



1 Imprints of evaporation and vegetation type in diurnal 2 temperature variations

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9

10 Abstract. Diurnal temperature variations are strongly shaped by the absorption of solar radiation, but
11 evaporation, or the latent heat flux, also plays an important role. Generally, evaporation cools. Its
12 relation to diurnal temperature variations, however, is unclear. This study investigates the diurnal
13 response of surface and air temperatures to evaporation for different vegetation types. We used the
14 warming rate of temperature to absorbed solar radiation in the morning under clear-sky conditions and
15 evaluated how the warming rates change for different evaporative fractions. Results for 51 FLUXNET
16 sites show that the diurnal variation of air temperature carries very weak imprints of evaporation across
17 all vegetation types. However, surface temperature warming rates of short vegetation decrease
18 significantly by $\sim 23 \times 10^{-3} \text{ K/W m}^{-2}$ from dry to wet conditions. Contrarily, warming rates of surface
19 and air temperatures are similar at forest sites and carry literally no imprints of evaporation. We
20 explain these contrasting patterns with a surface energy balance model. The model reveals a strong
21 sensitivity of the warming rates to evaporative fraction and aerodynamic conductance. However, for
22 forests the sensitivity to evaporative fraction is strongly reduced by 74 % due to their large
23 aerodynamic conductance. The remaining imprint is reduced further by $\sim 50\%$ through their enhanced
24 aerodynamic conductance under dry conditions. Our model then compares the individual contributions
25 of solar radiation, evaporation and vegetation types in shaping the diurnal temperature range. These
26 findings have implications for the interpretation of land-atmosphere interactions and the influences of
27 water limitation and vegetation on diurnal temperatures, which is of key importance for ecological
28 functioning. We conclude that diurnal temperature variations may be useful to predict evaporation for
29 short vegetation. In forests, however, the diurnal variations in temperatures are mainly governed by
30 their aerodynamic properties resulting in no imprint of evaporation in diurnal temperature variations.

31

32 1 Introduction

33 Temperature is one of the most widely monitored variables in meteorology. Besides being important
34 for our day-to-day activities, temperature serves as a primary attribute for the understanding of Earth
35 system processes. The diurnal variation of temperature is considered informative in climate science, as
36 described by the diurnal temperature range (DTR), which is basically the difference between daily
37 maximum and minimum temperatures. Information on the diurnal temperature range has facilitated a
38 broad spectrum of research including agriculture, health welfare, climate change and ecological
39 studies.



40 Over land the diurnal variation of temperature is mainly driven by the solar energy input (Bristow and
41 Campbell, 1984). Liu et al., (2004) shows a high correlation of 0.88 between the annual solar radiation
42 and DTR in China. Likewise, Makowski et al., (2009) found their annual correlation to be 0.87 for
43 Europe. Their obvious and still intricate association is also important in determining the influence of
44 solar dimming and brightening on diurnal temperature variations (Wang and Dickinson, 2013; Wild,
45 2005).

46 Solar radiation is the dominant, but not the only, factor shaping the diurnal temperature variation.
47 Available energy at the surface is partitioned into latent and sensible heat flux. A higher latent heat flux
48 signifies higher evaporation, which reduces the temperature through evaporative cooling, an effect that
49 can be seen in global climate model sensitivities to land evaporation (Shukla and Mintz, 1982).
50 Another climate model-based analysis (Mearns et al., 1995) shows that differences in evaporation
51 explain 52 % of the variance in DTR in the summer season for the USA. Similarly, climate model
52 simulations also show the high sensitivity of DTR to evaporation especially in the summer season
53 when evaporation is not energy limited (Lindvall and Svensson, 2015). Consequently, methods to
54 estimate evaporation use air temperature (Blaney and Cridlle, 1950; Hargreaves and Samani, 1985;
55 Thornthwaite, 1948) and remotely sensed surface temperature (Anderson et al., 2012; Boegh et al.,
56 2002; Jackson et al., 1999; Kustas and Norman, 1999; Price, 1982; Su et al., 2007). Most of the surface
57 energy balance based estimates of evaporation use DTR as an input (Baier and Robertson, 1965;
58 Vinukollu et al., 2011; Yao et al., 2013).

59 Clouds, precipitation, and atmospheric composition are also important factors that determine DTR (Dai
60 et al., 1999; Stenchikov and Robock, 1995). One can exclude their contribution to some extent by
61 considering only clear sky days to more clearly identify the role of evaporation on DTR. Furthermore,
62 the partitioning of the turbulent heat fluxes into sensible and latent heat is also affected by vegetation
63 type. Taller vegetation has a higher aerodynamic conductance that facilitate mass and heat exchange
64 between land and atmosphere (Jarvis and McNaughton, 1986). The greater conductance in forests
65 reduces their DTR by reducing their maximum temperature (Bevan et al., 2014; Gallo, 1996; Jackson
66 and Forster, 2010). Few studies captured the impact of aerodynamic properties of vegetation on
67 temperature, for example, in terms of the decomposed temperature metric theory (Juang et al., 2007;
68 Luyssaert et al., 2014) and the theory of intrinsic biophysical mechanism (Lee et al., 2011; Zhao et al.,
69 2014). Generally, the lower temperatures of forests are associated with their mean evaporative
70 environment, although this may be affected by periods of dry and wet conditions.

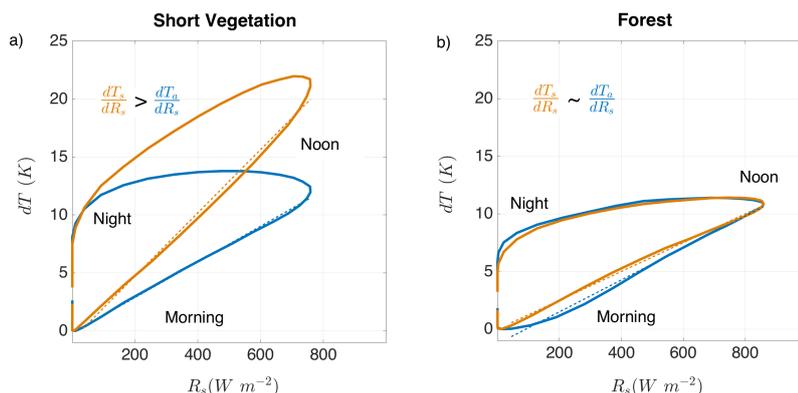
71 In this study we investigate how the diurnal variation in surface and air temperature responds to
72 changes in evaporative conditions in different vegetation types. Clearly, DTR is not independent of
73 solar radiation, which is why we develop an alternative characteristic, the warming rate that eliminates
74 the contribution of solar radiation. To illustrate this, observed diurnal air and surface temperatures are
75 plotted against absorbed solar radiation for a cropland and forest site in Figure 1. The diurnal evolution
76 of temperature is mainly governed by the absorbed solar radiation (R_s); this is discernible from the
77 linear increase in the morning ($20 \text{ W m}^{-2} \leq R_s \leq R_{s,\text{max}}$), as described by the slope. This dependence is



78 accounted for by what we refer to as the warming rate, the increase in temperature due to a unit
79 increase in the absorbed solar radiation, expressed as the derivative dT_a/dR_s for air temperature and
80 dT_s/dR_s for surface temperature with units of $K/W m^{-2}$. One can approximate the warming rate by the
81 ratio of DTR to maximum solar radiation, so that the warming rate can be seen as an efficient
82 characteristic that captures effects on DTR that are not caused by solar radiation. In this study, we use
83 linear regressions of observed data from the morning to noon to calculate warming rates.

84

85



86

87 **Figure 1** Mean diurnal hysteresis formed by plotting the diurnal temperature anomaly (y-axis) against
88 absorbed solar radiation (x-axis) for summer clear sky days. Surface temperature is depicted in orange
89 and air temperature in blue. (a) A short vegetation cropland site (US-ARM) in Southern Great Plains
90 Lamont OK, United States. (b) A forest site (CA-TP4) in Ontario, Canada. The dashed lines are the
91 linear regression of the observations falling in the morning slope of the hysteresis that corresponds to
92 the warming rate (dT/dR_s) of air (T_a) and surface temperature (T_s).

93

94 The temperature warming rate provides insights on the effects of vegetation on the diurnal variation of
95 temperatures. Figure 1a shows a greater surface temperature warming compared to air temperature for
96 a cropland site. Contrarily, the warming rates of the two temperatures are similar for a forest site
97 (Figure 1b). This indicates the strong coupling of diurnal air and surface temperatures in forests
98 compared to short vegetation.

