The Horizontal Spectrum of Vertical Velocities near the Tropopause from Global to Gravity Wave Scales

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ABSTRACT

Vertical motions are fundamental for atmospheric dynamics. Compared to horizontal motions, the horizontal spectrum of vertical velocity w is less well known. Here, w spectra are related to spectra of horizontal motions in the free atmosphere near the tropopause from global to gravity wave scales. At large scales, w is related to vertically averaged horizontal divergent motions by continuity. At small scales, the velocity energy spectra reach anisotropy as in stably stratified turbulence. Combining these limits approximates the w spectrum from global to small scales. The w spectrum is flat at large scales when the divergent spectrum shows a -2 slope, reaches a maximum at mesoscales after transition to -5/3 slopes, and then approaches a fraction of horizontal kinetic energy. The ratio of vertical kinetic energy to potential energy increases quadratically with wavenumber at large scales. It exceeds unity at small scales in stratified turbulence. Global and regional simulations and two recent aircraft measurement field campaigns support these relationships within 30% deviations. Energy exchange between horizontal and vertical motions may contribute to slope changes in the spectra. The model allows for checking measurement validity. Isotropy at large and small scales varies between the datasets. The fraction of divergent energy is 40%-70% in the measurements, with higher values in the stratosphere. Spectra above the tropopause are often steeper over mountains than over oceans, partly with two -5/3 subranges. A total of 80% of w variance near the tropopause occurs at scales between about 0.5 and 80 km.

1. Introduction

Horizontal wind and temperature measurements from long-distance commercial aircraft flights at midlatitudes in the upper troposphere and lower stratosphere show a characteristic spectrum of atmosphere dynamics (Nastrom and Gage 1985; Lindborg 1999). Not all measurements show the same characteristics (Gao and Meriwether 1998; Cho et al. 1999), but global high-resolution atmosphere models show similar canonical spectra (Hamilton et al. 2008; Burgess et al. 2013; Skamarock et al. 2014; Malardel and Wedi 2016). The large-scale geostrophic motions are interpreted as quasi two-dimensional (2D) with kinetic and potential energy spectra exhibiting spectral steepness near -3 while the mesoscale motions often show a flatter spectrum, with slopes near -5/3 (Gage and Nastrom 1986). The slope

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transition scale L occurs between about 300 and 800 km wavelengths, differing in the stratosphere and troposphere (Bierdel et al. 2016). The quasigeostrophic motions at scales $\lambda > L$ are mainly rotational with low horizontal divergence (Charney 1971; Gill 1982). Mesoscales ($\lambda < L$) have about equal shares of rotational and divergent horizontal motions (Blažica et al. 2013; Lindborg 2015). Finally, turbulence at small scales is near isotropic and mainly rotational (Tennekes and Lumley 1972). The canonical spectrum is relevant, for example, for numerical weather predictability, which is lower for a flatter spectrum because of shorter turnover time scales (Lorenz 1969). Considerable progress has been made in the understanding of these spectra, suggesting multiple reasons for specific spectral shapes (Li and Lindborg 2018; Selz et al. 2019).

The horizontal spectrum of vertical velocity has been discussed relatively little (Craig and Selz 2018), mainly because of high variability, missing universal theory, and only few measurements of w spectra covering the mesoscales (Bacmeister et al. 1996; Gao and Meriwether 1998; Callies et al. 2016). Vertical wind drives conversion between potential and kinetic energy, gravity

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waves, and turbulence at increasingly smaller scales (Lorenz 1960; Fritts and Alexander 2003), fundamental for atmospheric dynamics. Together with the spectrum of horizontal motions, the w spectrum describes the atmospheric flow energetics as a function of spatial scales (Skamarock et al. 2019), determines the required resolution in numerical models (Frehlich and Sharman 2008), and the amount of subgrid-scale contributions (Lilly 1967). The spectrum may influence cloud microphysics (Barahona et al. 2017; Kärcher et al. 2019) and horizontal cloud scales. Frequency spectra [e.g., from radar (Li et al. 2018) and balloons (Podglajen et al. 2016)] may be comparable to wavenumber spectra when Taylor's hypothesis of frozen motions applies (Bacmeister et al. 1996; Craig and Selz 2018). Vertical wind at large horizontal scales is mostly weak and difficult to determine (Holton 2012; Bony and Stevens 2019). Models and observations often show a maximum of w variance at the smallest resolved scales, leaving the ultimate explanation of this maximum open (Sharman et al. 2012; Zentek et al. 2016; Smith and Kruse 2017).

The atmosphere in the upper troposphere and lower stratosphere is often strongly stratified, with maximum stratification in the tropopause inversion layer (Birner et al. 2002; Gisinger et al. 2017). Stratified turbulence exhibits vertical velocities often far smaller than horizontal ones and layered horizontal motions with strong vertically rotational horizontal components (Lilly 1983). Horizontal and vertical kinetic and potential energies of linear inertial gravity waves are connected by dispersion laws at frequencies between the Coriolis and the Brunt-Väisälä frequency (Gill 1982). Strong vertical winds are observed regionally in nonlinear gravity waves and turbulence, for example, over mountains and convection (Sharman et al. 2012). Mountain waves can be simulated numerically in agreement with observations (Smith and Kruse 2017). For small-scale turbulence, at scales smaller than the Ozmidov scale (Riley and Lindborg 2008), the spectrum of vertical and horizontal velocities is linked by isotropy and mass continuity (Tennekes and Lumley 1972).

Recently, wind and temperature have been measured in long-distance flights with well-instrumented jet aircraft operating in the lower stratosphere and upper troposphere during field experiments over the South Pacific near New Zealand (Fritts et al. 2016) and over the North Atlantic near Iceland (Schäfler et al. 2018). Accurate wind measurements are demanding in particular for fast-flying jets because the wind vector is the relatively small difference between the measured velocity vector of the aircraft with respect to Earth and the measured velocity vector of the air with respect to the aircraft (Lenschow 1986). The measurements combine position and attitude data from inertial navigation platforms and

satellite navigation systems, and air velocity and angleof-attack data from static and dynamic pressure and temperature sensors at the aircraft fuselage and nose (Cooper et al. 2014; Mallaun et al. 2015; Giez et al. 2017). These techniques allow to derive spectra of vertical energy and momentum fluxes in mountain gravity waves (Smith et al. 2016). Occasionally, w spectra were observed for selected flight legs showing peaks at wavelengths of order 4–12 km, which are partially explainable with mesoscale models (Smith and Kruse 2017). Aircraft typically do not fly steady and straight, but undergo ridged body oscillations at frequencies between 0.01 and 0.1 Hz. The vertical body velocity is often larger than vertical air velocity, depending on aircraft design, speed, and autopilot control (Nelson 1998). The aircraft oscillations may be interfering with wind measurements at 4-10km wavelengths. Aircraft measurement disturbances and vertical motions at constant pressure levels were noted before, but regarded as unimportant for horizontal velocity spectra at large wavelengths (Nastrom and Gage 1985; Gao and Meriwether 1998; Frehlich and Sharman 2010). As we will show, some w spectra show maxima ("peaks") at horizontal scales corresponding to flight height, in particular near the tropopause over mountains. A wide maximum in w spectra at about 10– 20 km wavelengths was observed also in airborne Doppler lidar observations of tropospheric mountain waves (Witschas et al. 2017). In other cases, in particular over oceans, the measured w spectra are flatter without such a maximum (Gao and Meriwether 1998).

In this paper we suggest a hypothesis that connects the horizontal spectrum of vertical kinetic energy with horizontal spectra of horizontal kinetic and potential energy and discuss possible implications. Published model results (Skamarock et al. 2014; Selz et al. 2019) and an analysis of two recent airborne field campaigns (Fritts et al. 2016; Schäfler et al. 2018) are used to discuss our understanding of the *w* spectra from models and related measurements.

