,

13 1997). Subgrid scale fluxes are modeled following Deardroff (1980). The base calculations

14 are performed on a $100 \times 100 \times 100$ point mesh with a step time of 0.3 s.

15 Figure 7 shows the instantaneous cloud-field snapshots of the pseudo albedo (see definition in

16 Sect. 4) at four hours simulated by (a) the UCLA-0 model (picture taken from Stevens et al.,

17 2005), (b) the BRAMS model, both configured following the DYCOMS2-RF01 case (Stevens

18 et al., 2005) and (c) from 3DCLOUD with assimilation of meteorological profiles based on

19 the DYCOMS-RF01 case. Both BRAMS and 3DCLOUD cases are drawn from simulations

20 where $\Delta_x = \Delta_y = 40$ m and $\Delta_z = 12$ m. These three snapshots of cloud fields are characterized

21 by closed cellular convection with large cloud cover, as argued in Yamaguchi and Feingold

22 (2012), who did simulation of DYCOMS-RF01 case with the LES mode of the Advanced

23 Research WRF model. Figure 7 also shows the power spectra computed following the x and y

24 directions and then averaged, for BRAMS and 3DCLOUD optical depth fields. The

25 3DCLOUD optical depth spectral slope is close to -5/3 in the $[L_{out} : 1/(2\Delta x)]$ m^{-1}

26 wavenumber range, as expected, because of the statistical adjustment performed in the second

27 step of the 3DCLOUD algorithm. By contrast, the BRAMS optical depth spectral slope is

28 close to -5/3 only in the $[2 \times 10^{-3} : 5 \times 10^{-3} \approx 1/(5\Delta x)]$ m^{-1} wavenumber range. Depending

29 on their degree of sophistication, LES do not always guarantee cloud invariant scale

30 properties at the larger wavenumbers. Indeed, Bryan et al. (2003) have shown, that for the

31 finite-difference model, the vertical wind velocity spectral slope is steeper than -5/3 for scales

1   shorter than $6\Delta x$. Table 1 shows the computation performance of 3DCLOUD and BRAMS.

2   For this specific case, 3DCLOUD simulation is thirty times faster than BRAMS simulation.

3   **4   Examples of 3DCLOUD possibilities**

4

5   In this section, we present cloud fields generated by 3DCLOUD with the assimilation of

6   idealized meteorological profiles and fractional cloud coverage defined by the user. We also

7   show the effect of the outer scale L_{out} and the inhomogeneity parameter of optical depth ρ_τ

8   on the generated optical depth field. We also give an example of cirrus clouds with

9   fallstreaks. In order to have a spatial representation of the clouds as seen from above, we

10   choose to show the so-called pseudo-albedo α defined as:

11    $$\alpha = \frac{(1-g)\tau}{2+(1-g)\tau}$$ (23)

12   where the asymmetry parameter g is set to 0.86 and τ is the optical depth.

13   **4.1   Stratocumulus and cumulus fields with assimilation of meteorological**
14   **profiles based on DYCOMS2-RF01 and BOMEX cases.**

15   We choose to simulate stratocumulus and cumulus LWC in the context of DYCOMS2-RF01

16   and BOMEX cases. With this aim, we assimilate temperature and humidity initial profiles for

17   stratocumulus and cumulus given by Stevens et al. (2005) and Siebesma et al. (2003),

18   respectively. However, in order to mimic the sensible and latent heat, these profiles have to be

19   slightly modified. At sea surface ($z = 0$), for stratocumulus (DYCOMS2-RF01 case), the

20   liquid potential temperature and total mixing ratio are set to 290 K and 10 g/kg, respectively

21   (instead of 289 K and 9 g/kg, respectively). For cumulus (BOMEX case), the liquid potential

22   temperature is set to 299.7 K instead of 298.7 K. In addition, wind profiles assimilated by

23   3DCLOUD are those computed by the master ensemble at the end of simulation.

24   **4.1.1 Effects of numerical spatial resolution**

25   The effects of the numerical spatial resolution on 3DCLOUD simulations are presented

26   herein. Figure 8 shows pseudo-albedo and cross sections of the vertical velocity and cloud

27   water at the end of the simulation for the stratocumulus case based on the DYCOMS2-RF01

28   experiment. It also shows the mean profiles of potential temperature, liquid water mixing

1   ratio, horizontal velocity components, and vapour water mixing ratio for different numerical

2   spatial resolutions $\Delta x = \Delta y = 200$ m, 100 m, 50 m and 25 m. Horizontal extensions $L_x = L_y$ are

3   set to 10 km and vertical resolution Δz to 24 m for all simulations. Figure 9 is the same as

4   Fig. 8, but for the cumulus case with assimilation of meteorological profiles based on the

5   BOMEX case. The vertical resolution Δz is set to 38.5 m in this last case.

6   It is obvious that change in the horizontal mesh leads to a more pleasant and detailed flow

7   visualization but there is no significant impact on the mean statistics of the simulated

8   temperature vertical profile, water vapour mixing ratio and wind velocity. The water mixing

9   ratio simulated by 3DCLOUD for the DYCOMS2-RF01 case is very close to the mean profile

10   averaged over the fourth hour and provided by the master ensemble, even if the vertical

11   resolution used in this section is only $\Delta z = 38.5$ m compared to $\Delta z = 5$ m in Sect. 3.1. For the

12   BOMEX case, the water mixing ratio simulated by 3DCLOUD changes as a function of the

13   numerical spatial resolution. This behaviour is quite understandable as results drawn on Fig. 8

14   and Fig. 9 are snapshots at the end of the 3DCLOUD simulation and not average results over

15   1 hour as done on Fig. 3 and Fig. 5. Moreover, BOMEX meteorological conditions cause time

16   dependent cumulus fields, contrary to DYCOMS2 meteorological conditions that cause more

17   stationary stratocumulus fields.

18   In addition, it is expected for the BOMEX case, that cloud spacing converges at high spatial

19   resolution. In order to investigate it, we defined an estimator of the cloud spacing called the

20   mean distance D_{mean}. To compute D_{mean} , the 3D LWC is vertically projected on the 2D x-

21   y plan in order to obtain the 2D binary image of the cloud coverage with free cloud areas set

22   to 0 and cloudy areas set to 1. Then we compute the mean distance between the cloud cell for

23   the x and y directions to obtain D_{mean}. Figure 10 shows time series of D_{mean} for different

24   horizontal spatial resolution ($\Delta x = \Delta y = $ 192, 100, 50, 25, 12.5 and 8.3 m) with a constant

25   vertical resolution ($\Delta z = 38.5$ m), for cumulus cloud fields simulated by 3DCLOUD after

26   assimilation of the BOMEX case meteorological profiles. The main difference between these

27   simulations and the BOMEX case simulation is the smaller horizontal extension $L_x = L_y = 5$

28   km instead of 10 km in order to access high numerical spatial resolution $\Delta x = 8.3$ m ($N_x = $

29    $N_y = 600$, $N_Z = 70$). Cumulus clouds appear 10 to 20 min after the beginning of the

30   simulation. After 1 h of simulation, D_{mean} is relatively constant with time, meaning that

31   3DCLOUD has converged. The mean distance averaged over the last half-hour of the 2h

32   simulation $\overline{D_{mean}}$, is also presented in Fig. 10 as a function of the numerical spatial

1   resolution Δx. $\overline{D_{mean}}$ is relatively constant for a spatial resolution Δx smaller than 20 m,

2   showing that BOMEX cloud spacing converges for spatial resolution close to $\Delta x = 10$ m, a

3   value smaller than $\Delta x = 25$ m used in Fig. 9d.

4   Table 1 shows the time step, process time for one time step and process time for 2h-

5   simulation with 3DCLOUD model, as a function of the numerical resolution. DYCOMS2-

6   RF01 and BOMEX cases are presented. The process time for 2h-simulation is indicated

7   because 3DCLOUD algorithm convergence is achieved after 2 h (or less) of simulation for

8   stratocumulus, cumulus and cirrus regimes (see Fig. 10 for cumulus case). For both cases, the

9   smaller the spatial resolution, the smaller the step time and the larger the process time. A

10   comparison between 3DCLOUD and BRAMS LES computation time for a specific

11   DYCOMS2-RF01 case is added (see Sect. 3.4). For this specific case, 3DCLOUD simulation

12   is thirty times faster than BRAMS simulation. Note that 3DCLOUD (Matlab code) runs on a

13   personal computer with Intel Xeon E5520 (2.26 GHz) and BRAMS (Fortran code) runs on a

14   PowerEdge R720 with Intel Xeon E5-2670 (2.60 GHz), both of them having a single-

15   processor configuration.

