



Scalar turbulent behavior in the roughness sublayer of an Amazonian forest

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Abstract. An important current problem in micrometeorology is the characterization of turbulence in the roughness sublayer (RSL), where most of the measurements above tall forests are made. There, scalar turbulent fluctuations display significant departures from the predictions of Monin–Obukhov similarity theory (MOST). In this work, we analyze turbulence data of virtual temperature, carbon dioxide and water vapor in the RSL above an Amazonian Forest (with a canopy height of 40 m), measured at 39.4 and 81.6 m above the ground under unstable conditions. We found that dimensionless statistics related to the rate of dissipation of turbulence kinetic energy (TKE) and the scalar variance display significant departures of MOST as expected, whereas the vertical velocity variance follows MOST much more closely. Much better agreement between the dimensionless statistics with the Obukhov similarity variable, however, was found for the subset of measurements made at low zenith angle Z , in the range $0 < |Z| < 20$. We conjecture that this improvement is due to the relationship between sunlight incidence and “activation/deactivation” of scalar sinks and fonts vertically distributed in the forest. Finally, we evaluated the relaxation coefficient of Relaxed Eddy Accumulation: it is also affected by zenith angle, with considerable improvement in the range $0 < |Z| < 20$, and its values fall within the range reported in the literature for the unstable Surface Layer. In general, our results indicate the possibility of better stability-derived flux estimates for low zenith angle ranges.



1 Introduction

In the Atmospheric Surface Layer above the roughness sublayer (RSL) height z_* (approximately
20 three times the height of the roughness obstacles h — Cellier and Brunet 1992), flux estimates based
on mean concentration measurements are made with the help of Monin–Obukhov Similarity Theory
(MOST) and the corresponding similarity functions. It is now well known, however, that MOST-
based similarity functions often fail, to various degrees, in the roughness sublayer. In this region,
beyond the classical governing variables found in MOST, there are several more intervening vari-
25 ables, such as tree spacing and vegetation density, among others (Garratt, 1980; Foken et al., 2012;
Arnqvist and Bergström, 2015). This is the region where most mean concentration measurements
above forests are made, and such is the case with the 82-m tower data analyzed in this work.

In principle, the availability of mean concentration data would make flux-gradient methods a
natural choice to estimate scalar fluxes above the forest. Unfortunately, the difficulty of applying
30 MOST in the RSL adds considerable uncertainty to this approach.

Maybe one of the earliest reports of the failure of flux-gradient methods when measurements are
performed too close to the roughness elements was made by Thom et al. (1975), who compared flux-
gradient and energy-budget Bowen ratio methods over a Pine forest, and found that the dimensionless
gradients Φ_M and Φ_H of MOST are underestimated under such conditions. This was generally
35 confirmed by Garratt (1978), who estimated $z_* = 4.5h$ for the momentum flux and $z_* = 3h$ for the
sensible heat flux.

In the roughness sublayer, scalar and velocity gradients are weaker than above, leading to higher
values of the corresponding turbulent diffusivities (Cellier and Brunet, 1992). Under neutral condi-
tions, the momentum turbulent diffusivity increases by a factor 1.1–1.5, and the sensible heat
40 turbulent diffusivity by 2–3. The turbulent Prandtl number correspondingly decreases from to close
to 1 to approximately 0.5 at canopy top (Finnigan, 2000).

Cellier and Brunet (1992) propose a dimensionless factor γ to account for the increase in turbulent
diffusivity. Following this suggestion, Schween et al. (1997), using data measured over a 12-m tall
oak and pine tree forest, found $\gamma_\theta = 2.2$. They pointed out that the different behavior in the RSL may
45 be due to flux originating below the zero-plane displacement height, since in-canopy air may have
significantly different characteristics from above-canopy air. As we will see, zenith angle analyses
in the present work give support to this consideration.

Garratt (1980) proposed that for heights $z < z_*$ the vertical gradients depend on an additional
length scale z_s , due to the turbulent wake generated by the trees. The author investigated data from a
50 surface covered by scattered trees and shrubs in Australia, referred to as sub-tropical scrub or savan-
nah, of average height 8 m and occupying about 25% of the total surface area. He analyzed data in
the range $5 < d/z_0 < 85$ (d is the zero-plane displacement height and z_0 is the surface roughness),
suggesting an additional dependence of the dimensionless gradients on the variable z/z_* . Consid-
ering the canopy physical characteristics, he found that a relevant length scale is the tree spacing



55 δ , and proposed $z_* = 3\delta$. Regarding Garratt's wake production assumption, it has been argued that this effect dies out rapidly above the canopy and that it is not the main cause for the lack of MOST-compliance in the roughness sublayer (Mölder et al., 1999).

Other attempts to organize roughness sublayer data include the use of z/z_* by Cellier (1986), and Mölder et al. (1999)'s proposal of a function $(z/z_*)^n$ multiplying the dimensionless gradients: 60 Mölder et al. found $n = 1$ for scalars and $n = 0.6$ for momentum; they claim that the use of this factor produces acceptable results. Still, even with this correction, the resulting dimensionless functions in Mölder et al. (1999) display a much larger scatter than what is usually found above the roughness sublayer. Such roughness sublayer "dissimilarity" is not restricted to flux-gradient relationships: the dimensionless standard deviation of a scalar a , σ_a/a_* , has been found to be equally affected (Padro, 65 1993; Katul and Hsieh, 1999; von Randow et al., 2006; Williams et al., 2007; Dias et al., 2009).

In this paper, we analyze roughness-sublayer data collected under the scope of the ATTO project (Amazon Tall Tower Observatory), a German-Brazilian project undertaken under the leadership of Max-Planck-Institute (Germany), Instituto Nacional de Pesquisas da Amazônia (INPA) and Universidade Estadual do Amazonas (UEA) (Brazil). A 325-m tall tower has been erected in a forest 70 site 150 km NE of Manaus, and is currently undergoing instrumentation. Preliminary measurements have been made at an 82-m tall tower built at the site, and some analyses from the measured micrometeorological data are described here.

The main purpose of ATTO is to better understand the role of the Amazonian biome in the context of Global Climatic Changes. Specifically, the project aims at better understanding and modeling of 75 gaseous exchanges between the forest and the atmosphere (Andreae et al., 2015). For many scalars of interest, such as volatile organic compounds (VOCs), the high-frequency measurements needed in the Eddy Covariance Method are still difficult to make (particularly in long-term campaigns), leaving their flux estimates to methods based on the measurement of their mean concentrations and gradients.

80 Given the importance of correctly estimating trace gas fluxes over the Amazon forest, the lack of a theory for the roughness sublayer is clearly a major obstacle in the understanding of surface-atmosphere interactions with far-ranging implications on the regional and global hydrology, ecology, and climate.

