

1 **Title:** Evaluation of methods for estimating daily reference crop evapotranspiration at a site in the
2 humid alpine meadow

3 **Running title:** Comparison of reference evapotranspiration with lysimeter measurements

4 **Authors:** Licong Dai^{1,2,4}, Ruiyu Fu³, Xiaowei Guo^{1,2*}, Yangong Du^{1,2}, Fawei Zhang^{1,2}, Yikang
5 Li^{1,2}, Guangmin Cao^{1,2*}

6 **Affiliations:**

7 ¹ Qinghai Provincial Key Laboratory of Restoration Ecology for Cold Region, Northwest Institute
8 of Plateau Biology, Chinese Academy of Sciences, Xining 810001, China

9 ² Key Laboratory of Adaptation and Evolution of Plateau Biota, Northwest Institute of Plateau
10 Biology, Chinese Academy of Sciences, Xining, Qinghai 810001, China

11 ³ School of Geography and Planning, Nanning Normal University, Nanning 530001, China

12 ⁴ University of Chinese Academy of Science, Beijing 100049, China

13 * Corresponding author: Yangong Du and Guangmin Cao

14 **E-mail:** caogm@nwipb.cas.cn for Guangmin Cao and xwguo1206@163.com for XiaoweiGuo

15 **Author contributions:** L Dai and X Guo performed the research, analyzed data, and wrote the
16 paper; F Zhang, R Fu, Y Li and Y Du analyzed data; G Cao conceived the study.

17 **Conflict of interest:** The authors declare no conflict of interest

18 **Abstract**

19 Evapotranspiration as the key component of the terrestrial water cycle, an accurate assessment of
20 evapotranspiration is of great importance for water irrigation management. Although many
21 applicable ET models have been developed, most models are mainly focused on low altitude
22 regions, with little attention on alpine ecosystem worldwide. In this paper, we evaluate the
23 performance of 14 evapotranspiration (ET₀) models by comparison with large weigh lysimeter

24 measurements. Specifically, we use Bowen ratio-energy balance method, three combination
25 models, seven radiation-based models and three temperature-based models driven with data from
26 June 2017 to December 2018 in a humid alpine meadow, northeastern Qinghai-Tibetan Plateau.
27 The daily actual evapotranspiration was obtained by large weighing lysimeters located in an alpine
28 *Kobresia* meadow. We found that the performances of the 14 ET_0 models, ranked on the basis of
29 their RMSE (root mean square error), decreased in the order: Bowen> Priestley-Taylor> DeBruin-
30 Keijman> 1963Penman> FAO-24 Penman> FAO-56 Penman-Monteith > IRMAK1>
31 Makkink(1957) > Makkink(1967)>Makkink> IRMAK2 > Hargreaves>Hargreaves1>Hargreaves2.
32 For the combination models, FAO 24 Penman yielded the highest correlation, followed by Pen-63
33 and FAO-56 PM. For radiation-based models, PT and DK obtained the highest correlation,
34 followed by Makkink, Makkink(1967) and Makkink (1957), IRMAK1 and IRMAK2. For
35 temperature-based models, HAR, HAR1 and HAR2 obtained the same correlation. Overall, the
36 Bowen performed best, with RMSEs 0.98, followed by combination models (ranged from 1.19 to
37 1.27 mm d^{-1} and averaged 1.22 mm d^{-1}), radiation-based models (ranged from 1.02 to 1.42 mm
38 d^{-1} and averaged 1.27 mm d^{-1}) and temperature-based models (ranged from 1.47 to 1.48 mmd^{-1} and
39 averaged 1.47 mm d^{-1}). Furthermore, all models tended to underestimate measured ET_a during
40 periods of larger evaporative demand (i.e. growing season) and overestimate measured ET_a during
41 lower evaporative demand (i.e. non-growing season). Our results could provide a new sight for the
42 accurate assessment of evapotranspiration in an alpine ecosystem.

43 **Keywords:** alpine meadow; lysimeter measurement; Bowen ratio-energy balance method;
44 combination models; radiation-based models; temperature-based models

45 **1 Introduction**

46 Evapotranspiration (ET) is one of the key parameters in the simultaneous processes of heat
47 and water transfer to the atmosphere via transpiration and evaporation in the soil–plant–
48 atmosphere system (Sentelhas, Gillespie, & Santos, 2010), thereby playing an important role in
49 water balance calculations, water allocation and water irrigation management. Thus, an accurate
50 assessment of evapotranspiration could improve water management strategies and promote the
51 efficient use of water resources, especially in these regions suffering water shortages (Sun et al.,
52 2011).

53 To-date, direct measurements of ET have been achieved by a variety of methods such as the
54 Bowen Ratio Energy Balance System (A. Irmak & Irmak, 2008; S. Irmak, Allen, & Whitty, 2003;
55 S. Irmak, Howell, Allen, Payero, & Martin, 2005; Si et al., 2005), lysimeters (Jia, Dukes, Jacobs,
56 & Irmak, 2006; Valipour, 2015) and the eddy covariance technique (Novick et al., 2009; Zhang et
57 al., 2018). Alternatively, ET can be indirectly assessed by applying various reference
58 evapotranspiration equations. Several models of reference evapotranspiration have become widely
59 used for the calculation of ET, and can be classified into three types: radiation-based models
60 (Hargreaves & Samani, 1985), temperature-based models (Trajkovic, Gocic, Pongracz, &
61 Bartholy, 2019), and combination models (Penman, 1963). While the development of these models
62 has undoubtedly benefited the calculation of ET, it still difficult to choose the optimal one due to
63 the availability of the observed data together with most models have not been evaluated against
64 lysimeter measurements across a range of regions and climates (Kiefer, Andresen, Doubler, &
65 Pollyea, 2019; Liu et al., 2017). To select the best performing models, many studies have been
66 conducted to assess model performance under various climates. For instance, the Food and
67 Agriculture Organization of the United Nations (FAO) recommend the Penman–Monteith FAO-56