16

4.1.2 Assimilation of the fractional cloud coverage C

18   Results shown in Fig. 11 are the same as Fig. 8 but with the addition of the cloud coverage

19   assimilation $C = 99\%$, $C = 80\%$, $C = 50\%$ and $C = 20\%$. Horizontal extensions L_x and L_y

20   are set to 10 km and vertical resolution Δz is set to 24 m.

21   They show that 3DCLOUD is able to assimilate correctly fractional cloud coverage of

22   stratocumulus for very different values of C, even though the extreme example with

23   $C = 20\%$ is a fair weather cumulus field rather than a stratocumulus. For each value of C

24   assimilated, it is interesting to note that cloud base and cloud top heights are still localised

25   around 600 m and 800 m, respectively. Temperature vertical profiles are almost unchanged.

26   The water mixing ratio vertical profiles decrease with the assimilated C value.

4.1.3 Effect of the outer scale L_{out} and inhomogeneity parameter ρ_τ on the

28   **optical depth field**

29   We saw that 3DCLOUD can, at the end of step 1, simulate stratocumulus and cumulus fields

30   with enough coherent statistics profiles. However, optical depth (for stratocumulus and

1　cumulus) or IWC (for cirrus) generated during step 1 of 3DCLOUD does not show scale

2　invariant properties observed in real cloud and often characterised by the spectral exponent

3　β_{1D} close to -5/3. As described in Sect. 2.2, it is the main task of the step 2 of 3DCLOUD, in

4　addition to the adjustment of the mean and standard deviation of optical thickness or IWC.

5　We focus on the DYCOMS2-RF01 case at the spatial resolution $\Delta x = 50$ m (Fig. 8c). The

6　effective radius R_{eff} is set to 10µm to compute optical depth from liquid water content. The

7　mean optical depth of this initial field is set to 10 and we change the inhomogeneity parameter

8　ρ_τ and the outer scale L_{out} to 0.2 and 1 km, respectively, for case 1 to 0.7 and 1 km for case 2

9　and to 0.7 and 10 km for case 3. Figure 12 shows pseudo-albedo, mean power spectra,

10　probability density function of optical depth fields, mean vertical profiles of the horizontally

11　averaged optical depth for the three cases and a volume rendering of optical depth.

12　First, we notice that the pseudo-albedo of the initial optical depth field (see Fig. 12a) is

13　smoother than the pseudo-albedo of case 1, 2, and 3. Between cases 1 and 2, we clearly see an

14　increase in heterogeneity as case 1 is a quasi-homogenous stratocumulus with a small value of

15　$\rho_\tau = 0.2$ and case 2 is more inhomogeneous with a larger value of $\rho_\tau = 0.7$. Between cases 2

16　and 3, we can see the effect of the outer scale. In accordance with smooth variations, the

17　spectral slope of the initial optical depth is close to -3 for the $\left[10^{-3} : 10^{-2} \right]$ m^{-1} wavenumber

18　range (Fig. 12e). Cases 1, 2 and 3 present the proper spectral slope value of -5/3. For cases 1

19　and 2, this slope is obtained for the $\left[10^{-3} : 10^{-2} \right]$ m^{-1} wavenumber range, which is coherent

20　with the imposed value of outer scale $L_{out} = 1$ km. For case 3, $L_{out} = 10$ km, so the spectral

21　slope should be -5/3 on the $\left[10^{-4} : 10^{-2} \right]$ m^{-1} wavenumber range. However, we note that this

22　spectral slope value is achieved only for the $\left[5 \times 10^{-3} : 10^{-2} \right]$ m^{-1} wavenumber range, because

23　we keep the phase angles unchanged in the 3DCLOUD algorithm.

24　 In Fig. 12f, we represent the optical depth distributions. The initial optical depth distribution

25　does not follow a common distribution, whereas the optical depth distribution for cases 1 and

26　2 are log-normal. Indeed, in the 3DCLOUD algorithm, a gamma distribution for the optical

27　depth is imposed. For case 2 and 3, optical depth distributions are very close, even if the outer

28　scales are different. Thus, changing the L_{out} value does not affect significantly the shape of

1 optical depth distribution. In Fig. 12g, we can see that the horizontal mean optical depth

2 profiles are quasi identical for all cases.

3 These results show undeniably the flexibility of the 3DCLOUD algorithm. Indeed, in step 2,

4 3DCLOUD is able, by mapping a theoretical gamma-distributed optical depth onto the optical

5 depth field simulated at step1, to adjust, quasi-independently, the optical depth mean value,

6 the inhomogenenity parameter value of optical depth and the spectral slope value of optical

7 depth for $\left[1/L_{out} : 1/2\Delta x\right]$ m^{-1} wavenumber range.

8

4.2 Cirrus fields examples

4.2.1 Cirrus fields with assimilation of idealized meteorological profiles

11 Including some modifications presented in Sect. 2.1.1, 3DCLOUD is also able to generate

12 cirrus cloud. We briefly present in this section an example of ice water content (IWC) of

13 cirrus with fallstreaks. For cirrus, we chose to generate IWC field instead of optical depth

14 field as for stratocumulus or cumulus.

15 Figure 13 shows idealized vertical profiles of potential temperature, relative humidity, and

16 horizontal velocity components assimilated by 3DCLOUD as well as the ice water path (IWP)

17 simulated at step 1 by 3DCLOUD. It also shows the IWP simulated by 3DCLOUD during

18 step 2, the initial and corrected mean power spectra, the initial and corrected probability

19 density functions and the IWC volume rendering. Horizontal extensions $L_x = L_y$ and vertical

20 extension L_z are set to 10 km and 12.5 km, respectively. Horizontal resolutions $\Delta x = \Delta y$ and

21 vertical resolution Δz are set to 24 m and 83.3 m, respectively. $\overline{\text{IWC}}$ is set to the value

22 obtained at the end of step 1 (0.54 mg/m^3). The inhomogeneity parameter ρ_{IWC} is set to 1.0

23 and the outer scale L_{out} to 1 km.

24 Initial meteorological profiles assimilated by 3DCLOUD have been constructed in such a way

25 that thin cirrus is generated between 9.5 km and 10.5 km with fallstreaks. Vertical profiles of

26 potential temperature, and especially their vertical gradients under and above the cirrus are

27 based on those proposed by Liu et al. (2003). In order to generate instabilities due to radiative

28 cooling (not simulated by 3DCLOUD), we imposed a null vertical gradient of the potential

29 temperature near the cirrus top height. We imposed a mean relative humidity with respect to

1   ice (RHI) of 104 % between 9.5 km and 10.5 km. Just above the cloud, RHI is set to 50%

2   then 20% near 12 km in altitude. Under the cirrus, RHI decreases with height to 85% near 8

3   km. To generate fallstreaks, we imposed larger wind shear inside the cirrus than under the

4   cirrus.

5   In Fig. 13, we note that IWP obtained after step 1 is smoother than IWP obtained after step 2,

6   and that the initial IWC spectral slope value after step 1 is much smaller (around -5.5) than

7   the corrected IWC spectral slope after step 2 (around -1.6) in the $\left[10^{-3} : 2 \times 10^{-2} \right]$ m^{-1}

8   wavenumber range. For wavenumber smaller than $1/L_{out} = 10^{-3}$ m^{-1}, the power spectra are

9   constant. The corrected IWC probability distribution is exponential-like distribution after step

10   2. This is due to the larger value of $\rho_{IWC} = 1.0$ used in this example, compared to $\rho_\tau = 0.7$

11   used for stratocumulus in Sect 4.1.3.