2. The hypothesis

Vertical wind w is obviously connected to horizontal wind components u and v by mass continuity, which for an incompressible fluid in Cartesian coordinates (x, y, z) implies,

$$\frac{\partial w}{\partial z} = -\left(\frac{\partial u}{\partial x} + \frac{\partial v}{\partial y}\right). \tag{1}$$

Integrating Eq. (1) over height from the ground (z = 0) to a height z = h shows that w(h) is related to horizontal winds kinematically (Holton 2012):

$$w(h) - w(0) = -\int_0^h \left(\frac{\partial u}{\partial x} + \frac{\partial v}{\partial y}\right) = -h\left(\frac{\partial \overline{u}}{\partial x} + \frac{\partial \overline{v}}{\partial y}\right). \quad (2)$$

Here, the bars in \overline{u} and \overline{v} denote the vertical average. We start with the simple case of zero surface height and w(0) = 0. The vertical wind w at ground drops from the mass budget for vertically impermeable rigid ground, but Eq. (2) would require the height h within the differentials for variable surface height. Also density variations from adiabatic pressure changes and from internal heat sources are generally important (Egger and Hoinka 2019). We leave variable density and orography for later discussion. As noted by Margules in 1904 (Lynch 2003), vertical wind derived from the continuity equation is very error prone at large scales (Holton 2012). The Helmholtz decomposition (Wippermann 1957; Lorenz 1960; Koshyk and Hamilton 2001) is therefore essential for this approach. It allows splitting the horizontal wind vector into a rotational and a divergent part. By definition, only the divergent part contributes to the vertical wind. Equation (2) implies that Fourier modes \hat{w} in $w = \sum \hat{w} \exp(i\mathbf{k}\mathbf{x})$ are related to Fourier modes of \overline{u} and \overline{v} with derivatives in the divergence term transforming into multiplications with the corresponding wavenumber components. Hence, the kinetic energy of w(h) in wavenumber space is related to the sum of the energies of \overline{u} and \overline{v} :

$$\hat{w}\hat{w}^* = h^2 \left[k_x^2 \widehat{u} \widehat{u}^* + k_x k_y (\widehat{u} \widehat{v}^* + \widehat{v} \widehat{u}^*) + k_y^2 \widehat{v} \widehat{v}^* \right], \quad (3)$$

with asterisks indicating complex conjugate values. Because of missing information on the scaling of anisotropy (Nastrom et al. 1997; Bühler et al. 2017), we assume horizontal isotropy (vertical axisymmetry) in which the spectra depend only on the wavenumber magnitude $k = (k_x^2 + k_y^2)^{1/2}$ with zero mean correlations between \widehat{u} and \widehat{v} . Let E_w and E_D be the ensemble-mean (denoted by angle brackets) spectral energy densities per wavenumber interval dk of $\langle \hat{w}\hat{w}^* \rangle/2$ and of the vertically averaged divergent horizontal velocities $\langle \widehat{u}\widehat{u}^* + \widehat{v}\widehat{v}^* \rangle/2$ integrated over rings $2\pi k$ dk at given h. Because of Eq. (3), these spectra are related as

$$E_{w}(k,h) = h^{2}k^{2}E_{D}(k,h).$$
 (4)

This relationship should be exact in a horizontally isotropic atmosphere with constant density at height h over a flat surface. To allow for applications to measured or modeled spectra of divergent kinetic energy $E_d(k,h)$ at given height h, we relate $E_D(k,h)$ to $E_d(k,h)$ at this height. The horizontal spectra E_D of vertically averaged velocities and E_d of velocities at a fixed height differ in magnitude. Their ratio may be expressed by a filter function:

$$E_D(k,h) = F(hk)E_d(k,h). \tag{5}$$

Here, we suggest an approximate filter function:

$$F(kh) = \frac{1}{\beta^{-2} + \alpha^{-1}(h\,k)^2},\tag{6}$$

with two yet open parameters α and β . Inserting Eqs. (5) and (6) into Eq. (4) results into

$$E_{w}(k,h) = \frac{\alpha (h_{\text{eff}} k)^{2}}{\alpha + (h_{\text{off}} k)^{2}} E_{d}(k,h).$$
 (7)

The filter function is constructed such that Eq. (7) simplifies for very small and very large wavenumbers. In particular,

$$E_{w}(k,h) = (h_{\text{eff}} k)^{2} E_{d}(k,h), \quad hk \ll 1$$
 (8)

at small wavenumbers, independent of α . In the opposite limit.

$$E_{w}(k,h) = \alpha E_{d}(k,h) = \alpha d E_{h}(k,h); \quad hk \gg 1, \quad (9)$$

independent of β . Here we have introduced $E_d = dE_h$, where $E_h(k, h)$ is the measurable spectrum of horizontal kinetic energy (u and v) and d an estimated fraction of divergent motions in this spectrum.

The physics behind this filter function is that mass continuity controls the relationship between E_w and E_D , Eq. (4). For a vertically uniform flow in a layer of depth h with long-wave motions, $E_D \approx E_d$. For vertically variable long-wave motions, E_D differs from E_d by the factor β^2 . The factor will be the smaller the more variable the horizontal divergent layered motions. Here, the effective depth controlling w is $h_{\text{eff}} = \beta h$; $h_{\rm eff}$ measures the depth of layers with effectively uniform divergent flow. At large heights and large wavenumbers, E_w is no longer controlled by mass continuity from the layer mean divergence but more by local dynamics. We assume that this dynamics, in the ensemble average over many realizations, tends to approach near isotropy, so that E_w approaches αE_d up to a remaining anisotropy of factor α , but without the factor $(hk)^2$. These properties are approximated by the filter function F. The ratio d varies with k and h and with the flow dynamics because divergent and rotational motions follow different conservation laws (Wippermann 1957; Tennekes and Lumley 1972). The value of d is small for low wavenumbers and is variable and on the order of 0.5 at mesoscales (Li and Lindborg 2018).

Subsequently we relate α and β to known properties of the velocity field. The ensemble mean kinetic energy E_D of vertically averaged horizontal wind (e.g., from \overline{u}),

is related to the vertical correlations between winds at altitudes z' and z'' by

$$\langle \overline{u}^2 \rangle = \left\langle \left[\frac{1}{h} \int_0^h u(z) \, dz \right]^2 \right\rangle = \frac{1}{h^2} \int_0^h \int_0^h \langle u(z') u(z'') \rangle \, dz' \, dz'', \tag{10}$$

and similarly for \bar{v} . At large scales, $E_D/E_d \approx \beta^2$, see Eqs. (5) and (6). Hence, β^2 could be computed for known vertical profiles of horizontal wind or for given ensemble-mean correlations. For example, $\beta = 1/2$ for a wind profile increasing linearly from 0 at ground. For vertically more variable divergent wind fields, β can be considerably smaller. The correlations are the larger the steeper the spectra of horizontal velocity fluctuations decay with vertical wavenumber, causing large β for geostrophic motions with -3 spectra in both the horizontal and vertical wavenumbers (Charney 1971).

The values of $\alpha = E_w/E_d$ and $\alpha d = E_w/E_h$ depend on the high wavenumber spectra; see Eq. (9). For given E_h , E_w depends on the product αd ; see Eq. (9). In incompressible fully turbulent inertial-subrange turbulence (Kolmogorov 1941), E_w/E_h approaches a fixed value of 4/7 (Hinze 1959). For linear hydrostatic gravity waves in uniformly stratified atmospheres for high intrinsic frequencies near the Brunt–Väisälä frequency, E_w/E_h approaches one; see the appendix. However, in the atmosphere isotropic turbulence and linear gravity waves occur only occasionally. Instead, the free atmosphere is mostly stably stratified with occasional turbulent patches and $E_w/E_h < 1$ (Riley and Lindborg 2008).

From a fit of the model to the data from all individual flights presented below, for fixed d=1/2, we find mean values and standard deviations of $\alpha=0.5\pm0.2$ and $\beta=0.11\pm0.09$. From comparisons to global and regional models we find that β varies with height. Our hypothesis is that Eq. (7), and the concept behind it, approximate the relationship between vertical and divergent horizontal motions, and that $\alpha=0.5$ and β between 0.05 and 0.5 provides useful approximations for a range of applications.

