



1 A dataset of microclimate and radiation and energy fluxes from the Lake Taihu Eddy

2 Flux Network

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21 Abstract

- 22 Eddy covariance data are widely used for the investigation of surface-air interactions.
- 23 Although numerical datasets exist in public depositories for upland ecosystems, few research
- 24 groups have released eddy covariance data collected over lakes. In this paper, we describe a
- 25 dataset from the Lake Taihu Eddy Flux Network, a network consisting of seven lake sites and
- one land site. Lake Taihu is the third largest freshwater lake (area 2,400 km²) in China, under
- the influence of subtropical climate. The dataset spans the period from June 2010 to
- 28 December 2018. Data variables are recorded at half-hourly intervals and include
- 29 micrometeorology (air temperature, humidity, wind speed, wind direction, rainfall, and
- 30 water/soil temperature profile), the four components of surface radiation balance, friction
- velocity, and sensible and latent heat fluxes. Except for rainfall and wind direction, all other
- variables are gap-filled, with each datapoint marked by a quality flag. Several areas of
- research can potentially benefit from the publication of this dataset, including evaluation of
- 34 mesoscale weather forecast models, development of lake-air flux parameterizations,
- 35 investigation of climatic controls on lake evaporation, validation of remote sensing surface
- data products, and global synthesis on lake-air interactions. The dataset is publicly available
- 37 at <u>https://yncenter.sites.yale.edu/data-access</u> and from Harvard Dataverse
- 38 <u>https://doi.org/10.7910/DVN/HEWCWM</u> (Zhang et al., 2020)
- 39





41 **1. Introduction**

42	Inland lakes and reservoirs are a vital freshwater resource for the society. Globally, there are
43	more than 27 million water bodies with size greater than 0.01 km^2 , occupying a total of 3.5%
44	of the Earth's land surface area (Downing et al., 2006; Verpoorter et al., 2014). Accurate
45	observation of the lake microclimate and lake-air interactions will help to better manage this
46	water resource and to better predict how it may be affected by environmental changes.
47	Towards that end, an increasing number of studies have deployed the eddy covariance (EC)
48	methodology to monitor physical state (temperature, wind, humidity) and process variables
49	(momentum flux, and radiation and energy fluxes) in the lake environment (Vesala et al.,
50	2006; Blanken et al., 2011; Nordbo et al., 2011; Wang et al., 2014; Li et al., 2015; Yusup and
51	Liu, 2016; Du et al., 2018; Hamdani et al., 2018; Xiao et al., 2018; Wang et al., 2019). Unlike
52	EC studies in upland ecosystems, however, data from these lake studies are rarely published
53	as data papers or are archived in public data depositories accessible by the broader scientific
54	community. For example, of the nearly 500 sites that have contributed EC and
55	micrometeorological data to AmeriFlux, a public data depository
56	(https://ameriflux.lbl.gov/data/data-availability/), none is a lake site. Although a few
57	scientific groups have provided data supplements to their scientific papers on lake-air fluxes
58	(e. g., Charusombat et al., 2018; Franz et al., 2018; Zhao and Liu, 2018), we are not aware of
59	a data paper devoted to systematic description and archival of EC lake observations.
60	
61	In this paper, we describe the dataset from the Lake Taihu Eddy Flux Network (Lee et al.,

62 2014). Established in 2010, the network currently consists of six active lake sites, one