Moreover, given the always present need to take into account site-specific features in any micrometeorological study, we attempt here to provide a general analysis of roughness sublayer-related 85 questions at the ATTO site prior to the construction of the main tower. We expect that once measurements at the main tower become available, a better understanding of the questions preliminarily assessed here will be possible.

In this work, the sections are organized as follows. In Sect. 2 we describe the experimental site in 90 Amazonian Forest and data measurement; we also show the steps of data quality control. Theoretical concepts used to develop this research are reviewed in Sect. 3, where we describe the dissipation



rate, vertical velocity skewness, spectral analysis and relaxed eddy accumulation in light of MOST; in Sect. 4 we discuss these results. Variance method results in the roughness sublayer and the zenith angle influence on several turbulence statistics are shown in Sect. 5, followed by an analysis of scalar
95 similarity indices and their implications for flux estimation in Sect. 6. Finally, in Sect. 7 we make our final considerations.

2 Data measurement and quality control

2.1 Experimental site

The study area is located at Reserva de Desenvolvimento Sustentável Uatumã (RDSU) (Uatumã
100 sustainable development reservation), in the counties of São Sebastião Uatumã and Itapiranga, in the Northeastern of Amazonas state, Brazil. The site is 150 km Northeast of the state capital Manaus, between the coordinates $59^{\circ}10' - 58^{\circ}4'W$ and $2^{\circ}27' - 2^{\circ}4'S$.

In the forest, between 200 and 250 tree species per ha can be found, with a mean height of 40 m and with some individuals reaching 50 m. The site itself is located on a plateau (*terra firme*), with
105 altitude 130 m.

The micrometeorological data were measured at an 82-m tower with a rectangular cross section of $2.5 \times 1 \text{ m}^2$ at the site ($2^{\circ}8'40''S$, $59^{\circ}0'10''W$). Micrometeorological instrumentation was installed at the 23, 39.4 and 81.6 m levels (above ground).

In this work, we analyze pilot data from the 39.4 m and 81.6 m heights, measured during April,
110 2012. The data analyzed are the three wind components u , v and w measured by two sonic anemometers (R3, *Gill Instruments Ltd.* at 39.4 m; CSAT3, *Campbell Scientific Inc.*, at 81.6 m), the sonic temperature (which we assume to be the same as the virtual temperature θ_v), and the mass concentrations (mass of the species/total mass) of CO_2 , c , and H_2O , q , calculated from the corresponding mass densities (mass of the species/volume) measured by two IRGA's (LI-7500A, *LI-COR Inc.*).

115 2.2 Quality control

The 10Hz data were analyzed in 30-min. data blocks ("runs"). Incomplete runs were excluded, and spikes were removed following Vickers and Mahrt (1997). For the next phase of quality control, fluctuations were extracted around a running mean. After that, each run was sub-divided into 15 two-minute subruns, and a local (i.e. 2-min.) standard deviation was calculated. Whenever this value
120 was less than a threshold (pre-stipulated based on the sensor accuracy), the whole 30-min. run was excluded.

As a result, 21.5% of the 81.6-m level and 50.2% of the 39.4-m level runs were left. However, after this test, some strongly non-stationary time series remained, mainly in scalar data, even when linear detrending was applied. For this reason, these remaining data were further checked with two
125 tests (both after removal of the linear trend). The first was the Reverse Arrangement Test (Bendat



and Piersol 1986, p. 97; Dias et al. 2004), performed on $N = 50$ averages of the 30-min. run and a significance level $\alpha = 0.05$. The second test was defined by us based on a visual scrutiny of the data. It consisted of calculating the difference between the maximum and minimum values of the running mean within each 30 min. run. The run was discarded whenever this difference exceeded $\Delta\theta = 1.7$
130 ° C, $\Delta c = 0.11 \text{ g kg}^{-1}$ and $\Delta q = 3 \text{ g kg}^{-1}$. These further tests reduced data availability to 15.8% at the 81.6 m level and 40.7% at the 39.4-m level. For these runs, a 2-D coordinate rotation was applied for the final analyses.

3 Theoretical background

In this section, we briefly review some results, which are used in the next section to analyze the data.

135 3.1 Dissipation rate

The dimensionless dissipation rate of turbulence kinetic energy (TKE) is given by (Kaimal and Finnigan, 1994)

$$\phi_\epsilon = \frac{\kappa(z-d)\epsilon}{u_*^3}, \quad (1)$$

where u_* is the friction velocity, κ is the Von Karman constant and ϵ is the rate of dissipation of
140 TKE. In MOST, a function still widely used to predict ϕ_ϵ is (Kaimal et al., 1972)

$$\phi_\epsilon = \begin{cases} (1 + 0.5|\zeta|^{2/3})^{3/2}, & -2 \leq \zeta \leq 0, \\ (1 + 5\zeta), & 0 \leq \zeta \leq 1, \end{cases} \quad (2)$$

where ζ is the Monin–Obukhov stability parameter. As we shall see, in the roughness sublayer the dissipation rate deviates from the prediction by Eq. (2). In this work, we assess this depart from MOST by extending an index proposed by Mammarella et al. (2008):

$$145 \chi = \frac{u_*^3/\epsilon}{\kappa(z-d)/\phi_\epsilon} - 1 = \frac{L_\epsilon}{\kappa(z-d)} - 1 \quad (3)$$

In Eq. (3), L_ϵ is the length scale calculated from the friction velocity and the rate of dissipation of TKE, and adjusted to the effect of buoyancy: it can be regarded as an integral scale of the flow. It can readily be verified that $L_\epsilon/(\kappa(z-d)) = 1 \Rightarrow \chi = 0$ indicates that the dissipation data follow MOST perfectly. Originally, Mammarella et al. (2008) proposed χ to be used only under near-neutral
150 conditions. As our data comprise too few near-neutral runs, it was necessary to take into account stability by including ϕ_ϵ in the denominator of Eq. (3).



3.2 Vertical Velocity Skewness

One characteristic property of canopy turbulence, its intermittency, emerges when we consider higher-order velocity moments. In the surface layer, the skewnesses of u and w are

$$155 \quad \text{Sk}_w = \frac{\overline{w'^3}}{\sigma_w^3}, \quad (4)$$

$$\text{Sk}_u = \frac{\overline{u'^3}}{\sigma_w^3}. \quad (5)$$

In convective or near-neutral conditions, Sk_w is typically observed to be negative in the roughness sublayer and positive in the inertial sublayer (Raupach and Thom, 1981; Fitzjarrald et al., 1990; Kaimal and Finnigan, 1994; von Randow et al., 2006), whereas Sk_u tends to be positive in all situations. At some level above the canopy Sk_w changes sign, and it seems reasonable to regard this level to be a measure of the roughness sublayer height. According to Fitzjarrald et al. (1990), the negative Sk_w values above canopies are largely due to the fact that there is something below the “surface” [sic] in canopy layers, and there can be downward turbulent transport of vertical velocity variance associated with the drop in the Turbulence Kinetic Energy (TKE) as one goes into the canopy. Kaimal and Finnigan (1994) attribute the considerable scatter in published results primarily to morphological differences between canopies, but at any rate this combination of strongly positive u -skewness and strongly negative w -skewness indicates that the turbulence is dominated by intermittent downward moving gusts in the roughness sublayer.