The hypothetical spectral relationship is illustrated in Fig. 1, which shows schematically horizontal energy spectra for divergent and total (divergent plus rotational) horizontal velocities and for vertical velocity w. The horizontal energy spectra are sketched following simulated and observed results (Chen and Tribbia 1981; Nastrom and Gage 1985; Skamarock et al. 2014). Because of Eq. (7), the slope of the w spectrum equals the slope of the divergent energy spectrum plus two at large scales, and both spectra approach equal slopes at small scales. Any slope change in the divergent horizontal and the vertical velocity spectra occurs at about the same

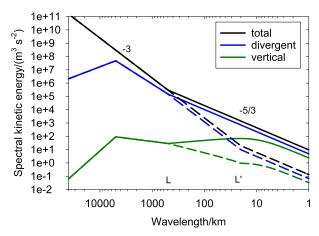


FIG. 1. Schematic of kinetic energy spectra of total and divergent horizontal velocities E_h and E_d and of vertical velocity E_w for a case with transition from -3 to -5/3 spectra in E_h at wavelengths $L=500\,\mathrm{km}$ (solid lines) and at $L'=20\,\mathrm{km}$ (dashed) with $h=10\,\mathrm{km}$, $\alpha=1/2$, and $\beta=0.11$.

transition scale L. The w spectrum starts small at the longest wavelengths (near the Earth's circumference), may reach a local maximum at wavelengths $\lambda < L$ of baroclinic waves with local maximum in the spectrum of divergent velocities, gets flat for divergent energy spectrum with slope -2, and approaches near isotropy at small scales. A maximum in w spectra is to be expected at mesoscales if the divergent wind spectrum has a -5/3 slope, and such a maximum should be absent for a divergent wind spectrum steeper than -2. For a wide -5/3 divergent spectrum, the vertical wind may reach a 1/3 slope. The mesoscale maximum is higher and wider for larger spectral slope transition scale L (solid curves) than for smaller L' (dashed curves). A more variable horizontal divergence spectrum may be accompanied by peaks in the w spectrum. Broad maxima in the w spectrum extending to the smallest mesoscales contribute strongly to the integral $\int E_w dk$ (i.e., total vertical velocity variance). In contrast, flat w spectra at large scales (small k) contribute little to the total velocity variance.

It remains to be seen how well this hypothetical model fits reality in the vertical energy carrying scales. Since the horizontal spectra at one height h cannot fully characterize the flow field below or above in its 3D details, we cannot expect that Eq. (7) is valid for individual cases. The given uncertainties become important in particular for low vertical correlations of the divergent horizontal motions at various heights, implying low β , and for events departing strongly from the assumed near isotropy at small scales. It will be interesting to see, how large the ensembles have to be to obtain statistically robust correlations. To test the hypothesis, we compare

the "modeled spectrum" from Eq. (7) with reference results from previous numerical simulations in section 3 and with aircraft measurements in section 4.

3. Comparisons with global and regional simulated spectra

The first test uses spectra derived with the Model for Prediction Across Scales (MPAS) in a global highresolution atmospheric simulation with 3 km grid scales (Skamarock et al. 2014). The published results include plots of 10-day-mean 2D energy spectra versus spherical wavenumber for the rotational, divergent and total kinetic energy from horizontal velocities and of the vertical velocity, separately in the troposphere (h =8.5–10.5 km) and stratosphere (16–18 km), which we copied into Fig. 2. The top panel shows that the total kinetic energy spectra from MPAS exhibit the canonical spectrum structure with approximately -3 slopes at large scales and -5/3 slopes at mesoscales, making these spectra suitable for our study. The large-scale motions are dominated by energy from rotational motions, as expected (Gill 1982). The divergent part gets increasingly important at larger wavenumbers.

The divergent MPAS spectra are used as input for estimating the w spectra, using Eq. (7) with $\alpha = 1/2$ and $\beta = 0.5 \text{ and } 0.05, h = 9.5 \text{ and } 17 \text{ km } (h_{\text{eff}} \approx 5 \text{ and } 1 \text{ km}),$ for the troposphere and stratosphere. For the selected β values, the model and the MPAS spectra have about same integral energy $\int E_w dk$. The results are plotted together with the MPAS w spectra in the bottom panel of Fig. 2. The modeled w spectra show shapes similar to the reference spectra. In particular, both have a wide maximum at high wavenumbers where the divergent horizontal velocity spectrum has a rather flat slope, with steepness between -2 and -5/3. The modeled spectra also show, at least in the troposphere, a weak maximum value between global and synoptic scales near the spherical wavenumber 10 (wavelength = $2000 \,\mathrm{km}$). The MPAS spectra decay rather steeply at high wavenumbers because of numerical dissipation. The modeled w spectrum decays similarly. The model spectrum is weakly sensitive to α because of small kh in the MPAS data. The stratospheric β value is considerably less than one, as expected because the stratospheric wind is often weakly correlated with tropospheric winds. Perfect agreement could not be expected because the atmosphere is neither homogeneous nor horizontally isotropic. Also the model does not account for density variations and contributions from surface orography.

A second test uses spectra obtained from a regional numerical weather prediction model COSMO-DE applied for Europe (Selz et al. 2019) for a domain of about

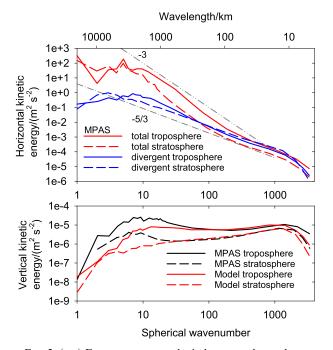


FIG. 2. (top) Energy spectra vs spherical wavenumber and wavelength for total (red) and divergent (blue) velocity components from the 3 km MPAS global simulation for the troposphere (solid) and the stratosphere (dashed), copied from Fig. 2 in Skamarock et al. (2014). Dashed–dotted lines represent the -3 and -5/3 slopes. (bottom) The w spectra in the troposphere (solid) and stratosphere (dashed). The black curves are copies from Fig. 5 in Skamarock et al. (2014), and the red curves are from the present model, Eq. (7).

 $1200 \,\mathrm{km} \times 1300 \,\mathrm{km}$ with 2.8 km horizontal resolution. The analysis data were obtained from the German Weather Service for 3 years, 2014–16, with 8 time steps per day at 3 h intervals, 8768 time steps in total. Mean 2D spectra of kinetic energy for 200, 300, and 500 hPa have been extracted from the data by T. Selz (2019, personal communication), see Fig. 3. The total horizontal kinetic energy follows the canonical spectrum with about -3 slope at wavelengths <200 km and slightly steeper than a -5/3 slope at wavelengths from about 20 to $100 \,\mathrm{km}$. The kinetic energy is largest at 300 hPa. The spectrum of divergent motions follows a flatter spectrum, with slopes in between -2 and -1.8 between 2000 and 20 km wavelengths. The vertical kinetic energy spectrum is flat in that scale range, and decreases with slopes steeper than -3 between 20 and 10 km. Spatial wavelengths less than about five grid spacings or 14 km may be not well resolved by the model (Craig and Selz 2018) and therefore are not shown. The fraction of divergent energy is small at large scales and reaches a maximum between 20 and 100 km, exceeding 70% at 200 hPa and 50% at lower heights. The model, Eq. (7), has been applied with the height corresponding to the pressure levels in the standard atmosphere. The parameter β was selected

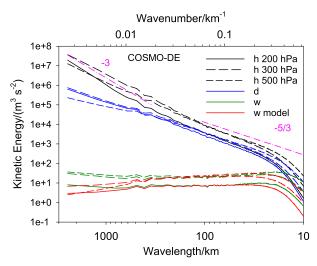


FIG. 3. Mean spectra of kinetic energy of total horizontal velocities h (sum of divergent and rotational contribution), divergent horizontal velocities d, and vertical velocity w at 200, 300, and 500 hPa (i.e., 11.8, 9.2, and 5.6 km height in the standard atmosphere) from 3 years of COSMO-DE data (Selz et al. 2019), courtesy of Tobias Selz. The dash–dotted lines are the -3 and -5/3 parts of the NGL spectrum The red curve is the model spectrum, Eq. (7), for $\alpha = 1/2$ and $\beta = 0.071$, 0.127, and 0.229, $h_{\rm eff} = 0.82$, 1.2, and 1.3 km, for 200, 300, and 500 hPa, respectively.