63	inactive lake site, and one active land site. Lake Taihu is the third largest freshwater lake
64	(area 2,400 km ²) in China. Data variables are recorded at half-hourly intervals and the
65	measurement has continued for over eight years. Several areas of research can potentially
66	benefit from the publication of this dataset, including evaluation of mesoscale weather
67	forecast models, development of lake-air flux parameterizations, investigation of climatic
68	controls on lake evaporation, validation of remote sensing surface data products, and global
69	synthesis on lake-air interactions.
70	
71	This paper is organized as follows. Section 2 is a brief overview of the sites and the
72	instruments used by the network. This is followed, in Section 3, with a description of data
73	quality measures deployed during the field monitoring. Section 4 provides the essential
74	information about the dataset, including data variables, gap-filling methods, and data quality
75	flags. Results of post-field evaluation of the data quality are given in Section 5.
76	
77	Users of this dataset may be interested in the relevant papers published by our group. Lee et
78	al. (2014) gave an overview of the Lake Taihu Eddy Flux Network. Using the data collected
79	at a subset of the sites and during the early phase of the network, Wang et al. (2014)
80	investigated the spatial variability of energy and momentum fluxes across the lake. Xiao et al.
81	(2013) improved the bulk parameterizations of heat, water and momentum fluxes for shallow
82	lakes. Deng et al. (2013) and Hu et al. (2017) modified the CLM lake simulator (Subin et al.,
83	2012) to improve its prediction of the lake evaporation. Wang et al. (2017) and Zhang et al.
84	(2019b) evaluated the performance of two mesoscale models of the lake-land breeze. More





85	recently, Xiao et al. (2020, manuscript in review) investigated drivers of the interannual
86	variability of the lake evaporation observed at one of the lake sites (BFG). The value of the
87	dataset is enhanced by these peer-reviewed publications because they have helped us to
88	continuously improve our measurement and data processing protocols. For example, we have
89	used the locally-calibrated bulk parameterizations of Xiao et al. (2013) to gap-fill the flux
90	variables.
91	
92	2. Sites and Instrumentation
93	2.1 Sites and data periods
94	Table 1 shows the basic site information and Figure 1 is a map that gives the relative position
95	of Lake Taihu in China and locations of the EC measurement sites. Also shown in Figure 1
96	are WMO baseline weather stations around the lake, whose data can be obtained from
97	National Meteorological Information Center in China (http://data.cma.cn/site/index.html).
98	The lake, located between the latitudinal range of $30^{\circ}5'40''$ N to $31^{\circ}32'58''$ N and
99	longitudinal range of $119^{\circ}52'32''$ E to $120^{\circ}36'10''$ E, has a total area of 2400 km ² and an
100	average depth of 1.9 m. The climate is subtropical monsoon, with an annual mean
101	temperature of 16.2°C and annual total precipitation of 1122 mm. The lake is ice-free
102	throughout the year.
103	
104	The EC network consists of seven lake sites and one land site. The lake sites (Meiliangwan
105	(MLW), Dapukou (DPK), Bifenggang (BFG), Xiaoleishan (XLS), Pingtaishan (PTS),
106	Dongtaihu (DTH), Meiliangwan2 (MLW2)) are distributed according to biological





107	characteristics and across eutrophication gradients of the lake. The MLW site, located in
108	Meiliangwan Bay near the north shore of Lake Taihu, was the first site in operation; the
109	measurement began in June 2010 and was replaced by MLW2 in 2018, at 10 km southwest of
110	MLW. Both MLW and MLW2 sites are located in the lake eutrophic zone. BFG is located in
111	the east part of Lake Taihu in relatively clean water inhabited by submerged vegetation with
112	a growth season from April to November. DTH is located in the shallow water (mean depth
113	of 1.3 m) in the southeast part of the lake. After more than 20 years of crab aquaculture, this
114	zone was returned to unmanaged state in December 2018 in order to improve water quality.
115	The observation at DTH enables the examination of lake-air exchange processes in the
116	transition from human management to a natural state. PTS is situated in the middle of Lake
117	Taihu where occasional algal blooms occur and no aquatic vegetation is present. DPK is
118	located near the west shore, in a relatively deep (depth 2.5 m) super eutrophic zone due to
119	heavy influence of agricultural and urban runoffs. XLS is located in the relatively clean and
120	vegetation-free zone in the southeast. Finally, DS is a land site surrounded by rice agriculture,
121	serving as a land reference for the lake sites. The MLW site is situated at a distance of 200 m
122	from the north shore of the lake. All the other lake sites in the lake are at a distance of more
123	than 1 km away from the land.
124	

125 2.2 Instrumentation

Each site is equipped with an EC system for long-term, continuous monitoring of the surface momentum, sensible heat, latent heat and carbon dioxide fluxes. The EC system consists of a sonic anemometer/ thermometer (Model CSAT3A; Campbell Scientific, Logan, UT, USA)