68 combination equation (PM-56) as the standard equation for estimating reference
69 evapotranspiration (Richard G Allen, Pereira, Raes, & Smith, 1998), and this has been widely used
70 worldwide when compared with other equations (Cai, Liu, Lei, & Pereira, 2007). The advantages
71 of the Penman–Monteith equation are that it does not require any local calibration because it
72 incorporates both physiological and aerodynamic parameters, and it has been well tested by a
73 variety of lysimeters (Trajkovic, 2009). Although many models have been widely used to estimate
74 ET, it should be noted that most previous models have only been evaluated with reference to FAO-
75 56 PM (Cao, Li, Liu, Zhong, & Zhao, 2015; Liu et al., 2017), with few being tested against
76 lysimeter measurements. Furthermore, the application of the PM-56 equation needs many
77 meteorological inputs, such as wind speed, temperature, humidity and solar radiation, that are
78 often not available in regions with harsh environments (Martel, Glenn, Wilson, & Kröbel, 2018;
79 Tabari, Kisi, Ezani, & Talaei, 2012). Thus, it is essential to develop a relatively accurate reference
80 evapotranspiration model that requires fewer meteorological parameters, to allow more simplified
81 estimates of ET than those of PM-56, applicable across a range of climatic conditions (Tabari &
82 Talaei, 2011). So far, many models have been developed. For example, Tabari (2010) assessed
83 four reference evapotranspiration models in an arid climate, and found that the Turc model
84 performed the best. Meanwhile, the Hargreaves equation performed best in semiarid regions
85 (Sabziparvar & Tabari, 2010). Liu et al. (2017) compared 16 models for reference
86 evapotranspiration against weighing lysimeter measurements, and found that the combination
87 models performed best for estimating ET in semiarid regions. Overall, most previous studies have
88 been conducted in low-humidity conditions at low altitude (i.e. arid and semiarid regions) (Liu et
89 al., 2017; Sentelhas et al., 2010), with few studies in humid climates, particularly in alpine

90 ecosystems.

91 The Qinghai-Tibetan Plateau (QTP), with an average altitude of 4000 m, is the world's
92 highest alpine ecosystem and is also known as the "Asian tower", playing an important role in
93 ensuring the safety of water resources in China and southeast Asia (Dai, Guo, Zhang, et al., 2019;
94 Zou et al., 2017). The alpine meadow almost account for almost 60% of the plateau area (Dai, Ke,
95 et al., 2019); therefore, accurate assessment of ET in an alpine ecosystem are not only provides
96 new insights into the water cycle, but also benefit the formulation of water resource management
97 strategies. Furthermore, given the uncertainty and confusion in the selection of ET equations
98 across different regions and climates, it is critical to thoroughly understand the performance of the
99 various models in a humid alpine meadow across worldwide (Zhang et al., 2018). The objectives
100 of the study was to evaluate the performance of 14 evapotranspiration (ET_0) models by
101 comparison with large weigh lysimeter measurements, with the aims of selecting the best fit model
102 in applications in this climate zone worldwide to estimate the ET.

103 **2 Materials and methods**

104 **2.1 Study area**

105 The study was conducted at the Haibei National Field Research Station, Qinghai, China
106 ($37^{\circ}37' N$, $101^{\circ}19'E$), which is situated on the Northeastern QTP at an elevation of 3200 m a.m.s.l
107 (Fig.1a). This area is characterized by a plateau continental monsoon climate, with well-developed
108 seasonally frozen ground. The average annual air temperature is $-1.7^{\circ}C$, with the maximum
109 monthly temperature in July ($10.1^{\circ}C$) and minimum monthly temperature in January ($-15.0^{\circ}C$).
110 The annual precipitation is about 580 mm, of which 80% falls in the growing season (i.e. from
111 May to September), lead to a high water content (close to field capacity) in the soil during

112 growing season, thus the evapotranspiration during growing season can be considered as reference
113 evapotranspiration (ET_0) according to FAO indications (Richard G Allen et al., 1998). The average
114 annual pan evaporation is approximately 1191.4 mm (Zhang et al., 2018). The soil type around the
115 lysimeter system is classified as Mat-Gryic Cambisol, which belongs to a clay loam, and has a
116 thickness of approximately 60 – 80 cm (Dai, Guo, Du, et al., 2019), the basic soil property was
117 show in Table 1. The grass crop is dominated by perennial sedge and graminoid species, including
118 *Kobresia humilis*, *Stipa aliena* and *Elymus nutans*, which together constitute 60% - 80% of plant
119 cover around the lysimeter system (Fig.1b).

120 **2.2 ET measurement and data quality control**

121 The actual ET was measured by large-scale weighing lysimeters (height 2 m, diameter 1 m,
122 and resolution 1g) (Fig.1c), which was recorded with a data logger (CR1000, Campbell, USA).
123 The soil in the two lysimeters was repacked soil, the time step of lysimeters measurement was 30
124 min. The grass crop was keep the consistent conditions to so that the measured data as
125 representative as possible. To ensure data quality, all negative or abnormal ET values caused by
126 falling soil particles were discarded; the abnormal ET values were these outlier that more than
127 three times of average ET, this screening process yielded 393 days of data spanning June 2017 to
128 December 2018. According to in-situ phenological observations of the foliage of dominant plants,
129 we defined the growing season as that from May 1st to September 30th, while the period from
130 October 1st to April 30th of the following year is defined as the non-growing season (Zhang et al.,
131 2018).

132 **2.3 Meteorological data collection**

133 All meteorological variables needed to calculate ET_0 using the various models were obtained

134 or estimated from the weather station at Haibei Station, and included relative humidity (RH), wind
135 speed (Field, Behrenfeld, Randerson, & Falkowski), net radiation (R_n), total radiation (R_s),
136 extraterrestrial solar radiation (R_a), soil heat flux (G), maximum air temperature (T_{max}), vapor
137 pressure deficit (VPD), minimum air temperature (T_{min}) and mean air temperature (T). The
138 radiation were measured by four radiometers (CNR4, Kipp&Zonen, Netherlands) at 1.5 m height;
139 relative humidity, wind speed and mean air temperature were measured at 1.5 m height (HMP45C,
140 Vaisala, Finland), and wind speed was converted to 2 m height for calculating ET_0 . The soil heat
141 flux (G) was measured by three heat flux plates (HFT-3, Campbell, USA), which were separately
142 buried 5 cm beneath the surface. Half-hourly means of meteorological data were stored by a data
143 logger (9210 XLITE, Sutron, USA). It should be note that the distance between weather station
144 and lysimeters was less than 10 m.