99

100 Certainly, it is intriguing to find out how evaporation alters this coupling. In our earlier work (Panwar
101 et al., 2019) we looked at the temperature warming rate for a cropland site in the Southern Great Plains.
102 We observed that the warming rate of surface temperature decreases from dry (less-evaporative) to wet
103 (evaporative) conditions but the warming rate of air temperature remained unaffected by evaporation.
104 Combining the boundary layer information and heat budget expression we explained that the diurnal
105 variation of air temperature does not contain the imprints of evaporation due to the compensating role
106 of boundary layer development. If this is a general finding, then the surface temperature warming rate



107 can be used for estimating evaporation of short vegetation. However, it is also interesting to see how
108 the evaporative cooling effect competes with the cooling effect by a higher aerodynamic conductance.

109

110 In this study, we approach two major questions to advance our understanding of diurnal temperature
111 variations: a) Do the diurnal variations of surface and air temperature respond to evaporation? and b)
112 What is the role of the different aerodynamic conductance of vegetation in altering these responses?
113 Our previous work (Panwar et al., 2019) already shows the stronger imprints of evaporation in diurnal
114 surface temperature variations. Here we examine the generality of this finding in short vegetation.
115 Additionally, to understand the role of aerodynamic conductance in modifying these imprints we
116 analyze data from the taller and more complex vegetation like savanna and forests.

117

118 We first present a model based on the surface energy balance to provide an expression for the surface
119 temperature warming rate and its response to evaporation and aerodynamic conductance (all variables
120 used are summarized in Table A1). To evaluate our model, we used observations from 51 FLUXNET
121 sites that include short vegetation, savanna and forests. Surface and air temperature warming rates,
122 aerodynamic conductances and their response to evaporation are quantified for each site. We then use
123 these findings with our model to explain and reproduce observed temperature warming rates and their
124 response to evaporation. The cooling effect of evaporation and its relation to aerodynamic conductance
125 is quantified for each vegetation type. Combining the warming rates with the information on solar
126 radiation, we conclude the study by demonstrating the contribution of solar radiation, evaporation and
127 aerodynamic conductance in shaping the DTR using our observational analysis and model.

128

129 **2 Modeling temperature-warming rate**

130 Surface and air temperatures possess a strong diurnal variation that is driven by the absorption of solar
131 radiation. The amplitude of this variation is also affected by other components of surface energy
132 balance, among which the partitioning of turbulent heat fluxes into latent and sensible heat is
133 important. Generally, the surface energy balance is written as

134

$$135 R_s = R_{l,net} + LE + H + G . \quad \text{Eq. (1)}$$

136

137 Here, R_s is the absorbed solar radiation at the surface, $R_{l,net}$ is the net longwave radiation, LE is the
138 latent heat flux (with L being the latent heat of vaporization and E the evaporation rate), H is the
139 sensible heat flux and G is the ground heat flux. For simplification of the surface energy balance we
140 linearize $R_{l,net}$ using the first order terms, such that $R_{l,net} = R_o + k_r (T_s - T_{ref})$. Here, R_o is the net
141 radiation at a reference temperature T_{ref} . The second term, $k_r = 4 \sigma T_{ref}^3$ is the linearization constant.
142 Incorporating this simplification of $R_{l,net}$ in Eq. (1), the surface energy balance can be rearranged to
143 yield an expression for T_s ,

144



$$T_s = T_{ref} + \frac{R_s - R_o - LE - H - G}{k_r} \quad \text{Eq. (2)}$$

145

146 The warming rate of surface temperature is obtained by taking the derivative of Eq. 2 with respect to
147 absorbed solar radiation, R_s . The warming rate of surface temperature is given by

148

$$\frac{dT_s}{dR_s} = \frac{1}{k_r} - \frac{1}{k_r} \frac{d(H + LE)}{dR_s} \quad \text{Eq. (3)}$$

149

150 Since, R_o and T_{ref} are assumed to be constants and do not vary diurnally with R_s , they disappear in
151 Eq. (3). Additionally it is assumed that the diurnal change in G in response to R_s is negligible
152 ($dG/dR_s \sim 0$) compared to other components of surface energy balance. This assumption is valid since
153 we are considering vegetated sites for our study, although we are aware that for non-vegetated surfaces
154 G can represent a noticeable share of absorbed solar radiation (Clothier et al., 1986; Kustas and
155 Daughtry, 1990).

156

157 We describe the evaporative conditions by the evaporative fraction (f_e), the ratio of the latent heat
158 flux (LE) to the total turbulent heat fluxes ($H + LE$). Given this, the term $H + LE$ in Eq. (3) can be
159 written as $H/(1 - f_e)$. Furthermore, the sensible heat flux can be expressed in terms of the
160 aerodynamic conductance as $H = c_p \rho g_a (T_s - T_a)$, where $c_p = 1005 \text{ J/kg K}$ is the specific heat capacity
161 of air, $\rho = 1.23 \text{ kg m}^{-3}$ is air density and g_a is the aerodynamic conductance. On including these
162 replacements in Eq. (3) we get an approximation for the surface temperature warming rate

163

$$\frac{dT_s}{dR_s} = \frac{(1 - f_e) + c_p \rho g_a (dT_a/dR_s)}{k_r (1 - f_e) + c_p \rho g_a} \quad \text{Eq. (4)}$$

164

165 where dT_a/dR_s is the air temperature warming rate. We can further simplify this expression by
166 considering the two terms in the denominator of Eq. (4). Considering $T_{ref} \sim 288 \text{ K}$, the term
167 $k_r (1 - f_e)$ varies by $\sim 4.87 \text{ Wm}^{-2}\text{K}^{-1}$ to $\sim 0.54 \text{ Wm}^{-2}\text{K}^{-1}$ from dry ($f_e \sim 0.1$) to wet ($f_e \sim 0.9$) conditions,
168 which is much smaller in magnitude compared to the term $c_p \rho g_a$ that is $\sim 60 \text{ Wm}^{-2}\text{K}^{-1}$ for a typical
169 cropland site ($g_a = 0.05 \text{ m s}^{-1}$) and $250 \text{ W m}^{-2}\text{K}^{-1}$ for a typical forest site ($g_a = 0.2 \text{ m s}^{-1}$). Because of
170 these magnitudes, the term $k_r (1 - f_e)$ can be neglected. This leads to a further simplification of the
171 warming rate to

172

$$\frac{dT_s}{dR_s} \approx \frac{(1 - f_e)}{c_p \rho g_a} + \frac{dT_a}{dR_s} \quad \text{Eq. (5)}$$

173

174 Eq. (5) shows that morning to noon warming of surface temperature is a function of evaporative
175 fraction, aerodynamic conductance and also of the warming rate of air temperature.

176



177 Finally, the sensitivity of the warming rate to changes in evaporative conditions is obtained by taking
178 the derivative of Eq. (5) with respect to evaporative fraction (f_e). To express these derivatives with
179 respect to evaporative fraction, we use the apostrophe ($d/df_e = ')$). Therefore, dT_s'/dR_s and
180 dT_a'/dR_s represent the change in surface and air temperature warming rates due to a unit change in
181 evaporative fraction. Similarly, g'_a is the change in aerodynamic conductance from dry to wet
182 evaporative conditions. We obtain:

183

$$\frac{dT_s'}{dR_s} = -\frac{1}{c_p \cdot \rho \cdot g_a} - \frac{1 - f_e}{c_p \cdot \rho \cdot g_a} \cdot \frac{g'_a}{g_a} + \frac{dT_a'}{dR_s} \quad \text{Eq. (6)}$$

184

185 This model provides two important expressions that we test with observations. The first expression is
186 the warming rate of surface temperature, described by Eq. (5), which requires the information of the
187 warming rate of air temperature, aerodynamic conductance and evaporative fraction. On multiplying
188 these two equations with daily maximum solar radiation shall provide an approximation of DTR that
189 can also be validated with the observational data. The second expression is the response of the surface
190 temperature warming rate to evaporation, shown in Eq. (6), which is a negative quantity provided
191 dT_a'/dR_s is small (or negative). The negative sign means that the surface temperature warming rate
192 decreases with increase in evaporative fraction. The amplitude of this decrease mainly depends on the
193 characteristic aerodynamic conductance (g_a) of vegetation (the first term on the right hand side of Eq.
194 6) and also on its relative sensitivity to evaporative fraction (g'_a/g_a , the second term on the right hand
195 side).