- 129 and a CO₂/H₂O infrared analyzer (Model 7500A, LI-COR, Inc., Lincoln, NE, USA at DS,
- 130 MLW, MLW2 and DPK; Model EC150, Campbell Scientific at other sites). The EC
- instrument is at a height of 3.5 to 20 m above the water or the soil surface. Other
- 132 measurements include air humidity and air temperature (Model HMP45D/HMP155A;
- 133 Vaisala, Inc, Helsinki, Finland), wind speed and wind direction (Model 03002; R. M. Young
- 134 Company, Traverse City, MI, USA) and four components of the net radiation (Model CNR4;
- 135 Kipp & Zonen B. V., Delft, the Netherlands). At the lake sites, water temperature profile was
- measured with temperature probes (Model 109-L; Campbell Scientific) at the water depth of
- 137 20, 50, 100, and 150 cm and in the sediment at about 5 cm below bottom of the water column.
- 138 At the DS land site, soil temperature profile was measured with the same type of probes at
- the depths of 5, 10, 20 cm. The MLW and the DS sites are supported by A/C power and other
- sites are powered by battery packs connected to solar panels.
- 141
- 142 The methane flux was measured at MLW, BFG and DTH for selected periods, in addition to
- 143 the standard variables described above, using a flux-gradient system (at MLW; Xiao et al.)
- 144 2014) and an open-path eddy covariance system (at BFG and DTH, Zhang et al., 2019a). The
- (arbon dioxide and methane flux data are not included in the current version of the data)
- release but will be added at a later time after the data quality has been fully examined and the
- 147 data gaps filled.

148

All the variables are reported as 30-min averages. The EC covariance data are expressed in the natural coordinate system (Lee et al., 2004). Additionally, a small density correction has





been applied to the water vapor flux according to Webb et al. (1980).

152

- 153 **3 Data Quality Control during Field Monitoring**
- 154 Every site in the Lake Taihu Eddy Flux Network is equipped with a wireless transmission
- module for real-time monitoring and for data transmission. Time series of all 30-min
- variables are examined weekly and abnormal behaviors are flagged for site operators. Each
- site is visited every one to two months to perform instrument repair and maintenance and to
- download 10 Hz EC data. The data coverage rates are summarized in Table 2, where the
- 159 percentage values represent the proportions of data with quality flag 0 (Table 3).
- 160
- 161 The four-way net radiometers at MLW and XLS were compared in the field against a
- laboratory standard of the same type in the summer of 2018 to check their long-term stability.
- 163 These two sites were chosen because they have been in operation for more than five years.
- 164 Additionally, the radiometer at MLW was relocated to MLW2 after MLW had been
- discontinued. The laboratory standard, which had been calibrated at the manufacturer prior to
- this performance evaluation, was mounted next to the field instrument for about 10 days at
- 167 each site, covering overcast to clear-sky conditions. The mean bias error was smaller than 1
- W m⁻² for all the radiation components. It was -0.81, -0.81, 0.79 and -0.44 W m⁻² for the
- 169 downward shortwave, upward shortwave, downward longwave and upward longwave
- radiation flux at MLW, respectively. The corresponding values were 0.91, 0.40, 0.69 and
- $171 \quad 0.77 \text{ W m}^{-2}$ for XLS. (Comparison experiments are being planned for the other sites.)





173	The EC gas analyzers were calibrated every one to two years. The zero-point calibration was
174	carried out with high-purity nitrogen gas, the CO ₂ span calibration was made with standard
175	carbon dioxide gases (in the concentration range of 389 to 525 ppm) provided by the National
176	Institute of Meteorology (NIM), China and certified to an accuracy of 1% , and the H ₂ O span
177	calibration was made with a portable dew-point generator (LI-610; LI-COR, Inc.).
178	
179	4. Gap-filling Methods and <mark>Data Quality Flags</mark>
180	We use five-point moving average to screen outliers. If the deviation from the moving
181	average is greater than two standard deviations, the data point is discarded. If a gap length is
182	30 min to 1 h, the gap is filled by linear interpolation. Larger gaps in meteorological variables,
183	radiation components and water temperature are filled with linear regression involving
184	observation of the same variable at another site. This spatial interpolation consists of three
185	steps. First, linear correlation is calculated using the valid data at the target site and at all
186	other sites for the month during which the data gap occurred. Second, the observation at the
187	site with the highest linear correlation is used to establish a linear regression equation. Third,
188	the gap at the target site is filled with the linear regression and the observation at that site.
189	
190	Radiation data gaps at the DS land site require special treatment. The radiometer at DS eddy
191	flux site ended in January 2013. Subsequent measurements of the radiation component are
192	provided by a radiometer belonging to the Dongshan WMO weather station at a distance of
193	50 m from the eddy covariance tower (Figure 1). While large gaps in meteorological

