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Study of the overturning length scales at the Spanish planetary boundary layer

Pilar López[1] and José L. Cano[2]

[1]Applied Mathematics Dpt. (Biomathematics), Complutense University of Madrid, Madrid, Spain
[2]Astronomy, Astrophysical and Atmospheric Science Dpt., Complutense University of Madrid, Madrid, Spain

Correspondence to: Pilar López (maplopez@bio.ucm.es)

Abstract. The focus of this paper is to analyse the behaviour of the maximum Thorpe displacement $(d_T)_{max}$ and the Thorpe scale L_T at the atmospheric boundary layer (ABL), extending previous research with new data and improving our studies related to the novel use of the Thorpe method applied to ABL. The maximum Thorpe displacements vary between -900 and $950\,m$ for the different field campaigns. The Thorpe scale L_T ranges between 0.2 and $680\,m$ for the different data sets which cover different stratified mixing conditions (turbulence shear-driven and convective regions). We analyse the relationship between $(d_T)_{max}$ and the Thorpe scale L_T and we deduce that they verify a power law. We also deduce that there is a difference in exponents of the power laws for convective conditions and shear-driven conditions. These different power laws could identify overturns created under different mechanisms.

1 Introduction

The atmospheric boundary layer (or ABL) is almost always turbulent. In the absence of turbulence, atmospheric temperature profiles become increasingly monotonic, due to the smoothing effect of molecular diffusion. Turbulence gives rise to an effective eddy diffusivity and together with other causes (such as fluid instabilities or internal wave breaking) makes vertical overturns appear as inversions in measured temperature profiles. These overturns produce small-scale turbulent mixing which is of great relevance for many processes ranging from medium to local scales. Unfortunately, measuring at small scales is very difficult. To overcome this disadvantage it is interesting to use theories and parameter-izations which are based on larger scales, for example, the theories of turbulent stirring which often depend on hypotheses about the length scales of turbulent eddies. Vertical overturns, produced by turbulence in density stratified fluids such as lakes or the ABL, can often be quantified by the Thorpe displacements d_T and the Thorpe scale L_T (Thorpe, 1977).

Next we present the atmospheric data used for the analysis. In Sect. 3 we present the Thorpe method and the definitions of the scale descriptors used. In Sect. 4, the results of Thorpe displacements, the maximum Thorpe displacement, and the Thorpe scale L_T at the ABL are presented and discussed.

2 Atmospheric data sets and meteorological instrumentation

The results presented in this paper are based on three ABL field campaigns carried out in Spain and called Almaraz94-95, Sables98, and Sables2006. ABL data from 98 zeppelin-shaped tethered balloon soundings ranging from 150 to 1000 m were carried out in the Almaraz94-95 field campaign made in Almaraz (Cáceres, Spain). The ABL profiles were obtained from 25 to 29 September 1995 in the time intervals 06:00–12:00 and 15:00–00:00 GMT, and from 5 to 10 June 1994 in the time intervals 05:00–12:00 and 17:00–00:00 GMT. The Almaraz94-95 experiment collects data over a whole day and, therefore, covers different stratified conditions and mixing conditions – from shear-driven turbulence to convective regions. For more details, see López et al. (2008). Sables98 (Stable Atmospheric Boundary Layer Experiment in Spain) took place over the northern Spanish plateau in the period 10–28 September 1998. The campaign site was the CIBA (Research Centre for the Lower At-

mosphere). Two meteorological masts (10 and 100 m) were available at CIBA with high-precision meteorological instruments (Cuxart et al., 2000). Additionally, a triangular array of cup anemometers was installed for the purpose of detecting wave events and a tethered balloon was operated at night-time. A detailed description can be consulted in Cuxart et al. (2000). The Sables98 field campaign only collects data over the night and, therefore, under neutral to stable conditions. The Sables2006 field campaign took place from 19 June to 5 July 2006 at the CIBA. As in Sables98, different instrumentation was available on a tower of 100 m; a surface triangular array of microbarometers was also deployed and a tethered balloon was used to get vertical profiles up to 1000 m. As in Sables98, the Sables2006 field campaign also collects data over the night. Therefore, the Sables98 and Sables2006 experiments let us analyse the behaviour of overturns under stable conditions, while Almaraz94-95 is under unstable conditions (and also stable ones). These three sets of data were selected for this analysis because they cover different mixing conditions (turbulence shear-driven and convective regions).