12

4.2.2 Cirrus field and wind shear

14   We investigate briefly the aspect of cloud organization due to wind shear with 3DCLOUD

15   model and with other stochastic models. We focus on the work of Marsham and Dobbie

16   (2005) and of Hogan and Kew (2005). These two studies are very pertinent together. Indeed,

17   based on RADAR retrievals of IWC from the Chilbolton 94 GHz RADAR on 27 December

18   1999, which shows a strongly sheared ice cloud (named hereafter RC99 case), Marsham and

19   Dobbie (2005) investigated shear effects by simulating the RC99 case with the UK Met office

20   LES. In contrast, Hogan and Kew (2005) used their Cloudgen model, a 3D stochastic cloud

21   model being able to simulate the structural properties of ice clouds. To configure 3DCLOUD

22   in order to simulate the RC99 case, we assimilate meteorological profiles (potential

23   temperature, horizontal wind velocity) based on those drawn in Fig 2 in Marsham and Dobbie

24   (2005). We run also the RC99a case with no wind (and therefore no wind shear), and the

25   RC99b case where the potential temperature profile (drawn in Fig 15 in Marsham and

26   Dobbie, 2005) reduces atmospheric stability in order to give more extensive Kelvin-

27   Helmholtz wave braking. All our simulations are done with $N_x = N_y = 200$ and $N_z = 66$

28   and $\Delta_x = \Delta_y = 250$ m and $\Delta_z = 120$ m. Horizontal extensions are $L_x = L_y = 50$ km and

29   vertical extension is $L_z = 8$ km between 4 km and 12 km. Note that 3DCLOUD, Marsham

30   and Dobbie (2005) and Hogan and Kew (2005) numerical resolution are $\Delta_x = \Delta_y = 250$ m,

1    $\Delta_x = 100$m and $\Delta_x \approx 780$ m, respectively. Note also that 3DCLOUD, Marsham and Dobbie

2   (2005) and Hogan and Kew (2005) horizontal extensions are $L_x = L_y = 50$ km, $L_x = 50$ km

3   and $L_x = L_y = 200$ km.

4   Figure 14 shows a 2D vertical slice of 3DCLOUD IWC at an angle parallel to the wind of the

5   RC99a case, the RC99 case and the RC99b case. Note that the 3DCLOUD fields are smooth

6   because they are obtained at the first step of the algorithm. These three snapshots are very

7   similar to those presented in Marsham and Dobbie (2005), allowing us to confirm that our

8   basic atmospheric equations are correctly solved. In Fig. 14a, we can see small structures (few

9   km) at above 7 km, due to radiative cooling at the cloud top and latent heat release in the

10   updraughts. Below, we can observe fallstreaks advected (or not if there is no wind) relative to

11   their source in the convective layer by the shear. The shear homogenizes the fallstreaks.

12   Figure 14b clearly shows that 3DCLOUD simulations at the first step of the algorithm

13   homogenize the fallstreaks a lot, certainly too much compared to the RADAR retrievals of

14   IWC from the Chilbolton 94 GHz RADAR on 27 December 1999 (see Fig. 1 in Hogan and

15   Kew, 2005). Figure 14c shows the RC99b case where 3DCLOUD model is able to simulate

16   Kelvin-Helmholtz wave breaking, a dynamic aspect difficult to simulate with purely

17   stochastic models.

18   Figures 15a and 15b are the same as Fig. 14b but 3DCLOUD fields are obtained at the second

19   step of the algorithm, with 1D spectral slope close to -5/3 from the outer scale $L_{out} = 15$ km

20   to the numerical scale $\Delta x = 250$ m. In Fig. 15a, the mean value of 3D IWC $\overline{IWC} = 0.07$ gm^{-3}

21   and inhomogeneity parameter $\rho_{IWC} = 0.4$ for cloudy voxels are those computed from the 3D

22   IWC field obtained at step one of the 3DCLOUD algorithm. In Fig. 15b, inhomogeneity

23   parameter ρ_{IWC} is a function of height. Its values are roughly estimated from the Fig. 2c in

24   Hogan and Kew (2005). Compared to Hogan and Kew (2005) simulations, 3DCLOUD

25   snapshots show more details in the convective layer above 7 km as it simulates relatively well

26   the convective cloud structures thanks to imposed numerical spatial resolution. By contrast to

27   Fig. 15a, Hogan and Kew (2005) simulations show more details in the layer under 7 km,

28   where the shear-induce mixing is important. In order to obtain such details in the layer under

29   7 km with 3DCLOUD, we have to constrain it with ancillary data: those provided by the

30   RADAR retrievals of IWC and shown on Fig. 2c in Hogan and Kew (2005). Indeed, if we

31   force, in the second step of 3DCLOUD algorithm, the vertical inhomogeneity parameter to

32   match the one estimated crudely from the RADAR retrievals, we obtain Fig. 15b. Details in

the layer under 7 km of this snapshot are quite similar to those obtained by Hogan and Kew (2005).

Finally, Fig. 16, which is similar to Fig. 2 in Hogan and Kew (2005), shows vertical profiles of the cloud fraction, the mean of logarithm of IWC for the cloudy voxels and the standard deviation of logarithm of IWC for cloudy voxels computed from two 3DCLOUD fields obtained at the second step of the algorithm for RC99 case. As for Fig. 15, inhomogeneity parameter is $\rho_{IWC} = 0.4$ for the first cloud field and depends on height and it is derived from RADAR retrievals for the second case. For both cloud fields, the cloud coverage is equal to 1 between 5 km and 10 km. RADAR retrievals of IWC shows that the cloud coverage is equal to 1 only between 5.5 km and 7 km, and decreases to 0 at 10 km. However, as discussed in Sect. 2.1.2, the current version of 3DCLOUD is not able to readily simulate fractional cloud coverage in the cirrus regime. For both cloud fields, vertical profiles of the mean IWC are quite similar and consistent with those retrieved from RADAR. The inhomogeneity parameter vertical profile simulated by the current version of 3DCLOUD is too small, leading to a smoothing of the layer under 7 km that can be improved if the vertical profile of the inhomogeneity parameter is known.

5   Conclusion

3DCLOUD is a flexible three-dimensional cloud generator developed to simulate with a personal computer and under Matlab environment, synthetic but realistic stratocumulus, cumulus and cirrus cloud fields. Simplified dynamic and thermodynamic laws allow the generation of realistic liquid or ice water content from meteorological profiles. The stochastic process with the Fourier framework allow us to provide ice water content or optical depth sharing similar statistical properties to those observed in real clouds such as the inhomogeneity parameter (set by the user) and the invariant scale properties characterised by a spectral slope close to -5/3 from the smaller scale (set by spatial resolution of grid computation) to the outer scale (set by the user). In order to simulate cloud structures, 3DCLOUD solves simplified basic atmospheric equations and assimilates the cloud coverage set by the user (only for the stratocumulus and cumulus regimes) and meteorological profiles (pressure, humidity, wind velocity) defined by the user.

The 3DCLOUD outputs were compared to LES ones for three classical test cases. We chose the case of DYCOMS2-RF01, the case of BOMEX , and the case of ICMCP. For these cases, results show that 3DCLOUD outputs are relatively consistent with LES outputs, and confirm that the chosen basic atmospheric equations of 3DCLOUD are solved correctly. We also show that, under the condition that the user provides coherent meteorological profiles, 3DCLOUD algorithm is able to assimilate them and generate realistic cloud structures.

3DCLOUD is a very interesting research tool to better understand 3D interactions between cloudy atmosphere and atmospheric radiation, which is of primary importance in order to make progress in the direct radiative problem (GCM context) and in the inverse radiative problem (remote sensing context, development of the next generation of atmospheric sensors). For example, 3DCLOUD was used to quantify the impact of stratocumulus heterogeneities on polarized radiation measurements performed by POLDER/PARASOL (Cornet et al., 2013) as well as the influence of cirrus heterogeneities on brightness temperature measured by IIR/CALIPSO (Fauchez et al., 2013, 2014).

We still have to develop a stochastic process to generate 3D field of cloud effective radius. In a longer term, investigations will focus on the generation of 3D mixed phase cloud and eventually on the simulation of 3D rain rate. Another task will be to provide a FORTRAN code of 3DCLOUD, assumed to be faster than the current Matlab code.

Code availability

The source code of the 3DCLOUD algorithm is available online at http://wwwobs.univ-bpclermont.fr/atmos/fr/restricted

Please contact the authors for the password.

Acknowledgements

References

Alexandrov, M. D., Marshak, A., and Ackerman, A. S.: Cellular statistical models of broken cloud fields. Part I: Theory, J. Atmos. Sci., 67, 2125-2151, doi:10.1175/2010JAS3364.1, 2010.

Asai, T.: A numerical study of the air-mass transformation over the Japan Sea in winter, J. Met. Soc. Japan, 43, 1-15, 1965.

Barker, H. W. and Liu, D. M.: Inferring optical depth of broken clouds from Landsat data, J. Climate, 8, 2620-2630, 1995.