194 variables (air temperature, relative humidity, wind speed and air pressure), downward solar





195	radiation and downward longwave radiation are filled with the spatial interpolation method,
196	large gaps in upward shortwave radiation and upward longwave radiation cannot be filled
197	with data from other lake sites even with linear regression. In the case of the upward
198	shortwave radiation, the data gaps were filled using the relationship between downward
199	shortwave radiation and the monthly mean albedo. In the case of upward longwave radiation,
200	the data gaps were filled by a regression equation between the upward longwave radiation
201	and the fourth power of soil temperature at 5-cm depth. Compared to the original data, the
202	gap-filled data do not capture the full diurnal variations but the daily-mean upward shortwave
203	and longwave radiation fluxes seem reasonable.
204	
205	Large data gaps in the EC variables (sensible heat flux, latent heat flux and friction velocity)
206	are filled with a hybrid method. If observations exist for the relevant state variable, the gap is
207	filled with the bulk transfer relationship using a locally-tuned transfer coefficient (Xiao et al.,
208	2013). For example, the relationship for filling gaps in the sensible heat flux is
209	$H = \rho_a c_p C_H U (T_s - T_a)$
210	where ρ_a is air density, c_p is specific heat of air at constant pressure, C_H is the transfer
211	coefficient for sensible heat, T_a is air temperature and T_s is water surface temperature. If data
212	for the state variable is missing, the spatial interpolation method is used to fill the gaps in
213	these EC variables.
214	
215	The spatial interpolation method described above occasionally causes a sudden jump at the

beginning or end of a data gap. To harmonize the data, we apply a 5-point moving averaging





to the gap-filled time series. If a data point deviates by 2 times of the standard deviation from
the moving average, it is replaced by linear interpolation using the two adjacent data points.

220	Each data point is assigned a quality flag to distinguish original measurements and gap-filled
221	values and gap-filling methods (Table 3). Flag 0 indicates high-quality original data. Other
222	flag values indicate gap-filled data or missing values. Flag 1 indicates that the data was filled
223	by temporal interpolation. Flag 2 indicates that the data was filled by the spatial interpolation
224	method. Flag 3 for the EC variables indicates that the data was filled by the bulk relationship.
225	We also use Flag 3 to mark the upward shortwave and longwave radiation data filled with the
226	albedo and the surface temperature relationship, respectively, for the DS land site. Missing
227	values occur on some situations, which are marked with Flag 4. Figure 3 is an example
228	showing the gap-filled time series of several variables at BFG along with the flag status.
229	
230	Rainfall data has not been quantity-controlled or gap-filled. Because of the episodic nature of
231	rainstorms and high spatial variability of rainfall, it is not appropriate to fill data gaps with
232	the time or spatial interpolation method. The total rain amount is likely biased low because
233	no wind screens are used to protect the rain gages from the influence of wind which is much
234	higher on the lake than on land (Figure 4 below). On several site visits, the drain opening to
235	the tipping bucket was found to be partially blocked by debris. Rain amount at a constant and
236	low rate and excessively long rain duration are evidence of such blockage. The flag status of
236 237	low rate and excessively long rain duration are evidence of such blockage. The flag status of 0 for the rainfall variable simply indicates that the field measurement is available, but it does