3 Thorpe method and overturn length scales

Thorpe devised an objective technique for evaluating a vertical length scale associated with overturns in a stratified flow (Thorpe, 1977; Itsweire, 1984; Gavrilov et al., 2005). Thorpe's technique consists of rearranging a density profile (which contains gravitationally unstable inversions) so that each fluid particle is statically stable. If the sample at depth z_n must be moved to depth z_m to generate the stable profile, the Thorpe displacement d_T is $z_m - z_n$ (Thorpe, 1977; López et al., 2008, 2016). The Thorpe displacement d_T is not necessarily the real space actually travelled by the fluid sample. It is an estimate of the vertical distance from the given vertical profile to the statically stable one that each fluid particle has to move upward or downward to its position in the stable monotonic profile (Thorpe, 1977; Dillon, 1982). Over most of a typical profile, the local stratification will be stable and the Thorpe displacement zero. A turbulent event is, therefore, defined as a region of continuously nonzero d_T, that is, overturns are defined as a profile section for which $\sum_i d_{T_i} = 0$, while $d_{T_i} \neq 0$ for most i (Dillon, 1982; Peters et al., 1995).

The maximum of the Thorpe displacement scale $(d_T)_{max} = \max[d_T(z)]$ represents the larger overturns which might have occurred at earlier times when buoyancy effects were negligible (Thorpe, 1977; Dillon, 1982; Itsweire, 1984), and it could be considered an appropriate measure of the overturning scale.

The Thorpe scale L_T is the root mean square (rms) of the Thorpe displacements $(L_T)_{rms} = L_T = \langle d_T^2(z) \rangle^{1/2}$. Therefore, it is a statistical measure of the vertical size of overturning eddies (Thorpe, 1977; Dillon, 1982; Itsweire, 1984; Fer

et al., 2004) and is proportional to the mean eddy size as long as the mean horizontal potential temperature gradient is much smaller than the vertical gradient. For our field ABL measurements, we can consider the ABL to be horizontally homogenous because the average horizontal temperature gradient (4×10^{-4} (K m^{-1})) is smaller than the average vertical temperature gradient (2×10^{-2} (K m^{-1})) (López et al., 2016).

Because of the expensive nature of collecting data at microscale resolution, there is a great interest in using parameterizations for small-scale dynamics which are based on larger scales – as L_T or $(d_T)_{max}$. Therefore, it is very important to analyse the relationship between L_T and $(d_T)_{max}$ for selecting the most appropriate overturning scale.

4 Quantitative results

Our methodology is based on reordering 111 measured potential temperature profiles, which may contain inversions, to the corresponding stable monotonic profiles. Then, the vertical profiles of the displacement length scales $d_T(z)$ or Thorpe displacement profiles can be calculated by using a bubble sort algorithm with ordering beginning at the shallowest depth (Thorpe, 1977; Dillon, 1982; Itsweire, 1984; López et al., 2008, 2016). This simple sorting algorithm works by repeatedly stepping through the data list to be sorted, comparing each pair of adjacent items, and swapping them if they are in the wrong order (López et al., 2016).

