A meta-analysis of open-path eddy covariance observations of apparent CO₂ flux in cold conditions in the FLUXNET network

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1 Abstract

Open-path eddy covariance systems are widely used for measuring the CO₂ flux between the 2 land and the atmosphere. A common problem is that they often yield negative fluxes or 3 physiologically unreasonable CO₂ uptake fluxes in the non-growing season under cold 4 conditions. In this study, we performed a meta-analysis of the eddy flux data from 64 5 6 FLUXNET sites and analyzed the relationship between the observed CO₂ flux and the 7 sensible heat flux. In theory, these two fluxes should be independent of each other in the cold 8 conditions (air temperature lower than 0 °C) when photosynthesis is suppressed. However, 9 our results show that a significant and negative linear relationship existed between these two fluxes at 37 of the sites. The mean linear slope value is $-0.008 \pm 0.001 \text{ }\mu\text{mol }\text{m}^{-2} \text{ s}^{-1}$ per W 10 m^{-2} among the 64 sites analyzed. The slope value was not significantly different among the 11 three gas analyzer models (LI-7500, LI-7500A, IRGASON/EC150) used at these sites, 12 indicating that self-heating may not be the only reason for the apparent wintertime net CO₂ 13 uptake. These results suggest a systematic bias towards larger carbon uptakes in the 14 FLUXNET sites that deploy open-path EC systems. 15 16

17

18 Keywords: Eddy flux; Self-heating; Spectroscopic effects; Air temperature; Absolute
19 humidity; CO₂ density

21 **1. Introduction**

The eddy-covariance (EC) technique is widely used for measuring exchanges of carbon 22 23 dioxide between terrestrial ecosystems and the atmosphere. In the global EC network (FLUXNET), about half of the sites located north of 40° N deploy open-path CO₂/H₂O 24 analyzers for flux measurements. Compared to closed-path analyzers, open-path analyzers 25 need less power and less maintenance, and thus are better suited for remote locations. 26 However, a physiologically unreasonable CO₂ uptake phenomenon has been reported 27 frequently for cold seasons when no photosynthetic activities exist (Hirata et al. 2005; Welp 28 et al. 2007; Lafleur and Humphreys 2008; Järvi et al. 2009; Wang et al. 2016) or for 29 environments where CO₂ uptake is not expected (Liu et al. 2012; Ma et al. 2014; Ono et al. 30 2008; Wohlfahrt et al. 2008). Specifically, for desert ecosystems, Schlesinger (2017) argued 31 that abiotic CO₂ uptake mechanisms like atmospheric pressure pumping, carbonate 32 dissolution, and percolation of soil water through the vadose zone can not adequately explain 33 the observation of EC systems. Following the micrometeorological sign convention, the 34 uptake phenomenon is marked by a negative CO₂ flux or a flux directed towards the surface. 35 36 According to Amiro et al. (2006) and other research groups (e.g., Burba et al. 2008; Helbig et al. 2016), the problem is not caused by software or hardware malfunction. Accurate CO₂ flux 37 measurement is important for evaluating the global CO₂ cycle, terrestrial ecosystem 38 39 responses to climate change, and modelling studies. Long-term integrated carbon flux will suffer a systematic bias if corrections are not made to the apparent flux in the cold season 40 (Burba et al. 2008) or even in the whole year (Helbig et al. 2016). 41

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43 Several explanations for the apparent negative flux are found in the published literature.

Self-heating of an open-path instrument is considered as a major reason for the negative flux 44 (Burba et al. 2005; Burba et al. 2008; Grelle and Burba 2007). Because of heat release by the 45 46 instrument's electronics and solar radiation heating of the instrument supporting frame, the air in the sensing volume of the analyzer may be warmer than the undisturbed air. Density 47 48 corrections using the sensible heat flux measured outside the sensing volume cannot fully remove the effects of temperature fluctuations on CO₂ density fluctuations in the sensing 49 volume. It is estimated that in mild temperatures (about 12 °C), the sensible heat flux in 50 ambient air is 14% lower than the flux involving the actual temperature fluctuations in the 51 52 sensing volume, and this rate of underestimation should increase as temperature decreases (Burba et al. 2008). The apparent cold-season CO₂ uptake was first observed with an older 53 analyzer model manufactured by Li-Cor Inc (model LI-7500, Li-Cor Inc, Lincoln, Nebraska). 54 To reduce the self-heating effect, the manufacturer has introduced an improved model 55 (LI-7500A) with reduced power consumption during cold periods (Burba 2013). However, 56 using a LI-7500A analyzer in their open-path EC system, Wang et al. (2016) found that there 57 58 is still an apparent CO_2 uptake in a desert ecosystem in low temperatures (mean air temperature -6.7 °C), implying that other sources of wintertime bias exists for LI-7500A, if 59 60 this gas analyzer indeed avoids the self-heating effect according to its producer.