3.3 Spectral analysis

170 Consider the temperature spectrum in the inertial subrange, in the form

$$k^{5/3} E_{\theta\theta}(k) = \alpha_{\theta\theta} \epsilon^{-1/3} N, \quad (6)$$

where $\alpha_{\theta\theta} = 0.8$, and N is the rate of dissipation of semi-temperature variance (Kaimal and Finnigan, 1994, p. 37).

On dimensional grounds, under the validity of MOST, a similarity function exists that describes the temperature spectrum in the inertial subrange, viz.

$$175 \quad g_{\theta}(\zeta) = \frac{\alpha_{\theta\theta} \epsilon^{-1/3} N (z-d)^{2/3}}{\theta_*^2}. \quad (7)$$

Again, we seek to determine to what degree g_{θ} calculated with roughness sublayer data obeys MOST scaling. The usefulness of this indicator lies in its limited frequency-range: both ϵ and N are inertial-subrange variables in the sense that they are obtained from a straightforward analysis of the inertial subrange of the velocity and temperature spectra. Therefore, g_{θ} is sensitive only to the highest range of frequencies (roughly $> 0.02\text{Hz}$). If the dissimilarity displayed by “bulk” dimensionless statistics such as σ_{θ}/θ_* is due to the larger scales only, then g_{θ} and similar variables should obey



MOST much more closely than the former. If on the other hand the dissimilarity is spread out through all frequency ranges, then g_θ should display the same sort of non-conformance to MOST as the other
 185 “bulk” statistics.

3.4 The Relaxed Eddy Accumulation method and related analyses

The original Eddy Accumulation method was proposed by Desjardins (1972) with the objective of estimating the turbulent flux of chemicals not easily measured at high frequency (see Ren et al., 2011). In the method, conditional sampling is performed on the gas, which is directed to either of
 190 two reservoirs according to the sign of the vertical wind velocity, w , by means of fast opening/closing valves. Evident advantages are the fact that the method dispenses with high-frequency concentration measurements, and that only one-level measurements are needed (Tsai et al., 2012). Following this method, the flux is calculated by (Businger and Oncley, 1990):

$$\begin{aligned} \overline{w^+c} + \overline{w^-c} &= (\overline{w^+} + \overline{w^-})\bar{c} + \overline{w^+c'} + \overline{w^-c'}, \\ 195 \quad &= \overline{w^+c'} + \overline{w^-c'} = \overline{w'c'}, \end{aligned} \quad (8)$$

since $\overline{w^+} + \overline{w^-} = \bar{w} = 0$.

Businger and Oncley (1990) proposed the Relaxed Eddy Accumulation (REA) method, which uses the mean concentrations in each of the two conditionally sampled reservoirs, \bar{c}^+ and \bar{c}^- , to calculate the flux as

$$200 \quad \overline{w'c'} = b\sigma_w(\bar{c}^+ - \bar{c}^-), \quad (9)$$

where σ_w is the standard deviation of the vertical velocity and b is the relaxation coefficient, that is empirically verified to be a dimensionless constant (it could be a MOST similarity function) of the order of 0.6 (Businger and Oncley, 1990; Katul et al., 1996).

Although initially developed and tested for the measurement of CO_2 , water vapor and sensible
 205 heat fluxes (Pattey et al., 1993; Bash and Miller, 2007), the REA method is often extended for the measurement of other gaseous/scalar fluxes, usually with direct sensible or latent heat fluxes used as proxies to estimate b . For example, Zhu et al. (2000) used it to calculate ammonia fluxes, and Mochizuki et al. (2014) for isoprene and monoterpene fluxes; Matsuda et al. (2015) used it to estimate sulfate and $\text{PM}_{2.5}$ fluxes; Bowling et al. (1998) used it to calculate isoprene fluxes, Bash
 210 and Miller (2007) and Sommar et al. (2013) for mercury fluxes and Moravek et al. (2014) used for peroxyacetyl nitrate fluxes.

In the ATTO project, we will be interested in the fluxes of several chemicals whose high-frequency measurement is still too laborious, too expensive, or downright impossible. Among these compounds are the VOCs released by the forests, the monoterpenes and isoprene being the most abundant fol-
 215 lowed by alcohols, carbonyls, acids, aldehydes, ketones and esters. If applicable, then, the REA method will be an invaluable tool.



Table 1. REA coefficients measured above the roughness sublayer, where z is the measurement height (in meters) and numbers between brackets are the height of the canopies.

Author(s)	Businger and Oncley (1990)	Katul et al. (1996)	Baker et al. (1992)
z (m)	4	5	–
Cover	crops	Corn (2.4 m) Grass (0.10 m)	soybean field
b_θ	0.6	0.58 ± 0.11	–
b_q	0.6	0.58 ± 0.15	0.56
b_c	–	0.56 ± 0.06	0.56
b_{O_3}	–	0.56 ± 0.06	–

Table 2. REA coefficients measured within the roughness sublayer (Gao, 1995).

z (m)	43.1	34.2	18	14.4	10.5	5.9
b_θ	0.58 ± 0.04	0.55 ± 0.04	0.51 ± 0.03	–	0.61 ± 0.06	0.62 ± 0.09
b_q	–	0.55 ± 0.02	0.51 ± 0.05	0.61 ± 0.05	–	–

However, strictly speaking, to be valid the method requires that the same value of b apply to all scalars. This is of course the same as the well-known hypothesis of perfect similarity between scalars (Hill, 1989; Dias and Brutsaert, 1996), which often fails under unstable conditions due to phenomena
 220 originating above the surface layer and to the transport of scalar variance from above (Cancelli et al., 2012, 2014). Furthermore, the physics of the roughness sublayer proper may complicate this picture even more.

Therefore, before applying the REA method to measurements made close to the canopy over a forest, it is important to assess the validity of scalar similarity and the equality of the b 's for all
 225 scalars. A review of several REA studies by Tsai et al. (2012) showed that b can vary as much as between 0.2 and 0.9, revealing the importance of its correct estimation. In Table 1 we give some values found in the literature outside of the roughness sublayer, and in Table 2 values found by Gao (1995) for the roughness sublayer for different values of the leaf-area index.