196

197 3 Data and method

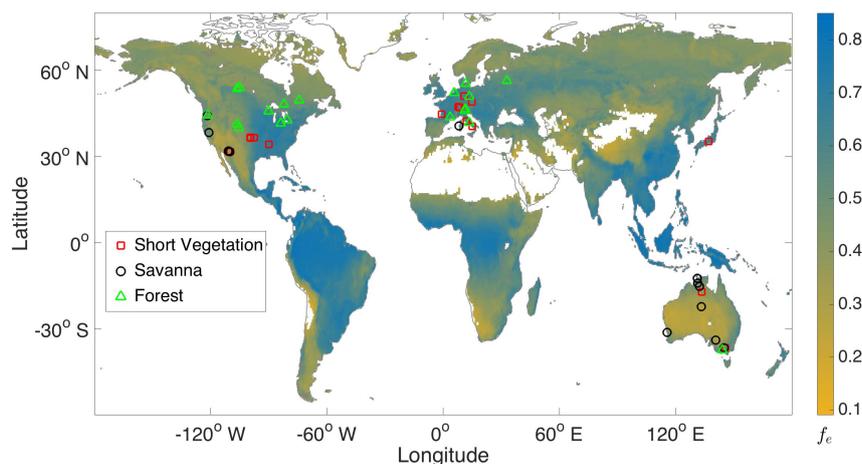
198 We use observations from 51 FLUXNET sites representing different vegetation types. The FLUXNET
199 data consists of sensible and latent heat fluxes using the standard eddy covariance method and provides
200 half hourly radiation and meteorological data (Baldocchi et al., 2001). The selected 51 sites contain
201 data of the surface energy balance components and temperatures for more than four years. To avoid the
202 effect of energy limitation on evaporation only summer days are considered. Summer is defined here as
203 days having their daily mean incoming solar radiation at the surface greater than the median of the
204 annual distribution. This approach standardizes the definition of summer days for sites at different
205 latitudes and provides the days with comparable solar energy input for the individual sites.

206

207 Furthermore, among summer days only clear sky days are considered to avoid the influence of clouds
208 on temperatures. The process of obtaining summer days already filters out the days with high duration
209 of cloud covers that result in reduced mean incoming solar radiation. An additional filter to remove
210 cloudy days is applied that is based on the quantile regression method using surface solar radiation and
211 potential solar radiation (Renner et al., 2019). This method was applied only from morning to noon, so
212 that if the day has clouds in the evening, it is still considered as a clear sky day. This does not influence
213 warming rates since they are calculated only from the morning to noontime variation of temperature.



214



215

216 **Figure 2** Geographical locations of FLUXNET sites used in this study. The vegetation type at each site
 217 is shown by the symbols. The color bar shows the mean annual evaporative fraction (f_e) derived from
 218 FLUXCOM data (2001 to 2013).

219

220 The vegetation type of each site is classified using the International Geosphere-Biosphere Programme
 221 (IGBP) Data and Information System (Loveland and Belward, 1997). The IGBP land cover product is
 222 available at 1 km resolution and was derived from the Advance Very High Resolution Radiometer
 223 (AVHRR). Detailed information of each site with their location, number of days used in the analysis,
 224 land use type and references is provided in the Appendix (Table A2). Vegetation are classified into
 225 three types that is based on their typical vegetation height and coverage, see Table 1. Shorter vegetation
 226 like croplands, grasslands, and shrublands are grouped into the ‘short vegetation’ type. Savanna
 227 ecosystems are complex with heterogeneous vegetation height, which basically delineates the transition
 228 of short vegetation to forests, and are grouped into the ‘savanna’ type. All forest types, including
 229 deciduous broadleaf, evergreen broadleaf, evergreen needleleaf and mixed, are grouped in the ‘forest’
 230 type.

231

232 The geographic location of the selected 51 sites is shown in Figure 2. The color bar represents the
 233 mean annual evaporative fraction derived from FLUXCOM data (Jung et al., 2019; Tramontana et al.,
 234 2016). Selected sites represent a wide range of ecosystems that is ideal for studying the generality of
 235 the response of warming rates to differences in evaporative conditions and vegetation type.

236

237 **Table 1.** Land use types of the different sites considered here and their grouping into the short
 238 vegetation, savanna and forest types.

Vegetation types	Land use type	Number of sites
Short Vegetation	Cropland	12



	Grassland	6
	Shrubland	5
Savanna	Savanna	4
	Woody Savanna	5
Forest	Deciduous broadleaf forest	4
	Evergreen broadleaf forest	1
	Evergreen needle leaf forest	9
	Mixed forest	5

239

240 The evaporative condition is quantified by evaporative fraction. One of the advantages of evaporative
241 fraction is its stability for daylight hours such that it can be assumed to be constant over a day
242 (Shuttleworth et al., 1989). Daily evaporative fraction is obtained by the linear regression of half hourly
243 morning to noon values of the ratio of the latent heat flux to the total turbulent heat fluxes. Similarly, a
244 linear regression of half hourly warming rate and evaporative fraction values is used to quantify the
245 response of the warming rate to evaporative fraction.

246

247 We use the term air temperature for the temperature measured above the canopy. Surface temperature
248 is calculated from the upwelling longwave radiation using the Stefan-Boltzmann law, such that the
249 surface temperature is the skin temperature of the vegetation. The aerodynamic conductance (g_a) is
250 obtained from the observed frictional velocity (u_*) and wind speed (u) by $g_a = u_*^2/u$ (see, e.g.,
251 Verma (1989)). For simplicity, the conductance of heat fluxes and momentum are assumed to be
252 identical (Mallick et al., 2016).

253

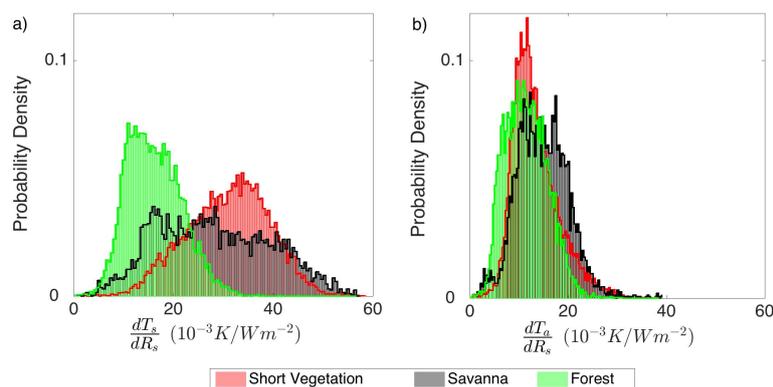
254 **4 Results**

255 4.1 Observational analysis

256

257 The primary advantage of warming rate over DTR is its suitability to compare sites with different solar
258 energy input. This is apparent from Figure 3, where we show the probability density distribution of the
259 observed daily warming rates of surface (a) and air temperatures (b) for short vegetation, savanna, and
260 forest. We look at the surface and air temperature warming rates to determine if they carry any
261 information on vegetation type. In general, the surface temperature warming rate of short vegetation is
262 larger by almost a factor of two compared to the surface temperature warming rate of forests. Savanna
263 covers the range in surface temperature warming rates, reflecting their characteristics being positioned
264 between short vegetation and forests. Hence, the vegetation type clearly affects the surface temperature
265 warming rate. Surprisingly, this is not true for air temperature warming rates. Short vegetation, savanna
266 and forests show similar distributions of air temperature warming rate. The air temperature warming
267 rate of short vegetation is smaller than its surface temperature warming rate. Conversely, in forests, the
268 magnitudes are similar, indicating the strong coupling between surface and air temperature.

269



270

271

272 **Figure 3** Probability density distribution of observed (a) surface temperature warming rates and (b) air
273 temperature warming rates for short vegetation (red), savanna (grey) and forest (green).