145 **2.4 Reference evapotranspiration models**

146 A total of 14 often-used ET_0 models were selected for comparison, including Bowen ratio-
147 energy balance method, three combination models (1963 Penman, FAO24 Penman and FAO-56
148 PM), seven radiation models (Priestley-Taylor, De Bruin-Keijman, Makkink, Makkink (1957),
149 Makkink (1967), IRMAK1 and IRMAK2) and three temperature-based models (Hargreaves,
150 Hargreaves1, Hargreaves2), to compare their performance by using lysimeter measurement. The
151 specific equations and parameters of these models are listed in Table 2.

152 **2.5 Evaluation criteria**

153 Here, the ET_a measured by the large-scale weighing lysimeters was the actual ET, and the
154 performances of the ET_0 equations were compared to these lysimeter system estimates on a daily
155 time step. The pair-wise comparisons were conducted using general linear regression. For further

156 comparison, the root mean squared error (RMSE), percentage error of estimate (PE), mean
 157 absolute error (MAE) and coefficient of determination (R^2) were used for the evaluation of
 158 reference evapotranspiration models. The RMSE, PE, MAE and R^2 are defined as:

$$RMSE = \sqrt{\frac{\sum_{i=1}^n (P_i - O_i)^2}{n}}$$

159

$$PE = \left| \frac{\bar{P} - \bar{O}}{\bar{O}} \right| \times 100\%$$

160

$$MAE = \frac{\sum_{i=1}^n (P_i - O_i)}{n}$$

161

$$R^2 = \frac{[\sum_{i=1}^n (P_i - \bar{P})(O_i - \bar{O})]^2}{\sum_{i=1}^n (P_i - \bar{P})^2 \sum_{i=1}^n (O_i - \bar{O})^2}$$

162

163 where P_i are the predicted values, O_i are observed values, \bar{P} and \bar{O} are the average of P_i and O_i , and
 164 n is the total number of data.

165 2.6 Statistical analysis

166 To achieve the best comparison between the models and measurements, we need to select the
 167 dominant meteorological factors affecting the measured ET. Given that it may not be appropriate
 168 to explore results based solely on the coefficient of independent variables in multiple regression
 169 analysis, owing to the strong collinearities and nonlinearities among meteorological factors, we
 170 adopted a boosted regression trees (BRT) model to quantitatively evaluate the relative influences
 171 of meteorological variables on measured ET. In the past, the BRT method has been widely used to
 172 improve the performance of a single model through by fitting a large number of models,
 173 ultimately yielding an overall prediction (Martínez-Rincón, Ortega-García, & Vaca-Rodríguez,
 174 2012). Most importantly, the BRT can evaluate the relative influence of an independent variable

175 on a dependent variable, without transformations, and can cope well with non-linear relationships.
176 Furthermore, the BRT displays good performance in dealing with stronger collinearities and
177 nonlinearities. Thus, the BRT was adopted to evaluate the individual influences of controlling
178 factors on measured ET. All statistical analyses were conducted in R software version 3.03 (R
179 Development Core Team, 2006), and all figures were plotted by Origin 9.0.

180 **3 Results**

181 **3.1 Seasonal variation of measured ET and environmental variables**

182 The measured ET showed a clear seasonal pattern, the growing season ET was significantly
183 higher than that in the non-growing season ($P < 0.05$) (Fig.2a). The average measured daily ET
184 during the study period was $2.33 \pm 0.11 \text{ mm d}^{-1}$, with average daily measured ET of $4.14 \pm 0.13 \text{ mm}$
185 and $0.65 \pm 0.05 \text{ mm d}^{-1}$ (here and throughout the remainder of the paper, values are expressed as
186 mean ± 1 SD) during the growing season and non-growing season, respectively (Fig.2a).
187 Environmental variables showed a similar seasonal pattern, with the maximum and minimum
188 values in the growing season and non-growing season (Fig. 2). The average daily R_n , R_s , T_a , VPD
189 and RH during the growing season were $9.53 \pm 0.28 \text{ MJ m}^{-2}$, $18.50 \pm 0.38 \text{ MJ m}^{-2}$ and 10.15 ± 0.21
190 $^{\circ}\text{C}$, $0.33 \pm 0.01 \text{ kPa}$, $74.88 \pm 0.62 \%$, respectively (Fig.2). The average daily R_n , R_s , T_a , VPD and
191 RH during the non-growing season were $3.28 \pm 0.15 \text{ MJ m}^{-2}$, $12.65 \pm 0.25 \text{ MJ m}^{-2}$ and -3.67 ± 0.42
192 $^{\circ}\text{C}$, $0.23 \pm 0.01 \text{ kPa}$, $57.16 \pm 0.86 \%$, respectively.

193 **3.2 Comparison of daily ET between reference evapotranspiration models and lysimeter** 194 **measurements during the study period**

195 The comparison of 14 evapotranspiration equations against the lysimeter measurements is
196 presented in Fig.3 and Table 3, showing that the relationship between daily ET_0 calculated by the
197 reference evapotranspiration equations and lysimeter measurements are each significantly

198 ($P < 0.01$), with higher coefficients of determination (R^2) ranging from 0.59 to 0.86. For the
199 combination models, FAO 24 Penman yielded the highest correlation, followed by FAO-56 PM
200 and Pen-63. For radiation-based models, PT obtained the highest correlation, followed by DK,
201 Makkink(1967) and Makkink (1957), IRMAK1, Makkink and IRMAK2. For temperature-based
202 models, HAR obtained the highest correlation, HAR1 and HAR2 obtained the same correlation.
203 The daily estimates of combination models generally underestimated the ET values measured by
204 lysimeter, with MAE -0.26 mm d^{-1} to -0.01 mm d^{-1} . However, the radiation-based models (except
205 PT and IRMAK1) and temperature-based models generally temperature-based models, with MAE
206 ranging from -0.14 mm d^{-1} to 0.50 mm d^{-1} for radiation-based models, and ranging from 0.40 mm
207 d^{-1} to 0.59 mm d^{-1} for temperature-based models.