4 Results

230 4.1 Dissipation rates

The rates of dissipation of TKE for each run were calculated from the longitudinal spectra on the basis of Kolmogorov's local isotropy theory (Kolmogorov, 1941). For each run, the inertial subrange

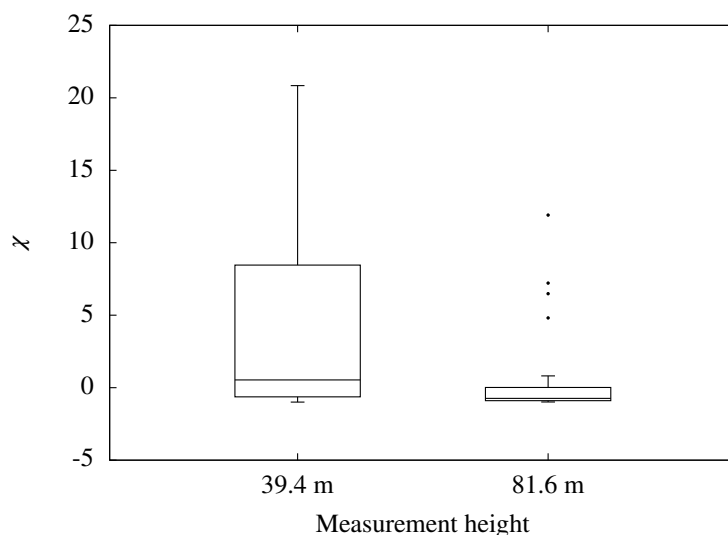


Figure 1. Tukey Boxplots (Frigge et al., 1989) of the χ variable defined at the 39.4 and 81.6 m levels. Some outliers at the 39.4 m are not shown.

was identified and a linear regression with an imposed $-5/3$ slope was calculated to determine ϵ . With ϵ , we calculated ϕ_ϵ in Eq. (1) and the index χ in Eq. (3).

235 The results can be seen in Fig. 1. As mentioned before, in the region of validity of MOST, we would expect the χ 's to cluster around 0.

We find that, close to the canopy top, the integral scale L_ϵ far exceeds $\kappa(z-d)$: this means that the latter is not a good estimate of the integral scale, as often found in the roughness sublayer. These results are in agreement with the findings by Rao et al. (1973) e Mammarella et al. (2008).

240 At $z/h_c = 2.04$ (h_c is the canopy height) the spread of χ around zero is much smaller, suggesting that, as expected, we are reaching the upper limit of the roughness sublayer at these heights, again in agreement with what is usually found in the literature.

4.2 Vertical Velocity Skewness

In Fig. 2, we show the vertical velocity skewness for the levels 39.4 m and 81.6 m, as a function
245 of the stability parameter ζ . The results confirm those found above with the dissipation rate in Sect. (4.1): roughness sublayer characteristics are still evident at the upper level, being obviously more pronounced at the lower level, where Sk_w reaches close to -0.8 . It is noteworthy that at the 39.4-m level, in spite of the prevalence of negative values of Sk_w , positive values typically associated with the surface layer can also be found. This suggests that the roughness sublayer height is actually

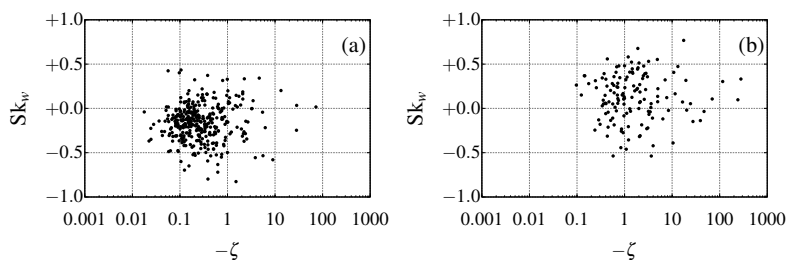


Figure 2. Vertical velocity Skewness: (a) 39.4 m and (b) 81.6 m.

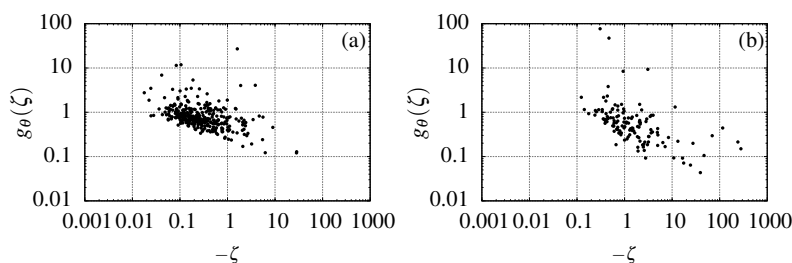


Figure 3. Inertial subrange similarity for temperature spectra: (a) 39.4 m and (b) 81.6 m.

250 varying from run to run, or that the physical picture is more complicated, with the possibility of
positive skewnesses inside the roughness layer. Clearly, the subject needs further research.

4.3 Inertial subrange similarity

Similarly to the analysis of the longitudinal velocity spectra, we identified for each run the inertial
subrange of the temperature spectrum and fitted a linear regression with a $-5/3$ slope, in order to
255 extract the rate of dissipation of semi-variance of temperature, N . From the latter, and ϵ , the value
of g_{θ} in Eq. (7) was calculated. It is plotted against ζ for the two levels, in Fig. 3-a and 3-b. There
is a clear pattern of decreasing g_{θ} with ζ , but still there is large scatter typical of roughness sublayer
results.

This analysis suggests that inertial-subrange scales, approximately in the range 0.02–0.8 Hz, are
260 also influenced by roughness sublayer effects: in other words, restricting the analysis to a range of
smaller scales does not improve appreciably the predictions of MOST. Similar plots and results were
also obtained for the other scalars (water vapor and CO_2), but are not shown here.



5 Variance method results

5.1 General behavior in the roughness sublayer

265 The “variance method”, pioneered by Tillman (1972), is widely used in micrometeorology for several purposes, including quality control procedures (Foken and Wichura, 1996; Thomas and Foken, 2002; Lee et al., 2004) and flux estimation (Hong et al., 2008). Here, we consider similarity functions of the type $\phi_a = \sigma_a/a_*$, where σ_a is the scalar standard deviation, and a_* the scalar’s turbulent scale.

270 Figures 4 and 5 show ϕ_w , ϕ_c , ϕ_q and ϕ_θ for both measurement levels. Only negative CO₂ fluxes ($c_* < 0$) and positive latent and sensible heat fluxes ($q_* > 0$, $\theta_* > 0$) were considered, for unstable conditions $\zeta < 0$. In the figures, we plot empirical $\phi_a(\zeta)$ functions from experimental data for which good MOST agreement was observed (Katul et al., 1995).

Once more, the large scatter typical of roughness sublayer data is found: notice that the scatter is much larger in ϕ_θ , ϕ_c and ϕ_q than in ϕ_w . Williams et al. (2007) suggest that this lack of MOST-compliance may be associated with heterogeneity of sources and sinks inside the canopy, contributing to the larger standard deviation (relative to the scalar turbulent scale).