274

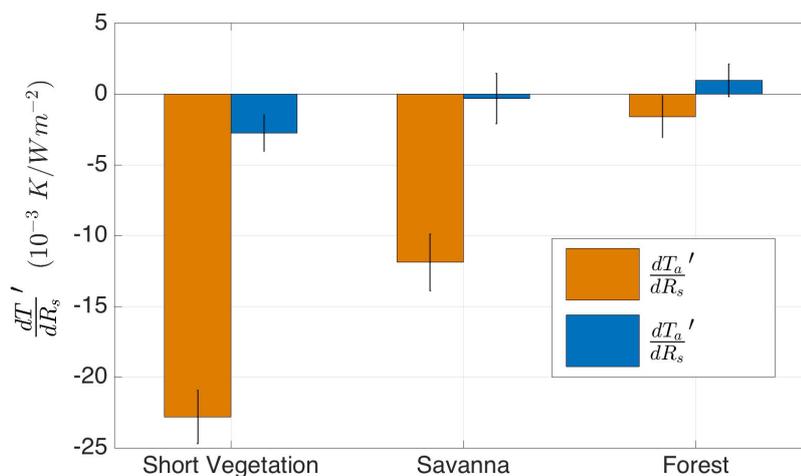
275 Site-specific information on warming rates is provided in Figure A1 in the Appendix. Within the short
276 vegetation type, grassland and shrubland sites show much greater surface temperature warming rates
277 than air temperature warming rates. This distinction could be attributed to the site-specific evaporative
278 conditions. Evaporative conditions at some sites have a certain general tendency. For instance, most of
279 the shrubland sites are drier, cropland sites are generally wetter and forest sites show intermediate
280 evaporative fractions. Such an uneven distribution of evaporative conditions could impact the warming
281 rates, such that it is higher for dry and lower for wet sites. On the other hand, despite these differences
282 in the mean, the sites contain days with a good range of evaporative fractions (see Figure A2 in
283 Appendix). The range of evaporative fractions is important to calculate the sensitivity of warming rates
284 to evaporative fraction.

285

286 Next, we quantify the response of surface and air temperature warming rates to changes in evaporative
287 fraction from dry to wet conditions. The warming rate response to evaporative fraction is obtained from
288 the linear regression of daily warming rates to daily evaporative fractions for each site. Figure 4 shows
289 the mean response of the surface (orange) and air (blue) temperature warming rate to evaporative
290 fraction for short vegetation, savanna and forest. For site-specific responses, see Figure A2 in the
291 Appendix. It is noticeable that regardless of the magnitudes of the warming rates and different mean
292 evaporative conditions, the response of warming rates to evaporative fraction is almost consistent for
293 the different vegetation types. For instance, the surface temperature warming rate of short vegetation
294 shows a consistent decrease of $\sim 23 \times 10^{-3} K/W m^{-2}$ from dry to wet days. However, the air temperature
295 warming rate decreases only by $\sim 5 \times 10^{-3} K/W m^{-2}$. In our earlier work, similar responses were
296 observed for a cropland site (See Appendix, site 8). We find a similarly weak response for savanna and
297 forests. In savanna, the surface temperature warming rate still decreases by $\sim 12 \times 10^{-3} K/W m^{-2}$ from dry
298 to wet conditions, but the air temperature warming rate remains almost the same. In forests, both,



299 surface and air temperature warming rates, show very weak to almost no response to evaporative
300 fraction, although with some variations as reflected by the error bars.



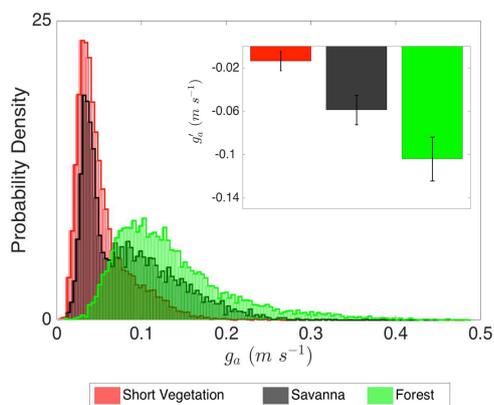
301
302 **Figure 4** Bar plots of the observed mean response of surface (dT_s'/dR_s) and air (dT_a'/dR_s)
303 temperature warming rates to changes in evaporative fraction for short vegetation, savanna and forests.
304 The error bars represent the standard error in the mean of all sites in the respective type.

305
306 Besides evaporation, the aerodynamic conductance also influences the diurnal variation of temperature.
307 The aerodynamic conductance governs the ventilation of energy and mass from the surface to the
308 atmosphere (Thom, 1972). Figure 5 shows the mean aerodynamic conductances for the vegetation
309 types. The mean aerodynamic conductance is usually a characteristic of vegetation height but
310 variations might occur due to changing evaporative conditions. In general, it is observed that the
311 aerodynamic conductance of short vegetation is much lower than the aerodynamic conductance of
312 forest. Savannas show relatively higher aerodynamic conductances compared to short vegetation. Some
313 woody savannas have comparable aerodynamic conductances to forests. Forests have generally high
314 aerodynamic conductances.

315
316 In addition to the mean aerodynamic conductance we also observed its response to evaporative
317 fraction. The change in aerodynamic conductance due to the change in evaporative fraction is denoted
318 by g'_a that is derived from the linear regression of their observed daily values. The negative sign of
319 g'_a reflects the decrease in g_a from dry to wet days so that the aerodynamic conductance is enhanced
320 on days with low evaporative fraction. Site-specific values of g_a and g'_a are provided in Figure A3 in
321 the Appendix. For all vegetation types the aerodynamic conductance increases on dry days. For short
322 vegetation this increase is about ~50 % such that the characteristic aerodynamic conductance of 0.041
323 m s⁻¹ increases to 0.055 m s⁻¹ on dry days. For forests, the aerodynamic conductance for most of the
324 sites increases by ~100% on dry days. This enhancement becomes considerably important for forests
325 because it increases their already large aerodynamic conductance, for instance from 0.12 m s⁻¹ to 0.24



326 m s^{-1} . However, the main distinction between forest and short vegetation remains their mean
 327 aerodynamic conductance whereas the enhanced aerodynamic conductance is just a secondary factor
 328 whose impact on warming rate is further analyzed using our model.



329
 330 **Figure 5** Probability density distribution of the aerodynamic conductance (g_a) derived from
 331 observations at the 51 FLUXNET sites for short vegetation, savanna and forests. The inset plot shows
 332 the mean sensitivity of aerodynamic conductance to evaporative fraction (f_e) for the three types, g'_a .
 333 The error bar represents the standard error in the mean regression of g_a and f_e .

334
 335 **Table 3.** First quartile (Q1), median and third quartile (Q3) for the observed distribution of dT_s/dR_s ,
 336 dT_a/dR_s and g_a for short vegetation, savanna and forest.

Vegetation		dT_s/dR_s ($10^{-3} \text{ K/W m}^{-2}$)	dT_a/dR_s ($10^{-3} \text{ K/W m}^{-2}$)	g_a (m s^{-1})
Short Vegetation	Q1	25.1	9.9	0.029
	Median	31.4	12.3	0.041
	Q3	36.7	15.7	0.063
Savanna	Q1	18.6	10.9	0.037
	Median	27.1	14.4	0.068
	Q3	36.8	18.1	0.115
Forest	Q1	11.8	8.1	0.085
	Median	15.5	11.1	0.118
	Q3	19.7	14.3	0.164

338
 339 To summarize our observational analysis, we show that the diurnal variation of surface temperature of
 340 short vegetation carries stronger imprints of evaporative conditions compared to the diurnal variation
 341 of air temperature. In forests, the diurnal variations of both, surface and air temperature do not respond
 342 to evaporative conditions. Observations also demonstrate characteristic high aerodynamic
 343 conductances of forests compared to short vegetation. Additionally, we showed an enhancement of



344 aerodynamic conductance on dry days that is relatively stronger for forests compared to short
345 vegetation.

346

347 To explain these findings we hypothesize that the high aerodynamic conductance of forest and its
348 enhancement on dry conditions lowers the diurnal warming of surface temperature. Consequently, the
349 warming rates of surface temperature of forests are less sensitive to evaporation. Our hypothesis is
350 based on the observational based findings and the interpretation of the model equations where one can
351 determine the contribution of g_a , f_e and g'_a in shaping the warming rates. This can already be
352 anticipated from Eq. 6, evaluated with the median values provided in Table 3. The first term on the
353 right-hand side of Eq. 6 is about $-21 \times 10^{-3} \text{ K}/(\text{W m}^{-2})$ for short vegetation, but only $-7 \times 10^{-3} \text{ K}/(\text{W m}^{-2})$
354 for forests, similar to what is shown in Figure 4. In the next section we verify our hypothesis using the
355 modeled expression for surface temperature warming rate and its response to evaporative conditions.
356 Along with the model evaluation we quantify the contribution of aerodynamic conductance and its
357 enhancement in compensating the imprints of evaporation on warming rates for surface temperature for
358 short vegetation, savannas and forests.

359

360 4.2 Model interpretation

361

362 In this section we estimate the surface temperature warming rate and its response to evaporative
363 fraction using our model, which is then compared to observations. Then we use the model to quantify
364 the contribution of evaporative fraction and of aerodynamic conductance to the diurnal temperature
365 range.