208 During the whole study period, the RMSE of Bowen ratio-energy balance method was 0.98,
209 the RMSE of combination models ranged from 1.19 to 1.27 mm d^{-1} and averaged 1.22 mm d^{-1} ,
210 Furthermore, the RMSE of FAO-56 PM increased from 1.22 to 1.29 mm d^{-1} as r_s changed from 20
211 to 60 s m^{-1} (Fig.4),. The RMSE for radiation-based models ranged from 1.02 to 1.42 mm d^{-1} and
212 averaged 1.27 mm d^{-1} , and the RMSE for temperature-based models ranged from 1.47 to 1.48
213 mmd^{-1} and averaged 1.47 mm d^{-1} . Based on the RMSE, the performances of the 14
214 evapotranspiration models decreased in the order: Bowen>PT>DK>Pen-63>FAO-24 Pen>FAO-56
215 PM> IRMAK1>Makkink (1957) >Makkink (1967) > Makkink > IRMAK2>HAR>HAR1>HAR2.
216 The best model (Good, Noone, & Bowen) was, respectively, 34% and 30% more accurate than the
217 poorest (HAR2) and the commonly used FAO-56 PM equation. Furthermore, the Pen-63 and
218 FAO-24 Pen demonstrated better performance than the commonly used FAO-56 PM equation.
219 Overall, for the whole study period, the Bowen yielded the best performance, followed by the

220 combination models, radiation-based models and temperature-based models.

221 **3.3 Comparison of daily ET between reference evapotranspiration models and lysimeter** 222 **measurements during the growing season**

223 During the growing season, the daily ET_0 calculated by 14 evapotranspiration equations was
224 significantly correlated with the lysimeter measurements ($P < 0.01$), with R^2 ranging from 0.32 to
225 0.64 (Fig.5 and Table 4). Of the combination models, FAO-56 PM obtained the highest R^2 ,
226 followed by FAO-24 Pen and Pen-63. Of the radiation-based models, PT obtained the highest R^2 ,
227 followed by DK, IRMAK1, Makkink(1967), Makkink (1957), IRMAK2 and Makkink. It should
228 be noted that Makkink(1967) and Makkink (1957) have the same R^2 . Of the temperature-based
229 models, HAR, HAR1 and HAR2 obtained the same R^2 . Interestingly, all models (except for
230 HAR1 and HAR2) generally underestimated ET during the growing season, values of MAE ranged
231 from -1.10 to -0.15 mm d^{-1} , with FAO-56 PM having the largest underestimate (by 26.59%) and
232 HAR the minimum underestimate (by 3.56%) (Table 4).

233 The RMSE of Bowen was 1.31, the RMSE for combination models ranged from 1.38 to 1.58
234 mm d^{-1} and averaged 1.47 mm d^{-1} , the RMSE for radiation-based models ranged from 1.19 to 1.55
235 mm d^{-1} and averaged 1.40 mm d^{-1} , and the RMSE for temperature-based models ranged from 1.42
236 to 1.43 mm d^{-1} and averaged 1.42 mm d^{-1} (Table 4). Based on the RMSE values, the performances
237 of the 14 evapotranspiration models follow the order: DK > PT > Makkink (1967) > Bowen > Pen-
238 63 > HAR1 > HAR2 > HAR > FAO-24 Pen > Makkink > IRMAK1 > Makkink(1957) > IRMAK2 > FAO-
239 56. Evidently, the best (DK) was 25% more accurate than the poorest (FAO-56). Overall, for the
240 growing season period, the Bowen yielded the best performance, followed by the radiation-based
241 models, temperature-based models and combination models.

242 **3.4 Comparison of daily ET between reference evapotranspiration models and lysimeter**
243 **measurements during the non-growing season**

244 During the non-growing season, the daily ET_0 calculated by the 14 evapotranspiration
245 equations was also significantly correlated with the lysimeter measurements ($P < 0.01$), but with
246 lower coefficients of determination (R^2) ranging from 0.17 to 0.64 (Fig.6 and Table 5). Of the
247 combination models, FAO-24 Pen obtained the highest R^2 , followed by Pen-63 and FAO-56 PM.
248 Of the radiation-based models, PT and DK obtained the highest R^2 , followed by Makkink(1967),
249 Makkink (1957), Makkink, IRMAK1 and IRMAK2. Of the temperature-based models, HAR
250 obtained the highest R^2 . Interestingly, all model (except Bowen) generally overestimated the ET
251 values measured by lysimeter during the non-growing season, with MBEs ranging from 0.40 to
252 1.20 $mm\ d^{-1}$ and averaging 0.78 $mm\ d^{-1}$; Makkink (1967) yielded the largest underestimate (by
253 185.83 %) and PT the minimum underestimate (by 61.06%)(Table 5).

254 The RMSE of Bowen was 0.52, the RMSE for combination models ranged from 0.86 to 1.00
255 $mm\ d^{-1}$ and averaged 0.92 $mm\ d^{-1}$, the RMSE for radiation-based models ranged from 0.80 to 1.44
256 $mm\ d^{-1}$ and averaged 1.12 $mm\ d^{-1}$, and the RMSE for temperature-based models ranged from 1.47
257 to 1.54 $mm\ d^{-1}$ and averaged 1.50 $mm\ d^{-1}$. Based on the RMSE, the performances of the reference
258 evapotranspiration models decreased in the order: Bowen>PT>DK> FAO-24Pen> FAO-56>Pen-
259 63>IRMAK1>Makkink(1957)>Makkink>IRMAK2>Makkink(1967)>HAR>HAR1>HAR2.

260 Evidently, the best model was 66.24% more accurate than the poorest (HAR2). Overall, for the
261 non-growing season period, the Bowen yielded the best performance, followed by the combination
262 models, radiation-based models and temperature-based models.