61

Another explanation for the uptake phenomenon is related to the spectroscopic effects.
Spectroscopic effects result from changes in the shape and the strength of CO₂ absorption
lines; these changes are caused by changes in temperature, water vapor and pressure. If these
effects are not properly dealt with, the fluctuations can be interpreted as changes in the

2011) These effects fall into two types: insufficient compensation due to high-frequency
2011). These effects full into two types: insufferent compensation due to high nequency
temperature fluctuations (Helbig et al. 2016) and spectroscopic cross-sensitivity (Kondo et al.
2014). In a class of analyzers (models IRGASON and EC150, Campbell Sci Inc., Logan,
Utah) that use EC operating software released before September, 2016, the CO ₂ density is
determined by a scaling law (Jamieson et al. 1963) with temperature measured by a
slow-responding thermometer mounted outside the analyzer's sensing volume. In an
experiment that compared an open-path EC system using an IRGASON analyzer and a
closed-path EC system, Bogoev et al. (2014) found that the CO ₂ flux measured with the
open-path EC is biased low and the low bias scales linearly with sensible heat flux. These
authors and Helbig et al. (2015 & 2016) showed that this spectroscopic effect can be
corrected by using fast-response air temperature measurements to perform the absorption line
calculation. After this correction, the linear relationship between the corrected CO ₂ flux and
the heat fluxes almost disappears. Unlike the self-heating effect, which is believed to be
limited only to cold seasons, this spectroscopic error exists in all seasons. At low
temperatures when the true CO ₂ flux is very small, the flux measured with an IRGASON may
appear negative (Wang et al. 2016). It is not known to what extent spectroscopic effects affect
the CO ₂ flux measured with other open-path analyzers.

Spectroscopic cross-sensitivity can arise from a pressure broadening effect and from
absorption line interference between CO₂ and H₂O. Generally, the absorption interference is
much smaller than the pressure broadening effect (Kondo et al. 2014). These effects are

88	corrected by using measurements of air pressure and H2O mole fraction and
89	manufacturer-determined correction coefficients. Kondo et al. (2014) found that for a
90	LI-7200 gas analyzer with the same design as the LI-7500, the manufactured correction
91	coefficient for spectroscopic cross-sensitivities results in an overestamation of CO2 mixing
92	ratio by about 0.9% at a H_2O mole fraction of 30.6 mmol mol ⁻¹ . Different from the effect
93	associated with high-frequency temperature fluctuations, this spectroscopic cross-sensitivity
94	biases the CO_2 flux towards more positive values, and the biases increase with the H_2O mole
95	fraction. Because this influence is much smaller than the other bias sources , in the following
96	we will not examine this problem.
97	
98	Finally, a negative CO ₂ flux in the cold season can result from errors propagated through the
99	density correction procedure. The WPL density correction requires that the CO ₂ density (ρ_c)
100	be measured precisely. But in field conditions, biases in ρ_c can be caused by thermal
101	expansion and contraction of the analyzer's frame on which the transducers are mounted, by
102	dirt contamination on the transducers, and by aging of the optical components (Fratini et al.
103	2014). An underestimation of ρ_c will cause CO ₂ flux to be too negative (Serrano-Ortiz et al.
104	2008). The bias in the CO_2 flux scales linearly with the sensible heat flux if the CO_2 density is
105	underestimated by a constant amount.
106	
107	In this study, we perform a meta-analysis of the eddy flux data from 64 sites located in North
108	America, Europe, Asian and Australia, and analyze the relationship between the apparent CO ₂

109 flux and the sensible heat flux in cold conditions. We aim to: 1) compare the negative bias

110	problem among different open-path gas analyzers, and 2) investigate the bias errors in
111	relation to climatic conditions (temperature, humidity) and to biases in the CO ₂ concentration.
112	
113	2. Data and Methods
114	2.1 Theoretical Consideration
115	Self-heating
116	The CO ₂ flux bias error arising from self-heating can be understood by examining the WPL
117	algorithm (Webb et al. 1980),
118	(1)
119	where $F_{c,a}$ is the CO ₂ flux after density correction (kg m ⁻² s ⁻¹), w is the vertical wind velocity
120	(m s ⁻¹), ρ_c and ρ_a are density of CO ₂ (kg m ⁻³) and dry air (kg m ⁻³), <i>T</i> is the air temperature
121	(K), M is the molecular mass (g mol ⁻¹), C_p is the specific heat of air (J kg ⁻¹ K ⁻¹), H is the
122	sensible heat flux (W m ⁻²) measured in ambient air outside the analyzer's sensing volume, E_0
123	is the H ₂ O flux (kg m ⁻² s ⁻¹), and subscripts a, v, and c represent dry air, water vapor, and CO ₂ ,
124	respectively. Let H_{real} be the real sensible heat inside the open path. To obtain the correct CO ₂
125	flux, F_c , Equation (1) should be modified to, (2)
126	A comparison of Equations (1) and (2) yields
127	(3)
128	
129	According to Burba et al. (2008), the sensible heat flux measured in the ambient air (H) is
130	highly correlated with the sensible heat flux inside the open path measured with a fine-wire
131	platinum resistor (H_{real}). Their linear relationship is,