4.1 Thorpe displacement profiles at ABL

Usually, the signature that might be expected for a large overturning eddy is sharp upper and lower boundaries with intense mixing inside – displacement fluctuations of a size comparable to the size of the disturbance itself are found in the interior. While common in surface layers strongly forced by the wind, these large features are not always found as in our ABL case (López et al., 2008, 2016). For our ABL studies, Thorpe displacements observed at profiles could be qualitatively classified into two groups, as Fig. 1 shows. The two graphs of Fig. 1 correspond to a campaign made on 25 September 1995. The left graph of Fig. 1 is at 07:00 GMT (stable conditions) and the right graph is at 17:00 GMT (convective conditions). The two kind of behaviours are as follows. First, the Thorpe displacements under neutral and stable stratification conditions are usually zero, except in a region with isolated Z patterns which would correspond to discrete patches (Fig. 1, left curve). These isolated overturns are very few well-defined sharp overturns which appear in sunset, night, and sunrise profiles. Secondly, we find other features that are smaller, some having an eddy-like shape similar to the larger disturbances, some a random mix of small-scale fluctuations without sharp boundaries (Fig. 1, right curve). These are the second group or non-zero Thorpe displacement regions with indistinct and distributed features which appear
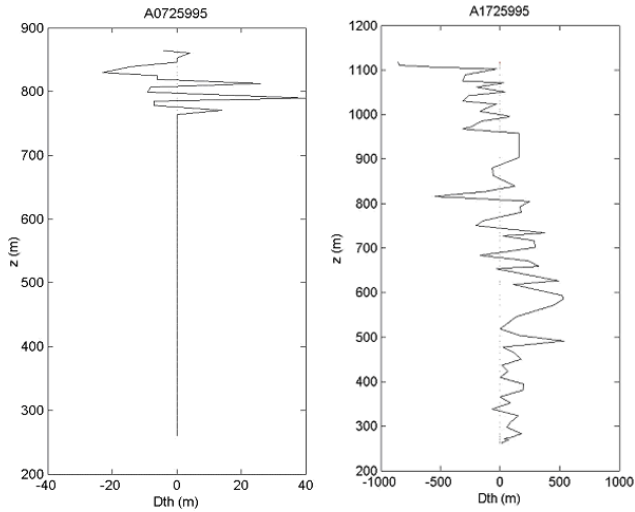
Figure 1. Left curve: Thorpe displacement Dth profile with an isolated patch corresponding to 07:00 GMT. Right curve: Thorpe displacement Dth profile with a random mix of fluctuations corresponding to 17:00 GMT.
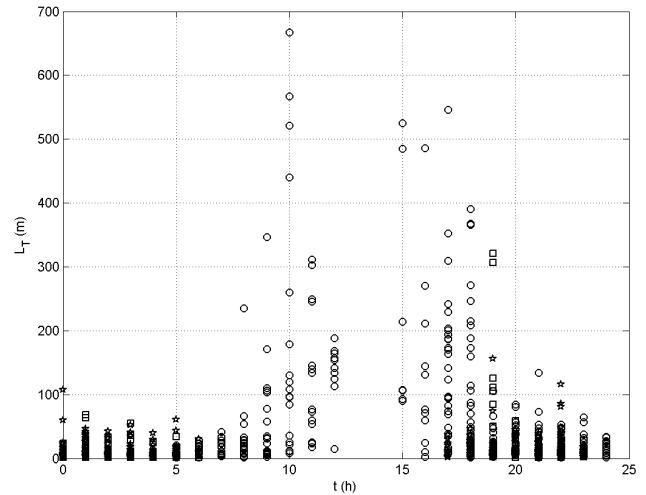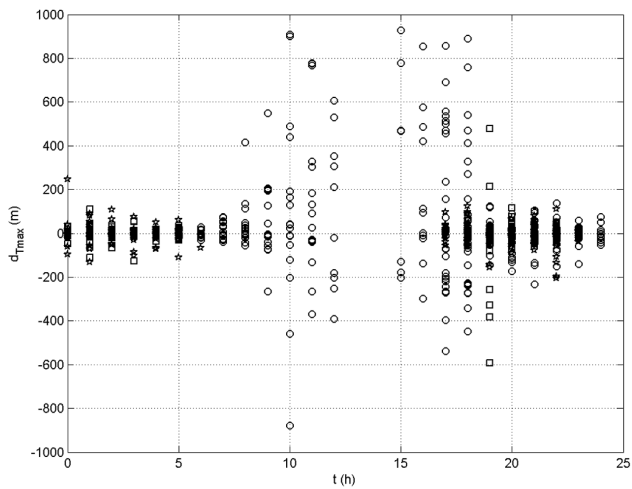

Figure 2. Time evolution of the maximum Thorpe displacements during a day cycle. The symbols are as follows: ○ is for Almaraz94-95 data, ⋆ is for Sables98 data, and □ is for Sables2006 data. The error of Thorpe displacements is ± 1 m.

under convective and/or neutral conditions (at noon, afternoon, and evening profiles). These Thorpe displacements are rarely zero for the whole profile. To verify this behaviour, see López et al. (2008) and López et al. (2016).

4.2 Time evolution of maximum Thorpe displacements and Thorpe scales

Figure 2 shows the time evolution of the maximum Thorpe displacement $(d_T)_{max}$ along a day for the three field campaigns. The scale $(d_T)_{max}$ is very small (approximately zero) under stable conditions from 23:00 to 06:00 GMT (between

Figure 3. Time evolution of the Thorpe scale during a day cycle. The symbols are as follows: ○ is for Almaraz94-95 data, ⋆ is for Sables98 data, and □ is for Sables2006 data. The error of Thorpe displacements is ± 1 m.