263 **3.5 Comparison of monthly averaged daily ET_0 between reference evapotranspiration**

264 **models and lysimeter measurements**

265 The estimations of 14 evapotranspiration models were consistent with the pattern shown by
266 lysimeter measurements (Fig. 7), with the peak in July. As already noted, the combination models
267 and Bowen underestimated the measurements from May to September, and overestimated the
268 measurements in the other months (Fig. 7a and 7b). The radiation-based models underestimated the
269 measurements from June to September, and overestimated the measurements in the other months
270 (Fig. 7c). However, the temperature-based models generally overestimated the measured ET
271 during most months (except July and August) (Fig. 7d). Overall, all models tended to
272 underestimate the measured ET during the growing season (with larger evaporative demand), and
273 overestimated ET during the non-growing season (with reduced evaporative demand).

274 **3.6 Dominant factors affecting the seasonal variation in lysimeter ET measurements**

275 BRT model indicated that the R_n was the dominant factor controlling the seasonal variation in
276 measured ET during the whole study period, accounting for 69.02% of total variability, followed
277 by VPD (7.13%), T (6.75%), RH (5.73%), R_a (5.15%), R_s (4.33%) and WS (1.85%) (Fig. 8a).
278 During the growing season, the R_n remained the main control of seasonal variation in measured
279 ET (Fig. 8b), accounting for 44.30% of total variability, followed by T (14.12%), R_s (12.56%), R_a
280 (8.34%), RH (7.80%), VPD (6.87%) and WS (6.02%). However, the seasonal variation of
281 measured ET in non-growing season was dominated by RH (Fig. 8c), accounting for 27.99% of
282 total variability, followed by R_n (20.99%), R_a (12.96%), VPD (12.56%), T (10.18%), R_s (9.39) and
283 WS (5.94%) (Fig. 8).

284 **4 Discussion**

285 **4.1 The performance comparison of combination models against lysimeter measurements**

286 Previous study have shown that the Penman family models are generally the most accurate

287 when evaluating ET across various climate scenarios and regions (Liu et al., 2017). Of the penman
288 models for ET_0 , the Penman–Monteith FAO 56 has been considered as the standard equation for
289 estimating evapotranspiration (Richard G Allen et al., 1998). For instance, Yoder et al. (2005)
290 found that the Penman–Monteith FAO 56 display the best performance in the humid southeast
291 United States. López-Urrea et al. (2006) test seven evapotranspiration equations using lysimeter
292 observations in a semiarid climate, and found the FAO-56 Penman–Monteith equation was the
293 most precise method compared with other evapotranspiration equations. Contrary to previous
294 studies, the Penman–Monteith FAO 56 was not the best in our study, we found Pen-63 and FAO-
295 24 Pen were more accurate (Pen-63 and FAO-24 Pen had smaller RMSE than Penman–Monteith
296 FAO 56 during the whole study period). Such similar results have been reported in many other
297 studies (Berengena & Gavilán, 2005; Martel et al., 2018). A more recent study also reported the
298 poor performance of FAO-56 PM when compared with data from 20 FLUXNET towers (Ershadi,
299 McCabe, Evans, Chaney, & Wood, 2014). Combining these results suggests that FAO-56 PM
300 might not be the only standard model for evaluating ET_0 , because it did not yield better accuracy
301 than the other penman models. Given the better performance of Pen-63 and FAO-24 than FAO-56
302 PM in this study, we may apply old Penman family models to our study region, especially
303 considering the Penman–Monteith FAO 56 requires many meteorological inputs, which limit its
304 use in areas with sparse data, especially in harsh environment (Tabari et al., 2012). Overall, the
305 poor performance of FAO-56 PM may attributed to the higher aerodynamic resistance (r_s), there is
306 increasing evidence indicated that the underestimation of FAO-56 PM suggested that the fixed r_s =
307 70 s m^{-1} in the equation is probably too large (Liu et al., 2017), this result was also confirmed by
308 our results that the values of RMSE increased from 1.22 to 1.29 mm d^{-1} as r_s changed from 20 to

309 60 s m^{-1} , and the RMSE was nearly unchanged when r_s varied between 0 and 20 s m^{-1} , the RMSE
310 was range from 1.12 to 1.13 mm d^{-1} (Fig. 4). Therefore, reducing value of r_s from 70 s m^{-1} to 0–20
311 s m^{-1} can improve daily estimates of the FAO-56 PM. Other studies also found that the r_s should be
312 a variable value rather than the fixed one (Richard G. Allen et al., 2006; Liu et al., 2017). For
313 instance, the r_s should be smaller when it was being underestimated and should be larger when it
314 was being overestimated (Ventura, Spano, Duce, & Snyder, 1999).

315 **4.2 The performance comparison of radiation and temperature models against lysimeter** 316 **measurements**

317 For the performance comparison of radiation models against lysimeter measurements, we
318 found that the PT models yielded the best performances of the radiation-based models, which was
319 in line with the previous study conducted in humid areas that the PT method exert the good
320 accuracy estimate for ET (Ershadi et al., 2014). There is increasing evidence indicated that the
321 input parameters was the dominated factors affecting their performance (Lang et al., 2017), we
322 thus conclude that the better performances of the PT models which might associated with the use
323 of only the most important meteorological factors affecting ET such as net radiation (R_n), was
324 supported by our results that the net radiation was the most important factor controlling measured
325 ET (Fig.8). Compare with PT, the other radiation models just use the R_s as the mainly driving
326 variable, the ET thus may overestimate because some R_s was reduced through reflecting into the
327 atmosphere due to the high albedo in this region (Zhang et al., 2018). Furthermore, each model
328 was developed from its specific underlying surface and climate conditions. For instance, the PT
329 was established in a humid climate condition, which was suitable for our humid alpine meadow.
330 Most importantly, the PT models required fewer meteorological inputs when compared with

331 combination models. Combine those factors, we can prefer to recommend the PT and DK model
332 for use in a humid alpine meadow on the northeastern Qinghai-Tibetan Plateau, especially when
333 considering the difficulty in obtaining ET in this harsh climate.