to best match the spectra at mesoscales. The data imply that the effective height $h_{\text{eff}} = \beta h$ decreases with height. The red curve in Fig. 3 shows that the model matches the COSMO-DE spectra over a wide range of scales (about 10–500 km) not perfectly but within about 20% deviations. At large scales, the model expects a slight increase with wavenumber because the divergent spectra decrease less than quadratic, while the COSMO-DE w spectra show a weak decrease, possibly because of the boundary conditions used in the regional model. Some of the model deviations may result from the orography, convection, and latent heat release, and the assumed 2D isotropy. The COSMO-DE w spectra at mesoscales agree reasonably with the model and best at scales from about 20 to $500 \,\mathrm{km}$. The basic hypothesis that E_w is connected to E_d at mesoscales by a factor $(h_{\text{eff}}k)^2$ is supported by this comparison.

4. Comparisons with aircraft measured mesoscale spectra

a. The aircraft measurements

For comparison to observations, we use aircraft measurements from two field campaigns: The Deep Propagating Gravity Wave Experiment (DEEPWAVE; abbreviated D) provided 25 Hz wind and temperature data (https://doi.org/10.5065/D6BG2M1H) from 26 flights

with the National Science Foundation Gulfstream GV research aircraft, operated by the National Center for Atmospheric Research, to observe gravity waves from mountains and tropospheric weather disturbances mainly in the lowermost stratosphere over New Zealand and the adjacent South Pacific (Fritts et al. 2016). The North Atlantic Waveguide and Downstream Impact Experiment (NAWDEX; abbreviated N) provided 10 Hz data (https://halo-db.pa.op.dlr.de/mission/93) from 13 flights with the German High-Altitude and Long-Range (HALO) Gulfstream G550 research aircraft, operated by the Deutsches Zentrum für Luft- und Raumfahrt, to observe weather systems in the upper troposphere and lower stratosphere in the North Atlantic, partly over Iceland and Greenland (Schäffer et al. 2018). The data from D and N contain, respectively, 98 and 44 straight and level flight sections longer than 2048 s (maxima of 6380 and 10553 s, respectively); in total 120 h, see the online supplemental material. Vertical velocity reached maximum values of 9.6 and 7.6 m s⁻¹ for D and N, with rms fluctuations of 0.45 and 0.25 m s⁻¹, respectively. Fourier 1D power spectra are computed legwise by FFT after detrending with a linear least squares fit, and enforcing periodicity by a 10% Tukey filter (Sharman et al. 2014) from legs of equal duration selected from the flight sections, partially with 50% overlap. The number of equal-duration legs varies from 178 and 107 for 2048s to 14 and 9 for 4096s legs, for D and N, respectively. Taylor's hypothesis and the speed of the airplane V relative to the air or ground are used to convert the measured time series into spatial records. In the standard analysis, V is the leg-mean true airspeed TAS. Because of the large speed ($\sim 230 \,\mathrm{m\,s^{-1}}$), TAS and ground speed (GS) are not much different. The alongand transverse wind velocity components and their spectra (E_1, E_2) are computed using leg-mean heading. Averaged angles (and heading instead of track angle) give less scatter than local values. As observed often (Schumann et al. 1995), low turbulence and strong stratification prevail in these data near the tropopause, with leg-mean Ozmidov scales < 10 m, so that the inertial subrange of isotropic turbulence is not resolved with 10 or 25 Hz data (Lilly 1983; Riley and Lindborg 2008). Nevertheless, mean dissipation rates ε are estimated to separate between high- and low-turbulence cases by fitting the Kolmogorov spectrum to the data. The fit uses data for frequencies from 0.27 to 1 Hz to avoid measurement noise at higher frequencies. Each set of legs is divided into two subsets with leg mean values below or above the median value of a selected parameter (e.g., for low and high ε). The spectra are spectral energy densities S per wavenumber interval dk, in units of m³ s⁻², with $\int S dk$ equal to the respective

leg-mean energies. The potential energy spectrum E_p is computed from half the temperature variance scaled with gravity, leg-mean temperature, and Brunt–Väisälä frequency; see Eq. (A2). Mean spectra are linear averages of the frequency spectra and converted to wavenumber spectra with $k = 2\pi f/V$ using the mean velocity V on average over the leg subsets. Significant mean spectra, with 90% confidence limits (indicated by error bars in the plots) below about 20% at the largest scales, are obtained for 2048 s legs.

b. Results and comparisons

The observed spectra for both field campaigns (D and N) and for low and high dissipation rates ϵ are shown in Fig. 4 for a wide range in the mesoscales, with wavelengths λ from ~ 500 to 0.1 km. We selected the results for 2048 s legs because they are statistically robust and include data from most of the flights in the two campaigns (see supplemental material). Before we discuss the model performance for these observations, we inspect the spectra more generally.

The magnitudes and the shapes of the measured spectra differ between low and high ε and D and N. The D data have higher variances because of a larger fraction of flights in mountain waves. The spectral slope at $10-100 \,\mathrm{km}$ wavelengths is -2.3 and -1.9 for low and high ε for D and -2.2 and -2.0 for N, steeper than -5/3. The mean spectra E_h of horizontal velocities approach the Naström-Gage-Lindborg (NGL) 1D spectrum at low wavenumbers. The transition from -3 to flatter slopes occurs in the NGL spectrum near $\lambda \approx 460 \,\mathrm{km}$ (see the appendix). So the slope of the NGL spectrum is already close to -5/3 in the scale range plotted. The Kolmogorov spectrum with the median dissipation rate derived from the data is about a factor 10 lower. We see that the separation into low and high ε values causes a clear separation of the horizontal velocity spectra at small scales but little separation at large scales. Hence turbulence dissipation is not always the dominant parameter controlling mesoscale spectra; divergence of energy fluxes contribute also. The potential energy spectrum E_p is parallel to but below the horizontal kinetic energy spectrum in the whole scale range. This will be discussed below. Both spectra seem to show two -5/3 ranges, one at lower and one at higher mesoscales, with an effective ε about a factor of 10 higher at larger scales. For D, the large-scale part is even flatter than -5/3. Such a splitting was observed before (Dewan 1979; Jasperson et al. 1990; Callies et al. 2016), also by R. D. Sharman and R. G. Frehlich (2012, personal communication) who explained this splitting by energy transfer from nonlinear gravity waves. We see that the splitting is different for D and N. Possible reasons are discussed below.

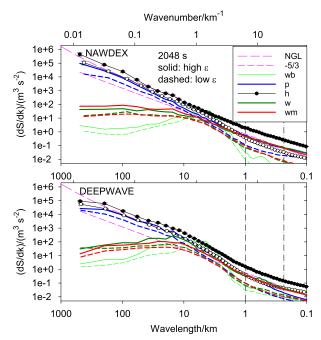


FIG. 4. Mean spectra of kinetic energy per wavenumber interval dS/dk vs wavelength λ and wavenumber $k = 2\pi/\lambda$, averaged over 20 log intervals per decade at high wavenumbers, as derived from (top) 107 NAWDEX legs (N) and (bottom) 178 DEEPWAVE legs (D) of 2048 s duration each, for dissipation rates ε below (dashed) and above (solid) the median values of 3.1 and $2.2 \times 10^{-6} \,\mathrm{m}^2\,\mathrm{s}^{-3}$ for N and D, respectively, with ε derived in the wavenumber interval indicated by vertical lines. Dashed purple lines represents the 1D Naström-Gage-Lindborg (NGL) spectrum (Lindborg 1999), and the -5/3 spectrum of Kolmogorov (1941) for median ε . Colored curves represent spectra of measured potential energy (blue), horizontal kinetic energy (from u and v) with full and open symbols and error bars for confidence intervals (black), vertical air kinetic energy w (dark green) and body kinetic energy w_b (light green), and the modeled vertical air velocity spectrum w_m (red). The model, Eq. (7), is applied for mean flight heights h (11.7 and 12.2 km for the low and high ε in D cases and 9.5 and 13.0 km for the low and high ε in N cases), d = 1/2, $\alpha = 1/2$, and $\beta = 0.11$.