366

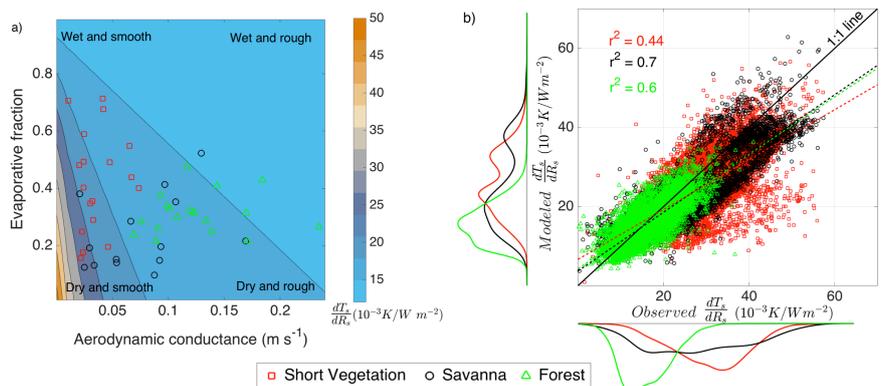
367 To model the warming rate we use Eq. (5), in which the vegetation type is captured by g_a , and
368 evaporative conditions by f_e . The model sensitivity of surface temperature warming rate to evaporative
369 fraction and aerodynamic conductance is shown in Figure 5a. The model shows a stronger gradient of
370 the warming rate with evaporative fraction for low aerodynamic conductances. As in the observations
371 warming rates for low aerodynamic conductances are greater compared to high aerodynamic
372 conductances. Observed warming rates for short vegetation, savannas and forests are also plotted in
373 Figure 5a, using their mean evaporative fractions and aerodynamic conductances, respectively. Note
374 that both, evaporative fraction and aerodynamic conductance, can vary. This implies that the position
375 of the sites can change vertically and somewhat horizontally with changes in evaporative conditions.

376

377 Almost all of the short vegetation sites have low aerodynamic conductances where the warming rate
378 increases with decreasing evaporative fraction. Contrarily, the forest sites show no such strong
379 variation in warming rate. This is consistent with the study by Diak and Whipple (1993), who showed
380 a similar dependency of the diurnal range of surface temperature on the Bowen ratio and surface
381 roughness length using a boundary layer model simulation. Our model can capture these patterns solely
382 with surface energy balance information and requires no information of the boundary layer. This



383 indicates that the diurnal variation in surface temperature is dominantly governed by the exchange at
 384 the surface, particularly aerodynamic conductance and evaporative fraction.
 385



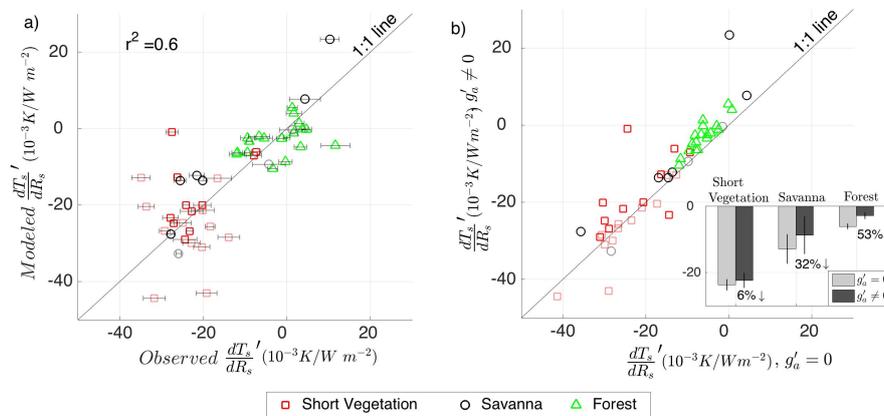
386
 387 **Figure 6** a) Modeled surface temperature warming rate (dT_s/dR_s) for different aerodynamic
 388 conductances (g_a , x-axis) and evaporative conditions (f_e , y-axis). The color bar shows the magnitude
 389 of the warming rate. The symbols correspond to the different sites, using their mean aerodynamic
 390 conductance and evaporative fraction. b) Modeled versus observed daily warming rates, dT_s/dR_s , for
 391 each site for the three vegetation types. The histograms show the distribution and spread. The
 392 coefficient of determination (r^2) is depicted for the linear regressions (dashed lines).
 393

394 We next tested the model by estimating the daily surface temperature warming rate for each site using
 395 Eq. (5) from daily values of observed f_e , g_a and dT_a/dR_s . Since dT_a/dR_s is similar for all sites, the
 396 diurnal variation of air temperature does not seem to depend on the diurnal variation of surface
 397 temperature, and vice versa. Figure 6b shows the comparison of the modeled surface temperature
 398 warming rates to those derived from observations. The model performs very well for all sites for the
 399 given information. The coefficient of determination (r^2) is also high for savanna and forests, pointing at
 400 the functionality of our model for complex and taller vegetation. However, short vegetation shows
 401 slightly weaker r^2 because our model underestimates the surface temperature warming rate at a few
 402 short vegetation sites. We speculate that these are the sites with non-vegetated surfaces where the
 403 ground heat flux contribution to diurnal surface temperature variations is significant (Saltzman and
 404 Pollack, 1977) which is currently neglected in our model.
 405

406 It is apparent from Figure 6 that the response of the surface temperature warming rate to evaporative
 407 fraction is predominantly governed by the aerodynamic conductance. The expression for dT_s'/dR_s in
 408 Eq. (6) quantifies this. Note that here we do not assume a constant aerodynamic conductance since g_a
 409 in the observation is enhanced on dry days. Our model reproduces the response of the warming rates to
 410 evaporative conditions ($r^2 = 0.6$) for all types, Figure 7a. Additionally, it even captures the ranges in
 411 dT_s'/dR_s for the specific vegetation types. Certain deviations exist because there are some biases in
 412 the number of wet and dry days in the observations that is reflected in the horizontal error bars. The



413 other possible root for a bias is the absence of a clear relation between g_a and f_e at some sites, these
 414 sites are indicated in lighter shades.
 415
 416



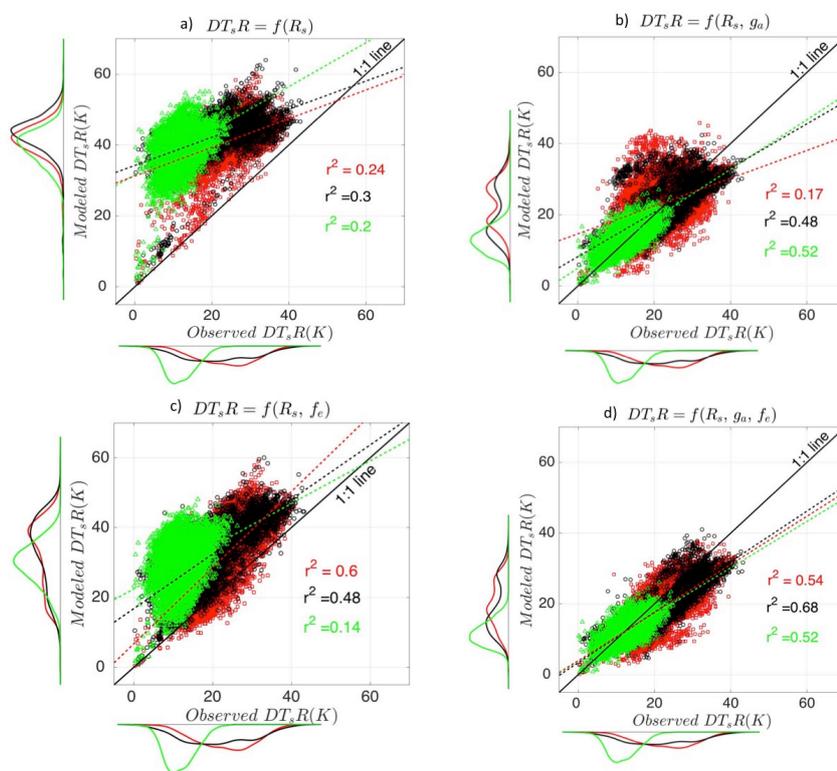
417
 418
 419 **Figure 7** a) Model evaluation of the response of surface temperature warming rates to evaporative
 420 conditions (dT_s'/dR_s) with those derived from observations for each site. b) Comparison of modeled
 421 dT_s'/dR_s for two cases: the first case assumes the aerodynamic conductance to be insensitive to
 422 evaporative fraction ($g'_a=0$, x-axis). The second case includes the sensitivity of aerodynamic
 423 conductance to evaporative fraction ($g'_a \neq 0$, in y-axis). The inset bar plot compares the mean
 424 contribution for the two cases with the error bars representing standard error of the mean. Sites with
 425 non-significant g'_a are marked by lighter shades.