334 For the performance comparison of temperature models against lysimeter measurements,
335 previous study reported that the Hargreaves versions equation as one of the most simple empirical
336 methods were widely used for the ET estimation due to its less meteorological data input,
337 especially considering not all the meteorological data required in the standard PM-56 model
338 (Jensen, Burman, & Allen, 1990). To further select the best Hargreaves versions equation, we
339 compare the performance of original (HAR) and two modified versions (HAR1 and HAR2) of the
340 Hargreaves equations, and found that the original HAR model had the lowest error (RMSE =1.47
341 mm d⁻¹, MAE=0.40 mm d⁻¹ and PE=17.37 %), which was consistent with previous studies
342 conducted on humid region (Tabari, 2010) but contrast to these study conducted in arid region that
343 the modified Hargreaves equation display a more accurate estimation of evapotranspiration
344 compared with the original Hargreaves equation (Ravazzani, Corbari, Morella, Gianoli, &
345 Mancini, 2012). Overall, the temperature models display a poor performance compared with
346 radiation models owing to the Hargreaves method was established in semiarid areas (Tabari,
347 2010), thus a local calibrations was required to improve the accuracy of Hargreaves method in
348 other region.

349 **4.3 The performance comparison of all the models**

350 By comparing the four type models, we found that the Bowen yielded the best performance,
351 followed by the combination models, radiation-based models and temperature-based models
352 (Table 3). Overall, most radiation-based models generally underestimated the measured ET during

353 the whole study period, whereas the temperature-based models tended to overestimate ET. This is
354 consistent with previous studies where the Makkink and Priestley-Taylor models generally
355 underestimated ET (Priestley & Taylor, 1972; Xu & Singh, 2002), while the Hargreaves equations
356 often overestimate ET in cold-humid conditions and requires a local calibration (Berti, Tardivo,
357 Chiaudani, Rech, & Borin, 2014). Given that the study region in our study belongs to humid
358 alpine meadow, thus ET tended to be overestimated. An alternative explanation for the poor
359 performance of the Hargreaves model in humid regions may owing to the Hargreaves method was
360 established in semiarid areas (Tabari, 2010), and the R_a parameter used in the Hargreaves model,
361 which is based on the maximum possible radiation value and does not take the atmospheric
362 transmissivity into account. However, the atmosphere transmissivity in humid regions is affected
363 by many factors, such as atmospheric moisture; thus, the solar radiation reaching the surface is
364 significantly reduced due to the high atmospheric moisture content (Temesgen, Allen, & Jensen,
365 1999), resulting in the overestimation of solar radiation, ultimately leading to an overestimation of
366 ET by the Hargreaves method.

367 Furthermore, there were also common features of all four groups of models. All the models
368 tended to underestimate the measured ET during the growing season (with larger evaporative
369 demand), and overestimated ET during the non-growing season (with reduced evaporative
370 demand), which was consistent with a previous study conducted in a semi-arid region (Liu et al.,
371 2017). Furthermore, we found that the measured ET and calculated ET_0 were less correlated during
372 non-growing season than during growing season. These discrepancies may relate to the dominant
373 component between transpiration and evaporation. The transpiration was the dominant during
374 growing season, almost account for 75% of evapotranspiration, whereas the evaporation was the

375 dominant component during non-growing season in the same study site (Zhang et al., 2018).
376 Considering the evaporation process was much complex and affected by many environmental
377 factors compared with transpiration process, ultimately lead to a poor correlation between
378 measured ET and calculated ET_0 during non-growing season. Therefore, both Hargreave's
379 equations and other models need further local or region calibration before being applied to a given
380 region (Xu & Singh, 2002). Besides, it should be noted that the data used in this study just
381 obtained from a year and a single weather station, which may insufficient to represent the whole
382 humid climate or the alpine ecosystem but represent a specific site. Thus, a longer period and
383 more lysimeter systems should be used in the alpine ecosystem in the future to obtain more
384 accurate estimates of evapotranspiration over the northeastern Qinghai-Tibetan Plateau.

385 **5 Conclusion**

386 This study is the first to document information on the comparison of evapotranspiration
387 models against lysimeter measurements in a humid alpine meadow. we found that the Bowen
388 ratio-energy balance method performed the best, followed by combination models, radiation-based
389 models and temperature-based models. In addition, the combination models tended to
390 underestimate measured ET, whereas temperature-based models and most radiation-based models
391 tended to overestimate measured ET during the whole study period. Specifically, all models
392 tended to underestimate ET during the growing season and overestimate ET during the non-
393 growing season, suggesting that these models should be calibrated or modified by local lysimeter
394 data when extrapolated to other regions. Furthermore, the 1963 Penman and FAO-24 Penman
395 models demonstrated better performances than the recommended FAO-56 Penman-Monteith
396 (PM), suggesting that older Penman equations may be superior to the standard FAO-56 Penman-

397 Monteith model, especially considering the good performance of the 1963 Penman model in this
398 study. Given the outstanding performance of Priestley-Taylor model, which require few
399 meteorological inputs, we thus recommend that these two models can be used in other alpine
400 meadows that have similar climates to that of the study region, to improve ET estimation.

401 **Conflict of interest**

402 The authors declare that there are no conflicts of interest associated in this manuscript.

403 **Data Availability Statement**

404 The data that support the findings of this study are available on request from the
405 corresponding author.

406 **Acknowledgments**

407 This work was supported by the National Natural Science Foundation of China (41730752,
408 31700395, 31770532 and 31400483), the comprehensive management and application
409 demonstration of small watershed in Hainan prefecture (2019-SF-152).

410 **References**

- 411 Allen, R. G., Pereira, L. S., Raes, D., & Smith, M. (1998). Crop evapotranspiration-Guidelines for
412 computing crop water requirements-FAO Irrigation and drainage paper 56. *Fao, Rome*,
413 *300(9)*, D05109.
- 414 Allen, R. G., Pruitt, W. O., Wright, J. L., Howell, T. A., Ventura, F., Snyder, R., . . . Yrisarry, J. B.
415 (2006). A recommendation on standardized surface resistance for hourly calculation of
416 reference ETo by the FAO56 Penman-Monteith method. *Agricultural Water Management*, *81*.
- 417 Berengena, J., & Gavilán, P. (2005). Reference evapotranspiration estimation in a highly advective
418 semiarid environment. *Journal of Irrigation and Drainage Engineering*, *131(2)*, 147-163.
- 419 Berti, A., Tardivo, G., Chiaudani, A., Rech, F., & Borin, M. (2014). Assessing reference

420 evapotranspiration by the Hargreaves method in north-eastern Italy. *Agricultural Water*
421 *Management*, 140, 20-25.