The vertical aircraft body velocity spectrum shows a large peak, even exceeding horizontal velocity spectra, near $\lambda = 5.0$ and 5.8 km for N and $\lambda = 5.6$ and 6.7 km for D, at low and high ϵ . This peak is a consequence of speed and autopilot-dependent mixed phugoid and pitch mode body oscillations (Nelson 1998), occurring in turbulent and in calm air. An analysis of these oscillations needs information about nonlinear contributions from the autopilot control loops that are not available publically. The vertical air velocity spectra are generally smooth over the whole scale range in spite of the large body motions, indicating precise measurements. Still, even the averaged air w spectra show local maxima that may contain small contributions from uncertainties in the data. A more detailed analysis of these uncertainties is

beyond the scope of this paper. From now on, w spectra refer to vertical air motions.

We see that the w spectra are flat at large scales, and decay about monotonously at small scales for N, while the w spectra for D data show a wide maximum at 19 and 15 km for high ε . Splitting the ensemble of legs into subsets with high and low w variance, we find more pronounced maxima near about 10–20 km for higher w variance. For lower w variance, the spectra tend to get flatter without such local maximum. The w spectra remain significantly (factor of 2–8) below the spectra of horizontal energy even at the smallest scales, without tendency to local isotropy.

It is remarkable that the model spectrum E_{wm} shows very similar behavior. The model spectrum $E_{\rm wm}$ is the red curve in Fig. 4. It got computed from the measured horizontal velocity spectrum and leg-subset mean heights h, Eq. (7), assuming 50% contributions from divergent velocities. The mean and standard deviations of the spectral ratio E_{wm}/E_w on average over all the bins shown are 1.12 ± 0.55 and 0.91 ± 0.23 for D and 1.06 ± 0.16 and 1.05 ± 0.34 for N for low and high ε , respectively. For a split of the legs into low and high flight levels relative to the local tropopause (from numerical weather prediction data), the ratio E_{wm}/E_w is shown in Fig. 5. The mean ratio values are 0.96 ± 0.26 and 0.96 \pm 0.31 for D and 1.0 \pm 0.28 and 1.05 \pm 0.24 for N for TRO and STR in this figure. The ratio is remarkably close to one for both experiments with standard deviations below about 30% over the whole spectral range on average. The agreement would degrade, in particular at small scales, if we would replace $\alpha = 1/2$ by larger values. Hence, the observations support $\alpha d = 1/4$ more than $\alpha d = 1$ or 4/7 for linear inertial gravity waves or locally isotropic turbulence, respectively. As expected, β and the effective height $h_{\rm eff}$ may differ depending on the large-scale flow situation. For individual legs, the model/measurement ratio ranges between 0.2 and 5. The average ratio E_{wm}/E_w is 1 ± 0.3 . This is supported by analysis of spectra with other averaging times and other parameters used to separate leg subsets (besides dissipation ε : vertical air and aircraft velocity variances, height above tropopause, wind speed, head wind, crosswind, a "waviness parameter" derived from cospectra and quadraure wT spectra, Scorer parameter (Gisinger et al. 2017), spectral slope, and surface height).

The agreement is surprisingly good if one considers that these are 1D spectra, not 2D, as assumed in the derivation of Eq. (7). The contributions from divergent and vertical waves perpendicular to the flight direction are not dominant at mesoscales in these measurements. The relatively small difference between 1D and 2D

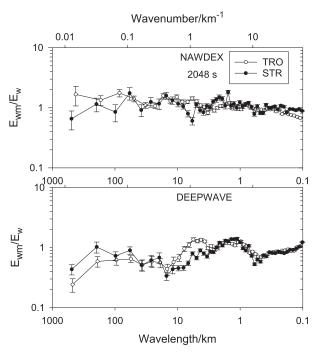


FIG. 5. Ratio of modeled to measured vertical kinetic energies for 2048 s legs from (top) N and (bottom) D for equal-sized subsets of legs with low and high height relative to the tropopause (TRO has open symbols; STR has full symbols).

spectra at mesoscales is also supported by model results (Zentek et al. 2016).

The degree of 2D isotropy may be estimated from Fig. 6 showing the fraction $E_2/(E_1 + E_2)$ of spectral energies from transversal motions E_2 relative to longitudinal and transversal horizontal contributions. In 2D or 3D isotropic turbulence, 1D spectra of longitudinal (subscript 1) and transversal (subscript 2) velocities versus longitudinal wavenumber k_1 are related by isotropy and continuity [Lindborg's (2007) Eq. (38), and Hinze's (1959) Eqs. (3)–(71)] and for power-law spectra, $E(k) \sim k^{-n}, n \ge 1$, it follows that $E_2/(E_1 + E_2) = 3/4$ in 2D for n = 3, and $E_2/(E_1 + E_2) = 4/7$ for n = 5/3 in 3D. The colored lines in Fig. 6 show these values. We see that the data from N come reasonably close to isotropy for very large and very small wavenumbers, but the alongtrack variance dominates at mesoscales in these data. So the measured fields are not truly isotropic, which in fact could hardly be expected for a limited set of flights and for the likely nonisotropic nature of large-scale atmospheric dynamics. The data from D deviate from isotropy over the whole scale range. Low $E_2/(E_1 + E_2)$ ratios may indicate enhanced E_1 from Doppler shifting in nonsteady irrotational gravity waves, which should increase with aircraft speed (Cho et al. 1999). However, the small-scale isotropy test gives about the same result regardless of whether the aircraft flew in or against the

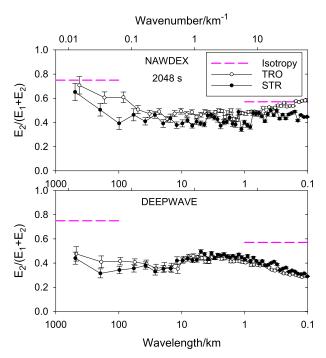


FIG. 6. Local isotropy test in terms of the ratio of horizontal transversal (E_2) to sum of transversal and longitudinal (E_1) velocity spectra for (top) N and (bottom) D and TRO (open) and STR (full), as in Fig. 5. The two horizontal purple lines represent the expected fractions in cases of locally isotropic and incompressible flow in 2D turbulence with a -3 slope at large scales and in 3D turbulence with a -5/3 spectrum at small scales.

main wind direction or perpendicular to it, so that measurement issues cannot be excluded.

The assumption that the divergent part amounts to 50% in the mesoscales needs some checking and is of general interest. In spite of the obvious deviations from isotropy and unknown deviations from 2D turbulence, the measured E_1 and E_2 spectra are used, as in previous studies (Bierdel et al. 2016; Callies et al. 2016; Li and Lindborg 2018), to estimate the divergent (E_d) and rotational (E_r) horizontal kinetic energy spectra with the Helmholtz decomposition for 2D isotropic turbulence of Lindborg (2015). Previous studies found divergence fractions from 1/3 to 2/3 (d/r ratios from 1/2 to 2), varying among the datasets used but tending to higher values in the stratosphere where high-frequency wavy motions dominate (Blažica et al. 2013; Bierdel et al. 2016; Callies et al. 2016; Li and Lindborg 2018). The results in Fig. 7 reveal spectral divergence energy fractions varying with wavenumber and datasets. The N data show a more uniform behavior and suggest a near 40% fraction in the troposphere and near 70% in the stratosphere. For D, these results are more variable; here, the method partially fails to provide physically reasonable results. The derived E_r values are partly negative because of

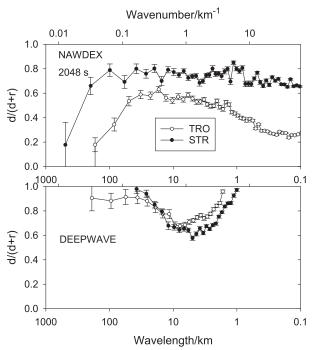


FIG. 7. Fraction of divergent horizontal kinetic energy, for (top) N and (bottom) D, and TRO (open) and STR (full), as in Fig. 5.

negative E_2 – E_1 contributions (not plotted). Overall, the data roughly support the assumption of d near 50% at mesoscales.