This tendency of ϕ_a data in the roughness sublayer to lie above the corresponding MOST functions is generally observed in field experiments (Cava et al., 2008; Dias et al., 2009), but a definitive explanation for it is still lacking.

280 The ϕ_w data, on the other hand, show the opposite trend (they fall *below* the corresponding MOST function for the surface sublayer), with much less scatter than in the scalar case.

5.2 Zenith angle influence

It is known that the zenith angle (Z) can influence transfer characteristics between a vegetated surface and the atmosphere, e.g. the scalar roughness length (Sugita and Brutsaert, 1996). Iwata et al. (2010) note that, above tall vegetation, the vertical distribution of sources and sinks of scalars can vary both seasonally and during the day, depending on how deep light penetrates into the canopy. Therefore, the zenith angle is an obvious candidate as a possible effect on scalar transfer between the canopy and the atmosphere. In the following, we re-do the ϕ_a analysis for three different classes of zenith angle (which were found by trial-and-error to produce best results for the central class):

290 $0 < |Z| \leq 20$, $20 < |Z| \leq 60$ and $60 < |Z| \leq 90$.

In Fig. 6 (for the 39.4-m level), there is considerable improvement in the similarity relationships for all scalars when the sun is high, in the $0 < |Z| \leq 20$ class of central zenith angles. Results are better for temperature than for CO₂ and H₂O, but this may be related to intrinsic difficulties of measuring the latter: dust and vapor condensation on the IRGA mirror surfaces, and the effects of sensor separation, are possible causes (Tsai et al., 2012). Similar results were found for the 81.6-m level.

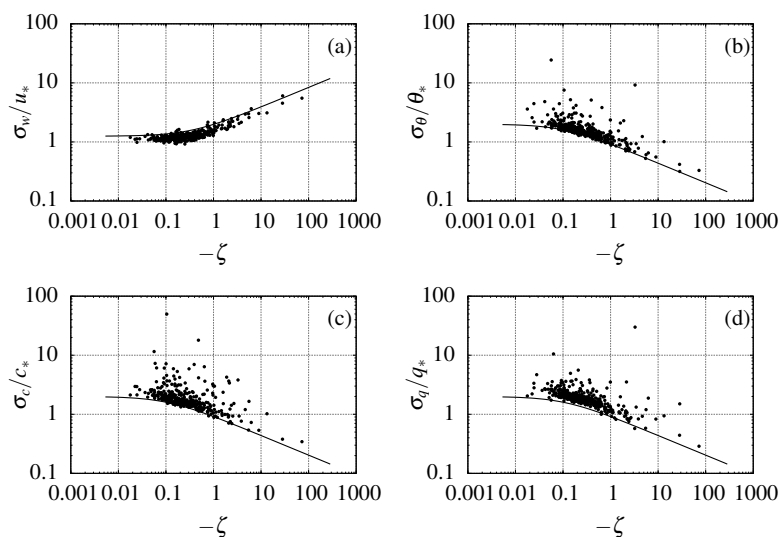


Figure 4. Dimensionless standard deviation for vertical velocity (a), temperature (b), CO₂ (c) and water vapor (d) — 39.4 m.

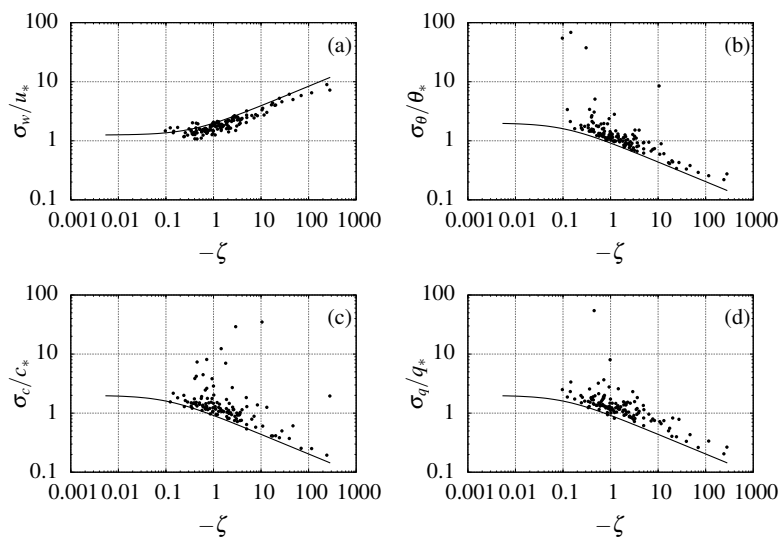


Figure 5. Dimensionless standard deviation for vertical velocity (a), temperature (b), CO₂ (c) and water vapor (d) — 81.6 m.

These results are encouraging: at least in the hours around noon, similarity relationships as good as those observed in the surface layer over low vegetation can be obtained. This opens up the possibility

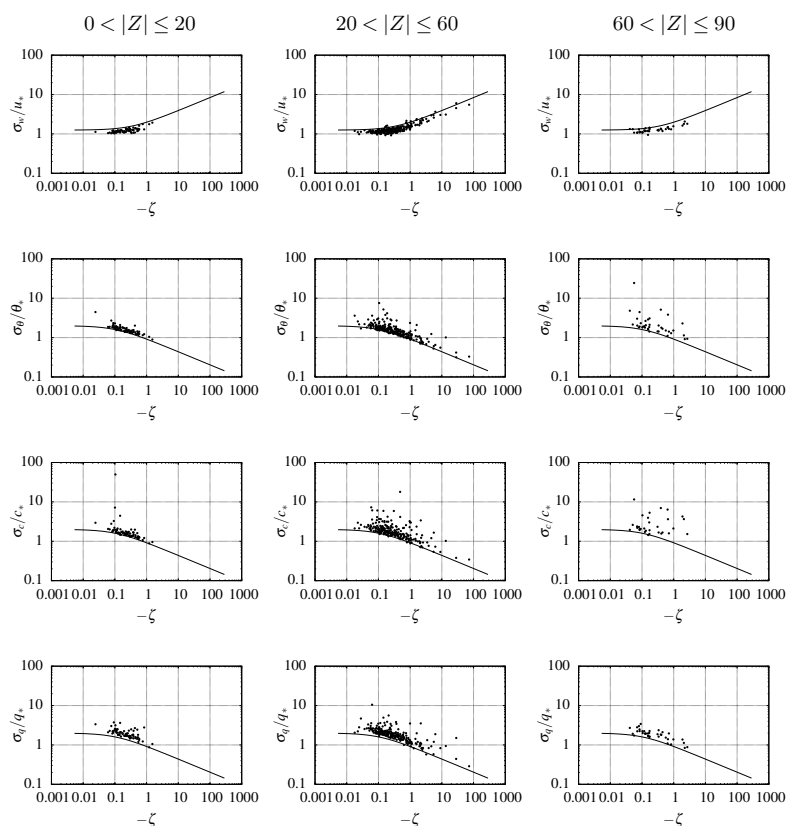


Figure 6. Dimensionless standard deviation with solar angle – 39.4 m. First row shows the velocity similarity function, second shows temperature function and third and fourth show CO₂ and water vapor, respectively.