426
 427 Figure 7b shows the contribution of the enhanced aerodynamic contribution on drier days in
 428 compensating the response of warming rates to evaporative conditions. For this, we compare the
 429 modeled dT_s'/dR_s with and without the inclusion of enhanced aerodynamic conductance term (the
 430 second term on the right-hand side of Eq. 6), such that dT_s'/dR_s when $g'_a = 0$ only captures the
 431 contribution of mean aerodynamic conductance and dT_s'/dR_s when $g'_a \neq 0$ additionally shows the
 432 contribution of the enhanced aerodynamic conductance on drier days. For the comparison of the two
 433 cases it is important to recognize that the more negative the values of dT_s'/dR_s , the stronger the
 434 imprint of evaporation is in the diurnal variation of temperature.

435
 436 In general, for most of the sites the enhanced aerodynamic conductance plays a small, but noticeable
 437 role in weakening the response of K of the warming rate to evaporative fraction. This is evident since the
 438 data points lie above the 1:1 line and tend to be less negative for the case when $g'_a \neq 0$. This effect is,
 439 however, more consistent for forests compared to short vegetation and savannas (see the inset bar plot
 440 which summarizes the mean dT_s'/dR_s for two cases). For short vegetation sites, dT_s'/dR_s decreases
 441 only by 6% when $g'_a \neq 0$ is considered. For savannas, the decrease is 32%, and it is highest with 53%



442 for forests. This suggests that along with the inherent high aerodynamic conductance of forests, its
 443 enhanced aerodynamic conductance is also responsible for the absence of evaporation imprints in the
 444 diurnal variation of temperature. The higher aerodynamic conductance of forests is responsible for
 445 reducing 74 % of the imprints of evaporation in diurnal surface temperature when compared to the
 446 short vegetation.
 447



448
 449 **Figure 8** Comparison of model estimates of the diurnal surface temperature range (DT_sR) for short
 450 vegetation, savanna and forests with observations for four scenarios: a) DT_sR is only a function of solar
 451 radiation, b) DT_sR is a function of solar radiation and aerodynamic conductance, c) DT_sR is a function
 452 of solar radiation and evaporative fraction, d) DT_sR is a function of solar radiation, aerodynamic
 453 conductance and evaporative fraction. Dashed lines show the linear regression between model and
 454 observation.

455
 456 We next link our model for surface temperature warming rates back to the diurnal variation in surface
 457 temperature. To understand how solar radiation, the aerodynamic conductivity of the different
 458 vegetation types and evaporative fraction individually influence the diurnal variation in temperature,
 459 we can obtain the diurnal surface temperature range (DT_sR) by multiplying the expression for warming
 460 rate given by Eq. 5 with the daily maximum in absorbed solar radiation. To quantify the sensitivity of
 461 DT_sR to its three main contributors, we considered four cases. In first case, we assume that the diurnal



462 variation in surface temperature is solely driven by solar radiation, such that there is no evaporation
463 ($f_e = 0$) and the surface has no vegetation, represented by a very low aerodynamic conductance of
464 $g_a = 0.02$. Figure 8a shows that in this scenario, DT_sR is overestimated for all vegetation types with
465 poor $r^2 \leq 0.3$. This greater warming indicates that vegetation and evaporation cools surface
466 temperatures. In second case we added the information on aerodynamic conductance of each vegetation
467 type along with solar radiation (Figure 8b). The DT_sR estimates for forests ($r^2=0.54$) and to some extent
468 for savanna ($r^2=0.48$) is considerably improved, but not for short vegetation ($r^2=0.17$). Nevertheless, in
469 this case DT_sR is much cooler and closer to the observed values, indicating the importance of
470 aerodynamic conductance in cooling the diurnal temperature. Aerodynamic conductance alone does not
471 explain the scatter in DT_sR in short vegetation. In third case we kept the information on daily
472 evaporative conditions but assumed a very low $g_a = 0.02$ (Figure 8c). Contrarily to Figure 8b, DT_sR in
473 short vegetation is captured much better ($r^2=0.6$), but the magnitudes are overestimated. Similar to
474 short vegetation, DT_sR is also overestimated for savanna and forest. For forests, r^2 is very low, because
475 the aerodynamic conductance is the key property affecting DT_sR . Finally; we added the information on
476 all the components of the model, solar radiation, aerodynamic conductance and evaporative fractions
477 (Figure 8d). Compared to the previous three cases the estimates are much closer to the observation with
478 a good r^2 for all vegetation types. This sensitivity analysis shows that vegetation type and evaporation
479 play significant roles in driving the diurnal variation in surface temperature. Evaporation is important
480 to capture the spread whereas aerodynamic conductance is important to capture the magnitudes of
481 diurnal variation of surface temperature, particularly for forest sites.

482 **6 Discussion**

483
484 We demonstrate a robust way of characterizing the diurnal variation of temperature using their morning
485 to noon warming rates, which are derived from the half hourly temperatures and solar radiation. The
486 warming rate is suitable for the comparison of locations with different solar energy input whereas other
487 metrics like diurnal temperature range depends on solar radiation (Makowski et al., 2009).
488 Consequently, temperature warming rates for specific vegetation types are comparable for sites at
489 different geographic locations. Our surface energy balance model can reproduce the warming rate and
490 shows the physical significance of evaporation and aerodynamic conductance. The model can capture
491 the diurnal variation of temperature quite well. These approximations can further be improved by a
492 more detailed formulation of net longwave radiation (which could, for instance, include optical
493 properties of the atmosphere) and the ground heat flux. Warming rates are also sensitive to clouds and
494 might not capture the information of evaporation and vegetation on cloudy days. Also, we did not
495 provide a way to calculate warming rates of air temperature. These could represent topics for future
496 research.

497
498 One of the main findings of our study is the different response of diurnal surface and air temperature to
499 evaporation. The air temperature warming rate does not contain any imprints of evaporation whereas,
500 for short vegetation, the surface temperature warming rate decreases strongly with evaporative fraction.
501 This finding is consistent with our previous work where we explained the role of boundary layer in



502 compensating imprints of evaporative conditions in the diurnal variation of air temperature. We found
503 that the diurnal variation of air temperature is similar for all vegetation types irrespective of their
504 aerodynamic conductance and evaporative conditions. We anticipate that our hypothesis of the
505 compensating effect of boundary layer might also be true for forests, but this would need further
506 research.

507

508 The notion that diurnal surface and air temperature variations respond differently to evaporation should
509 be considered when developing air temperature products from remotely sensed surface temperature
510 (Cresswell et al., 1999; Fu et al., 2011; Hengl et al., 2012; Jang et al., 2004; Kilibarda et al., 2014; Zhu
511 et al., 2013). Typically, these products are primarily based on the assumption that surface temperature
512 is proxy of air temperature. Generally, these approaches overestimate daytime air temperature (Oyler et
513 al., 2016; Zhang et al., 2011). This finding is consistent with our results, which show a greater
514 variation of surface temperature depending on vegetation type and evaporative fraction (cf. Eq. 5).

515

516 Our study shows that surface and air temperature warming rates are similar in forests, which indicates
517 the strong coupling between the two temperatures. This finding is in agreement with the previous study
518 by Li et al., 2015 and Mildrexler et al., 2011, where evaporative cooling and high aerodynamic
519 conductance of forests were identified as the responsible factors for the strong coupling of surface and
520 air temperature. However, we also show that this coupling remains persistent irrespective of the
521 evaporative conditions of the forest. Using our model and observations we show that the aerodynamic
522 conductance of forest increases on dry days resulting in reduced warming of surface temperature and
523 hence its stronger coupling to air temperature. These findings complement the recent studies on the
524 convective effect where its role in lowering the surface temperature for a semi-arid forest is discussed
525 (Banerjee et al., 2017, 2018; Brugger et al., 2019; Kröniger et al., 2018; Rotenberg and Yakir, 2010).
526 Our demonstration of enhanced aerodynamic conductance on dry days is similar to what these authors
527 describe as the convective effect.