Barker, H. W., Wiellicki B. A., and Parker, L.: A parameterization for computing grid-averaged solar fluxes for inhomogeneous marine boundary layer clouds. Part II: validation using satellite data, J. Atmos. Sci., 53, 2304–2316, 1996.

Battaglia, A. and Mantovani, S.: Forward Monte Carlo computations of fully polarized microwave radiation in non isotropic media, J. Quant. Spectrosc. Rad. Transf., 95, 285-308, 2005.

Battaglia, A. and Tanelli, S.: DOMUS: DOppler MUltiple-Scattering Simulator, IEEE Trans. Geosci Rem. Sens., 49, 442-450, 2011.

Benassi, A., Szczap F., Davis, A. B., Masbou, M., Cornet, C., and Bleuyard, P.: Thermal radiative fluxes through inhomogeneous cloud fields: A sensitivity study using a new stochastic cloud generator, Atmos. Res., 72, 291-315, 2004.

Bony, S. R., Colman, R., and Kattsov, V. M.: How well do we understand and evaluate climate change feedback processes ?, J. Climate, 19, 3445-3482, 2006.

Bryan, G. H., Wyngaard, J. C., and Fritsch, J. M.: Resolution requirements for the simulation of deep moist convection, Mon. Wea. Rev., 131, 2394-2416, 2003

Cahalan, R. F.: Bounded cascade clouds albedo and effective thickness, Nonlinear Proc. Geophys., **1**, 156-167, 1994.

Cahalan, R. F., Ridgway, W., Wiscombe, W. J., Bell, T. L., and Snider, J. B.: The albedo of fractal stratocumulus clouds, J. Atmos. Sci., 51, 2434–2455, 1994.

Carlin, B., Fu Q., Lohmann U., Mace G. G., Sassen K., and Comstock, J. M.: High cloud horizontal inhomogeneity and solar albedo bias, J. Climate, 15, 2321-2339, 2002.

Chen, C. and Cotton, W. R.: A one-dimensional simulation of the stratocumulus-capped mixed layer, Bound-Layer Meteor., 25, 289–321, 1983.

Cheng, W. Y. Y., Wu, T., and Cotton, W. R.: Large-eddy simulations of the 26 November 1991 FIRE II cirrus case, J. Atmos. Sci., 58, 1017-1034, 2001.

Clothiaux, E. E., Barker H. W., and Korolev, A. V.: Observing clouds and their optical properties, in : 3D Radiative Transfer in Cloudy Atmospheres, edited by Marshak A. and Davis A. B., Springer, 93-152, 2004

Collins, W. D. and Satoh, M.: Simulating global clouds, past, present, and future, in: Clouds in the Perturbed Climate System, edited by Heintzenberg, J. and Charlson R. J., MIT press, 469-486, 2009.

Cotton, W. R., Pielke, R. A., Walko, R. L., Liston, G. E., Tremback, C. J., Jiang, H., McAnelly, R. L., Harrington, J. Y., Nicholls, M. E., Carrio, G. G., and McFadden, J.P.: RAMS 2001: Current status and future directions, Meteorol. Atmos. Phys., 82, 5-29, 2001.

Cornet, C., Buriez, J.-C., Riedi, J., Isaka, H., and Guillemet, B.: Case study of inhomogeneous cloud parameter retrieval from MODIS data, Geophys. Res. Lett., 32, L13807, doi:10.1029/2005GL022791, 2005

Cornet, C., Labonnote, L.-C., and Szczap, F.: Three-dimensional polarized Monte Carlo atmospheric radiative transfer model (3DMCPOL): 3D effects on polarized visible reflectances of a cirrus cloud, J. Quant. Spectr. Radiative Transfer, 111, 174-186, doi:10.1016/j.jqsrt.2009.06.013, 2010.

Cornet, C., Szczap, F., Labonnote, L.-C., Fauchez, T., Parol, F., Thieuleux, F., Riedi, J., Dubuisson, P. and Ferlay, N.: Evaluation of cloud heterogeneity effects on total and polarized visible radiances as measured by POLDER/PARASOL and consequences for retrieved cloud properties , AIP Conf. Proc., 1531, 99-102, doi: 10.1063/1.4804717 , 2013.

Davis, A. B., Marshak, A., Cahalan, R. F., and Wiscombe, W. J.: Multifractal characterizations of non-stationarity and intermittency in geophysical fields, observed, retrieved or simulated. J. Geophys. Res., 54, 241–260, 1994.

Davis, A. B., Marshak, A., Wiscombe, W. J., and Cahalan, R. F.: Scale invariance of liquid water distribution in marine stratocumulus: Part I. Spectral properties and stationarity issues. J. Atmos. Sci. 53, 1538–1558, 1996.

Davis, A. B., Marshak, A., Cahalan, R. F., and Wiscombe, W. J.: The Landsat scale break in Stratocumulus as a three-dimensional radiative transfer effect: implication for cloud remote sensing. J. Atmos. Sci. 54, 241–259, 1997.

Davis, A. B., Marshak, A., Gerber, H., and Wiscombe, W. J.: Horizontal structure of marine boundary layer clouds from centimeter to kilometer scales. J. Geophys. Res. 104, 6123–6144, 1999.

Deardorff, J. W.: Stratocumulus-capped mixed layers derived from a three-dimensional model, Bound.-Layer Meteor., 18, 495–527, 1980.

Di Giuseppe, F. and Tompkins, A. M.: Effect of spatial organisation on solar radiative transfer in three dimensional idealised stratocumulus cloud fields, J. Atmos. Sci., 60, 1774-1794, 2003.

Emanuel K. A.: Atmospheric Convection, Oxford university press, 1994.

Evans, K. F. and Wiscombe, W. J.: An algorithm for generating stochastic cloud fields from radar profile statistics, Atmos. Res., 72, 263-289, 2004.

Fauchez, T., Cornet, C., Szczap F., and Dubuisson, P.: Assessment of cloud heterogeneities effects on brightness temperatures simulated with a 3D Monte-Carlo code in the thermal infrared, AIP Conf. Proc. Vol. 1531, 75-78, DOI: 10.1063/1.4804711, 2013

Fauchez, T., Cornet, C., Szczap, F., Dubuisson, P., and Rosambert, T.: Impacts of cirrus clouds heterogeneities on TOA thermal infrared radiation, accepted to Atmos. Chem. Phys., 2014.

Hill, P. G., Hogan R. J., Manners, J., and Petch, J. C.: Parametrising the horizontal inhomogeneity of ice water content using CloudSat observations, Q. J. R. Meteorol. Soc., 138, 1784-1793, 2012.

Hogan, R. J. and Kew, S. F.: A 3D stochastic cloud model for investigating the radiative properties of inhomogeneous cirrus clouds, Q. J. R. Meteorol. Soc., 131, 2585-2608, 2005.

Holton, J. R.: An Introduction to Dynamic Meteorology, Fourth edition, Elsevier academic press, 2004.

Houze Jr., R. A.: Cloud Dynamics, Academic press, San Diego, 1993.

1 Illingworth, A. and Bony, S.R.: Observational strategies at meso- and large scales to reduce
2 critical uncertainties in future cloud changes, in : Clouds in the Perturbed Climate System,
3 edited by Heintzenberg, J. and Charlson R. J., MIT press, 511-530, 2009.

4 Kärcher, B. and Spichtinger, P.: Cloud-controling factors of cirrus, in : Clouds in the
5 Perturbed Climate System, edited by Heintzenberg, J. and Charlson R. J., MIT press, 235-
6 268, 2009.

7 Kobayashi, T.: Effects due to cloud geometry on biases in the albedo derived from radiance
8 measurements, J. Climate, 6, 120-128, 1993.

9 Lafont, D. and Guillemet, B.: Subpixel fractional cloud cover and inhomogeneity effects on
10 microwave beam-filling error, Atmos. Res., 72,149-168, doi: 10.1016/j.atmosres.2004.03.013,
11 2004.

12 Liou, K. N., An introduction to atmospheric radiation, Academic press, 2002.

13 Liu, H.-C., Wang, P. K., and Schlesinger, R. E.: A numerical study of cirrus clouds. Part II:
14 effects of ambient temperature, stability, radiation, ice microphysics, and microdynamics on
15 cirrus evolution, J. Atmos. Sci., 60, 1097–1119, 2003.

16 Marshak, A., Davis, A., Wiscombe, W. J. , Ridgway, W., and Cahalan, R.: Biases in
17 shortwave column absorption in the presence of fractal clouds, J. Climate, 11, 431-446, 1998.