300 of retrieving fluxes, by means of a host of standard MOST methods for these hours of the day. The results also require explanation. It is not immediately clear why these low zenith angles produce best results, but at least two (entirely phenomenological) explanations seem possible. One is vertical heterogeneity of sources and sinks, such as highlighted by Tsai et al. (2012): obviously, these sources and sinks may be more heterogeneous in the vertical under lower sunlight penetration. Horizontal
 305 heterogeneity, on the other hand, may also be playing a role: the vegetation height is not uniform, and a patchwork of shaded regions is clearly visible, from the tower, at higher values of the zenith angle. This may be enough to “activate”/“deactivate” sources of heat, CO₂ and H₂O at the nominal source level $z - d$, producing what is effectively a non-homogeneous horizontal surface with local advection effects, which may be very hard to identify with standard techniques.

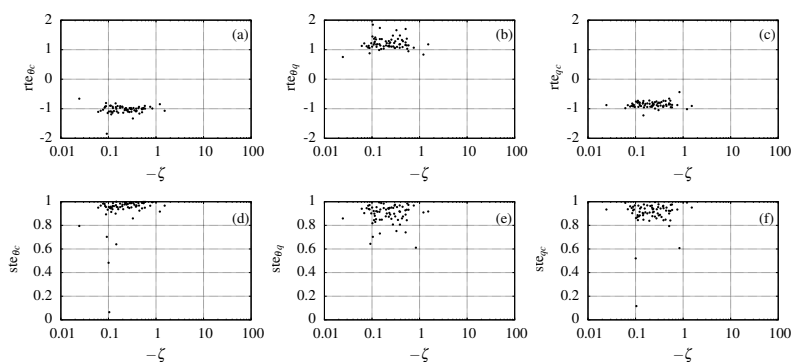


Figure 7. Scalar flux similarity indices r_{te} (above) and st_{e} (below) for $0^\circ < |Z| \leq 20^\circ$; θ - c (a), (d); θ - q (b), (e); and q - c (c), (f). For r_{te} , some data points were off the scale.

310 6 Scalar similarity indices and implications for flux estimation

6.1 Transport efficiencies

Not surprisingly, it turns out that success or failure of MOST-predictions in the roughness sublayer is also related to the degree of scalar similarity, although it seems that this aspect of RSL turbulence has not yet been fully explored. Two simple indices that are able to describe similarity between the
 315 fluxes of two scalars a and b are:

$$r_{te_{ab}} = \frac{r_{wa}}{r_{wb}}, \quad (10)$$

$$st_{e_{ab}} = 1 - \frac{||r_{wa}| - |r_{wb}||}{|r_{wa}| + |r_{wb}|}, \quad 0 \leq st_{e_{ab}} \leq 1, \quad (11)$$

where r_{wa} is the correlation coefficient between the vertical velocity fluctuation w' and the scalar fluctuation a' . Different from the correlation coefficient between the two scalars, r_{ab} , $r_{te_{ab}}$ is a better
 320 descriptor of scalar flux similarity (Cancelli et al., 2012). $st_{e_{ab}}$, proposed by Cancelli et al. (2012), has a similar interpretation, but is explicitly designed so that, unlike $r_{te_{ab}}$, it is always bounded from above by 1.

Both $r_{te_{ab}}$ and $st_{e_{ab}}$ were calculated for all pairs of scalars. Next, we discuss the results for 39.4 m, shown in Fig. 7, 8 and 9 (the 81.6-m results are similar). Again, we find best scalar similarity for the
 325 $0 < |Z| \leq 20$ range of zenith angles.

For most of the runs in the range $0 < |Z| \leq 20$, st_e is between 0.8 and 1.0. The most similar pair of scalar fluxes was CO_2 -temperature (c.f. Fig. 7-a and 7-d). We conjecture that this is related to the fact that the signs of the entrainment fluxes for these two scalars at the top of the atmospheric boundary layer are usually the same (negative), whereas the opposite is true for water vapor flux
 330 (positive, meaning detrainment instead of entrainment). This relationship between entrainment

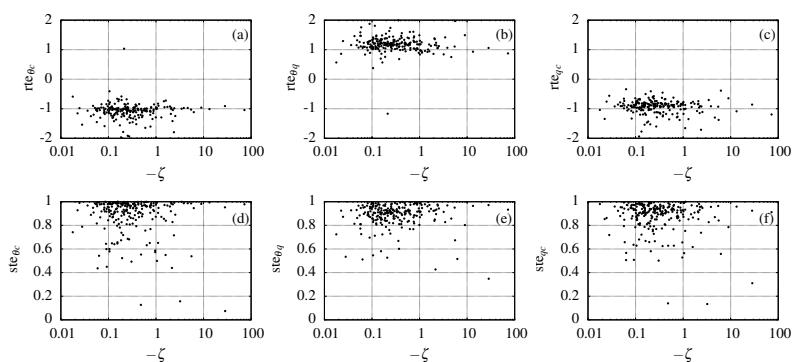


Figure 8. Scalar flux similarity indices r_{te} (above) and s_{te} (below) for $20^\circ < |Z| \leq 20^\circ$; θ - c (a), (d); θ - q (b), (e); and q - c (c), (f). For r_{te} , some data points were off the scale.

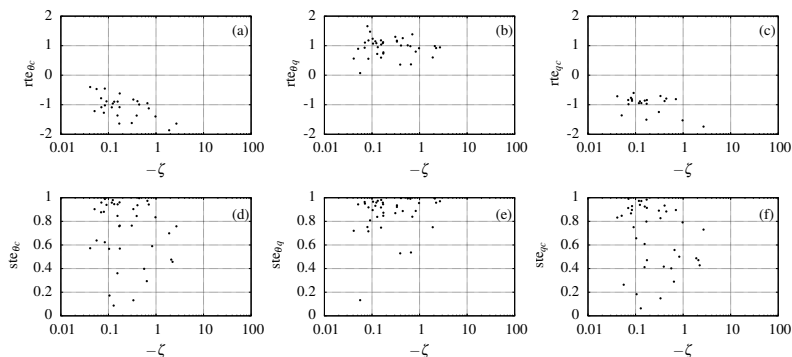


Figure 9. Scalar flux similarity indices r_{te} (above) and s_{te} (below) for $60^\circ < |Z| \leq 90^\circ$; θ - c (a), (d); θ - q (b), (e); and q - c (c), (f). For r_{te} , some data points were off the scale.

flux and similarity indices was observed, for example, in the LES simulations of Cancelli et al. (2014).