132	(4)
133	where =0.86 and =2.67 W m ^{-2} .
134	
135	The Equation (3) can be rewritten as
136	
137	(5)
138	where
139	(6)
140	(7)
141	
142	For =0.86, =2.67, the slope parameter <i>b</i> in Equation (6) is approximately $-0.008 \mu\text{mol }\text{m}^{-2}\text{ s}^{-1}$
143	per W m ⁻² after unit conversion (from kg m ⁻² s ⁻¹ per W m ⁻² to μ mol m ⁻² s ⁻¹ per W m ⁻² ,
144	divided by CO ₂ molar mass), and the parameter <i>a</i> is approximately $-0.16 \mu mol m^{-2} s^{-1}$.
145	
146	Equation (5) predicts that negative $F_{c,a}$ is more likely to occur at times of high sensible heat
147	flux. An implicit assumption is that sensor self-heating is dominated by solar radiation rather
148	than heating caused by the sensor electronics. We also hypothesize that b is more negative for
149	colder sites since the self-heating effect is expected to be more severe.
150	
151	Insufficient compensation for spectroscopic effects
152	Spectroscopic effects affect every instrument that measures absorption or IRGASON/EC150
153	gas analyzers, which are subject to insufficient compensation for spectroscopic effects due to

high-frequency temperature fluctuations, the observed flux $F_{c,a}$ and the true flux F_c also follows the linear relationship with the ambient sensible heat flux (Bogoev et al. 2014; Helbig et al. 2015; Helbig et al. 2016), as given by Equation (5). A theoretical slope value can be described as (Wang et al. 2016),

158 (8)

Under typical atmospheric conditions, the theoretical value for *b* is about $-0.025 \ \mu mol \ m^{-2}$

 $160 ext{ s}^{-1} ext{ per W m}^{-2}$. A comparison of an IRGASON open-path EC versus a closed-path EC reveals

161 that the actual slope is about half of the theoretical value, at $-0.014 \ \mu mol \ m^{-2} \ s^{-1}$ per W m⁻²

162 (Bogoev et al. 2014). Subsequent experiments by Helbig et al. (2015 & 2016) in boreal forest,

163 grassland and cropland sites showed that the slope value ranges from -0.014 to -0.020 µmol

 $164 \text{ m}^{-2} \text{ s}^{-1} \text{ per W m}^{-2}$ and that the intercept value (a) ranges from -0.300 to $0.080 \text{ }\mu\text{mol m}^{-2} \text{ s}^{-1}$.

165

166 **Biases in CO₂ density**

Bias errors in the CO₂ density ρ_c can also be examined in the framework of the WPL theory. Let be the true mean CO₂ density, and be the measurement bias. The measured flux $F_{c,a}$ after the WPL correction is,

170

171

(9)

172 with the slope parameter given by

173 (10)

174

175 In Equation (9), *LE* is the latent heat flux and λ is the latent heat of vaporization. The

magnitude of *LE* is comparable with *H*, and the term before *LE* is equal to 0.004 (μ mol m⁻² s⁻¹ per W m⁻²), which is one order of magnitude smaller than 0.05 (μ mol m⁻² s⁻¹ per W m⁻²) in Equation (10). Thus, we ignored this term to simplify the analysis.

179

In the situation where is underestimated by 10% (; Serrano-Ortiz et al. 2008), the slope *b* is approximately $-0.005 \ \mu mol \ m^{-2} \ s^{-1} \ per \ W \ m^{-2}$.

182

183 **2.2 Data sources and data processing**

The eddy flux data analyzed in the present study were obtained from the AmeriFlux, the FLUXNET, the ChinaFlux, and the Chinese Heihe databases. The majority of the sites are located north of 40° N (Figure 1; Tables S1 & S2). These sites have continuous eddy flux records for at least 5 days when the air temperature is below the freezing point in the winter. There are a total of 64 sites, including 57 sites using LI-7500, 6 sites using LI-7500A and one site using IRGASON. The eddy flux data at 28 sites were obtained from the FLUXNET. These data have been gap filled and through a Ustar-threshold filtering following the

191 FLUXNET data processing pipeline. The other eddy flux data, obtained from the AmeriFlux

192 (20 sites), the ChinaFlux (7 sites) and the Heihe databases (9 sites), were not gap-filled and

193 without the Ustar-threshold filtering.

194

195 When the air temperature is below 0 $^{\circ}$ C in the winter (January, February and December in the

196 Northern Hemisphere; June, July and August in Australia), the true carbon flux F_c should be

197 slightly positive due to ecosystem respiration, but should be very small and independent of

the sensible heat flux H due to suppression of photosynthesis. We assume that any correlation 198 between the measured flux $F_{c,a}$ and H is evidence of measurement errors. For each site, we 199 200 only used winter data when the half-hourly or hourly air temperature was below 0 $^{\circ}$ C, and applied the ordinary linear regression to the observed CO₂ flux and sensible heat flux, using 201 the sensible heat flux as the independent variable. The regression yields the slope parameter b202 and intercept parameter a. We then tried to discern patterns of the slope parameter among the 203 64 sites. Half hourly or hourly data influenced by precipitation were excluded from the 204 regression analysis. We also restricted the sensible heat flux to the range of -100 to 400 W 205 m^{-2} and the latent heat flux to the range of -200 to 700 W m^{-2} to avoid extreme values due to 206 unknown measurement errors. The CO₂ flux was limited in -10 to 10 µmol m⁻² s⁻¹. Prior to 207 the regression, the half-hourly or hourly data were averaged by every 20 W m⁻² bin of 208 209 sensible heat flux to reduce the effect of random measurement errors on the parameter estimation. 210