239

240	The data coverage begins from the start time of each site (Table 4) and the ends in December
241	2018. The time resolution is 30 min. The dataset includes microclimate variables (air pressure,
242	air temperature, relative humidity, wind speed, wind direction and rainfall), radiation fluxes
243	(upward and downward shortwave radiation, upward and downward longwave radiation),
244	water temperature at depth of 0.2 m, 0.5 m, 1.0 m and 1.5 m, and in the 5-cm sediment) and
245	eddy fluxes (friction velocity, and sensible heat and latent heat fluxes; Table 4). The time
246	stamp is Beijing time (UTC + 8 h) given by data columns 1 to 5 as year, month, day, hour,
247	and minute, and marks the end of the observation period. For example, time stamp "2012, 1,
<mark>248</mark>	12, 00' indicates that the data acquisition period is from 11:30 to 12:00 on January 1, 2012.
249	
250	Although the data table does not include the radiative surface temperature T_s , the user can

easily calculate it from the two longwave radiation fluxes, as

252
$$T_{s} = \left(\frac{L_{\uparrow} - (1 - \varepsilon)L_{\downarrow}}{\varepsilon\sigma}\right)^{\frac{1}{4}}$$

where σ is the Stefan-Boltzmann constant, ε is emissivity, and L_1 and L_1 are upward and downward longwave radiation flux, respectively. We use a value of 0.97 for lake surface emissivity in this calculation (Deng et al., 2013; Wang et al., 2014).

256

257 **5. Data Consistency Evaluation**

Figure 4 compares the annual mean air temperature, relative humidity, and wind speed at theTaihu eddy flux sites with those at the four WMO weather stations (Wuxi, Liyang, Huzhou





260	and Dongshan) around the lake (Figure 1). The error bars represent the maximum and
261	minimum values among the four WMO stations and the lines represent the mean values of
262	the four station measurements. The annual mean air temperature at DTH is 0.3°C higher than
263	the station mean. At other sites, air temperature is in close agreement of the weather station
264	data, in terms of both magnitude and inter-annual variability. The annual mean wind speed at
265	MLW, a site near the shoreline, is comparable with the station data. At other more exposed
266	sites, the wind speed is much higher than observed at the WMO stations. The annual mean
267	relative humidity RH shows a larger spread among the eddy flux sites than among the WMO
268	stations partly because the measurement height at the eddy flux sites is not standardized
269	(Table 1). The upward trends in RH over time at DPK and XLS seem to be related more to
270	aging of the sensor than to a real inter-annual variability. We have not fully investigated this
271	aging problem, but it is possible to rectify it by doing a detailed regression analysis against
272	the station data.
273	
274	Consistency of the energy flux variables can be evaluated with the energy balance closure.

Using observations made at a subset of the sites in the earlier years of the flux network,

Wang et al. (2014) reported a closure rate of 70 % to 110 % on the monthly basis, meaning

that the sum of the measured monthly sensible and the latent heat flux $H + \lambda E$ is 70 % to

278 110 % of the monthly available energy $R_n - G$, where R_n net radiation and G is heat storage

in the water column. By selecting days without data gaps, we found that the daily energy

balance closure is in the range between 66 % and 78 % for all the lake sites and all the years.

Such closure rates are typical of eddy covariance observations (Tanny et al., 2008; Wilson et





282 al., 2002).

283

284	We have shown that the monthly latent heat flux at the lake sites MLW, BFG and DPK
285	during July 2010 to August 2012 follows the Priestley-Taylor (PT) model prediction with the
286	original PT constant α of 1.26 and that at the DS land site is in agreement with the PT model
287	if the constant is lowered to 1.0 (Lee et al., 2014). Figure 5 demonstrates that the same
288	relationships hold for all the sites and all the observational months, indicating the overall
289	stability of our measurement systems and the robustness of our gap-filling procedure. The
290	reader is reminded that the monthly latent heat flux in Figure 5 has been adjusted to force
291	energy closure following the method recommended by Barr et al. (1994), Blanken et al.
292	(1997) and Twine et al. (2000). (The half-hourly flux data in the data archive have not been
293	adjusted for energy balance.)
294	
295	The Stefan-Boltzmann Law offers another way for checking data consistency. Because the
296	lake surface emits longwave radiation like a blackbody and because the annual mean air
297	temperature and the surface water temperature are nearly identical at this lake (Wang et al.,
298	2014), the change in the annual upward longwave radiation ΔL_{\uparrow} can be expressed as
299	$\Delta L_{\uparrow} = 4\sigma T_a^{3} \Delta T_a$
300	where T_a is annual mean air temperature, and Δ is the difference between the target year and
301	the year with the lowest air temperature observed at the site. All the five long-term lake sites
302	show good consistency between the longwave radiation and the air temperature observations

303 (Figure 6).