The ratios E_h/E_p and E_w/E_p of horizontal and vertical kinetic energies relative to potential energy are shown in Fig. 8 and compared with the 1D spectral model derived in the appendix. Equations (A4) and (A5) describe the ratio of potential energy to horizontal and vertical motions from linear gravity waves for given empirical spectra E_h and E_w . As input, we use the 2D spectrum, Eq. (A6), as in Lindborg (1999), with transition scale $L = 394 \,\mathrm{km}$, estimate the fraction of divergent energy from Eq. (A8), and estimate E_w from Eq. (7) with h from the data, $\beta = 0.11$, and $\alpha = 1/2$. The blue and red lines show the 1D spectra computed numerically from the 2D spectra. The data show a ratio E_h/E_p between 2 and 8. A ratio of $E_h/E_p > 1$ is expected for linear gravity waves and $E_h/E_p \approx 2$ for geostrophic turbulence (Charney 1971; Gage and Nastrom 1986). The model based on linear gravity waves underestimates E_h/E_p , by at least a factor of 2, presumably because of missing rotational wind components. The ratio E_w/E_p starts from very small values at global scales, increases about quadratically for scales > L, and approaches one within about 20% deviations in the data for small scales $\lambda < L$. At small scales, values of $E_{\rm w}/E_p>1$ occur, which are inconsistent with linear gravity waves and hence either nonlinear waves or turbulent motions. Generally, the

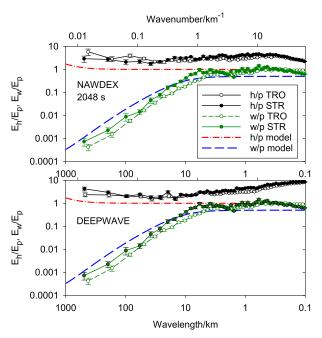


FIG. 8. Ratio of horizontal to potential energy E_h/E_p and of vertical to potential energy E_w/E_p , for (top) N and (bottom) D and TRO (open) and STR (full), as in Fig. 5. The red and the blue dashdotted curves represent the gravity wave model for E_h/E_p and E_w/E_p ; see Eqs. (A5)–(A8) and (7).

data show smooth changes with λ as modeled. This also supports the measurement quality.

In Fig. 9, results are compared for flights over mountains (land) and over oceans (zero surface height above sea level and more than 300 km distance from the geographic centers of New Zealand and Iceland). The results differ in that the splitting into two -5/3 spectra seems to be stronger over mountains than over oceans. The spectral energies are definitely higher over mountains than over oceans, as found earlier (Nastrom et al. 1987). The ratio $E_{\rm wm}/E_{\rm w}$ and, hence, the relationship between the horizontal and vertical spectra, differs only little between ocean and mountain cases.

Finally, Fig. 10 shows the spectra for the longest legs. Here, the wavelengths reach 1500 and 2000 km for D and N. We do not know of similarly long legs with w spectra from the literature. At the largest scales, the spectrum is based on just one wave, but the results get significant for $\lambda < 300$ km. The N data exhibit a significant -5/3 slope over three decades of scales, which is remarkable. The D spectra show again indications for two -5/3 ranges. The results suggest a transition scale L of order 300 km for N with rather low w variance. For the more turbulent D case, this transition seems to occur at larger wavelengths, possibly above 1500 km. Hence, the slope transition in the canonical spectrum varies depending on vertical velocity

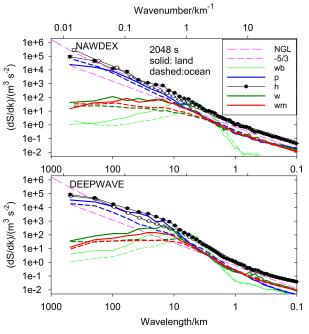


FIG. 9. As in Fig. 4, but separately for ocean (dashed) and land (solid). The averages include 125 (87) ocean and 53 (20) land legs for D (N) of 2048 s duration each. The mean flight heights are h=11.7 and 12.2 km for D over ocean and land and h=9.6 and 13.0 km for N over ocean and land. The mean ratio $E_{\rm wm}/E_{\rm w}$ is 0.94 \pm 0.26 and 0.91 \pm 0.31 for D and 1.06 \pm 0.28 and 0.87 \pm 0.28 for N.

variance at mesoscales. The modeled w spectrum (red curve) is close to the measured w spectrum (green) at least for scales smaller than about 200 km in both datasets.

5. Discussion and implications

The comparisons to global and regional models and to measurements show that the model is useful in characterizing the horizontal and vertical atmospheric flow energetics. The comparison with 2D spectra from a global model and a regional model confirm Eq. (7) in approximating the connection between E_d and E_w reasonably, at least in the mesoscales. The value of $h_{\rm eff}/h$ is smaller for the stratosphere than the troposphere because of small vertical correlations of horizontal motions in the stratosphere with the flow farther down, in particular, below the tropopause. The comparisons with observed 1D spectra at mesoscales near the tropopause suggest that Eq. (7) describes the relationship between measured spectra on average to an accuracy of about 30%. The relationship is best applicable for spectra on average over many long flights. Apparently the number of flights in N and D, 107 and 178 for 2048s leg lengths, were just sufficient in this sense. But also individual legs, like the longest flights of each campaign, with

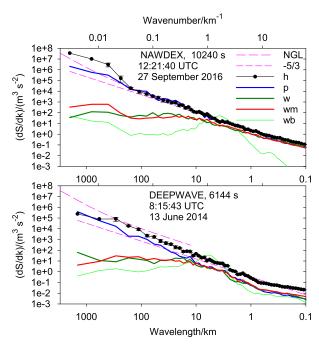


FIG. 10. As in Fig. 4, but for the longest legs of (top) $10\,240\,\mathrm{s}$ (2000 km at $h=8.7\,\mathrm{km}$, $\langle w'^2\rangle=0.024\,\mathrm{m}^2\,\mathrm{s}^{-2}$, $\varepsilon=17\times10^{-6}\,\mathrm{m}^2\,\mathrm{s}^{-3}$, $N=0.012\,\mathrm{s}^{-1}$, and $4.2\,\mathrm{km}$ below the tropopause) for N and (bottom) $6144\,\mathrm{s}$ ($1500\,\mathrm{km}$ at $h=12.3\,\mathrm{km}$, $\langle w'^2\rangle=0.068\,\mathrm{m}^2\,\mathrm{s}^{-2}$, $\varepsilon=1\times10^{-6}\,\mathrm{m}^2\,\mathrm{s}^{-3}$, $N=0.0266\,\mathrm{s}^{-1}$, and $1.2\,\mathrm{km}$ above the tropopause) for D. The leg start times are given. The values of E_{wm}/E_w are 1.41 ± 0.7 for D and 1.18 ± 1.2 for N.

1500–2000 km leg lengths, support the model within reasonable deviations.

Based on the 10% and 90% limits of the *w* energy distribution shown in Fig. 11, most of the vertical velocity energy near the tropopause occurs at mesoscales between about 0.5 and 80 km, with 50% at scales between 7 and 17 km. The result confirms that uncertainties in Eq. (7) originating from processes at large scales have little impact on the integral *w* variances. The global MPAS model with a 3 km grid resolves 50% contributions at scales below about 25 km. Apparently, grid resolution has still to be increased by factors of 3–10 to resolve 50%–90% of the *w* variances near the tropopause. An analysis of large-eddy simulations, as, for example, in Heinze et al. (2017), resolving mesoscale and turbulence scales, may be used to check these results.