528

529 Unlike forests, the surface temperature warming rate in short vegetation responds strongly to changes
530 in evaporative conditions. In observations, the warming rate decreases by $\sim 23 \times 10^{-3} \text{ K/W m}^{-2}$ from dry
531 to wet days. In general, this decrease is comparable for all the short vegetation sites and we anticipate
532 that some spread is due to their somewhat different aerodynamic properties. Another source of
533 ambiguity is the unequal distribution of days of different evaporative fractions which also influences
534 the computation of dT_s'/dR_s and g_a' . This constraint requires longer time series of observations to
535 obtain a greater sampling range of dry and wet days. Overall, our results nevertheless show that the
536 surface temperature warming rate is a promising indicator of evaporative fraction, especially for short
537 vegetation.

538

539 The other implication of our study is a better physical understanding of the processes that govern the
540 diurnal temperature range. Our model is capable of capturing the contribution of solar radiation,
541 vegetation and evaporation in shaping DTR. We show that the aerodynamic conductance of vegetation



542 is the key-cooling operator whereas evaporation explains the spread in DTR. These findings are
543 important when interpreting DTR for different ecosystems.

544

545 **7 Conclusions**

546

547 Temperature and evaporation are among the foremost-discussed variables in hydrology and climate
548 science. Our study contributes information on the relationship between diurnal temperature variations,
549 evaporative conditions, and vegetation. To measure the diurnal variation, we introduce the morning to
550 noontime warming rate of temperature. This rate has advantages over mean, maximum or minimum
551 temperatures for conducting a multisite analysis because it removes the effect of solar radiation. We
552 demonstrated that the warming rate and its response to evaporation is reproducible from the
553 simplification of surface energy balance. In doing so, we can address the two major questions that we
554 formulated in the introduction. First, our observational analysis shows no imprints of evaporation in
555 the air temperature warming rate across vegetation types. However, the surface temperature response to
556 evaporation is rather vegetation dependent, being stronger for short vegetation and absent in forests.
557 These findings provide insights for the second question about the role of aerodynamic conductance.
558 We showed that the aerodynamic conductance is very important in reducing the diurnal variation of
559 surface temperature. It is mostly the high aerodynamic conductance of forests, which compensates their
560 response to evaporative fraction. In addition, the aerodynamic conductance in itself is sensitive to
561 evaporative conditions. Using observational and model-reproduced findings we demonstrate that along
562 with the high aerodynamic conductance of forests their aerodynamic conductance roughly doubles on
563 dry days. The higher aerodynamic conductance results in more efficient transport of heat from the
564 surface to the atmosphere and compensates for the diurnal rise in surface temperature, which is
565 reflected in their lower surface temperature warming rate.

566

567 To conclude, our results imply that diurnal temperature variations can be understood and predicted by
568 relatively few factors, solar radiation, aerodynamic conductance and evaporative fraction. Surprisingly,
569 diurnal air temperature carries little information of vegetation type and evaporative conditions of the
570 land surface, while surface temperatures carry a stronger imprint of evaporation, but only for short
571 vegetation.

572

573 **Data availability:** For the spatial plot of evaporative fraction we used the FLUXCOM monthly data of
574 sensible and latent heat fluxes. The FLUXCOM data is available at <http://www.fluxcom.org/>. For
575 observational analysis we used FLUXNET data for 52 sites. More description of each site is provided
576 in the Appendix. For FLUXNET data please see the link <https://fluxnet.fluxdata.org/>.

577

578 **Author contribution:** All authors conceived the study. AP analysed data, AK derived the energy-
579 balance model that is further developed by AP. MR provided classification of cloud-free conditions.
580 All authors interpreted results. AP wrote the manuscript with input of MR and AK.

581



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583
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587
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798 **Appendix**

799 **Table A1** Abbreviation used

Symbol	Full form	Unit
DTR	Diurnal temperature range	K
DT_sR	Diurnal surface temperature range	K
R_s	Surface solar radiation	$W m^{-2}$
$R_{s,max}$	Maximum of surface solar radiation	$W m^{-2}$
T_a	2 m air temperature	K
T_s	Surface temperature, obtained from longwave radiation	K
$R_{L,net}$	Net longwave radiation	$W m^{-2}$
LE	Latent heat flux	$W m^{-2}$
H	Sensible heat flux	$W m^{-2}$
G	Ground heat flux	$W m^{-2}$
R_o	Net radiation at reference temperature	$W m^{-2}$
T_{ref}	Reference temperature	K
k_r	Linearized constant	$W m^{-2} K^{-1}$
σ	Stefan-Boltzmann constant	$W m^{-2} K^{-4}$
c_p	Specific heat capacity of the lower atmosphere	J/kg K
ρ	Density of the lower atmosphere	$kg m^{-3}$
g_a	Aerodynamic conductance	$m s^{-1}$
u	Wind speed	$m s^{-1}$
u_*	Frictional velocity	$m s^{-1}$
f_e	Evaporative fraction	-
$\frac{dT_s}{dR_s}$	Surface temperature warming rate	$K/W m^{-2}$
$\frac{dT_a}{dR_s}$	Air temperature warming rate	$K/W m^{-2}$
$\frac{dT_s'}{dR_s}$	Derivative of surface temperature warming rate to evaporative fraction	$K/W m^{-2}$
$\frac{dT_a'}{dR_s}$	Derivative of air temperature warming rate to evaporative fraction	$K/W m^{-2}$
g_a'	Derivative of aerodynamic conductance to evaporative fraction	$m s^{-1}$

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802 **Table A2** Description of sites used for this study

Site no.	IGBP land use	Site ID	Site name	Location		Number of days used	DOI
				Latitude	Longitude		
1	Croplands (CRO)	AU-Rig	Riggs Creek	-36.65	145.57	237	https://doi.org/10.18140/FLX/1440202
2		CH-Oe1	Oensingen1 grass	47.28	7.73	182	https://doi.org/10.18140/FLX/1440135
3		CZ-wet	CZECHWET	49.02	14.77	184	https://doi.org/10.18140/FLX/1440145
4		DE-Geb	Gebesee	51.10	10.91	285	https://doi.org/10.18140/FLX/1440146
5		IT-BCi	Borgo Cioffi	40.52	14.95	274	https://doi.org/10.18140/FLX/1440166



6		IT-CA2	Castel d'Asso2	42.37	12.02	143	https://doi.org/10.18140/FLX/1440231
7		JP-SMF	Seto Mixed Forest Site	35.25	137.06	164	https://doi.org/10.18140/FLX/1440239
8		US-ARM	ARM Southern Great Plains site	36.60	-97.48	648	https://doi.org/10.18140/FLX/1440066
9	Croplands /Natural Vegetation (CRO/NV)	CH-Cha	Chamau grassland	47.21	8.41	188	https://doi.org/10.18140/FLX/1440131
10		CH-Fru	Fruebuel grassland	47.11	8.53	260	https://doi.org/10.18140/FLX/1440133
11		FR-LBr	Le Bray (after 6/28/1998)	44.71	-0.76	265	https://doi.org/10.18140/FLX/1440163
12		US-Goo	'Goodwin Creek'	34.25	-89.87	206	https://doi.org/10.18140/FLX/1440070
13	Grasslands (GRA)	AU-Stp	Sturt Plains	-17.15	133.35	532	https://doi.org/10.18140/FLX/1440204
14		IT-MBo	Monte Bondone	46.01	11.04	480	https://doi.org/10.18140/FLX/1440170
15		US-AR1	ARM USDA UNL OSU Woodward Switchgrass 1	36.42	-99.42	242	https://doi.org/10.18140/FLX/1440103
16		US-AR2	ARM USDA UNL OSU Woodward Switchgrass 2	36.63	-99.59	225	https://doi.org/10.18140/FLX/1440104
17		US-SRG	Santa Rita Grassland	31.78	-110.82	696	https://doi.org/10.18140/FLX/1440114
18		US-Wkg	Walnut Gulch Kendall Grasslands	31.73	-109.94	1074	https://doi.org/10.18140/FLX/1440097
19	Shrublands (SH)	AU-ASM	Alice Springs	-22.28	133.24	477	https://doi.org/10.18140/FLX/1440194
20		US-SRC	Santa Rita Creosote	31.90	-110.83	621	https://doi.org/10.18140/FLX/1440098
21		US-SRM	Santa Rita Mesquite	31.82	-110.86	1121	https://doi.org/10.18140/FLX/1440090
22		US-Whs	Walnut Gulch Lucky Hills Shrubland	31.74	-110.05	558	https://doi.org/10.18140/FLX/1440097
23		AU-Cpr	Calperum	-34.00	140.58	284	https://doi.org/10.18140/FLX/1440195
24	Savannas (SA)	AU-DaP	Daly River Pasture	-14.06	131.31	439	https://doi.org/10.18140/FLX/1440123
25		AU-DaS	Daly River Savanna	-14.15	131.38	504	https://doi.org/10.18140/FLX/1440122
26		AU-Dry	Dry River	-15.25	132.37	466	https://doi.org/10.18140/FLX/1440197
27		AU-How	Howard Springs	-12.49	131.15	355	https://doi.org/10.18140/FLX/1440125
28	Woody Savannas (WSA)	AU-Gin	Gingin	-31.37	115.65	212	https://doi.org/10.18140/FLX/1440199
29		AU-Whr	Whroo	-36.67	145.02	206	https://doi.org/10.18140/FLX/1440206
30		IT-Noe	Sardinia/Arca di Noe	40.60	8.15	555	https://doi.org/10.18140/FLX/1440171
31		US-Me6	Metolius New Young Pine	44.32	-121.60	270	https://doi.org/10.18140/FLX/1440099
32		US-Var	Vaira Ranch	38.40	-120.95	1091	https://doi.org/10.18140/FLX/1440094
33	Deciduous Broadleaf Forest (DBF)	DK-Sor	Soroe- LilleBogeskov	55.48	11.64	169	https://doi.org/10.18140/FLX/1440155
34		IT-Col	Collelongo- Selva Piana	41.84	13.58	343	https://doi.org/10.18140/FLX/1440167
35		US-Oho	Oak Openings	41.55	-83.84	408	https://doi.org/10.18140/FLX/1440088
36		US-WCr	Willow Creek	45.80	-90.07	237	https://doi.org/10.18140/FLX/1440095
37	Evergreen Broadleaf Forest	AU-Wom	Wombat	-37.42	144.09	180	https://doi.org/10.18140/FLX/1440207
38		CA-Obs	SK-Southern Old Black Spruce	53.98	-105.11	620	https://doi.org/10.18140/FLX/1440044
39		CA-Qfo	Quebec Eastern Old Black Spruce (EOBS)	49.69	-74.34	194	https://doi.org/10.18140/FLX/1440045
40		DE-Tha	Tharandt- Anchor Station	50.96	13.56	268	https://doi.org/10.18140/FLX/1440152