304

305	6 Data availability
306	All data can be open-accessed online for download and use at https://yncenter.sites.yale.edu/
307	and from Harvard Dataverse (https://doi.org/10.7910/DVN/HEWCWM, Zhang et al., 2020).
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309 7 Summary
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- 310 The dataset described here consists of microclimate variables (air temperature, air humidity,
- 311 wind speed, wind direction, water or soil temperature profile, and rainfall), four components
- of the radiation balance, friction velocity, and sensible and latent heat fluxes observed at
- seven lake sites and one land site. The period of coverage is from June 2018 to December
- 2018. The observation interval is 30 min. Except for rainfall and wind direction, all other
- variables have been gap-filled. Every data point is tagged with a data quality flag to help the
- 316 user determine how to best use the data.

317

318 Author contribution

- 319 XL, WX and MZ directed the field program, ZZ performed data gap-filling and prepared the
- data for public release, CC, WW, CX, HC, JW, JZ, LJ, QL, WH, WZ, YL, YX, YW, YP, YH,
- 321 ZC and ZQ participated in field data collection, and ZZ, XL and MZ wrote the manuscript.

322

323 **Competing interests**

324 The authors declare no conflict of interest.

325

326 Acknowledgments





327	This work was supported by the National Key R&D Program of China (2019YFA0607202),
328	the National Natural Science Foundation of China (grant numbers 41575147, 41801093, and
329	41475141) and the Priority Academic Program Development of Jiangsu Higher Education
330	Institutions (grand number PAPD).
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332	





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Table 1. A list of sites in the Lake Taihu Eddy Flux Network

Site ID	MLW	DPK	BFG	XLS	PTS	MLW2	DTH	DS
Lat (°N)	31.4197	31.2661	31.1685	30.9972	31.2323	31.3818	31.0611	31.0799
Long (°E)	120.2139	119.9312	120.3972	120.1344	120.1086	120.1608	120.4704	120.4346
Start date	Jun 2010	Aug 2011	Dec 2011	Nov 2012	Jun 2013	Feb 2018	Nov 2017	Apr 2011
Biology	Eutrophic	Super eutrophic	Submerged macrophyte	Transitional	Mesotrophic	Eutrophic	Aquaculture	Cropland/ Rural residence
Met height (m)	3.5	8.0	8.5	9.4	8.5	6.0	4.5	10.0
$T_{\rm w}$ / $T_{\rm s}$ depths	20, 50, 100,	20, 50, 100,	20, 50, 100,	20, 50, 100,	20, 50, 100,	20, 50, 100,	20, 50,	5 10 20
(cm)	150, sediment	150, sediment	150, sediment	150, sediment	150, sediment	150, sediment	sediment	5, 10, 20
Radiation height (m)	1.5	1.5	1.5	1.5	1.5	1.5	1.5	3.0
EC height (m)	3.5	8.5	8.5	9.4	8.5	6.0	4.5	20

336





339 Table 2 .	Percent of data coverage
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	-	<u> </u>						
Variable type	MLW	DPK	BFG	XLS	PTS	DTH	MLW2	DS
Micrometeorology	93.3	81.1	97.6	97.0	97.5	98.1	90.3	91.7
Radiation flux	85.5	90.8	96.9	97.4	98.6	98.2	98.2	82.7
Water/soil temperature	83.4	81.3	94.0	91.1	90.3	87.7	22.4	98.4
Eddy flux	73.3	61.8	82.7	79.1	80.6	85.7	85.5	82.8