sunset and sunrise) for all the experiments. From 19:00 GMT, it is observed that scale $(d_T)_{max}$ decreases. The greatest values of $(d_T)_{max}$ appear under convective conditions from 09:00 to 19:00 GMT that are positive and negative. But the positive values of $(d_T)_{max}$ are greater than the negative ones. The positive $(d_T)_{max}$ has its greatest values about 950 m and the greatest negative $(d_T)_{max}$ is about 600 m (absolute value). These results mean the following. Thorpe displacements were defined as the difference between the final height and the initial height of the fluid particle, i.e. $d_T = (z_m)_{final} - (z_n)_{initial}$. If $d_T > 0 ((z_m)_{final} > (z_n)_{initial})$, the fluid particle has to go up to reach its stable position, and if $d_T < 0 ((z_m)_{final} < (z_n)_{initial})$, it has to go down to reach its stable point. From Fig. 2 we can deduce that fluid particles go upwards and downwards with a greater vertical distance under convective stratification conditions. Under stable stratification conditions – at night –, the fluid particles also move upwards and downwards, but with small values for the vertical distance travelled. Hence, it is clear that the maximum Thorpe displacement is always greater under convective conditions than under stable ones, independent of its sign. Therefore, the maximum Thorpe displacements are a parameter which could represent the dynamical behaviour of air particles and its relationship with the stratification conditions. Finally, there is a gap in Fig. 2 due to unregistered data between 13:00 and 14:00 GMT.

Figure 3 shows the time evolution of the Thorpe scale, L_T, during a day for the three field campaigns. The Thorpe scale L_T has small values (close to zero) under neutral and stable conditions from 20:00 to 09:00 GMT (between sunset and sunrise). This scale reaches its greatest values under convective conditions from 09:00 to 19:00 GMT. There are two distinct behaviours with high ($L_T > 100$ m) and low

($L_T < 100$ m) Thorpe scales. In most of the turbulent patches, the Thorpe scale does not exceed several tens of metres, and they appear under stable and neutral stratification conditions when the Thorpe displacements are also small and related to instantaneous density gradients. In contrast, under convective conditions, Thorpe scales are relatively large. They exceed hundreds of metres and they may be related to convective bursts. Hence, the Thorpe scale L_T is always greater under convective conditions than under stable ones, and it is a parameter which could also represent the dynamical behaviour of air particles. As in Fig. 2, there is a gap in Fig. 3 due to the unregistered data between 13:00 and 14:00 GMT. Both scales, the Thorpe scale L_T and the maximum Thorpe displacement $(d_T)_{max}$, have small values (close to zero) under neutral and stable conditions, and their greatest values appear under convective conditions. Therefore, it is reasonable to think about which of the two scales could represent better the dynamical behaviour of turbulent overturns.

Moreover, it is necessary to choose an appropriate overturning scale to characterize instabilities leading to turbulent mixing and the turbulent overturning motions themselves, and to look for a relationship with the Ozmidov scale at ABL data (Dillon, 1982; Lorke and Wüest, 2002; Fer et al., 2004). We could choose the Thorpe scale rather than the maximum Thorpe displacement because we only sample vertically, while the turbulence is three dimensional and, therefore, the Thorpe scale is more likely to be a statistically stable representation of the entire feature (Dillon, 1982). But the maximum of the Thorpe displacements is also considered to be an appropriate measure of the overturning scale and it is always greater than L_T (better detectable by a limited-resolution instrument). Different researchers have found a linear model between L_T and $(d_T)_{max}$ for profiles from the equatorial undercurrent (Moum, 1996; Peters et al., 1995) and a high linear correlation computed from the Banyoles99 field data where the ratio $(d_T)_{max}/L_T$ is approximately equal to 3 (Piera, 2004). A correlation must exist between L_T and $(d_T)_{max}$ because when computing the rms of a set of Thorpe displacements with high kurtosis distributions, the final result depends on the largest values (Piera, 2004; Stansfield et al., 2001). A similar linear correlation between L_T and $(d_T)_{max}$ has been found by other researchers: a ratio $(d_T)_{max}/L_T \approx 3.3$ is obtained in the oceanic thermocline (Moum, 1996), a ratio $(d_T)_{max}/L_T \approx 2.4$ is obtained from laboratory experiments (Itsweire et al., 1993) and, finally, the ratio $(d_T)_{max}/L_T$ is nearly 3 in numerical simulations (Smyth and Moum, 2000). But for microstructure profiles from strongly stratified lakes, a power law – such as $(d_T)_{max} \sim (L_T)^{0.85}$ – is found (Lorke and Wüest, 2002). This relation also holds for profiles from other lakes under very different conditions of mixing and stratification, with a strong correlation that holds over 4 orders of magnitude (Lorke and Wüest, 2002).

