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ADVECTION OF PASSIVE TRACERS IN THE ATMOSPHERE: BATCHELOR SCALING

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Extensive numerical experiments are performed on tracer dispersion in global reanalysis wind fields. Particle trajectories are computed both along an isobaric (500 hPa) and an isentropic (315 K) surface in a time interval of one year. Besides mean quantities such as advection of the center of mass and growth of tracer clouds, special attention is paid to the evaluation of particle pair separation dynamics. The characteristic behavior for intermediate time scales is *Batchelor's* dispersion along both surfaces, where the zonal extent of the tracer cloud increases *linearly* in time. The long-time evolution after 70–80 days exhibits a slower, *diffusive* dispersion (Taylor regime), in agreement with expectations. Richardson–Obukhov scaling (superdiffusion with an exponent of 3/2) could not be identified in the numerical tests. The results confirm the classical prediction by Batchelor that the initial pair-separation determines subsequent time evolution of tracers. The quantitative dependence on the initial distance differs however from the prediction of the theory.

Keywords: Chaotic advection; geostrophic turbulence; pair dispersion; scaling.

1. Introduction

Tracer transport and mixing processes substantially influence the distribution of chemical compounds in geophysical flows, therefore the quantifying and modeling of related phenomena have a long history [Weiss & Provenzale, 2008; Huang, 2010]. On the largest scales, the same governing equations are used both in the atmosphere and oceans, thus the tools to study tracer advection are also similar. Direct atmospheric experiments exist since the seventies, the first major balloon trajectory measurements were performed over the southern hemisphere: the EOLE at 200 mb [Morel & Bandeen, 1973; Morel & Larcheveque, 1974], and the TWERLE at 150 mb pressure level [Julian *et al.*, 1977; Er-El & Peskin, 1981]. Besides numerical simulations [Rood, 1987; Staniforth & Côté, 1991], laboratory models [Jánosi *et al.*, 2010] provide a deeper insight into the physical basis of the key processes.

Here we report the results of numerical experiments of tracer dispersion in realistic midtropospheric wind fields (see Fig. 1). The key quantities of interest are the mean drift of the "center of mass" of a cloud of tracers, and the spread of the cloud related to the average pair separation. The literature on pair dispersion both in 2D and 3D turbulent flows is pretty controversial, in spite of the intense research in the past decades (see [Salazar & Collins, 2009; Bourgoin *et al.*, 2006] and references therein). The main difficulty is that the dynamics obeys a series of crossover behavior from

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Fig. 1. Map sequence illustrating the dispersal of a small tracer cloud of 5×10^5 particles starting from 45° N, 0° in a cell of size $0.5^\circ \times 0.5^\circ$ at 00 h UTC, 01/01/2000, 500 hPa pressure level. Subsequent snapshots are separated by 7 days: (a) day 7, (b) day 14,... (f) day 42. The color bar indicates tracer density measured as particle number in a cell of $0.5^\circ \times 0.5^\circ$.

exponential to power-law dispersion with various exponent values, and the crossover points cannot be easily located since they depend on several factors. Among others, the initial separation is an essential parameter, thus an extended tracer cloud exhibits almost always a mixed dynamics (nearby points behave differently from pairs of large initial separation). Recent challenging experiments have confirmed the classical theory by Batchelor [1950] that the time evolution strongly depends on the initial pair separation [Bourgoin *et al.*, 2006].

The numerical experiments are based on the equations of passive scalar advection, where the background wind field is provided by the ERA-Interim data bank of the European Centre for Medium-Range Weather Forecasts (http://www.ecmwf.int/research/era/). The advection along

both isobaric and isentropic surfaces has almost the same dynamic features. The initial transients (short time dispersal) cannot be evaluated, because of the limited spatial and temporal resolution of the reanalysis wind field in the numerical tests. Pair separation is "ballistic" for intermediate time intervals, characterized by a linear growth of the zonal extent of the tracer cloud. This dynamics is not expected from theoretical considerations [Salazar & Collins, 2009], but it is identified in other numerical [Huber *et al.*, 2001; Haza *et al.*, 2008] and experimental works [Bourgoin *et al.*, 2006]. When strong correlations are produced by coherent structures in the flow decay at longer times, a gradual crossover to diffusive dispersal can be observed.