211

We used the CO₂ mole fraction data from CarbonTracker (version CT2016, global $3^{\circ} \times 2^{\circ}$ grid, level 1) to calculate the bias in the CO₂ density ρ_c measured by the EC open-path analyzers. CarbonTracker is a data inversion system aiming to calculate global CO₂ fluxes from high precision atmospheric CO₂ measurements. Peters et al. (2007) compared the optimized three dimensional CO₂ mole fraction fields produced by CarbonTracker with 13,000 independent CO₂ flask samples taken in the free troposphere, and found that the mean and standard deviation of the residuals are 0.07 ppm and 1.91 ppm, respectively.

219

220	To determine the bias error in the CO ₂ density, we first selected a CarbonTracker grid cell
221	that is closest to the measurement site. We then chose the CarbonTracker surface CO_2 mole
222	fraction for those days when the actual flux data were used for our analysis. The mean
223	CarbonTracker CO_2 mole fraction of those measurement days is regarded as the true CO_2
224	mole fraction. Finally, the bias ratio in CO ₂ mass density () is represented by the bias ratio of
225	the CO_2 mole fraction (CO_2 concentration is given by mole fraction in FLUXNET) since the
226	pressure and temperature effects are canceled out.

227

228 **3. Results and Discussion**

229 **3.1 Relationship between observed** $F_{c,a}$ and H

A significant (p < 0.05) and negative correlation exists between the observed $F_{c,a}$ and H at 37

of the 64 sites, and only 4 sites have significant and positive correlation. The mean coefficient

of determination (R^2) is 0.63 for all the sites. The regression statistics are shown in

233 Supplementary Table S1 and the scatter plots for individual sites are given in Figure S1.

Among the sites analyzed, 45 sites have R^2 value larger than 0.5.

235



radiation. The mean R^2 is 0.48 for the 39 sites that have the radiation data. Among these sites,

238 17 sites have \mathbb{R}^2 values larger than 0.5. The results show that *H* is a better independent

239 variable than solar radiation to do this analysis.

240

241 We analyzed the relationship of CO₂ flux and sensible heat measured by a closed-path

242	analyzer at site AT-Neu in the cold season, and we found that this relationship is insignificant
243	(p = 0.13, R^2 = 0.26). The result supports our hypothesis that the true carbon flux F_c should
244	be independent of the sensible heat flux H . On the other hand, the relationship for the
245	open-path measurement at this site is significant ($p = 0.017$, $R^2 = 0.71$).
246	
247	Figure 2 shows two examples. At a shrub land ecosystem in the Kubuqi desert in China (site
248	ID: CN-Kub_s; analyzer type: LI-7500; Figure 2a), the linear relationship can be described as
249	(11)
250	where $F_{c,a}$ is the observed carbon flux (µmol m ⁻² s ⁻¹), <i>H</i> is sensible heat (W m ⁻²). The error
251	bounds on the regression parameters are ± 1 standard error. In this case, the negative
252	correlation between $F_{c,a}$ and H is very strong, with an R ² value of 0.99 and the p-value
253	smaller than 0.0001.
254	
255	Not all the sites have such a nearly perfect linear relationship between $F_{c,a}$ and H as shown in
256	Figure 2a. For example, the relationship is much weaker for the Brook cropland site in the U.
257	S. (site ID: US-Br3; analyzer type: LI-7500; Figure 2b). The linear equation for this site is
258	given as
259	(12)
260	with $R^2 = 0.38$ and a p-value of 0.79. US-Br3 is one of the 15 sites that show a large scatter
261	around $H = 0$. Nine of them belong to the cold climate zone and the other 6 sites belong to the
262	mild temperature zone. None is found in the arid zone and the polar region. Perhaps the
263	scatter was caused by moisture interferences (dew formation or rain). Significant negative

264 correlation exists between the site-mean CO₂ flux and the site mean sensible heat flux
265 (Supplementary Figure S2).

266

The regression slope b ranges from -0.051 to 0.013 µmol m⁻² s⁻¹ per W m⁻² with an average 267 value of $-0.008 \ \mu\text{mol}\ \text{m}^{-2}\ \text{s}^{-1}$ per W m⁻² among the 64 sites analyzed. For the sites listed in 268 Table S1, 84% have a negative slope and 16% have a positive slope. The frequency histogram 269 shows that 38% of them range from $-0.015 \ \mu\text{mol}\ \text{m}^{-2}\ \text{s}^{-1}$ per W m⁻² to $-0.007 \ \mu\text{mol}\ \text{m}^{-2}\ \text{s}^{-1}$ 270 per W m⁻², and 67% of them range from $-0.015 \mu mol m^{-2} s^{-1}$ per W m⁻² to 0.00 $\mu mol m^{-2} s^{-1}$ 271 per W m⁻² (Figure 3). The mean slope of the 28 FLUXNET sites, whose data have been gap-272 filled and through a Ustar-threshold filtering, is $-0.009 \pm 0.0004 \text{ }\mu\text{mol} \text{ }m^{-2} \text{ }s^{-1} \text{ per W }m^{-2}$, and 273 is not significantly different from the mean slope of other sites $(-0.007 \pm 0.0003 \ \mu mol \ m^{-2} \ s^{-1})$ 274 per W m⁻²; p = 0.45). Here and hereafter, the variation range on b is expressed as \pm one 275 standard error. The difference between vegetation types considered is not statistically 276 significant (p = 0.70, Figure S3). 277