Hence, we analyse the relation between the L_T and $(d_T)_{max}$ scales for our ABL data. Figure 4 shows the max-
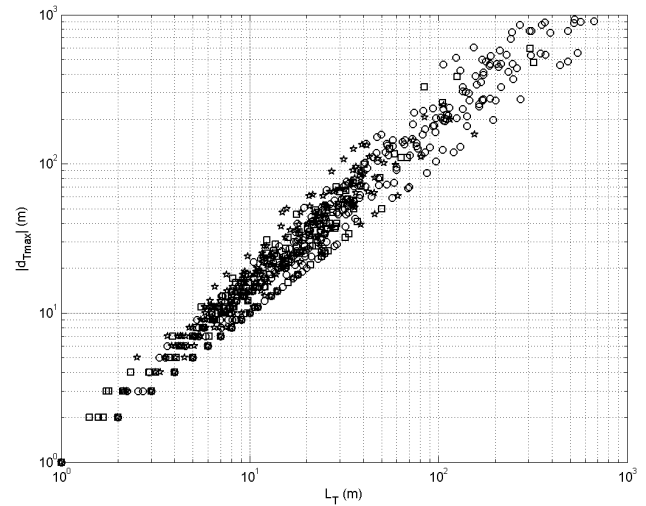

Figure 4. Absolute value of the maximum Thorpe displacement versus the Thorpe scale. The symbols are as follows: ○ is for Almaraz94-95 data, ⋆ is for Sables98 data, and □ is for Sables2006 data.

imum Thorpe displacement versus the Thorpe scale at a log scale, using the data of the three field campaigns. We observe that the linear model proposed by other authors (Moum, 1996; Peters et al., 1995; Piera, 2004; Itsweire et al., 1993; Smyth and Moum, 2000) does not agree with our ABL data.

Therefore, we could think that the nearly constant ratio $(d_T)_{max}/L_T$ obtained in a wide range of field and laboratory experiments is not verified in our ABL data (Fig. 4). And, hence, the shape of the Thorpe displacement distribution could change at the ABL. We also observe a strong correlation which holds over 3 orders of magnitude as in other studies from profiles in lakes (Lorke and Wüest, 2002). It is the first time that such a relationship between these two overturning length scales is found for ABL data (Fig. 4).

Like other authors, we could state that this high correlation indicates that the Thorpe scale is determined by the overturns near to the maximum Thorpe displacement. We find the following power law:

$$\left|(d_T)_{max}\right| \sim (L_T)^{1.14}, \tag{1}$$

which is similar to the one deduced by Lorke (Lorke and Wüest, 2002) from profiles in strongly stratified lakes. We perform a simple linear regression analysis. Of particular interest is the p value[1] associated with the analysis of vari-

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[1]The p value helps us to determine the significance of the results when we perform a hypothesis test which is used to test the validity of a claim that is made about a population. The p value is defined as the probability of obtaining a result equal to or more extreme than what was actually observed. We use a p value (always between 0 and 1) to weigh the strength of the evidence. A small p value (typically = 0.05) indicates strong evidence against the initial claim (null hypothesis).

ance[2], which tests the statistical significance of the fitted model. For our case the p value is less than 0.05 (operating at the 95 % confidence level), which indicates that the linear model between $|(d_T)_{max}|$ and L_T is statistically significant. Moreover, the R^2 coefficient[3] is 96.95 %, which indicates that the linear simple regression accounts for about 97 % of the variability in the maximum Thorpe displacement $|(d_T)_{max}|$ as a function of the Thorpe scale, L_T, statistically.

This relation between the maximum Thorpe displacement and the Thorpe scale by a power law has been deduced for the overall data (not separating the data from the three field campaigns). But we have used a data set with three different experiments under different mixing conditions. The SABLES98 and SABLES2006 experiments were conducted at night (turbulence by shear-driven) and ALMARAZ94-95 during a day cycle and, therefore, convective regions have not been excluded. Hence, we consider analysing whether this power law is different from night to day. The objective is to study whether it is possible to distinguish between the shear-driven overturns and the convective ones. First, we separate the data from the three experiments into two sets: data obtained overnight (from the Sables98, Sables2006, and Almaraz94-95 field campaigns), or the night data set, and data which have been obtained during the day (only from the Almaraz experiment), or the day data set. Then we perform a linear simple regression analysis with an adjustment by least squares for the two data sets. And, finally, we make a comparison of the regression lines relating $|(d_T)_{max}|$ and L_T at the two levels of our categorical factor (daytime and nighttime).

Figure 5 represents the maximum Thorpe displacement versus the Thorpe scale only for the daytime data set (from 07:00 to 19:00 GMT). We observe a strong correlation which holds over 3 orders of magnitude as was deduced for the whole data set and other studies (Lorke and Wüest, 2002).