The paper is organized as follows. Section 2 gives an overview of the ECMWF reanalysis wind

field and the numerical methods used in the trajectory calculations. The results are listed in Sec. 3, a detailed comparison with theoretical predictions is provided in Sec. 4. A summary is given in Sec. 5.

2. Data and Methods

The accuracy of numerically computed tracer trajectories in various flow fields has improved a lot in the past decades [Stohl & Seibert, 1998; Stohl, 1998; Weiss & Provenzale, 2008]. In principle, precise atmospheric trajectories can be calculated directly from high resolution wind observations by interpolating between the measuring locations and time instants. In practice, however, trajectory calculations are mostly based on gridded output of numerical models (weather forecasts or reanalyses) such as provided by the European Centre for Medium-Range Weather Forecasts (ECMWF) [Uppala *et al.*, 2005].

In this work, the third generation ECMWF reanalysis ERA-Interim data bank is exploited, which is almost up to date from 01/01/1989 [http:// www.ecmwf.int/research/era/do/get/era-interim]. Zonal (u) and meridional (v) wind velocity components of global geographic coverage are evaluated for the years 2000 and 2009. Four values are available each day for 00 h, 06 h, 12 h and 18 h UTC (Universal Time Coordinated) at each geographic location with a spatial resolution of $1.5^{\circ} \times 1.5^{\circ}$ (lat/long). Note that the gridded wind fields are rather smooth, subgrid scale turbulence or vertical convection are not resolved. Although the wind velocity at a given site and time is intended to represent an instantaneous value [Uppala et al., 2005], direct comparison with high resolution wind tower measurements indicates that the velocity should be regarded as a six-hour mean value [Kiss et al., 2009].

Contrary to the horizontal wind velocity vector (u, v), there are no routine observations for the vertical component w. Estimates can be produced by various meteorological models, but they are definitely less accurate than the fields of the horizontal wind. A plausible idea is to compute the 2D divergence field and estimate w from the results, however it is known that this procedure has a very large error at the limited resolution of the reanalyses [Sardeshmukh & Liebmann, 1993]. In the absence of reliable 3D wind fields, the usual procedure is to follow isobaric (constant pressure), or isentropic (constant potential temperature) surfaces with accommodating 2D wind velocity vectors. The latter has the advantage that atmospheric variables tend to be better correlated along isentropic surfaces than on constant pressure surfaces, however the potential temperature (entropy) of a given air parcel can significantly change when diabatic processes have an important role along a trajectory (e.g. in baroclinic stratification, which is common in mid-latitude tropospheric flows) [Stohl & Seibert, 1998; Stohl, 1998]. For this reason we have computed and compared the statistical properties of tracer trajectories along both a constant pressure and a constant potential temperature level.

Global ERA-Interim wind fields are available on 37 pressure and 15 potential temperature levels in the ranges 1000-1 hPa, and 265-850 K [http:// data-portal.ecmwf.int/data/d/interim_daily/]. Since we are interested in free atmospheric flow, two levels in the mid-troposphere are selected for subsequent analysis. Firstly, the pressure level at 500 hPa in the year 2000, and secondly, the isentropic level at 315 K in the year 2009. The former level is located in the range of 5–6 km altitude depending on latitude, season and weather. The 315 K potential temperature level usually intersects the 500 hPa at around $30-45^{\circ}$ latitude on both hemisphere, it has lower values around the equator (down to 700 hPa), and higher values for high latitudes (up to 300–200 hPa at the poles) (Fig. 2).

Trajectory calculation is based on the solution of the advection equation for a given infinitesimal air parcel [Ottino, 1989; Tél & Gruiz, 2006]:

$$\frac{d\mathbf{r}(t)}{dt} = \mathbf{v}(\mathbf{r}(t), t), \tag{1}$$



Fig. 2. Zonal mean pressure of the 315 K surface in the year 2009. The dashed line represents the isobaric surface used.

where $\mathbf{v}(\mathbf{r}(t), t)$ is the instantaneous velocity field at the position \mathbf{r} and time t and $\mathbf{r}(t)$ is the trajectory. Equation (1) can be solved analytically for simple flow fields only, more realistic situations require a numerical treatment based on some finite difference Taylor-series representation [Stohl, 1998].