422 Cai, J., Liu, Y., Lei, T., & Pereira, L. S. (2007). Estimating reference evapotranspiration with the FAO
423 Penman–Monteith equation using daily weather forecast messages. *Agricultural and Forest*
424 *Meteorology*, 145(1-2), 22-35.

425 Cao, J., Li, Y., Liu, X., Zhong, X., & Zhao, Y. (2015). Comparison of four combination methods for
426 reference crop evapotranspiration. *Chinese Journal of Agrometeorology*, 36(4), 428-436.

427 Dai, L., Guo, X., Du, Y., Zhang, F., Ke, X., Cao, Y., . . . Cao, G. (2019). The Response of Shallow
428 Groundwater Levels to Soil Freeze–Thaw Process on the Qinghai–Tibet Plateau.
429 *Groundwater*, 57(4), 602-611.

430 Dai, L., Guo, X., Zhang, F., Du, Y., Ke, X., Li, Y., . . . Shu, K. (2019). Seasonal dynamics and controls
431 of deep soil water infiltration in the seasonally-frozen region of the Qinghai-Tibet plateau.
432 *Journal of Hydrology*, 571, 740-748.

433 Dai, L., Ke, X., Guo, X., Du, Y., Zhang, F., Li, Y., . . . Shu, K. (2019). Responses of biomass allocation
434 across two vegetation types to climate fluctuations in the northern Qinghai–Tibet Plateau.
435 *Ecology and evolution*, 9(10), 6105-6115.

436 De Bruin, H., & Stricker, J. (2000). Evaporation of grass under non-restricted soil moisture conditions.
437 *Hydrological sciences journal*, 45(3), 391-406.

438 Droogers, P., & Allen, R. G. (2002). Estimating reference evapotranspiration under inaccurate data
439 conditions. *Irrigation and drainage systems*, 16(1), 33-45.

440 Ershadi, A., McCabe, M., Evans, J. P., Chaney, N. W., & Wood, E. F. (2014). Multi-site evaluation of
441 terrestrial evaporation models using FLUXNET data. *Agricultural and Forest Meteorology*,
442 187, 46-61.

443 FAO. (1998). Crop evapotranspiration Guidelines for computing crop water requirements. *Fao*
444 *Irrigation & Drainage Paper*, 56.

445 Field, C. B., Behrenfeld, M. J., Randerson, J. T., & Falkowski, P. (1998). Primary production of the
446 biosphere: integrating terrestrial and oceanic components. *Science*, 281(5374), 237-240.

447 Good, S. P., Noone, D., & Bowen, G. (2015). Hydrologic connectivity constrains partitioning of global
448 terrestrial water fluxes. *Science*, 349(6244), 175-177.

449 Hansen, S. (1984). Estimation of Potential and Actual Evapotranspiration: Paper presented at the
450 Nordic Hydrological Conference (Nyborg, Denmark, August-1984). *Hydrology Research*,
451 15(4-5), 205-212.

452 Hargreaves, G. H., & Samani, Z. A. (1985). Reference crop evapotranspiration from temperature.
453 *Applied engineering in agriculture*, 1(2), 96-99.

454 Irmak, A., & Irmak, S. (2008). Reference and crop evapotranspiration in South Central Nebraska. II:
455 Measurement and estimation of actual evapotranspiration for corn. *Journal of Irrigation and*
456 *Drainage Engineering*, 134(6), 700-715.

457 Irmak, S., Allen, R., & Whitty, E. (2003). Daily grass and alfalfa-reference evapotranspiration estimates
458 and alfalfa-to-grass evapotranspiration ratios in Florida. *Journal of Irrigation and Drainage*
459 *Engineering*, 129(5), 360-370.

460 Irmak, S., Howell, T., Allen, R., Payero, J., & Martin, D. (2005). Standardized ASCE Penman-
461 Monteith: Impact of sum-of-hourly vs. 24-hour timestep computations at reference weather
462 station sites. *Transactions of the ASAE*, 48(3), 1063-1077.

463 Jensen, M. E., Burman, R. D., & Allen, R. G. (1990). *Evapotranspiration and irrigation water*
464 *requirements*.

465 Jia, X., Dukes, M. D., Jacobs, J. M., & Irmak, S. (2006). Weighing lysimeters for evapotranspiration

466 research in a humid environment. *Transactions of the ASABE*, 49(2), 401-412.

467 Kiefer, M. T., Andresen, J. A., Doubler, D., & Pollyea, A. (2019). Development of a gridded reference
468 evapotranspiration dataset for the Great Lakes region. *Journal of Hydrology: Regional
469 Studies*, 24, 100606.

470 Lang, D., Zheng, J., Shi, J., Liao, F., Ma, X., Wang, W., . . . Zhang, M. (2017). A Comparative Study of
471 Potential Evapotranspiration Estimation by Eight Methods with FAO Penman–Monteith
472 Method in Southwestern China. *Water*, 9(10), 734.

473 Liu, X., Xu, C., Zhong, X., Li, Y., Yuan, X., & Cao, J. (2017). Comparison of 16 models for reference
474 crop evapotranspiration against weighing lysimeter measurement. *Agricultural Water
475 Management*, 184, 145-155.

476 López-Urrea, R., de Santa Olalla, F. M., Fabeiro, C., & Moratalla, A. (2006). Testing
477 evapotranspiration equations using lysimeter observations in a semiarid climate. *Agricultural
478 Water Management*, 85(1-2), 15-26.

479 Makkink, G. F. (1957). Testing the Penman formula by means of lysimeters. *Journal of the Institution
480 of Water Engineers*, 11, 277-288.