We perform the linear simple regression analysis. The p value associated with the analysis of variance is less than 0.05 (operating at the 95 % confidence level[4]), which indicates that the linear fit between $|(d_T)_{max}|$ and L_T is statistically significant as before. The R^2 coefficient represents

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[2]The analysis of variance (ANOVA) is a statistical tool that separates the total variability of a data set into two components: random (which do not have any statistical influence on the given data set) and systematic factors (which have some statistical effect on the data). The ANOVA test is used to determine the impact independent variables have on the dependent variable in a regression analysis.

[3]The R^2 coefficient is called the determination coefficient which represents the proportion of the variance (fluctuation) of one variable that is predictable from the other variable. It is a measure that allows us to determine how certain one can be in making predictions from a certain model. In our case, the coefficient of determination is a measure of how well the regression line represents the data.

[4]The confidence level is a measure of the reliability of a result. A confidence level of 95 % or 0.95 means that there is a probability of at least 95 % that the result is reliable.
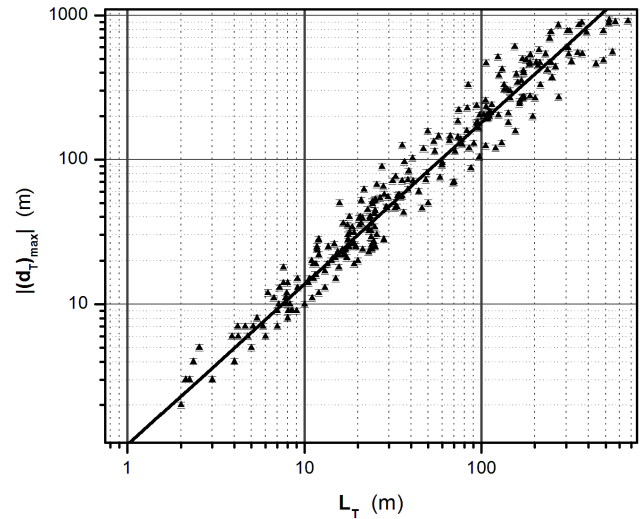

Figure 5. Absolute value of the maximum Thorpe displacement versus the Thorpe scale for the daytime data set (▲). The linear fit is indicated by the continuous black line.
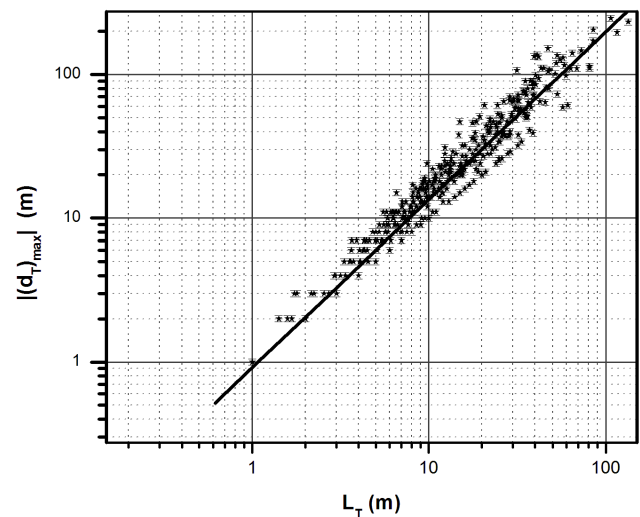

Figure 6. Absolute value of the maximum Thorpe displacement versus the Thorpe scale for the nighttime data set (○). The linear fit is indicated by the continuous black line.

the percentage of the variability in $|(d_T)_{max}|$, which has been explained by the fitted linear regression model and is about 97 %.

Figure 6 represents the maximum Thorpe displacement versus the Thorpe scale only for the nocturnal data set (from 20:00 to 06:00 GMT). We also observe a strong correlation which holds over 3 orders of magnitude as before (see Figs. 4 and 5).

Finally, we perform the linear simple regression analysis. The p value associated with the analysis of variance is less than 0.05 (operating at the 95 % confidence level), which indicates that the linear model is statistically significant as be-

fore. Moreover, the R^2 coefficient is 95.89, which means that the linear regression accounts for about 96 % of the variability in the maximum Thorpe displacement $|(d_{\mathrm{T}})_{\mathrm{max}}|$.

Therefore, we have deduced that the relation between the maximum Thorpe displacement $|(d_{\mathrm{T}})_{\mathrm{max}}|$ and the Thorpe scale L_{T} by a power law is different from day to night. For the nighttime data set, the power law is

$$|(d_{\mathrm{T}})_{\mathrm{max}}| \sim (L_{\mathrm{T}})^{1.17}. \tag{2}$$

And for the daytime data set, the relation is the following:

$$|(d_{\mathrm{T}})_{\mathrm{max}}| \sim (L_{\mathrm{T}})^{1.12}. \tag{3}$$

We observe that the kind of relation is the same (a power law) but the exponents are different. So we question whether these coefficients are statistically different and whether there is or is not a different behaviour of the overturn length scales between day and night.