18 Mayer, B.: Radiative transfer in the cloudy atmosphere, European Physical Journal
19 Conferences, **1**, 75-99, 2009.

20 Meyers, M. P., Walko, R. L., Harrington, J. Y., and Cotton, W. R.: New RAMS cloud
21 microphysics parameterization. Part II. The two-moment scheme, J. Atmos. Res., 45, 3-39,
22 1997.

23 Moeng, C. H.: A large-eddy-simulation model fort he study of planetary boundery-layer
24 turbulence, J. Atmos. Sci., 41, 2052-2062, 1984.

25 Penide, G., Giraud, V., Bouniol, D., Dubuisson, P., Duroure, C., Protat, A., and Cautenet, S.:
26 Numerical simulation of the 7 to 9 September 2006 AMMA mesoscale convective system:
27 Evaluation of the dynamics and cloud microphysics using synthetic observations. AMMA
28 special issue, Quart. J. Roy. Meteor. Soc., 136 (S1), 304-322, DOI 10.1002/qj.558, 2010.

Pincus, R. and Evans, K. F.: Computational cost and accuracy in calculating three-dimensional radiative transfer: Results for new implementations of Monte Carlo and SHDOM, J. Atmos. Sci., 66, 3131-3146, 2009.

Pielke, R. A., Cotton, W. R., Walko, R. L., Tremback, C. J., Lyons, W.A., Grasso, L. D., Nieholls, M. E., Moran, M. D., Wesley, D. A., Lee, T. J., and Copeland, J. H.: A comprehensive meteorological modelling system-RAMS, Meteorol. Atmos. Phys., 49, 69-91, 1992.

Pielke, R. A.: Mesoscale Meteorological Modeling, Academic press, 2002.

Riddaway, R. W.: Numerical methods, available at : www.ecmwf.int/newsevents/training/rcourse_notes/NUMERICAL_METHODS/NUMERICAL_METHODS/Numerical_methods6.html (last access: 10 January 2014), 2001.

Sassen, K., Wang L., Starr D. O. C, Comstock J. M., and Quante, M.: A midlatitude cirrus cloud climatology from the facility for atmospheric remote sensing. Part V: Cloud structural properties, J. Atmos. Sci., 64, 2483-2501, 2007.

Siebesma, A. P., Bretherton, C. S., Brown, A., Chlond, A., Cuxart, J., Duynkerke, P. G., Jiang, H., Khairoutdinov, M., Lewellen, D., H.-Moeng, C., Sanchez, E., Stevens, B., and Stevens, D. E.: A Large-Eddy Simulation intercomparison study of shallow cumulus convection , J. Atmos. Sci., 60, 1201-1219, 2003.

Starr, D. O. C. and Cox, S. K.: Cirrus Clouds. Part I: A cirrus cloud model, J. Atmos. Sci., 42, 2663–2681, 1985.

Starr, D. O. C., Benedetti, A., Boehm, M., Brown, P. R. A., Gierens, K. M., Girard, E., Giraud, V., Jakob, C., Jensen, E., Khvorostyanov, V. I., Koehler, M., Lare, A., Lin, R.-F., Maruyama, K., Montero, M., Tao, W.-K., Wang, Y., and Wilson, D.: Comparison of cirrus cloud models: a project of the GEWEX Cloud System Study (GCSS) working group on cirrus cloud system. Proceedings of the 13th International Conference on Cloud and Precipitation (ICCP), Reno, NV, USA, 14–18 August 2000.

Stevens, B., H-Moeng, C., Ackerman A. S., Bretherton,C. S., Chlond, A., de Roode, S., Edwards, J., Golaz, J. C., Jiang, H., Khairoutdinov, M., Kirkpatrick, M. P., Lewellen, D. C., Lock, A. L., Müller, F., Stevens, D. E., Whelan, E., and Zhu, P.: Evaluation of large-eddy simulations via observations of nocturnal marine stratocumulus, Mon. Wea. Rev., 133, 1443–1462, 2005.

1 Szczap, F., Isaka, H., Saute, M., Guillemet, B., and Ioltukhovski, A.: Effective radiative

2 properties of bounded cascade nonabsorbing clouds: Definition of the equivalent

3 homogeneous cloud approximation, J. Geophys. Res., 105, D16, 20,617-20,634,

4 doi:10.1029/2000JD900146, 2000.

5 Szczap, F., Cornet, C., Alqassem, A., Gour, Y., Labonnote, C.-L., and Jourdan, O.:

6 A 3D polarized Monte Carlo LIDAR system simulator for studying effects of cirrus

7 inhomogeneities on CALIOP/CALIPSO measurements, AIP Conf. Proc., 1531, 139-142, doi

8 10.1063/1.4804727 , 2013.

9 Várnai, T. and Marshak, A.: Observations of three-dimensional radiative effects that influence

10 MODIS cloud optical thickness retrievals, J. Atmos. Sci., 59, 1607-1618, 2002.

11 Várnai, T. and Marshak, A.: View angle dependence of cloud optical thickness retrieved by

12 MODIS, J. Geophys. Res., 112, D06203, doi:10.1029/2005JD006912, 2007

13 Venema, V., Meyer, S., García, S. G., Kniffka, A., Simmer, C., Crewell, S., Löhnert, U.,

14 Trautmann, T., and Macke A.: Surrogate cloud fields generated with the Iterative Amplitude

15 Adapted Fourier Transform algorithm, Tellus A, 58A, 104-120, 2006.

16 Yamaguchi, T., and Feingold, G.: Technical note: Large-eddy simulation of cloudy boundary

17 layer with the Advanced Research WRF model, JAMES, 4, doi:10.1029/2012MS000164,

18 2012.

19 Yang, H., Dobbie, S., Mace, G. G., Ross, A., and Quante, M.:GEWEX Cloud System Study

20 (GCSS) cirrus cloud working group: development of an observation-based case study for

21 model evaluation, Geosci. Model Dev., 5, 829-843, 2012.

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1 **Tables**

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3 Table 1. Time step, process time for one time step and process time for 2h-simulation with
4 3DCLOUD model, as a function of the numerical resolution. DYCOMS2-RF01 and BOMEX
5 cases are presented. A comparison between 3DCLOUD and BRAMS LES computation time
6 for a specific DYCOMS2-RF01 case is added. 3DCLOUD (Matlab code) runs on a personal
7 computer with Intel Xeon E5520 (2.26 GHz) and BRAMS (Fortran code) runs on a
8 PowerEdge R720 with Intel Xeon E5-2670 (2.60GHz), both of them having a single-
9 processor configuration.

Study case	Point mesh $N_x \times N_y \times N_z$	Horizontal numerical resolution Δx [m]	Time step [s]	Process time [s]	Process time for 2h-simulation [s]
DYCOMS2-RF01	$50 \times 50 \times 50$	200	10	0.4	290
	$100 \times 100 \times 50$	100	7	1.3	1340
	$200 \times 200 \times 50$	50	5	5	7200
	$400 \times 400 \times 50$	25	3	18	43200
BOMEX	$50 \times 50 \times 70$	200	30	0.7	170
	$100 \times 100 \times 70$	100	25	2.5	720
	$200 \times 200 \times 70$	50	20	10	3600
	$400 \times 400 \times 70$	25	14	40	20600
DYCOMS2-RF01 3DCLOUD	$100 \times 100 \times 100$	40	13	2.7	1500
BRAMS	$100 \times 100 \times 100$	40	0.3	2	48600

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1 **Figure captions**

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3 Figure 1. General flow chart of the stratocumulus, cumulus and cirrus generator 3DCloud.
4 Note that 3DCloud algorithm is divided in two distinct steps.

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6 Figure 2. Time series of (a) the mean cloud top height, (b) the mean cloud base height, (c) the
7 cloud coverage, and (d) the liquid water path. The DYCOM2-RF01 case is displayed.The
8 solid lines indicate 3DCLOUD results. The dotted lines indicate a mean over all LES results.
9 The light shading around this mean delimits the maximum and minimum values within the
10 master ensemble at any given time.

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12 Figure 3. Mean profiles averaged over the fourth hour of (a) the longwave net flux, (b) the
13 liquid water potential temperature, (c) the total water mixing ratio, (d) the liquid water mixing
14 ratio, (e) the horizontal velocity components, and (f) the air density. The DYCOM2-RF01
15 case is displayed. The solid lines indicate 3DCLOUD results. The dotted lines indicate a mean
16 over all LES results. The light shading around this mean delimits by the maximum and
17 minimum values within the master ensemble at any given height.