The similarity indices for the other zenith angle intervals are shown in Fig. 8 and 9. In the interval $20 < |Z| \leq 60$, the scalar fluxes still display a certain degree of similarity. In the interval $60 < |Z| \leq 90$, however, little similarity is observed. We note that these data often correspond to early morning and late afternoon periods, when scalar sources are usually “deactivated”. These results reinforce the picture of scalar fluxes emanating/being absorbed from different sources/sinks within the canopy and in the soil (c.f. Scanlon and Kustas (2010)).

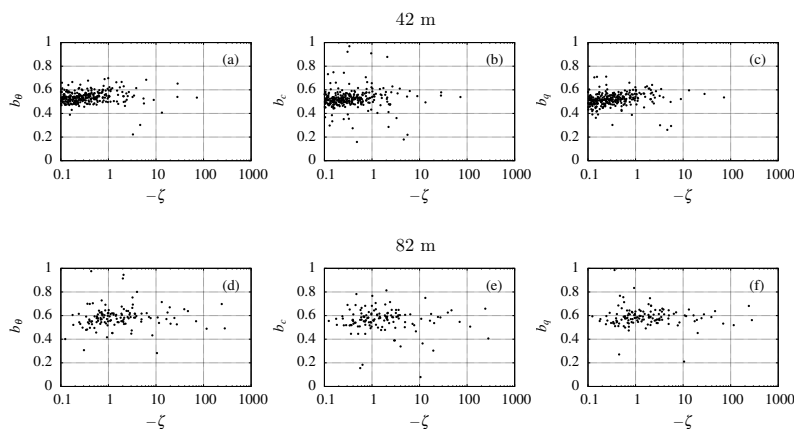


Figure 10. Relaxation coefficient from Eddy Accumulation Method, where (a) and (d) are temperature, (b) and (e) are CO_2 and (c) and (f) are water vapor.

6.2 Relaxed Eddy Accumulation

340 As reviewed in Sect. 3, the relaxed eddy accumulation is a valuable alternative for the measurement of scalars for which fast-response sensors are not available. This comes at the cost of the extra hypothesis of perfect scalar similarity.

Since there was no REA system installed at the tower, we have simulated the method using the eddy-covariance data. The fast scalar data were used to obtain updraft and downdraft REA samples
345 by conditional sampling of θ , c and q values for $w > 0$ and $w < 0$, respectively. These simulated samples were then averaged to obtain $\overline{\theta^+}$, $\overline{\theta^-}$, $\overline{c^+}$, $\overline{c^-}$, $\overline{q^+}$ and $\overline{q^-}$.

Given the results found in the previous subsection for rte and ste, it is natural to expect the REA method to perform better, again, in the range $0 < |Z| \leq 20$. In the following we analyze the relaxation coefficients b defined in Eq. (9), according to scalar and zenith angle.

350 The overall results, not classified according to zenith angle, are shown in Fig. 10, again for unstable conditions only. A few cases for which $b < 0$ were discarded, since the REA, similarly to flux-gradient methods, cannot cope with counter-gradient fluxes. Our data does not show any significant dependency of b on ζ , in agreement with Katul et al. (1996) and Businger and Oncley (1990).

In table 3, we give the (overall) means, medians and standard deviations of b for each scalar.
355 All the means are in the same range found by other authors (e.g. 0.51 – 0.62 by Katul et al. (1996)), which incidentally are values obtained for the surface layer *above* the roughness sublayer. The scatter in our data, however, is larger: at 81.6 m it reaches 0.56 for b_c .



Table 3. Statistics from REA in 39.4 and 81.6 m, where b_θ , b_c and b_q are the relaxation coefficients for temperature, CO₂ and water vapor, respectively.

	39.4 m			81.6 m		
	b_θ	b_c	b_q	b_θ	b_c	b_q
Mean	0.541	0.534	0.522	0.603	0.615	0.592
Median	0.534	0.520	0.517	0.580	0.573	0.583
Standard deviation	0.078	0.128	0.083	0.201	0.569	0.099

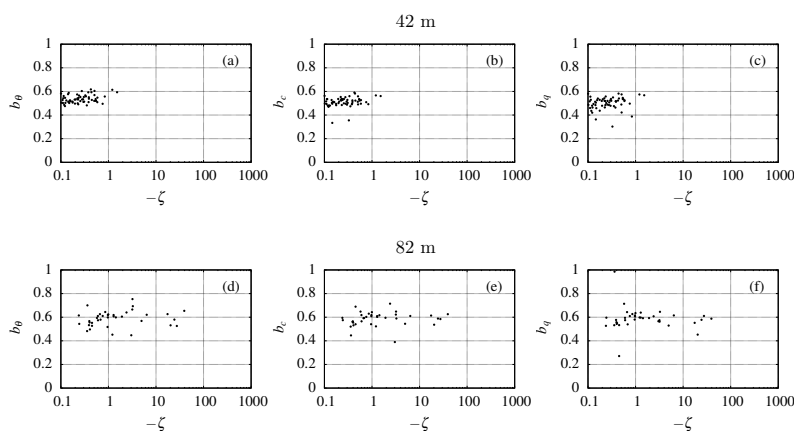


Figure 11. Relaxation coefficient variation with solar angle $-0 < |Z| \leq 20$. (a) and (d) are temperature, (b) and (e) are CO₂ and (c) and (f) are water vapor.

Our mean values of b are also somewhat smaller at 39.4 m (between 0.52 and 0.54) than at 81.6 m (between 0.59 and 0.61). This is similar to the results of Gao (1995) (shown in Table 2), where
 360 the mean $b_{\theta,q}$ values are 0.51 next to the canopy top (18 m) and 0.58 in 43.1 m (approximately two times the mean forest height average).

We classify the b -values according to zenith angle in Table 4. For each zenith angle interval, means and medians are quite similar among the three scalars. Moreover, for most cases in the intervals $0 < |Z| \leq 20$ and $20 < |Z| \leq 60$ the standard deviations are larger at 81.6 m.

365 For $0 < |Z| \leq 20$, the standard deviations of b_θ , b_c and b_q are significantly smaller (by a factor of 2) than for the other two classes, as well as for all data considered together. We note that in general the standard deviations for $60 < |Z| \leq 90$ are smaller than those for $20 < |Z| \leq 60$, but they are also more uncertain, due to the small number of points falling into that class. Overall, the better results for small zenith angles are confirmed for the REA method.