278

We compared the slope of the sites that deployed the CSAT3 anemometer to measure the turbulent velocity and those that deployed the Gill anemometer. The mean slope *b* is $-0.008 \pm$ 0.001 µmol m⁻² s⁻¹ per W m⁻² for the CSAT3 sites (number of sites n = 46) and $-0.008 \pm$ 0.002 µmol m⁻² s⁻¹ per W m⁻² for the Gill sites (n =17). The difference between these two groups is statistically insignificant (p = 0.92). For the site (Site ID: JP-SMF) deployed the DAT-540 anemometer (Kaijo, Japan), the slope *b* is $-0.010 \mu mol m^{-2} s^{-1}$ per W m⁻². It should be mentioned that biases in *H* can be also caused by sonic anemometer measurement errors, such as the angle of attack errors with Gill anemometers (Nakai et al. 2006; Nakai and
Shimoyama 2012) and transducer shadowing effects on CSAT3 (Horst et al. 2015) and
IRGASON (Horst et al. 2016).

289

290 The comparison among the three models of gas analyzer is given in Figure 4. The mean slope value for the sites that deployed LI-7500 is $-0.007 \pm 0.001 \mu mol m^{-2} s^{-1}$ per W m⁻² (n = 54), 291 excluding three sites (site ID: US-ICh, US-ICs and FR-Fon, with slope values of 0.011, 292 -0.008 and -0.009, respectively), which are suspected to have archived the flux data after 293 self-heating correction (Euskirchen et al. 2012; Delpierre et al., 2015). As for the LI-7500A 294 sites, the average slope value is $-0.012 \pm 0.002 \text{ }\mu\text{mol }\text{m}^{-2} \text{ s}^{-1}$ per W m⁻² (n = 6). The problem 295 did not go away with LI-7500A despite hardware improvement over the older version 296 297 LI-7500. Besides the result of the Tarim site (site ID: CN-Tarim2; Wang et al., 2016), we also acquired the slope values published for three other sites using IRGASON analyzers and one 298 site using an EC150 analyzer (Helbig et al. 2016). In that paper, the slopes were calculated 299 for $F_{\rm c}$ and the kinematic temperature flux, and we used air density and specific heat capacities 300 under typical atmospheric conditions to convert these values to slope values for the F_{c} and H. 301 The mean slope is $-0.013 \pm 0.001 \ \mu\text{mol} \ m^{-2} \ s^{-1}$ per W m⁻² for these sites (n = 5). The mean 302 (and standard error) R² of the LI-7500, the LI-7500A and the IRGASON/EC150 site group is 303 0.61 (\pm 0.04), 0.76 (\pm 0.06) and 0.77 (\pm 0.10) respectively (Figure 4). The differences in the 304 regression slope or R^2 among the three analyzer types are not statistically significant (p > 305 0.30). 306

The mean slope value and its standard error are compared among different climate zones 308 according to Köppen climate classification (Figure 5). The sites in the temperate zone have 309 the most negative mean slope value ($-0.012 \pm 0.003 \mu mol m^{-2} s^{-1}$ per W m⁻², n = 17), those 310 in the polar zone have the least negative mean value $(-0.001 \pm 0.003 \text{ }\mu\text{mol} \text{ }m^{-2} \text{ }s^{-1} \text{ }per \text{ }W \text{ }m^{-2},$ 311 n = 6), and sites in the arid zone ($-0.009 \pm 0.002 \mu mol m^{-2} s^{-1}$ per W m⁻², n = 16) and the 312 cold zone ($-0.006 \pm 0.002 \mu$ mol m⁻² s⁻¹ per W m⁻², n =25) fall between these two mean 313 values. The differences in the slope value may potentially reflect CO₂ flux measurement 314 biases in different background air temperature, humidity and CO₂ density in different climate 315 316 zones.

317

318 **3.2 Self-heating effect**

According to our analysis in section 2.1, if self-heating is the main reason to explain the linear relationship between the appareant CO₂ flux and *H* and if the linear relationship between *H* and H_{real} reported by Buraba et al. (2008) holds, the slope *b* should be -0.008 µmol m⁻² s⁻¹ per W m⁻². This expected value is very close to the mean value of -0.007 ± 0.001 µmol m⁻² s⁻¹ per W m⁻² of the 54 sites that deployed the LI-7500 analyzer (Figure 4).



temperature is slightly negative (-0.2) and the correlation is not significant (p = 0.13). The correlation with site mean absolute humidity is not statistically significant either (Figure 6b). This single-variable correlation can be confounded by bias errors in the CO₂ density. Thus, we first calculated *b* caused by $\delta \rho_c$ using Equation (10) and then subtracted this value from the regression slope for each site. The resulting slope residual still does not show a positive linear relationship with air temperature (Figure S4). In other words, our meta-analysis failed to uncover a climatic pattern regarding the self-heating effect.