481 Martel, M., Glenn, A., Wilson, H., & Kröbel, R. (2018). Simulation of actual evapotranspiration from
482 agricultural landscapes in the Canadian Prairies. *Journal of Hydrology: Regional Studies*, 15,
483 105-118.

484 Martínez-Rincón, R. O., Ortega-García, S., & Vaca-Rodríguez, J. G. (2012). Comparative performance
485 of generalized additive models and boosted regression trees for statistical modeling of
486 incidental catch of wahoo (*Acanthocybium solandri*) in the Mexican tuna purse-seine fishery.
487 *Ecological Modelling*, 233, 20-25.

488 Nemani, R. R., Keeling, C. D., Hashimoto, H., Jolly, W. M., Piper, S. C., Tucker, C. J., . . . Running, S.

489 W. (2003). Climate-driven increases in global terrestrial net primary production from 1982 to
490 1999. *Science*, 300(5625), 1560-1563.

491 Novick, K., Oren, R., Stoy, P., Juang, J.-Y., Siqueira, M., & Katul, G. (2009). The relationship between
492 reference canopy conductance and simplified hydraulic architecture. *Advances in Water
493 Resources*, 32(6), 809-819.

494 Penman, H. (1963). Vegetation and Hydrology Tech. Comm. No 53. *Commonwealth Bureau of Soils,
495 Harpenden, England, 125p.*

496 Poeplau, C., Don, A., Vesterdal, L., Leifeld, J., Van Wesemael, B., Schumacher, J., & Gensior, A.
497 (2011). Temporal dynamics of soil organic carbon after land-use change in the temperate
498 zone-carbon response functions as a model approach. *Global Change Biology*, 17(7), 2415-
499 2427.

500 Priestley, C. H. B., & Taylor, R. (1972). On the assessment of surface heat flux and evaporation using
501 large-scale parameters. *Monthly weather review*, 100(2), 81-92.

502 Ravazzani, G., Corbari, C., Morella, S., Gianoli, P., & Mancini, M. (2012). Modified Hargreaves-
503 Samani equation for the assessment of reference evapotranspiration in Alpine river basins.
504 *Journal of Irrigation and Drainage Engineering*, 138(7), 592-599.

505 Sabziparvar, A.-A., & Tabari, H. (2010). Regional estimation of reference evapotranspiration in arid
506 and semiarid regions. *Journal of Irrigation and Drainage Engineering*, 136(10), 724-731.

507 Sentelhas, P. C., Gillespie, T. J., & Santos, E. A. (2010). Evaluation of FAO Penman-Monteith and
508 alternative methods for estimating reference evapotranspiration with missing data in Southern
509 Ontario, Canada. *Agricultural Water Management*, 97(5), 635-644.

510 Si, J., Feng, Q., Zhang, X., Liu, W., Su, Y., & Zhang, Y. (2005). Growing season evapotranspiration
511 from *Tamarix ramosissima* stands under extreme arid conditions in northwest China.

512 *Environmental Geology*, 48(7), 861-870.

513 Sun, G., Alstad, K., Chen, J., Chen, S., Ford, C. R., Lin, G., . . . Miao, H. (2011). A general predictive
514 model for estimating monthly ecosystem evapotranspiration. *Ecohydrology*, 4(2), 245-255.

515 Tabari, H. (2010). Evaluation of Reference Crop Evapotranspiration Equations in Various Climates.
516 *Water Resources Management*, 24(10), 2311-2337.

517 Tabari, H., Kisi, O., Ezani, A., & Talaei, P. H. (2012). SVM, ANFIS, regression and climate based
518 models for reference evapotranspiration modeling using limited climatic data in a semi-arid
519 highland environment. *Journal of Hydrology*, 444, 78-89.

520 Tabari, H., & Talaei, P. H. (2011). Local Calibration of the Hargreaves and Priestley-Taylor Equations
521 for Estimating Reference Evapotranspiration in Arid and Cold Climates of Iran Based on the
522 Penman-Monteith Model. *Journal of Hydrologic Engineering*, 16(10), 837-845.

523 Temesgen, B., Allen, R., & Jensen, D. (1999). Adjusting temperature parameters to reflect well-watered
524 conditions. *Journal of Irrigation and Drainage Engineering*, 125(1), 26-33.

525 Trajkovic, S. (2009). Comparison of radial basis function networks and empirical equations for
526 converting from pan evaporation to reference evapotranspiration. *Hydrological Processes: An
527 International Journal*, 23(6), 874-880.

528 Trajkovic, S., Gocic, M., Pongracz, R., & Bartholy, J. (2019). Adjustment of Thornthwaite equation for
529 estimating evapotranspiration in Vojvodina. *Theoretical and Applied Climatology*, 138(3-4),
530 1231-1240.

531 Valipour, M. (2015). Retracted: Comparative Evaluation of Radiation-Based Methods for Estimation of
532 Potential Evapotranspiration. *Journal of Hydrologic Engineering*, 20(5), 04014068.

533 Ventura, F., Spano, D., Duce, P., & Snyder, R. L. (1999). An evaluation of common evapotranspiration
534 equations. *Irrigation ence*, 18(4), 163-170.

535 Xu, C.-Y., & Singh, V. (2002). Cross comparison of empirical equations for calculating potential
536 evapotranspiration with data from Switzerland. *Water resources management*, 16(3), 197-219.

537 Yoder, R., Odhiambo, L. O., & Wright, W. C. (2005). Evaluation of methods for estimating daily
538 reference crop evapotranspiration at a site in the humid southeast United States. *Applied*
539 *engineering in agriculture*, 21(2), 197-202.

540 Zhang, F., Li, H., Wang, W., Li, Y., Lin, L., Guo, X., . . . Cao, G. (2018). Net radiation rather than
541 surface moisture limits evapotranspiration over a humid alpine meadow on the northeastern
542 Qinghai–Tibetan Plateau. *Ecohydrology*, 11(2), e1925.

543 Zou, D., Zhao, L., Yu, S., Chen, J., Hu, G., Wu, T., . . . Pang, Q. (2017). A new map of permafrost
544 distribution on the Tibetan Plateau. *The Cryosphere*, 11(6), 2527.

545

546