Table 4. Statistics from REA for each zenith angle class

	39.4 m			81.6 m		
$0 < Z \leq 20$	b_θ	b_c	b_q	b_θ	b_c	b_q
Mean	0.537	0.515	0.503	0.588	0.585	0.587
Median	0.531	0.513	0.516	0.598	0.594	0.589
Standard deviation	0.034	0.050	0.049	0.065	0.058	0.093
$20 < Z \leq 60$	b_θ	b_c	b_q	b_θ	b_c	b_q
Mean	0.540	0.539	0.526	0.619	0.634	0.596
Median	0.535	0.524	0.516	0.578	0.561	0.581
Standard deviation	0.051	0.128	0.093	0.236	0.696	0.093
$60 < Z \leq 90$	b_θ	b_c	b_q	b_θ	b_c	b_q
Mean	0.564	0.539	0.538	0.690	0.591	0.652
Median	0.522	0.497	0.531	0.704	0.551	0.617
Standard deviation	0.184	0.201	0.043	0.134	0.179	0.090

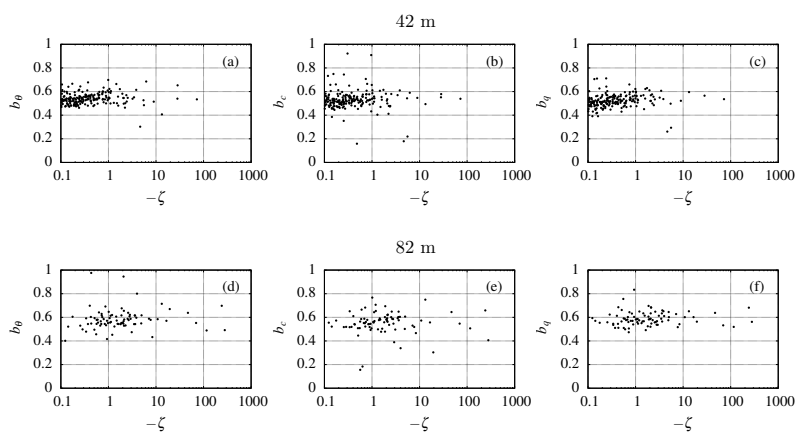


Figure 12. Relaxation coefficient variation with solar angle $-20 < |Z| \leq 60$. (a) and (d) are temperature, (b) and (e) are CO_2 and (c) and (f) are water vapor. Some data points were off the scale

370 7 Conclusions

An experimental study of the behavior of scalars in the roughness sublayer has been made, with the objective of assessing their departure from the predictions of MOST.

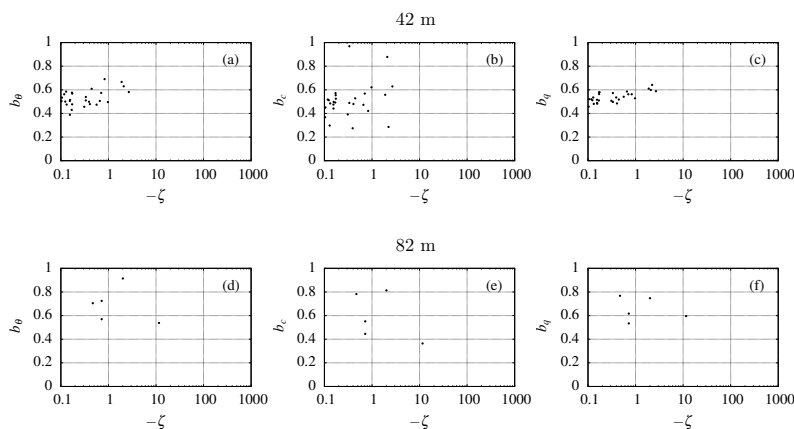


Figure 13. Relaxation coefficient variation with solar angle $-60 < |Z| \leq 90$. (a) and (d) are temperature, (b) and (e) are CO_2 and (c) and (f) are water vapor. Some data points were off the scale

The TKE dissipation rate departures are larger at the 39.4-m level and smaller at the 81.6-m level, suggesting a gradual transition out of the roughness sublayer. This is not confirmed, however, by
 375 all turbulence statistics that we analyzed. For example, the dimensionless scalar standard deviations ($\phi_{\theta,q,c}$) at 39.4 and 81.6 do not show significant differences. ϕ_w , on the other hand, remains much closer to the predictions of MOST at the two levels.

Moreover, an analysis of the scalar dissipation rates did not reveal any improvement in scalar behavior at smaller (i.e. inertial-subrange) scales, indicating that the observed departures from MOST
 380 are occurring at all scales.

A significant finding in this work is that the degree of departure from MOST predictions is related to the zenith angle. Fairly good adherence of $\phi_{\theta,q,c}$ to MOST (comparable to what is found in measurements above the RSL) was found for $0 < |Z| \leq 20$, with increasingly larger scatter as $|Z|$ increases. It is possible that this is related to strongly dissimilar scalar sources and sinks in the
 385 vertical when the sun is low, when different physical (i.e. heating) and biophysical (i.e. transpiration and photosynthesis) forcings are produced throughout the canopy. It is also possible that, at these higher zenithal angles, shading by the irregular tree heights produces a patchwork of different heat (and possibly water vapor and CO_2) sources, adding horizontal inhomogeneity.

The same pattern observed for the dimensionless scalar standard deviations appears with regard
 390 to the flux-similarity indices r_{te} and s_{te} . Again, they are closer to the theoretical values of ± 1 and $+1$ (respectively) at lower zenith angles, agreeing with the classical MOST predictions.

Finally, the b coefficients associated with the Relaxed Eddy Accumulation method were also affected by the zenith angle, with considerable improvement in the the range $0 < |Z| \leq 20$. We con-



395 firmed that b does not depend on stability ζ (for unstable conditions), and our values are in the same
range as previously observed values. The value of b close to the canopy (39.4 m) may turn out to be
slightly lower than above (81.6 m), similarly to what is reported by Gao (1995): a possible variation
with height within the RSL needs further research.

Data Availability

All data used in this study are kept in the ATTO Database at *Instituto de Pesquisas da Amazônia*.
400 Access should be requested to the ATTO Project Leaders.

Acknowledgements. We thank the Max Planck Society and the Instituto Nacional de Pesquisas da Amazonia for
continuous support. We acknowledge the support by the German Federal Ministry of Education and Research
(BMBF contract 01LB1001A) and the Brazilian Ministério da Ciência, Tecnologia e Inovação (MCTI/FINEP
contract 01.11.01248.00) as well as the Amazon State University (UEA), FAPEAM, LBA/INPA and SDS/CEUC/RDS-
405 Uatumã. Einara Zahn thanks Brazil's CAPES for her Master's scholarship. Leonardo Sá, Antônio Manzi and
Nelson L. Dias thank Brazil's National Research & Technology Development Council (CNPq) for their "Pro-
ductivity in Research" Grants 303728/2010-8, 312431/2013-9 and 303581/2013-1.



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