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19 Figure 4. Time series of (a) the cloud coverage and (b) the liquid water path. The BOMEX
20 case is displayed. The solid lines indicate 3DCLOUD results. The light shading delimits the
21 maximum and minimum values within the master ensemble at any given time.

22

23 Figure 5. Mean profiles averaged over the fifth hour of (a) cloud coverage, (b) the potential
24 temperature, (c) the water vapour mixing ratio, (d) the liquid water mixing ratio, (e) the
25 horizontal velocity components, and (f) the air density.The BOMEX case is displayed. The
26 solid lines indicate 3DCLOUD results. The light shading delimits the maximum and
27 minimum values within the master ensemble at any given height.

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29 Figure 6. Time series of vertically-integrated ice water path (IWP) from different cirrus
30 models, which participated in the Idealized Cirrus Model Comparison Project and from the

1 3DCLOUD model (thick blue lines). The upper panel is for the cold cirrus case and the
2 bottom one is for the warm cirrus case. Cyan line represents models with bin microphysics,
3 red line models with bulk microphysics, green line single column models, and thin black lines
4 models with heritage in the study of deep convection or boundary layer clouds. This figure is
5 made from the one taken from Starr et al. (2000) and Yang et al. (2012).

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7 Figure 7. The instantaneous cloud-field snapshots of the pseudo albedo at four hours
8 simulated by (a) the UCLA-0 model (picture taken from Stevens et al., 2005), (b) the BRAMS
9 model, both configured following the DYCOMS2-RF01 case (Stevens et al., 2005) and (c)
10 from 3DCLOUD with assimilation of meteorological profiles based on the DYCOMS-RF01
11 case. The UCLA-0 field is drawn from simulation where $N_x = N_y = 192$ and $\Delta_x = \Delta_y = 20$
12 m. Both BRAMS and 3DCLOUD are drawn from simulations where $N_x = N_y = N_z = 100$,
13 $\Delta_x = \Delta_y = 40$ m and $\Delta_z = 12$ m. Note that the 3DCLOUD field is obtained at the second step
14 of the algorithm, with the inhomogeneity parameter $\rho_\tau = 0.3$, mean optical depth $\bar{\tau} = 10$ and
15 $L_{out} = 2$ km. (d) is the optical depth power spectra computed following the x and the y
16 directions and then averaged, for BRAMS (points) and 3DCLOUD (circles). A theoretical
17 power spectrum with spectral slope $\beta = -5/3$ is added (black line).

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20 vertical velocity (shaded) and the cloud water (contoured), at the end of simulation, for the
21 stratocumulus simulated by 3DCLOUD with assimilation of meteorological profiles based on
22 the DYCOMS2-RF01 case. Different numerical spatial resolutions are presented with
23 $\Delta x = \Delta y$: (a) and (e) $\Delta x = 200$ m, (b) and (f) $\Delta x = 100$ m, (c) and (g) $\Delta x = 50$ m and (d) and
24 (h) $\Delta x = 25$ m. (i), (j), (k) and (l) mean profiles of the potential temperature, the liquid water
25 mixing ratio, the horizontal velocity components, and the vapour water mixing ratio. The
26 solid lines indicate meteorological profiles based on DYCOMS2-RF01 case and assimilated
27 by 3DCLOUD. Points, dotted lines, dashed lines and dash-dot lines indicate 3DCLOUD
28 results at the end of simulation for the different numerical spatial resolution $\Delta x = 200$ m,
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Figure 9. Same as Fig. 8, for cumulus cloud simulated by 3DCLOUD with assimilation of meteorological profiles based on the BOMEX case. We let 3DCLOUD iterating until 2h-simulation is done.

Figure 10. Time series of the mean distance between cloud areas for different horizontal numerical spatial resolutions (colored lines) with a constant vertical numerical spatial resolution ($\Delta z = 38.5$ m) and mean distance averaged over the last-half hour of a 2h simulation as a function of numerical spatial resolution (black line with circles). The cumulus cloud is simulated by 3DCLOUD with assimilation of meteorological profiles based on the BOMEX case. Horizontal extensions are $L_x = L_y = 5$ km and vertical extension is $L_z = 2700$ m.

Figure 11. Same as Fig. 8c ($\Delta x = 50$ m), for different assimilated values of the cloud coverage: (a) and (e) $C = 99\%$, (b) and (f) $C = 80\%$, (c) and (g) $C = 50\%$ and (d) and (h) $C = 20\%$.

Figure 12. (a) Pseudo albedo estimated from optical depth (initial field) simulated in the step 1 by 3DCLOUD for the DYCOMS2-RF01 case (see Fig. 8c), (b), (c) and (d) pseudo albedo adjusted in the step 2 of 3DCLOUD for different values of the inhomogeneity parameter ρ_τ and of the outer scale L_{out}. (e) mean power spectra of optical depth along x and y directions. The power spectra are scaled for better visualization. (f) probability density function of optical depth, (g) mean vertical profiles of horizontally averaged optical depth and (h) volume rendering of optical depth for the case 3. ρ_τ and L_{out} are 0.2 and 1 km for case 1, 0.7 and 1 km for case 2 and 0.7 and 10 km for case 3, respectively. Solid lines, dotted lines, dashed lines and dash-dot lines indicate initial field, case 1, case 2 and case 3 fields, respectively.

Figure 13. Idealized vertical profiles assimilated (dashed lines) and simulated (solid lines) by 3DCLOUD during step 1 of (a) the potential temperature and relative humidity and of (b) the horizontal velocity components and the ice water content (IWC), (c) ice water path (IWP) simulated by 3DCloud in step 1, (d) IWP simulated by 3DCloud in step 2, (e) mean power

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1   spectra of IWC along x and y directions after the step 1 and the step 2. (f) IWC probability

2   density functions after step 1 and step 2. (g) IWC volume rendering after step 2. ρ_{IWC} is set

3   to 1 and L_{out} is set to 1 km. Number of iterations is 1000.

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5   Figure 14. 2D vertical slice of 3DCLOUD ice water content (IWC gm^3) through a 3D

6   simulation at an angle parallel to the wind (a) RC99a case, (b) RC99 case and (c) RC99b case.

7   Fields are obtained from simulations where $N_x = N_y = 200$ and $N_z = 66$ and $\Delta_x = \Delta_y = 250$

8   m and $\Delta_z = 120$ m. Horizontal extensions are $L_x = L_y = 50$ km and vertical extension is

9    $L_z = 8$ km between 4 km and 12 km. Note that the 3DCLOUD fields are smooth because

10   obtained at the first step of the algorithm.

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12   Figure 15. Same as Fig. 14b but after the second step of the algorithm, where 1D spectral

13   slope is close to -5/3 from outer scale $L_{out} = 15$ km to the numerical scale $\Delta x = 250$ m.(a)

14   mean of 3D IWC is $\overline{IWC} = 0.07$ gm^{-3} and inhomogeneity parameter $\rho_{IWC} = 0.4$ for cloudy

15   voxels. (b) same as (a) but with inhomogeneity parameter ρ_{IWC}, function of height and

16   derived from the 27 December 1999 RADAR measurements between 10 and 12 UTC crudely

17   estimated from the Fig. 2c in Hogan and Kew (2005).

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19   Figure 16. Vertical profiles for the RC99 case of (a) cloud fraction, (b) mean of ln(IWC) for

20   the cloudy voxels and (c) standard deviation of ln(IWC) for cloudy voxels computed from

21   two 3DCLOUD fields obtained at the second step of the algorithm. The solid lines indicate

22   simulation where inhomogeneity parameter is $\rho_{IWC} = 0.4$ and the dotted lines indicate

23   simulation where the parameter ρ_{IWC}, function of height, is derived from the 27 December

24   1999 RADAR measurements between 10 and 12 UTC and crudely estimated from the Fig 2c

25   in Hogan and Kew (2005).