344

345	Table 3. A list of data quality flags			
	Flag Data quality description			
	0	Original data		
	1	Gap-filled with time interpolation		
	2	Gap-filled with spatial interpolation		
	3	Gap-filled with bulk relationship		
	4	NAN		

346





Table 4. A list of data columns and variable definitions

Column	Description	Variable name	Unit
1	Year	Year	_
2	Month	Month	-
3	Day	Day	-
4	Hour	HH	-
5	Minute	MM	-
6	Day of Year	DOY	-
7	Air pressure	Р	kPa
8	Quality flag of air pressure	P_flag	
9	Air temperature	Та	°C
10	Quality flag of air temperature	Ta_flag	
11	Relative humidity	RH	%
12	Quality flag of Relative humidity	RH_flag	
13	Wind speed	WS	m s ⁻¹
14	Quality flag of wind speed	WS_flag	
15	Wind direction	WD	Degree
16	Quality flag of wind direction	WD_flag	
17	Rainfall	R	mm
18	Quality flag of rainfall	R_flag	
19	Upward shortwave radiation	UR	W m ⁻²
20	Quality flag of upward shortwave radiation	UR_flag	
21	Downward shortwave radiation	DR	W m ⁻²
22	Quality flag of downward shortwave	DR_flag	
	radiation		
23	Upward longwave radiation	ULR	W m ⁻²
24	Quality flag of upward longwave radiation	ULR_flag	
25	Downward longwave radiation	DLR	W m ⁻²
26	Quality flag of downward longwave	DLR_flag	
	radiation		
27	Water temperature at 0.2 m	T _w _20	°C





28	Quality flag of Water temperature at 0.2 m	$T_w_{20}_{flag}$	
29	Water temperature at 0.5 m	T _w _50	°C
30	Quality flag of Water temperature at 0.5 m	$T_w_50_flag$	
31	Water temperature at 1.0 m	Tw_100	°C
32	Quality flag of Water temperature at 1.0 m	Tw_100_flag	
33	Water temperature at 1.5 m	T _w _150	°C
34	Quality flag of water temperature at 1.5 m	T _w _150_flag	
35	Sediment temperature	T _w _bot	°C
36	Quality flag of sediment temperature	T _w _bot_flag	
37	Friction velocity	U^{*}	m s ⁻¹
38	Quality flag of friction velocity	U*_flag	
39	Sensible heat flux	Н	W m ⁻²
40	Quality flag of sensible heat flux	H_flag	
41	Latent heat flux	LE	W m ⁻²
42	Quality flag of latent heat flux	LE_flag	

Notes: 1) Time marks end of an observation in Beijing time (UTC+8:00); 2) At the DS site, columns 27, 29,

and 31 represent soil temperature at 5, 10, and 20 cm, respectively.

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359 **Figure 1.** Map showing locations of Lake Taihu, eddy covariance sites (red bubbles) and

360 WMO weather stations (yellow triangles). The background is a natural color image from

361 LANDSAT 8 without correction for atmospheric interference.

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Figure 2. Comparison of four components of the radiation balance between the original 368 radiometer (horizontal axis) and a laboratory standard (vertical axis) at MLW and XLS. Refer 369 to Table 4 for variable definitions. 370

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Figure 3. Complete gap-filled time series for selected variables observed at BFG. Blue, black,
cyan and red dots represent quality flag 0, 1, 2, and 3, respectively. Variable definitions are
given in Table 4

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Figure 4. Annual mean air temperature (top), relative humidity (middle) and wind speed
(bottom) observed at the eddy flux sites (symbols) and at the four WMO weather stations
around the lake (line). Error bar represents the range of the annual means of the four WMO
stations.





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Figure 5. Comparison of observed monthly latent heat flux with Priestley-Taylor model prediction using the origional α coefficient of 1.26 and a modified coefficient of 1.03. Here R_n is net radiation, *G* is heat storage in the water column, Δ is the slope of the saturation vapor pressure curve, and γ is the psychrometric constant.







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Figure 6. The relatinship between changes in observed annual mean upward longwave

- 400 radiation flux and annual mean air temperature (dots). Solid lines represent the prediction of
- 401 the Stefan-Boltzmann Law.





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