337

338 **3.3 Comparison among analyzer types**

The slope among the sites that deployed either the IRGASON or the EC150 analyzer is all 339 negative and varies in a narrow range from -0.019 to -0.011μ mol m⁻² s⁻¹ per W m⁻², giving 340 a mean value of $-0.013 \ \mu\text{mol}\ \text{m}^{-2}\ \text{s}^{-1}$ per W m⁻² (Figure 4). This mean value is 86 % greater 341 in magnitude than the mean value obtained for the LI-7500 analyzers, but the difference is 342 not statistically significant (p = 0.14). Because the IRGASON is an integrated system whose 343 measurements of the CO₂ concentration and the temperature are made in the same sensing 344 volume, any sensor self-heating would be automatically detected and be removed by the WPL 345 correction procedure. Instead, the correlation between the CO₂ flux bias and the sensible heat 346 flux is a result of the spectroscopic effect. The same spectroscopic effect also exists for the 347 EC150 gas analyzer (Helbig et al. 2016). 348

349

As for LI-7500, Welles & McDermitt (2005), Fratini et al. (2014) and Helbig et al. (2016)

believed that the air temperature only marginally affects the broadband measurements, so the

spectroscopic effect can be ignored. Our results show that, for LI-7500, the slope varies in a wider range from -0.051 to 0.013μ mol m⁻² s⁻¹ per W m⁻², with a mean value of -0.007μ mol m⁻² s⁻¹ per W m⁻². The average slope value is less negative than the result of IRGASON/EC150. Thus, we infer that the spectroscopic effect, if there is any, should be weaker for LI-7500 than for IRGASON/EC150.

357

The slopes for the 6 sites that deployed the LI-7500A gas analyzer are all negative, ranging 358 from -0.017 to -0.003 µmol m⁻² s⁻¹ per W m⁻², with a mean value of -0.012 µmol m⁻² s⁻¹ 359 per W m⁻². This mean value is 71 % greater in magnitude than the mean value for LI-7500, 360 but the difference is also not significant (p = 0.16). According to the manufacturer (Burba 361 2013), LI-7500A has been improved over LI-7500 to reduce the self-heating effect. Our result 362 363 suggests that the apparent uptake problem still exists for this type of analyzers, at least for the 6 sites we have analyzed (Figure S1). This result suggests that surface heating may not be the 364 only reason for the apparent wintertime net CO₂ uptake. 365

366

367 Using an open-path EC system consisting of a LI-7500A analyzer and a Gill anemometer,

368 Wang et al. (2016) observed a midday uptake flux of -1.6μ mol m⁻² s⁻¹ at a desert ecosystem

369 in the winter (mean air temperature -6.7 °C). They estimated that if self-heating were the

370 cause, this negative flux would require an amount of self-heat equivalent to 31 W m^{-2} .

371 However, the mean difference between the sensible heat fluxes derived from the two sonic

anemometers is only about 0.4 W m^{-2} . In their study, the apparent uptake flux measured with

the LI-7500A analyzer is nearly the same as that measured with an IRGASON EC system. It

374	is interesting that the mean slope for LI-7500A is nearly the same as that for
375	IRGASON/EC150 (Figure 4). Similar to IRGASON/EC150, the open-path methane analyzer
376	described by Burba et al. (2010) may need correction to account the spectroscopic effect due
377	to temperature fluctuations. Our current understanding of the LI-7500A analyzers is
378	inadequate to draw a firm conclusion as to whether they have the same type of errors.
379	

380 3.4 Bias in gas concentration measurements

As explained in Section 2.1, errors in the CO₂ concentration measurements affect the CO₂ 381 382 flux measurement. Underestimation of ρ_c will result in a negative slope value in the regression of the measured CO₂ flux versus the sensible heat flux. In the current study, the 383 bias in the CO₂ concentration was calculated as the measured value minus the CarbonTracker 384 385 result. To examine how well the surface CO₂ concentration produced by CarbonTracker represents the true CO₂ concentration, we compared the monthly CarbonTracker 386 concentration in the winter (January, February and December) with the measurement made 387 with closed-path analyzers at Harvard Forest in the U.S. and at Old Aspen in Canada (site 388 IDs: US-Ha1 and CA-Oas). At these sites, the closed-path analyzers (model LI-6262) were 389 calibrated periodically against CO₂ standard gases traceable to the World Meteorological 390 Organization (WMO) standards (Bakwin et al. 2004; Krishnan et al. 2006), so the 391 concentration measurements are in high quality. In a scatter plot, the results lie near the 1:1 392 line, with R² equal to 0.77 and 0.89 for Harvard Forest and Old Aspen, respectively (Figure 393 S5). On average, the residual (CarbonTracker minus observation) is -1.9 ± 5.2 ppm (mean \pm 394 1 SD) at Harvard Forest and -0.1 ± 2.6 ppm (mean ± 1 SD) at Old Aspen. This comparison 395

supports the use of CarbonTracker CO₂ concentration as a benchmark to evaluate bias errors in ρ_c measured with open-path analyzers.