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1   **Figures**

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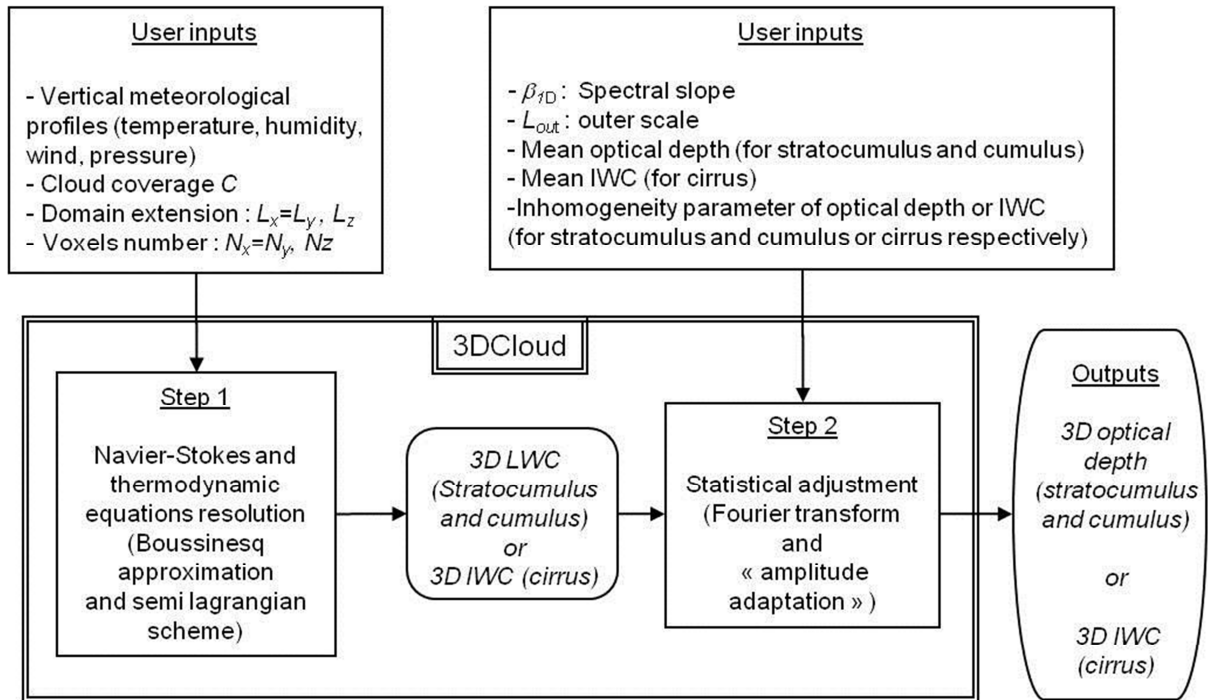
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Cold cirrus case