Our primary goal is to give a *statistical char*acterization of atmospheric dispersal instead of calculating accurate trajectories (which is hopeless in view of the chaotic nature of the advection dynamics), therefore, we use the simplest approximations wherever possible. The temporal and spatial resolutions of ERA-Interim wind fields represent the most significant limitation, therefore very accurate atmospheric trajectories cannot be expected even by the most advanced numerical methods. It is also meaningless to compare trajectories starting from the same initial location but advancing on the two different surfaces, therefore the investigated temporal periods are also separated.

The limited spatial and temporal resolutions require the implementation of some interpolation procedures for the wind field at the numerical solution of Eq. (1). Several methods are known and tested in the literature [Rood, 1987; Staniforth & Côté, 1991; Stohl *et al.*, 1995]. For the time variable t, the simple linear interpolation provides a sufficiently accurate solution [Stohl *et al.*, 1995]. Computationally more demanding algorithms are used for the spatial interpolation, such as the inverse quadratic $(1/r^2 \text{ weighting})$ or cubic spline approximations [Press *et al.*, 1992]. As for the numerical integration, the simple Euler method is implemented for the isobaric, and the fourth order Runge–Kutta algorithm for the isentropic trajectory computations [Press *et al.*, 1992]. Consistency is checked with various time-steps of 6, 15 and 30 min (see below).

Usually it is convenient to apply the regular latitude-longitude grid to characterize the advection in the atmosphere. But when cell sizes enter the statistics, it can be problematic, e.g. a cell of $1^{\circ} \times 1^{\circ}$ has a much smaller area near the poles than at the equator. In order to avoid such a high variability of the cell sizes in, e.g. tracer density statistics, an alternative tiling of almost equal-area tetragons is also used (Fig. 3), which is constructed in the following way. Two sides of the cells are constituted by latitudes of the same angle difference $\Delta \theta$, e.g. ϵ , as in the case of the regular latitude-longitude grid. At the equator the other two sides of a cell are assigned by meridians with the angle difference $\Delta \phi = \epsilon$. Since the area of a cell of a sphere of unit radius defined by meridians θ_1, θ_2 and latitudes ϕ_1, ϕ_2 is $(\theta_2 - \theta_1)(\sin \phi_2 - \sin \phi_1)$, we can construct cells the area of which are equal to that of a cell near the equator. The angle between the meridians of a cell becomes larger at higher latitudes (Fig. 3). Since a band of a latitude has to be filled with integer



Fig. 3. Tiling of the Earth surface by almost equal-area tetragons. Red color marks an extended tracer cloud advected from 45° N, 0° in a cell of size $10^{\circ} \times 10^{\circ}$ at 00 UTC 01/01/2009 on 315 K level, after a day.

number of cells, the cell sizes at different latitudes are not exactly the same.

3. Results

The time evolution of a small tracer cloud released from the middle of the northern hemisphere is illustrated in Fig. 1. Colors indicate the density computed as particle number in a given cell of $0.5^{\circ} \times 0.5^{\circ}$. Note that the meridional transport is strongly hindered across the equator, almost one year is necessary to obtain a near uniform tracer distribution on both hemispheres.

The following figures summarize the statistics of several numerical experiments. Note that the results are very similar for the isobaric and isentropic trajectories, thus we usually show plots for one case only, except when the definite goal is to illustrate close agreement. For the sake of simple identification, isobaric results are denoted by blue (turquoise error bars), while green indicates the results of isentropic calculations.

Figure 4 displays the time evolution of the center of mass (CM) in the zonal and the meridional directions. An ensemble of 96 numerical experiments was evaluated. The center of 12 different clusters were located at equal longitudinal spacing of 30° along the latitudinal circle 45°N, and each initial configuration was simulated from eight different initial time instants for 20 days with a cloud of 10^5 particles over an initial area of $1^{\circ} \times 1^{\circ}$. The mean CM drift velocity at 500 hPa has an almost constant eastward zonal component of $11^{\circ}/\text{day}$ (~870 km/day at 45° N) with a practically zero mean meridional component. The statistics is very similar over the 315 K isentropic surface, the mean zonal CM drift velocity is $14^{\circ}/\text{day}$ (~1107 km/day at 45° N) with zero meridional value. The larger value can be explained by the fact that the isentropic surface is situated at definitely higher altitudes over 45° N, than the 500 hPa isobar, thus a stronger eastward drift is expected.

This latter result is obtained from an ensemble averaging over 96 numerical experiments, where clusters of 130 particles distributed on a regular lattice inside of an initial cell of $10^{\circ} \times 10^{\circ}$ are studied with identical initial condition sets as above. The role of regular spacing was to avoid small tracer distances, we will return to this point in Sec. 4.