The model derivation ignored density variations and variable surface orography, which are of course important for the atmosphere. For variable surface height, as in the mountain waves observed during DEEPWAVE, mountains likely add a contribution depending on the surface slope, horizontal velocity near the surface, and vertical wave propagation. Orography certainly induces

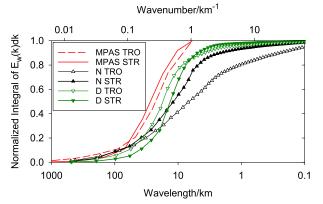


FIG. 11. Normalized integral of vertical kinetic energy, $\int_{k_{\rm min}}^k E_w(k') \, dk' \int_{k_{\rm min}}^{k_{\rm max}} E_w(k') \, dk'$, vs wavenumber k from numerical results (MPAS; Skamarock et al. 2014) and from measurements (N and D) on 2048 s legs, for tropospheric (TRO) and stratospheric (STR) cases. 50% of the energy is included for wavelengths $\lambda = 2\pi/k$ smaller than 23, 7, and 17 km (29, 12, and 14 km) for MPAS, N, and D in TRO (STR), respectively. The MPAS data include resolved w energy globally above 6 km scales. The data for D and N resolve scales from \sim 480 km down to 18 and 46 m, respectively, in this analysis.

large motion changes regionally, but mean w spectra may be less affected after averaging over a few long waves at small k and over the many short waves that occur at high k. Density variations from pressure changes and internal heat sources contribute to vertical velocity changes that we cannot quantify (Malardel and Wedi 2016; Craig and Selz 2018; Egger and Hoinka 2019). In the anelastic approximation, w would be the vertical wind weighted with density at height h, and \overline{u} and \overline{v} the vertically averaged horizontal mass fluxes. Here, surface contributions may get overestimated because of strong decrease of density with height (Egger and Hoinka 2019). Analyzing the flight data weighted with the measured densities shows small changes in the $E_{\rm wm}/E_{\rm w}$ ratio, but the data contain only a small spread of heights. Separation of the spectra into those over land and ocean has far larger impact.

The kinetic relationship implies a connection between the spectral slope of the horizontal velocities and the vertical velocity spectrum at least for incompressible and horizontally isotropic divergent motions. Besides in the numerical results shown in section 3, this relationship can also be identified in w and divergence spectra derived from numerical simulations of baroclinically unstable shear flows (Menchaca and Durran 2019). This kinematic connection may cause a dynamic connection between vertical and horizontal motions and affect the transition scale *L*. The flatter spectrum at mesoscales may be a consequence of energy transfer from horizontal to vertical divergent motions. This would be

consistent with studies noting that the transition in spectral slope of kinetic energy spectra occurs when the divergent component of the spectrum becomes similar in magnitude to the rotational component (Waite and Snyder 2009; Skamarock et al. 2019). Our data suggest that the spectral energy transfer rate is not only controlled by the classical ε . Other energy sources and losses at mesoscales may be more important than turbulence at the same level (e.g., by blocking and vortex shedding because of convection and mountains and by latent heat release in clouds) (Lilly 1983; Craig and Selz 2018). Note that spectral energy exchange between large and small scales occurs at time scales of order E/ε , which may be large compared to advection time scales L/U(e.g., over mountains with width L and wind speed U). Hence, energy maxima in the spectra at certain scales may originate mainly from energy induced by ambient disturbances (mountains, convection, etc.) at the same scales and less by upscale or downscale energy transfer.

We note that similar spectral relations with $E_w \approx (hk)^2 E_h$ for 2D spectra, have been found also for the boundary layer close to the Earth's surface (Peltier et al. 1996; Kang and Lenschow 2014; Tong and Nguyen 2015). The authors explained the connection with continuity as we do, with different quantitative models. Also other observations show similar spectral relationships (Gultepe and Starr 1995; Bacmeister et al. 1996; Gao and Meriwether 1998). A quadratic relationship between w and u spectra at low wavenumbers is also detectable in w spectra derived from potential temperature changes (Nastrom and Fritts 1992). Even strong downslope wind spectra show such behavior (Lilly 1978). The w spectra peaks at 4–16 km wavelengths obtained and analyzed by Zhang et al. (2015) and Callies et al. (2016) cannot be explained this way because the peaks have no counterparts in the measured horizontal velocity and potential energy spectra. There must have been other sources causing such vertical wind signals, such as divergent motions perpendicular to the flight direction (e.g., when passing mountain valleys with variable cross-stream winds) (Smith et al. 2016). Our results show that the smooth maximum of vertical velocity kinetic energy found in aircraft measurements at 5–15 km horizontal scales has good reasons. This supports the validity of such measurements. This is important to know when using long-range and high-speed research aircraft for atmospheric research.

6. Conclusions

A hypothesis has been introduced that connects vertical and horizontal divergent motions by mass continuity at large scales and by a trend toward locally

isotropic gravity wave or turbulence dynamics at small scales. The hypothesis is expressed in terms of a simple algebraic relationship between the horizontal spectrum of vertical velocities E_w at a given height h and the spectra of total and divergent horizontal winds at the same level, E_h and E_d . For large scales, E_w is close to $(h_{\rm eff}k)^2E_d$ where the effective height is related to geometric height by $h_{\text{eff}} = \beta h$, and β is a parameter of order 0.1 depending on vertical correlations between horizontal divergent motions. Comparisons with simulated and measured spectra, for a wide range of atmospheric conditions, in different parts of the world and for different measurements, support the hypothesis at mesoscales within about 30%. The small divergent motions at large scales and the high anisotropy at small scales differ from expectations for linear gravity waves. The ratio of vertical kinetic energy to potential energy increases quadratic with wavenumber at large scales and approaches a constant ratio at small scales but the ratio is larger than expected for linear gravity waves. Hence, the spectra cannot be explained by inertia-gravity waves, consistent with earlier findings.

The relation used relies on 2D isotropy, near incompressible continuity, and small surface contributions. These restrictions may be overcome with more extended theories and more knowledge on vertical correlations and anisotropy of horizontal divergent velocity components.

Spectra of vertical velocity and of potential energy are important additions to spectra of horizontal kinetic energy to characterize the atmospheric flow energetics. The near kinematic connection between vertical and horizontal divergent motions and with buoyancy likely has dynamical implications by causing energy transfer between horizontal and vertical motions. Further studies may quantify this energy transfer. Energy exchange between horizontal and vertical motions may be the key for understanding the split into divergent and rotational motions and the related flattening or steepening of the spectrum in stably stratified turbulence.

We used 120 h of constant-level flight data in the upper troposphere and lower stratosphere at scales from 0.1 to 2000 km horizontally from the DEEPWAVE and NAWDEX flight measurements. DEEPWAVE concentrated on mountain waves near New Zealand while NAWDEX addressed warm conveyer belts and frontal processes in the North Atlantic. In spite of these differences, the wind and temperature spectra from these experiments show remarkable similarities. The resolved smallest motions measured in mostly stably stratified layers near the tropopause show ratios E_w/E_h of about 0.1–0.5, far less than for isotropic turbulence. The horizontal divergence fraction was found to be 40%–70%

with higher values in the stratosphere, similar to earlier results. The data show significant deviations from 2D isotropy. Horizontal energy spectra with -5/3 slopes are observed in the troposphere over oceans. Near the tropopause, spectra over mountains are often steeper than over oceans, possibly with two -5/3 subranges. The mesoscale -5/3 spectrum is not controlled by local energy dissipation alone. Instead, energy exchange with equal-scale disturbances (e.g., by vertical motions) may be more important. The increase of vertical velocity kinetic energy over a broad range of scales over mountains is consistent with a flattening of the spectrum of horizontal velocity kinetic energy in the measurements.