These exponents are the slopes of the regression lines fitted to daytime and nighttime data sets (see Figs. 5 and 6). To know whether they are statistically different, we need to perform a comparison of regression lines. This procedure is a test to determine whether there are significant differences between the intercepts and the slopes at the different levels of our factor (day and night). This test fits two different regression lines to the nighttime and daytime data sets and performs two analyses of variance (one for each linear model and secondly for comparing the two regression lines). For the first analysis, the p value is less than 0.05, if we operate at the 5 % significance level, and indicates that the linear fit between $|(d_{\mathrm{T}})_{\mathrm{max}}|$ and L_{T} is statistically significant for daytime and nighttime data sets (t-statistic tests[5] have also been performed whose p values are less than 0.05, indicating that the model coefficients are significantly different from 0). The second analysis of variance is performed to determine whether there are significant differences between the slopes of the daytime and nighttime fitted lines. The F test[6] for slopes tests whether the slopes of the lines are all equal. Operating at the 1 % significance level[7], we find a p value (for slopes) which is less than 0.01, and, therefore, there are significant differences between the slopes of the daytime and nighttime lines (we get the same result for the intercepts).

There is one more question, that is, to analyse whether the power law fits the data better than a linear one in statistical terms. We have made a simple regression analysis to

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[5]A two-sample t test examines whether two samples are different, and it is a statistical analysis of two population means.

[6]The F test tests the statistical significance of the fitted model. A small p value (less than 0.05) indicates that a significant relationship of the form specified exists between two variables, y and x. It is most often used when comparing statistical models that have been fitted to a data set, in order to identify the model that fits best.

[7]In hypothesis testing, the significance level is the criterion used for rejecting the null hypothesis (a hypothesis about a population parameter). The significance level is the probability of rejecting the null hypothesis given that it is true.

construct three statistical models describing the dependence of $|(d_{\mathrm{T}})_{\mathrm{max}}|$ on L_{T} considering the three different situations, i.e. all the data, the daytime data, and the nighttime data set. The linear models were fitted using least squares and tests (analysis of variance) were run to determine the statistical significance of the fitted model.

For all three data sets, we got the same results. The analysis of variance indicated that a linear model between $|(d_{\mathrm{T}})_{\mathrm{max}}|$ and L_{T} is statistically significant (because the p value is less than 0.05). But the R^2 – or determination coefficient – which represents the percentage of the variability in $|(d_{\mathrm{T}})_{\mathrm{max}}|$ which has been explained by the fitted regression model is less in the power law fit (87.9 % for the whole data set, 84 % for the daytime data set, and 90.11 % for the nighttime data set) than in the linear one (96.95 % for the whole data set, 96.76 % for the daytime data set, and 95.89 % for the nighttime data set). As a consequence, the remainder of the unexplained variability is attributable to deviations around the line, which may be due to other factors, for example, to a failure of the linear model to fit the data adequately. We conclude that both models, the power law fit and the linear one, are statistically significant, but the power law fit has a better determination coefficient and accounts better for the variability in the maximum Thorpe displacement measurements. Therefore, we consider the power law fit to be the best fitted model for the three data sets.

Finally, we deduce that the two power relations between the maximum Thorpe displacement $|(d_{\mathrm{T}})_{\mathrm{max}}|$ and the Thorpe scale L_{T} for nighttime data (Eq. 2) and daytime data (Eq. 3) are significant different, with a 99 % confidence level. Therefore, we could classify overturns as day and night ones; that is, we could distinguish between convective and shear-driven mechanism originating overturns.

As mentioned before, although both scales $(d_{\mathrm{T}})_{\mathrm{max}}$ and L_{T} are alternative length scales to characterize turbulent overturns, it is reasonable to choose one of the two scales to represent better overturns. If there is a high linear correlation between the maximum Thorpe displacement $(d_{\mathrm{T}})_{\mathrm{max}}$ and the Thorpe scale L_{T}, the last one could be considered a better descriptor of the overturn properties, although it depends mainly on the values of $(d_{\mathrm{T}})_{\mathrm{max}}$, and the relative errors from both scales are approximately equal (Piera, 2004). But we have just deduced that the relationship between the maximum Thorpe displacement $(d_{\mathrm{T}})_{\mathrm{max}}$ and the Thorpe scale L_{T} does not follow a linear model in our ABL research, unless as a power law as with other authors (Lorke and Wüest, 2002). Consequently, there would not be a constant ratio $|(d_{\mathrm{T}})_{\mathrm{max}}|/L_{\mathrm{T}}$, which could suggest that the shape of the Thorpe displacement distribution could change. Therefore, it is necessary to study the probability density functions (pdfs) of the Thorpe displacements to understand better the relationship between $(d_{\mathrm{T}})_{\mathrm{max}}$ and L_{T}. Moreover, the Thorpe scale is mainly determined by larger overturns which are not very frequent (Stansfield et al., 2001), and it would be very useful to determine it based on the probability density function

of the Thorpe displacements. This pdf study would allow us to decide which of the two overturn length scales is a more representative measure of turbulent overturns.