The same ensemble was used to determine the zonal and meridional spreads of the tracer cloud. The edges of the clusters move with speeds different than that of the CM. The spreads $\theta(t)$ and $\phi(t)$ of the cloud is defined as the difference between the maximum and minimum angular coordinates of the cloud at a given time instant, in analogy with related laboratory experiments [Jánosi *et al.*, 2010]. The average behavior is illustrated in Fig. 5: the zonal spreading quickly converges to a ballistic behavior of linear growth (slopes are $31.6^{\circ}/day$ for the isobaric, and $33.3^{\circ}/day$ for the isentropic



Fig. 4. Average time evolution of the center of mass (CM) for a tracer cloud formed by 10^5 particles advected on the 500 hPa pressure level. The initial area was $1^{\circ} \times 1^{\circ}$, 12 different initial longitude positions distributed evenly along the latitude 45° N and eight initial dates in the year 2000 formed the ensemble of 96 numerical experiments. (a) The mean zonal drift of CM (blue) and the standard deviation (turquoise). (b) The mean meridional drift of CM (blue) and the standard deviation (turquoise).



Fig. 5. Ensemble mean angular spread of the cloud in (a) zonal $\theta(t)$ and (b) meridional $\phi(t)$ directions. θ_0 and ϕ_0 mark the initial angular differences. Tracer clusters were formed by 130 particles, isentropic trajectories were evaluated. Light green band indicates one standard deviation obtained from 96 numerical experiments (see text). Black dashed line marks the average encompassing time of the globe $t^* \approx 13$ days in (a), white dashed line is a linear fit in (a), and a square root fit in (b).

cases), while the meridional spread exhibits diffusive behavior with a $\sim \sqrt{t}$ growth after an initial transient of a couple of days.

It is interesting to compare the results of Fig. 5 with an alternative global measure of cloud size, the mean standard deviation of particle distances determined from the instantaneous center of mass. The results are plotted in Fig. 6. As in the case of the angular spread, the zonal mean standard deviation of particle distances increases linearly in time ($\approx 6^{\circ}/day$ on the 315 K isentropic surface), while the meridional spread is $\sim \sqrt{t}$. This is in agreement

with other studies [Pierrehumbert & Yang, 1993; Huber *et al.*, 2001].

Note that the measures of cloud extent defined by the edges or the standard deviation from the instantaneous center of mass provide a sufficient characterization for objects of compact shapes. However, tracer "clouds" in the atmosphere have a filamentary, fractal structure [Newell *et al.*, 1992; Pierrehumbert & Yang, 1993; Yang & Pierrehumbert, 1994] (see Fig. 1), thus other measures might easily have different time evolutions. Such a measure, the covered area fraction A is determined



Fig. 6. Time evolution of the mean standard deviation of (a) zonal (σ_{θ}) and (b) meridional (σ_{ϕ}) distances measured from the center of mass. 96 isentropic experiments were evaluated in 2009 with 130 particles, and initial cloud size $10^{\circ} \times 10^{\circ}$ (therefore the initial value is $\sigma_0 = 2.8767^{\circ}$ in both directions). White dashed line denotes (a) linear, and (b) square root fit from day five.



Fig. 7. Time evolution of the covered area fraction of a $10^{\circ} \times 10^{\circ}$ cloud started from 45° N, 0° , 315 K, 01/01/2009, 6.4×10^{5} particles. Light green line represents the covered area fraction calculated from the regular latitude-longitude grid, while dark green line illustrates that from the equalarea tiling (3). Dotted lines illustrate asymptotic diffusive tendencies ($\sim \sqrt{t}$), the inset illustrates linear growth during the first month.

by the number of cells containing at least a single particle, normalized by the total cell-number of the global mesh. Similarly to the laboratory experiments [Jánosi *et al.*, 2010], ballistic (linear) growth appears on an intermediate time interval of 10–20 days, see the inset in Fig. 7. The longtime growth after about two encompassing times (\sim 30 days) is close to be diffusive, as expected. The numerical value of covered area fraction depends on the definition of the cells, but in the first two months and after 5–6 months (by the time the particles of the cloud have covered almost both hemispheres) the difference between the curves is small. At intermediate times the difference is due to the fact that the particles of the cloud reach the polar region, where the size of the cells is quite different in the two tilings. Naturally, in the regular latitude-longitude mesh the particles can fill more lower number cells near the pole, than in the equal-area tiling, when only few particles are advected there.