398

Among the 51 sites having CO₂ concentration measurement records, the bias ratio ranges 399 400 from -18% to 6%, with a mean value of -5%. Of these sites, 45 sites show a negative bias. 401 The tendency to observe low biases with open-path analyzers in the cold season may be related to thermal contraction of the analyzer's optical path (Fratini et al. 2014) or lack of 402 frequent calibration. At these northern sites, it is common to perform instrument calibration in 403 the warm season when the sites are more accessible than in the cold season. 404 405 According to Equation (10), under typical atmosphere conditions, the regression slope b406 407 should be approximately equal to. At the averaged CO_2 concentration bias ratio of -0.05, the corresponding slope b is $-0.0025 \text{ }\mu\text{mol} \text{ }m^{-2} \text{ s}^{-1}$ per W m⁻², or about 30% of the mean 408 value of $-0.008 \pm 0.001 \ \mu mol \ m^{-2} \ s^{-1}$ per W m⁻² of the sites we analyzed. 409 410 The Pearson's correlation coefficient between the slope parameter b derived from 411 measurements and is 0.04. There is no significant linear relationship between b and the bias 412 in the CO₂ density (Figure 7, p = 0.99). The three outliers in Figure 7 are US-Wi5 (b = -0.051413 μ mol m⁻² s⁻¹ per W m⁻²), DE-RuS ($b = -0.037 \mu$ mol m⁻² s⁻¹ per W m⁻²) and AT-Neu ($b = -0.037 \mu$ mol m⁻² s⁻¹ per W m⁻²) 414 $-0.030 \ \mu mol \ m^{-2} \ s^{-1} \ per \ W \ m^{-2}$). 415 416

417 **3.5 Correcting wintertime CO₂ flux**

Our analysis suggests that a systematic bias exists at many open-path EC sites in the 418 FLUXNET network. A natural question is how to best correct the bias error. According to 419 Equations (5) and (9), the true flux F_c is related to the measured flux $F_{c,a}$ as 420 (13)421 The mean value of a among the 64 sites is 0.02 (\pm 0.10) µmol m⁻² s⁻¹, which is higher than 422 the value derived from self-heating theory (Section 2.1). In Bogoev et al. (2015), this 423 interception term is 0.067 µmol m⁻² s⁻¹ for an IRGASON analyzer. We consider this term as a 424 bias source which is not scaled with H and site specific. 425 426 Figure 8 shows the diurnal composite of the flux for a shrubland site in Northern China (site 427 ID: CN-Kub s) in December 2008 before and after correction using Equation (13) 428 (regression parameter values $b = -0.014 \ \mu \text{mol m}^{-2} \ \text{s}^{-1}$ per W m⁻² and $a = -0.28 \ \mu \text{mol m}^{-2} \ \text{s}^{-1}$). 429 The mean air temperature during this month was -9.2 °C. The analyzer model was LI-7500. 430 The original flux $F_{c,a}$ is negative in the daytime, with the most negative value (-2 µmol m⁻² 431 s^{-1}) occurring at noon when the sensible heat reaches the maximum value. The 24-hour mean 432 value is $-0.34 \,\mu\text{mol}\,\text{m}^{-2}\,\text{s}^{-1}$. After the correction, the flux F_c appears much more reasonable: 433 it varies in a very narrow range between $-0.25 \text{ }\mu\text{mol}\text{ }\text{m}^{-2}\text{ }\text{s}^{-1}$ and 0.37 $\mu\text{mol}\text{ }\text{m}^{-2}\text{ }\text{s}^{-1}$ in the 434 daytime, and the 24-hour mean value is slightly positive (0.11 μ mol m⁻² s⁻¹). 435 436 The correction procedure has a large impact on the cumulative carbon flux at this site. 437 Without the correction, the annual net ecosystem productivity (NEP) is 163 g C m^{-2} in 2008 438 (Figure 9). Here, a positive NEP indicates that the ecosystem is a sink for atmospheric CO₂, 439

440	and vice versa. If the correction is applied to the three winter months (December, January,
441	February), the annual NEP will change to 107 g C m ⁻² . If the correction equation is applied to
442	five months (November to March), the annual NEP will be 62 g C m ^{-2} (Figure 9). If the
443	correction is extended to the whole year, the annual NEP will be negative (-172 g C m^{-2}),
444	implying that the site is a carbon source. A similar sensitivity is documented by Amiro et al.
445	(2010), who showed that implementing the self-heating correction over varying time lengths
446	results in contrasting annual NEP values at a burned boreal forest site.
447	
448	At site AT-Neu (a grass land ecosystem in Austria), the annual NEP determined with an
449	open-path gas analyzer (LI-7500) is -72 g C m ⁻² in year 2003, while the result of a
450	closed-path gas analyzer is -119 g C m ⁻² . If the correction is applied to the three winter
451	months (December, January, February; $b = -0.03 \ \mu mol \ m^{-2} \ s^{-1}$ per W m ⁻² and $a = -0.39 \ \mu mol$

452 $m^{-2} s^{-1}$), the annual NEP will change to $-84 \text{ g C} m^{-2}$. When the correction is extended to the

453 whole year, the annual NEP will be -188 g C m⁻² (Figure 10). The result shows that some

biases with the open-path flux may still exist in the warm season, but the regression

455 parameters may be different from the values found for the cold season.