5 Conclusions

The paper presents results related to the time evolutions of the ABL turbulent parameters L_T and $(d_T)_{max}$ during a day with different levels of stability. Secondly, the paper adds insight into the problem of the relationship between these two overturning length scales at the ABL.

The Thorpe scale L_T and the maximum Thorpe displacement $(d_T)_{max}$ have small values under neutral and stable conditions, and their greatest values appear under convective conditions. The values of the Thorpe scale range in (1, 660) m, which are greater than effective values in the stratosphere which are $L_T \sim 1$–1.1 m (Gavrilov et al., 2005), values in mixing surface layers and seasonal thermoclines which are $L_T \sim 0.03$–1.90 m (Dillon, 1982), values in the vertical mixing process induced by internal tides which are $L_T \sim 0.2$–4.2 m (Kitade et al., 2003), or values in dense overflow which are $L_T \sim 1$–17 m (Fer et al., 2004). The greater values appear under convective conditions which could generate overturns of a larger scale. Under shear-driven conditions, our Thorpe scales are smaller than convective ones, ranging in (1, 100) m, but they are also greater if we compare them with the scales of other authors. Therefore, we deduce that there would be a relationship between the ABL processes which generate mixing and the overturn size and behaviour (for example, the terrain shape interacts with the ability of the ABL to produce local mixing very near the ground, and this could be affect overturns). This theme will need further field work where different conditions are met (combination of the boundary condition effects and of stability combining the 3-D and 2-D characteristics of scale-to-scale direct and inverse cascades, intermittency of the forcing, and scale-to-scale stratified turbulence cascades (Vindel et al., 2008; Yagüe et al., 2006)).

Equations (1) to (3) show that the relationship between the Thorpe scale L_T, and the maximum Thorpe displacement $(d_T)_{max}$ is a power law which has been statistically demonstrated. We must therefore conclude that the linear model proposed by other authors (Moum, 1996; Peters et al., 1995; Piera, 2004; Itsweire et al., 1993; Smyth and Moum, 2000) is not adequate for our ABL data. Research will continue on this interesting question which is related to the selection of a length scale for characterizing turbulent overturns. This last problem would be better analysed if we were to study the probability density function (pdf) of overturning length scales. The objective is to decide whether L_T is or is not statistically a more appropriate length scale than $(d_T)_{max}$. Moreover, it is interesting to verify the assumption that the Thorpe scales have a universal probability density function which could be used to verify how accurately the Thorpe

scales were computed and also to determine whether $(d_T)_{max}$ is statistically better than L_T as an overturning length scale. It is very likely that the pdf parameters depend on the governing background conditions generating Thorpe displacements, which are different in the boundary layers from those in the interior layers with intermittent mixing, or in convective conditions from shear-driven conditions. We also would like to verify whether the density probability function is decaying exponentially for increasing displacement length with a separate cut-off before $(d_T)_{max}$.

In the future, we will go on studying the power relationship between the maximum Thorpe displacement and the Thorpe scale corresponding to ABL data to verify the power law deduced in this paper. For this purpose, we will use a new set of ABL data from new field campaigns. We will analyse the probability density function of overturning length scales to clarify better the relation between $(d_T)_{max}$ and L_T and as a tool to choose the more appropriate turbulent patch length scale. Moreover, we would like to study the following hypothesis that if the Thorpe scale is greater than the integral scale, there would be a local convective process and, if it is not, there would be stratification.

Finally, there is another subject which is important to mention. In future research, we need to study better the overturn identification as in Piera et al. (2002). They propose a new method based on wavelet denoising and the analysis of Thorpe displacement profiles for turbulent patch identification. Although their method is for microstructure profiles (which is not our case), it reduces most of the noise present in the measured profiles (increasing the resolution of the overturn identification) and is very efficient even at very low-density gradients for turbulent patch identification. Another way to get overturn identification would be, for example, to use a 3- or 4-D parameter space formed by (L_O, L_T, L_{MO}) to locate mixing events and also to study the evolution of the processes.

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