In Fig. 7 one can see that after about 25– 30 days the extent of the cloud increases slower than before. Then, after 50–60 days, it grows again approximately as fast as in the first month. The cause of this slowdown may be that after about a month the cloud covers almost the whole Northern Hemisphere ($A \approx 40$ –50%), but when the particles cross the equator and spread also in the Southern Hemisphere the spreading is rather slow due to the trade-wind belt. When they ultimately cross the barrier between the tropics and extratropics [Pierrehumbert & Yang, 1993] also in the Southern Hemisphere, the covered area fraction increases again faster due to the stronger mixing effect of the midlatitude cyclones.

In Fig. 8, the circumnavigation (or zonal return) time t_c for each individual tracer particle



Fig. 8. (a) Empirical probability density distribution of the circumnavigation time t_c for three numerical experiments at 500 hPa with 10³ tracers and temporal resolutions of 6 min, 15 min, and 30 min (Euler integration). The initial condition was a uniform spatial distribution along the longitude 0° in the range of [80°S, 80°N] latitude. (b) The same as (a) on the isentropic level of 315 K with two time steps 22.5 min and 45 min (fourth order Runge–Kutta integration). In both cases, the decay obeys power-law for large t_c values (note the double-logarithmic scale).



Fig. 9. Asymptotic empirical probability density distribution of the latitude of return on an inverted scale. Seven different numerical experiments are performed with different spatial interpolation schemes, integration time-steps, and cluster sizes (see legend). The initial condition was a uniform meridional distribution along the longitude 0° in the range of [80°S, 80°N] latitude. (a) 500 hPa, and (b) 315 K.

is evaluated. The initial configuration was a material line of uniformly distributed tracers at 0° (longitude) over the latitude range of 80°S-80°N. The circumnavigation time t_c has a very broad distribution, as illustrated in Fig. 8. Note that the encompassing time t^* of a dye cloud is related to the width of the distribution of the single-particle circumnavigation time t_c , which reflects the difference between the fastest and slowest tracers in a given cloud. This quantity has a definite latitude dependence, as expected, because the meridional drift is a relatively slow diffusive process, and obviously a full circle takes a longer time near the equator than close to the poles. Note the consistency of the normalized histograms obtained at different time steps of the numerical integration. The tail of the empirical histograms obeys power-law decay with an exponent ≈ -2.9 for the isobaric, and ≈ -2.5 for the isontropic case. Due to the heavy tail, the mean value circumnavigation time $\langle t_c \rangle$ is larger than the mode (the most probable value) \hat{t}_c . It is remarkable that the mean value is almost the same $\langle t_c \rangle \approx 23.5$ days for isobaric, and $\langle t_c \rangle \approx 23.2$ days for isentropic trajectories), while the modes differ by almost a factor of two: $\hat{t}_c \approx 18.7$ days for isobaric, and $\hat{t}_c \approx 11.4$ days for isentropic runs.

Figure 9 illustrates an interesting focusing effect in the atmosphere: tracers that started from a uniform meridional distribution have a tendency to drift towards the mid-latitudes on both hemispheres during the advection. The results of seven experiments with different integration time-steps and spatial interpolation methods (inverse quadratic and cubic spline) are plotted together to demonstrate again statistical consistency. A cumulative distribution is obtained during one year of advection, by recording the latitude at each crossing of 0° longitude. A similar focusing effect was reported in [Pierrehumbert & Yang, 1993] by simulating trajectories over isentropic surfaces, however they studied tracer density distributions at fixed time differences.

4. Batchelor's Scaling

The motion of an incompressible turbulent fluid is described by the Navier–Stokes equation amended by the condition of zero divergence. According to Kolmogorov's (K41) similarity theory [Pope, 2000], the largest spatial and temporal scales are given by the energy-injection length scales \mathcal{L} and eddy turnover time $T_{\mathcal{L}}$, respectively, while the smallest scales are the Kolmogorov length $\eta = (\nu^3/\epsilon)^{1/4}$ and the Kolmogorov time $\tau = \sqrt{\nu/\epsilon}$ (where ν is the kinematic viscosity, and ϵ is the constant energy dissipation rate per unit mass). The interval (η, \mathcal{L}) is known as the inertial subrange, because viscous dissipation becomes important only at length scales $l < \eta$.