456

Our bias detection method assumes that the true CO_2 flux is independent of the sensible heat flux. Obviously, this assumption does not hold in the warm season, because a high rate of photosynthesis tends to occur at times of a high sensible heat flux, so we cannot use the same method to detect flux bias errors in the warm season. Inter-comparisons of open-path and closed-path measurements are necessary in order to determines the relationship between the flux bias errors and sensible heat flux in the warm season. In the case of IRGASON and
EC150 analyzers, the correction equation (Equation (13); Bogoev et al. 2014; Wang et al.
2016; Helbig et al. 2016) established in the cold season should also be applicable to other
months because the spectroscopic effect is the same all-year round.

466

467 **4. Conclusions**

In this study, we analyzed the CO_2 and sensible heat flux data collected at 64 eddy flux sites 468 in the cold season. A significant (p < 0.05) and negative linear relationship between the 469 observed CO₂ flux $F_{c,a}$ and the sensible heat flux H was found for 37 of the sites, suggesting a 470 systematic bias towards larger carbon uptakes in the FLUXNET network in the cold season. 471 The mean regression slope was -0.007 ± 0.001 , -0.012 ± 0.002 and -0.013 ± 0.001 µmol m⁻² 472 s⁻¹ per W m⁻² for LI-7500, LI7500A and IRGASON/EC150 gas analyzers, respectively. The 473 apparent uptake problem still exists for LI-7500A analyzers, even though LI-7500A has been 474 improved over LI-7500 to reduce the self-heating problem. This result suggests that self-475 476 heating may not be the only reason for the apparent wintertime net CO₂ uptake observed at many eddy flux sites. The slope value did not show statistically significant linear 477 relationships with local temperature and humidity. 478

479

480 On average, the CO₂ concentration measured at these sites (with open-path analyzers) in the

481 cold season is biased low by 5 % in comparison to the CarbonTracker surface CO₂

- 482 concentration. The corresponding slope value is $-0.0025 \ \mu mol \ m^{-2} \ s^{-1}$ per W m⁻² and about
- 483 30% of the mean value of the 64 sites, but the slope value is only weakly correlated with the

484 site mean concentration bias.

485

486	We have documented post-field corrections to the CO ₂ flux measured with the LI-7500 and
487	LI-7500A analyzers in the cold season. At present, we do not have evidence supporting the
488	application of these corrections for the whole year. Inter-comparisons of open-path and
489	closed-path eddy covariance measurements are necessary to investigate whether similar bias
490	errors exist in the warm season.

491

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Figure caption list

Figure 1. Locations of the 64 eddy flux sites used in this study.

Figure 2. Relationship between wintertime CO₂ flux ($F_{c,a}$) and sensible heat (H) at the Kubuqi shrub land site in China (site ID: CN-Kub_s; panel a) and at the Brooks cropland site in the U. S. (site ID: US-Br3; panel b). Grey dots represent half-hourly data, and black dots and error bars represent bin-average values and standard deviations. Error bounds of the regression coefficients are ± 1 standard error.

Figure 3. Distribution of the slope parameter *b* (μmol m⁻² s⁻¹ per W m⁻²) according to vegetation type: ENF, Evergreen Needleleaf Forest; DBF, Deciduous Broadleaf Forests; GRA, Grasslands; WET, Permanent Wetlands; CRO, Croplands; OSH, Open Shrub lands; BAR, Barrens.

Figure 4. Comparison among three gas analyzer types of the slope parameter *b* (μ mol m⁻² s⁻¹ per W m⁻²; gray bars) and the R² value (white bars) of the linear regression between wintertime CO₂ flux and sensible heat flux. Error bars are ± 1 standard error.

Figure 5. Same as Figure 4 except for comparison among climate zones. The letters (a and b) mean statistical differences at p < 0.05.

Figure 6. The regression slope parameter b (µmol m⁻² s⁻¹ per W m⁻²) versus site mean air temperature (panel a) and absolute humidity (panel b). Different symbols represent different vegetation types.

Figure 7. The regression slope parameter *b* (μ mol m⁻² s⁻¹ per W m⁻²) versus relative bias in the CO₂ concentration. Different symbols represent different analyzer types. The black line represents the theoretical relationship *y* = 0.05*x*.

Figure 8. Diurnal composite of the original uncorrected carbon flux (solid dots) and the corrected carbon flux (open dots) in December 2008 at the Kubuqi shrub land site (site ID: CN-Kub_s).

Figure 9. Cumulated net ecosystem productivity at the Kubuqi shrub land site (site ID:

CN-Kub_s). A positive NEP indicates that the ecosystem is a sink for atmospheric CO₂, and vice versa.

Figure 10. Cumulated net ecosystem productivity at the grass land site in Austria (site ID: AT-Neu). A positive NEP indicates that the ecosystem is a sink for atmospheric CO₂, and vice versa.



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Slope

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