41	Evergreen Needleleaf Forest (ENF)	IT-Lav	Lavarone (after 3/2002)	45.95	11.28	557	https://doi.org/10.18140/FLX/1440169
42		IT-Ren	Renon/Ritten (Bolzano)	46.58	11.43	362	https://doi.org/10.18140/FLX/1440173
43		NL-Loo	Loobos	52.16	5.74	401	https://doi.org/10.18140/FLX/1440178
44		US-GLE	GLEES	41.36	-106.23	514	https://doi.org/10.18140/FLX/1440069
45		US-Me2	Metolius Intermediate Pine	44.45	-121.55	450	https://doi.org/10.18140/FLX/1440079
46		US-NR1	Niwot Ridge (LTER NWT1)	40.03	-105.54	600	https://doi.org/10.18140/FLX/1440087
47	Mixed Forest (MF)	CA-Gro	ON-Groundhog River Mixedwood	48.21	-82.15	339	https://doi.org/10.18140/FLX/1440034
48		CA-Oas	SK-Old Aspen	53.62	-106.19	688	https://doi.org/10.18140/FLX/1440043
49		CA-TP4	ON-Turkey Point 1939 White Pine	42.70	-80.35	482	https://doi.org/10.18140/FLX/1440053
50		FR-Pue	Puechabon	43.74	3.59	535	https://doi.org/10.18140/FLX/1440164
51		RU-Fyo	Fedorovskoje-drained spruce stand	56.46	32.92	257	https://doi.org/10.18140/FLX/1440183

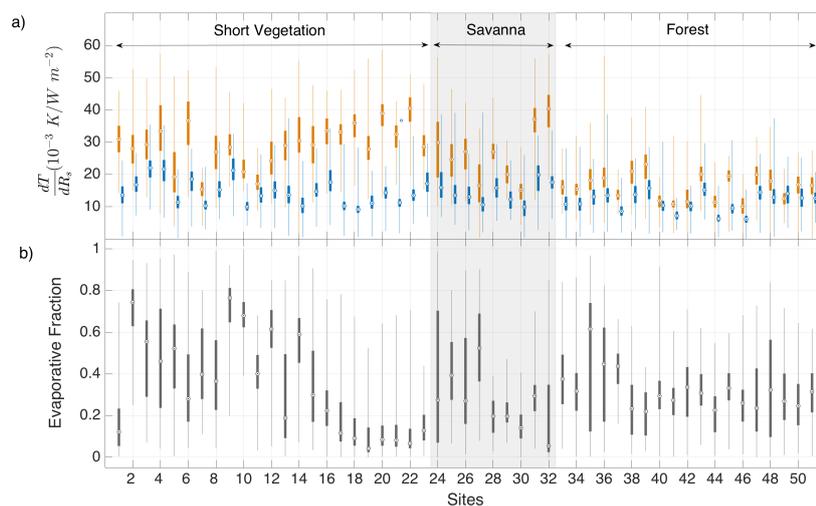
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806 **Observational analysis for each site**

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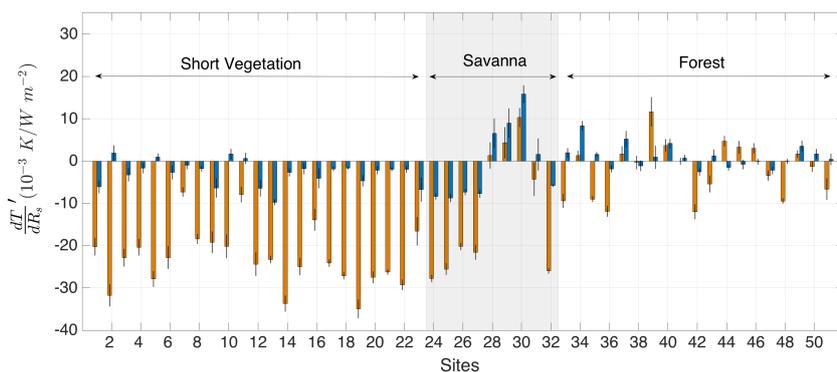


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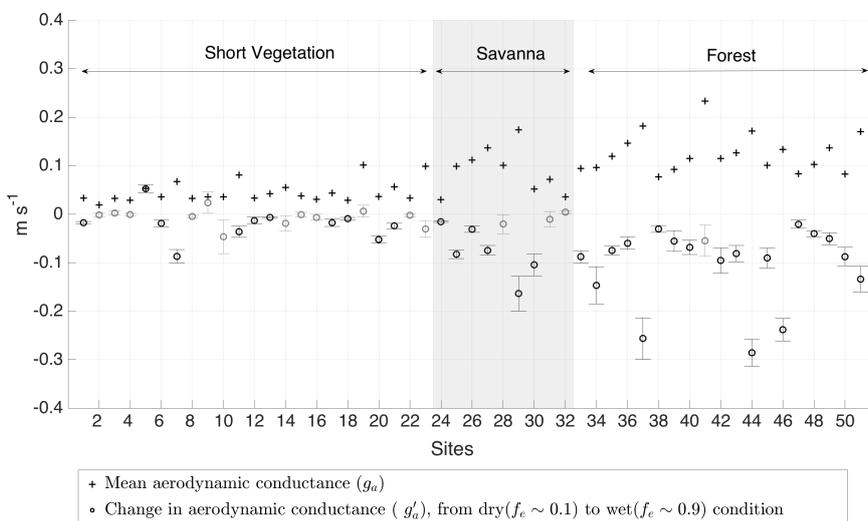
809 **Figure A1** (a) Box plot of surface (T_s , orange) and air (T_a , blue) temperature warming rates (dT/dR_s),
 810 (b) Box plot of evaporative fractions. The vegetation types are separated by grey and white shades. The
 811 circle in the box plot indicates the median and the top and bottom edges indicate the 75th and 25th
 812 percentiles, respectively. The whisker covers the range in the observation.

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 816 **Figure A2** Warming rate response to evaporation (dT' / dR_e) for surface (T_s , orange) and air (T_a ,
 817 blue) temperature. The vegetation types are separated by grey and white shades. The black bar
 818 represents the standard error in the linear regression of observed warming rate and evaporative fraction.
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 821 **Figure A3** Mean aerodynamic conductance (g_a) and response of aerodynamic conductance to
 822 evaporation (g'_a) for each site. Sites with non-significant g'_a in observation is marked by light shades.
 823 The vegetation types are separated by grey and white shades. The error bar represents the standard
 824 error in the observed linear regression of aerodynamic conductance and evaporative fraction.
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