Theoretical description of the mean square separation between two fluid elements $\langle r(t)^2 \rangle$ in the inertial subrange is dated back to 1926, when Richardson suggested that it should grow in time as t^3 [Bourgoin *et al.*, 2006; Ouellette *et al.*, 2006]. Obukhov specified that in homogeneous and isotropic 3D turbulence, the pair dispersion should grow as $\langle r(t)^2 \rangle = g \epsilon t^3$, where g is a universal constant. Batchelor refined this work [Batchelor, 1950] by considering the role of initial separation $r_0 \equiv r(t=0)$, and concluded that the mean square separation should grow as t^2 for times shorter than a characteristic timescale $t_0 = (r_0^2/\epsilon)^{1/3}$:

$$[\langle r(t)^2 \rangle - r_0^2] = \frac{11}{3} C_2 (\epsilon r_0)^{2/3} t^2, \qquad (2)$$

where $C_2 \approx 2.13$ is the scaling constant for the second-order Eulerian velocity structure function [Bourgoin *et al.*, 2006], and $t \ll t_0$. In the classical K41 theory of turbulence, t_0 may also be identified as the time for which the divergence of two fluid elements is determined by their initial relative velocity $\Delta \mathbf{v}_0$ in a given eddy of size r_0 . For times $t > t_0$, the growth of the pair separation is expected to follow the Richardson–Obukhov scaling, independently of r_0 . When the separation exceeds the size of largest coherent structures, diffusive dispersion is expected $\langle r(t)^2 \rangle \sim t$ (Taylor regime). In statistically stationary forced homogeneous and isotropic twodimensional turbulence, the Richardson–Obukhov t^3 scaling holds, however experiments found various empirical exponent values between 2 and 3 [Salazar & Collins, 2009; Tabeling, 2002].

An essential condition behind the validity of the Batchelor scaling Eq. (2) is the lack of correlations between the initial separation \mathbf{r}_0 and the relative velocity of the pair $\Delta \mathbf{v}_0$ [Ouellette *et al.*, 2006]. When this condition is not fulfilled, the correct scaling form of relative pair separation is based on the vectorial difference as

$$\langle |\mathbf{r}(t) - \mathbf{r}_0|^2 \rangle = \frac{11}{3} C_2 (\epsilon r_0)^{2/3} t^2.$$
 (3)

Note that these Batchelor forms bear important consequences related to the evaluation of experimental data. (i) For fixed energy dissipation rate per unit mass ϵ , the rate of pair separation depends on the initial value r_0 itself. (ii) The transition time to the Richardson–Obukhov scaling t_0 depends also on r_0 . This means that one cannot observe a "clean" dynamics in a cloud of tracers with a mixture of various initial separations. (iii) Identification of the Richardson–Obukhov t^3 scaling requires also a significant time scale separation between t_0 and the eddy turnover time $T_{\mathcal{L}}$ [Ouellette *et al.*, 2006]. Since the latter quantity is estimated about 3–5 days in the troposphere (considering scales of a typical midlatitude cyclone) [Gerber & Vallis, 2007], and values for ϵ are around $10^{-5} \text{ m}^2/\text{s}^3$ [Lilly, 1983], an initial pair separation of $r_0 \approx 300{-}500 \text{ km}$ easily produces t_0 more than a day.

Here we show representative results for 96–96 isobaric and isentropic dispersal simulations, where the average pair separation is determined by both the scalar and the vectorial definitions Eqs. (2) and (3). Individual clusters (12 locations and 8 starting time instants, as before) are formed by 130 regularly spaced tracers over a geographic area of $10^{\circ} \times 10^{\circ}$. The time dependence of pair separation was determined in initial distance bins of 50 km, where the central value of the bins are indicated in the legend of Figs. 10(a), 10(c) and 10(e) (thus bin "75 km" denotes $50 \text{ km} < r_0 < 100 \text{ km}$, etc., except the largest value where "975 km" means $r_0 > 950 \text{ km}$).

A special feature emerges for the vectorial difference in Eq. (3), because pair dispersion occurs over a spherical surface. The definition for $[\mathbf{r}(t) - \mathbf{r}_0]$ is unambiguous in 3D Euclidean sense, however pair distances must be determined along geodesics in the atmosphere at large enough separations. To avoid computational discrepancies, we adopted the following convention. One of the points of each pair is selected, and the angles are calculated in the comoving Euclidean frame of reference (zonal x and meridional y axes) with the origin at this point. The lengths of vectors are calculated along geodesics. We checked that the statistics conforms with Eq. (2) where scalar separations were computed by pure spherical geometry.