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APPENDIX

Model Equations for Linear Inertia-Gravity Waves and Modifications

This appendix relates various forms of kinetic and potential energy and tests their consistency with linear hydrostatic inertia–gravity wave theory. Linear inertia–gravity waves can be expressed as Fourier modes as a function of horizontal and vertical wavenumbers k_x and k_y , m, and the intrinsic frequency $\hat{\omega}$ (Gill 1982; Fritts and Alexander 2003). For $f < \hat{\omega} < N$, these scales satisfy the well-known dispersion law:

$$\hat{\omega}^2 = f^2 + \frac{N^2(k_x^2 + k_y^2)}{m^2},\tag{A1}$$

with Brunt–Väisälä frequency N and Coriolis parameter f. On average over wave phases, such waves have vertical energy W, horizontal kinetic energy K, and potential energy P,

$$W = \frac{1}{2}\overline{w^2}, \quad K = \frac{1}{2}\overline{u^2 + v^2}, \quad P = \frac{1}{2}\frac{g^2}{N^2T_0^2}\overline{T^{\prime 2}},$$
 (A2)

with gravity g and mean temperature T_0 , with energy ratios (Geller and Gong 2010):

$$\frac{K}{P} = \frac{1 + (f/\hat{\omega})^2}{1 - (f/\hat{\omega})^2} \quad \text{and} \quad \frac{W}{P} = \frac{\hat{\omega}^2}{N^2}.$$
 (A3)

Observations of measured energy ratios and frequency spectra are consistent with these relationships (Podglajen et al. 2016). As a consequence of Eq. (A3), hydrostatic linear gravity waves require W < P < K because otherwise the intrinsic frequency is outside the range $f < \hat{\omega} < N$. The equations also imply $W/K \to 0$ for $\hat{\omega} \to f$ and $W/K \approx 1$ for $\hat{\omega} = N \gg f$. For nonhydrostatic gravity waves, these ratios depend additionally on vertical wavenumber m (Geller and Gong 2010).

From the hydrostatic equations one can eliminate the intrinsic frequency $\hat{\omega}$ to derive

$$\frac{W}{P} = \frac{f^2}{N^2} \frac{K + P}{K - P},\tag{A4}$$

$$\frac{P}{K} = -\frac{1 + \frac{W}{K} \left(\frac{N}{f}\right)^2}{2} + \left\{ \left[\frac{1 + \frac{W}{K} \left(\frac{N}{f}\right)^2}{2} \right]^2 + \frac{W}{K} \left(\frac{N}{f}\right)^2 \right\}^{1/2}.$$
(A5)

Equation (A4) allows computing W for given K and P. The result also satisfies Eq. (A5). Since the model is linear, the relationships hold for each Fourier mode k in terms of spectral energy densities E(k): $P = E_p dk$, $W = E_w dk$, and $K = E_h dk$.

Let us assume, Eq. (7) may be used to compute E_w for given E_d . For completion, we then need E_h and E_d . We may assume that the spectrum E_h of total horizontal kinetic energy follows approximated measured spectra (Lindborg 1999, 2009),

$$E_h = ak^{-3} + bk^{-5/3}. (A6)$$

For 1D and 2D spectra, the coefficients are $a_{2\rm d}=2a_{1\rm d}$, $b_{2\rm d}=1.4b_{1\rm d}$, $a_{1\rm d}=9.1\times10^{-4}$, and $b_{1\rm d}=3\times10^{-10}$ (coefficients in units of meters and seconds). In log scales, the asymptotic -3 and -5/3 lines cross at the transition scale L:

$$L = \frac{2\pi}{\exp[(3/4)(\ln b - \ln a)]}.$$
 (A7)

The value of L is 457 km for the 1D and 394 km for the 2D spectra. Finally, we need E_d . In view of observations,

it is tempting to assume that the fraction of divergent motions is close to d of order $\frac{1}{2}$ at scales < L, and approaches zero at scales > L, and varies with k like

$$\frac{E_d}{E_h} = d\frac{kL}{1+kL}. (A8)$$

A value of zero for E_d/E_h at large scales is consistent with quasigeostrophic theory [see section 8.16 in Gill (1982)] and a linear relationship between E_d and E_h at $k \ll L$ would be consistent with the -3 slope of E_h and the -2 slope of E_d in observations (Chen and Tribbia 1981) and numerical results (Skamarock et al. 2014). However, for linear gravity waves, E_d/E_h is not a free parameter; it follows from the equations of motions.

Along the same line of arguments as in Gill (1982, section 8.4 therein), for the hydrostatic case, one can derive the mean horizontal kinetic energy of divergent motions $D = (1/2)(\overline{u_d^2 + v_d^2})$. For this purpose, we start from the linearized Boussinesq equations, apply the Helmholtz decomposition to split the velocities into rotational (u_r, v_r) and divergent velocities (u_d, v_d) , and determine the related streamfunction and velocity potential such that their Laplacians equal the 2D divergence and vorticity. This allows computing (u_d, v_d) and the mean ratio of energies of divergent motions D relative to vertical energy W and to total horizontal energy K:

$$\frac{D}{W} = \frac{m^2}{k^2 + l^2}$$
 and $\frac{D}{K} = \frac{1}{1 + (f/\hat{\omega})^2}$. (A9)

The rotational energy equals K - D. The first part would be about consistent with Eq. (7) for $\beta \approx 1/\pi$, if applied to vertical wavenumbers $m \approx \pi/h$. However, the second equation implies that D/K varies between 0.5 at low frequencies and 1 at high frequencies $\hat{\omega}/f$. In fact, replacing Eq. (A8) by $E_d/E_h = 0.5\{1 + (2kL)^2/[1 + (2kL)^2]\}$ gives results consistent with linear gravity waves but inconsistent with observations and global numerical results.

Hence, as found earlier (Lindborg 2007), the theory of linear gravity waves does not explain the observations. Still, Eqs. (7) and (A4)–(A8) provide a closed set of model equations, which are interesting to consider. In this model, only E_p is related to linear gravity waves, Eq. (A5), and the low-wavenumber part should be taken with care. The value of E_p is sensitive to the values of f and N; that is, E_p increases with increasing N/f. Typical values at midlatitudes near the tropopause are $f = 10^{-4} \, \mathrm{s}^{-1}$ and $N = 0.02 \, \mathrm{s}^{-1}$. Moreover, the results depend on α and β . The wave theory suggests $\alpha = 1/d$ because then $E_w \approx E_p \approx E_h$ for $\hat{\omega} = N \gg f$. However, the observations in stratified turbulence, for example, those shown in section 4, show significantly smaller ratios of E_w/E_h .

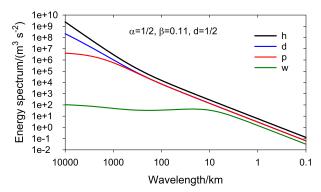


FIG. A1. The 2D spectra of total, divergent, vertical, and potential energy $(E_h, E_d, E_w, E_p,$ respectively) vs wavenumber k and wavelength $\lambda = 2\pi/k$, for prescribed E_h , modeled E_d and E_w , and E_p derived for linear hydrostatic inertia–gravity waves, for fixed values of α , β , and d (see figure for values) and for $h = 10 \, \mathrm{km}$.

Therefore, we leave the wave theory and assume d=1/2, $\alpha=1/2$, and $\beta\approx0.11$, $h=10\,\mathrm{km}$. For these parameters, the spectra are shown in Fig. A1. For comparison with the aircraft measurements, we need the 1D spectra. The 1D spectra can be obtained from the given 2D spectra:

$$E_{\rm 1d}(k_1) = \frac{2}{\pi} \int_0^\infty \frac{E_{\rm 2d}(k)}{k} dk_2, \quad k = (k_1^2 + k_2^2)^{1/2}.$$
 (A10)

Here, $E_{2d}(k)/(2\pi k)$ is the spectral energy density that is constant within a ring with circumference $2\pi k$ for 2D isotropic turbulence, and the prefactor is $2/\pi = 2 \times 2/(2\pi)$ because $E_{1d}(k_1)$ is defined for positive k_1 , and the integral is taken over positive k_2 wavenumbers only (E. Lindborg 2019, personal communication). Equation (A10) can be evaluated numerically. Computed 2D spectra differ from the 1D spectra by factors between 0.7 and 3 depending of the slope. The factor is largest for spectra with positive slope.

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