Warm cirrus case

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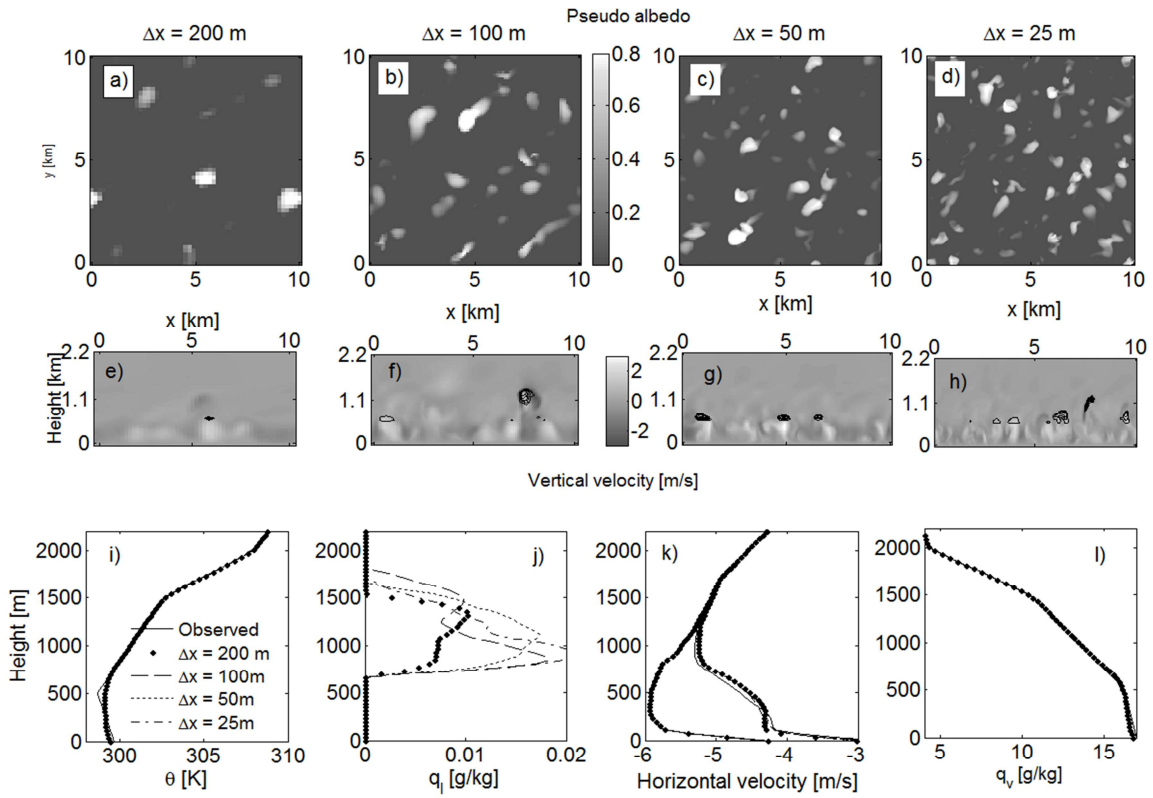
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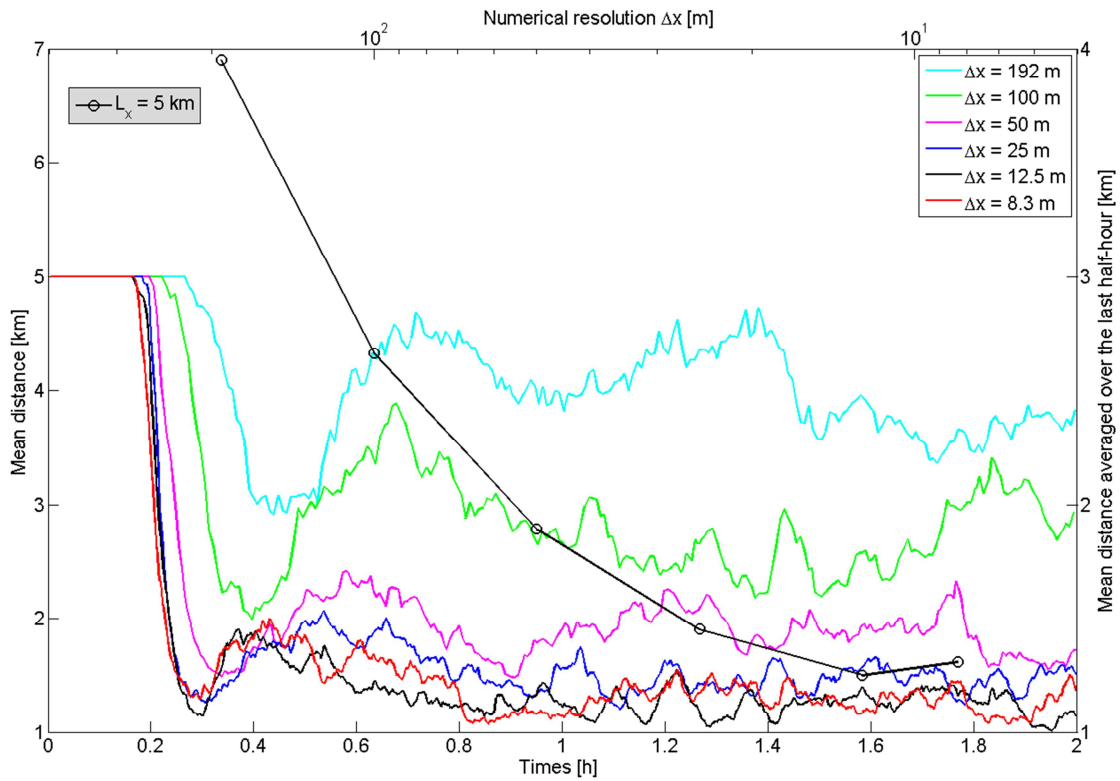
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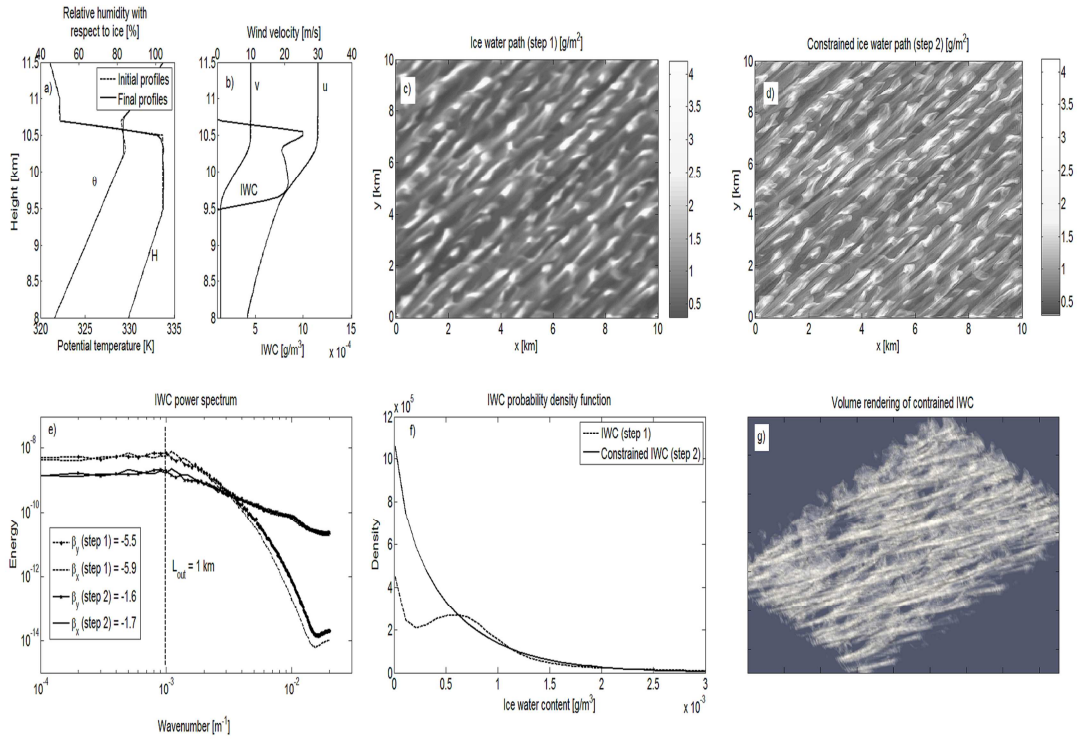
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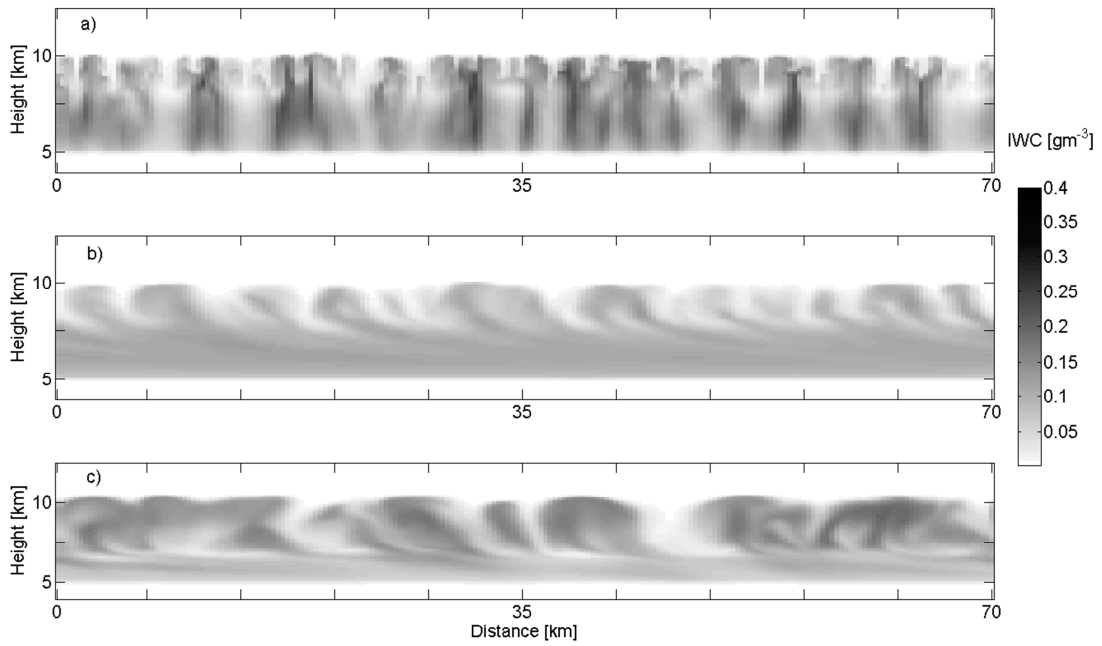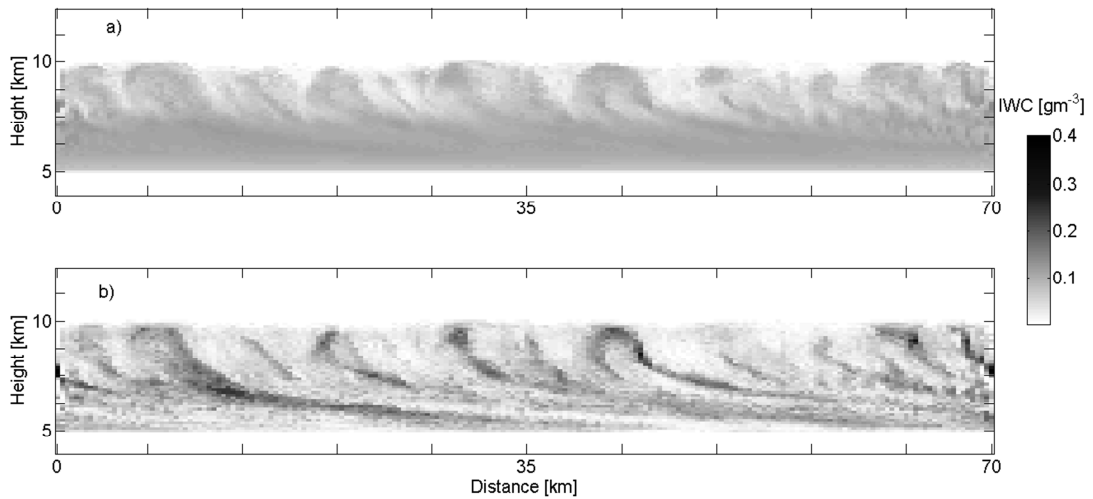
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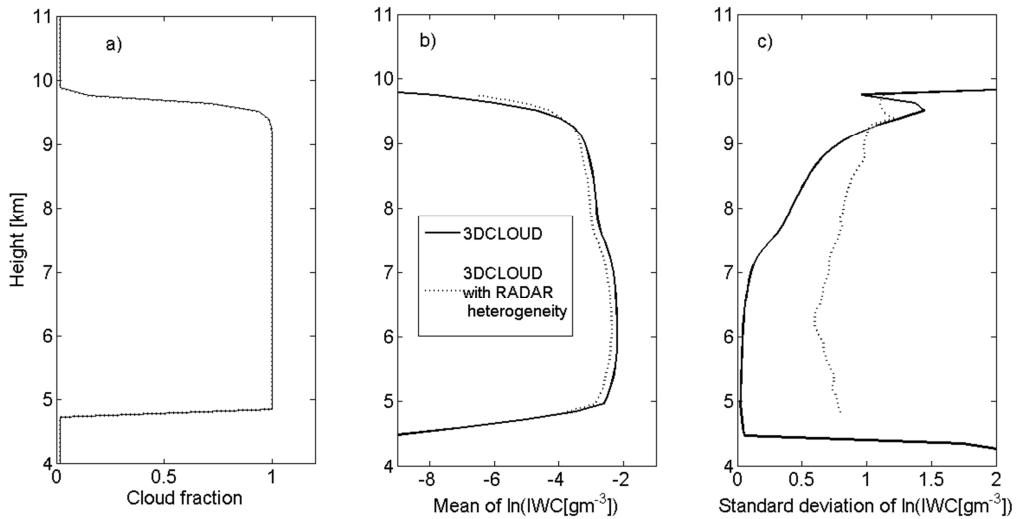7 estimated from the Fig. 2c in Hogan and Kew (2005).

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