The results in Figs. 10(a), 10(c) and 10(e) are consistent with the Batchelor hypothesis over almost two decades of the time axis: the mean squared separation increases with t^2 , and the slopes depend on the initial value r_0 . We emphasize that the behavior is almost the same for both the scalar and vectorial distances [Eqs. (2) and (3)], apart from numerical values for the slopes shown in Fig. 10(f). The empirical scaling with r_0 is somewhat surprising, because the found exponent $\sim 5/3$ is very far from the 2/3 of the theory [see Eqs. (2) and (3)], and this cannot be a simple consequence of numerical inaccuracies or statistical errors. We are fully aware of the fact that large scale dispersal in the atmosphere (and in the laboratory tank) can



the empirical scaling exponent is 1.66 ± 0.02 . (c) The same as (a) determined by the scalar pair separations [Eq. (2)]. (d) Fitted slopes as in (b), the empirical scaling exponent is 1.59 ± 0.03 . (e) The same as (c) for 96 isentropic dispersal experiments, scalar pair distances [Eq. (2)]. (f) The blue symbols and lines are the same as in for different initial separations (see legend, curves from top to bottom). The dashed line illustrates t^2 scaling. (b) Fitted slopes as a function of initial separation r_0 , (b) and (d), the green symbols are the slopes obtained from (e). The empirical scaling exponent for the latter is 1.67 ± 0.03 .

be different from what one expects in 3D homogeneous and isotropic turbulence, still the theoretical predictions concerning the time dependence are very similar in both cases.

5. Summary

Here we summarize our findings for the Batchelor scaling in a compact form.

- Important results of the numerical advection tests are that the mean pair separation strongly depends on the initial value r_0 and follows a ballistic growth [Figs. 10(a), 10(c) and 10(e)], similarly to the Batchelor hypothesis. A relevant consequence is that an extended tracer cloud containing different values of r_0 exhibits always a mixed behavior, therefore a direct relationship between pair separation and overall statistics cannot be easily formulated.
- The slopes of the mean pair separation squares versus t^2 lines do not follow the Batchelor scaling, the apparent exponent value is $\approx 5/3$ [Fig. 10(f)]. In contrast to Eqs. (2) and (3), simple dimensional considerations with this exponent suggest a combination like $U^{-3}(\epsilon r_0)^{5/3}$, where U has a dimension of velocity. However, the appearance of such a factor seems a challenge. Further work is needed to explain these observations.
- The quasigeostrophic turbulent wind field is far from being homogeneous and isotropic, large scale (irregular) eddies determine the dynamics. In addition, reanalysis fields do not resolve structures below the grid spacing. Consequently, the collective dynamics of an extended tracer cloud is spoiled by pairs of initial separation below gridsize. An obvious sign of this effect is the appearance of unrealistically long initial transients at too small cloud sizes, where the "true" dynamics unfolds only when the mean pair separation definitely exceeds the cell size as also reported in [Jánosi et al., 2010]. This observation might help to clarify earlier results on scaling behavior of tracers simulated in wind fields of limited resolutions [Huber et al., 2001; Pierrehumbert & Yang, 1993].
- In Sec. 3 of the 1950 key paper by Batchelor (The applicability of the similarity hypotheses to the turbulence in the atmosphere and ocean), a very conservative estimate for the atmospheric inertial subrange spans between 0.002–100 m. Indeed, this is far from the scales we used in

our simulations, where the lower limit is determined by the spatial resolution of the global wind field, and the upper limit is given by the estimate of typical eddy turnover time by Gerber and Vallis [2007]. Note however, that homogeneity and isotropy are statistical terms, and any given instantaneous snapshot of a turbulent field exhibits coherent structures on an extremely wide range. For the very reason, scaling laws can be formulated only for mean quantities (individual pair separation trajectories have fluctuation in a wide range clearly seen in Figs. 4–6). Our observation of the Batchelor type scaling is purely empirical, we did not use any additional assumption or filtering, apart from taking care of the initial pair separation. It is highly possible that the anomalous exponent value is an inherent property of quasigeostrophic turbulence, however we cannot give an easy explanation.

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