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1 Cool Farm Tool Water: A global on-line tool to assess water use  
2 in crop production.

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21 **Abstract**

22 The agricultural sector accounts for 70% of all water consumption and poses great pressure on ground  
23 water resources. Therefore, evaluating agricultural water consumption is highly important as it allows  
24 supply chain actors to identify practices which are associated with unsustainable water use, which risk  
25 depleting current water resources and impacting future production. However, these assessments are  
26 often not feasible for crop producers as data, models and experiments are required in order to conduct

27 them. This work introduces a new on-line agricultural water use assessment tool that provides the water  
28 footprint and irrigation requirements at field scale based on an enhanced FAO56 approach combined with  
29 a global climate, crop and soil databases. This has been included in the Cool Farm Tool – an online  
30 tool which already provides metrics for greenhouse gas emissions and biodiversity impacts and therefore  
31 allows for a more holistic assessment of environmental sustainability in farming and agricultural supply  
32 chains. The model is tested against field scale and state level water footprint data providing good results.  
33 The tool provides a practical, reliable way to assess agricultural water use, and offers a means to engage  
34 growers and stakeholders in identifying efficient water management practices.

## 35 **Highlights**

- 36 • CFTW facilitates water use assessments in crop production in food supply chains.
- 37 • Global on-line tool combining user input with a soil, daily climate and crop database.
- 38 • Comparison to observed evapo-transpiration ( $R^2 = 0.52$ ) shows good agreement.
- 39 • Comparison to observed total water footprint ( $R^2 = 0.96$ ) shows very good agreement.
- 40 • Using CFTW can reduce errors in water footprints by >70% compared to state estimates.

## 41 **Keywords**

42 water footprint, FAO56, crop water use, stakeholder involvement, water resource management, irrigation  
43 requirements

## 44 **1 Introduction**

45 With increasing global food demand, agricultural water use and consequent ground water depletion,  
46 improved farm water management is becoming increasingly critical (Godfray et al., 2010; Siebert et al.,  
47 2010; Tilman et al., 2011; Wada et al., 2012). A recent global modelling study by Jägermeyr et al. (2016)  
48 investigated different integrated crop water management interventions, including an increase of irrigated  
49 areas. The study indicates that production could be increased by 41% and thus the gap in future global  
50 food demand could be reduced by 50% - but not without further increasing irrigation water consumption.  
51 Therefore a solid understanding and estimation of crop water usage, crop water demand and the effect of  
52 different water management at farm level is crucial to enable the identification of improved management  
53 opportunities.

54 Several models, of varying complexity, have been developed in order to account for water use in crop  
55 production at the field scale (Baroni et al., 2010; Kroes et al., 2008; Raes et al., 2006; Ragab, 2002; Rosa  
56 et al., 2012; Smith et al., 1992; Steduto et al., 2009). Most of them use, to some extent, the approach  
57 presented in the "FAO irrigation and drainage paper No. 56 crop evapotranspiration" (FAO56) (Allen  
58 et al., 1998). However, these models are often not tailored to application by crop producers, due to (i),  
59 the unavailability of soil, crop and climate data required for the model, (ii), the use of terminology not

60 understood outside the research community, (iii), lack of an engaging user interface for some models, in  
61 addition to (iv), a lack of guidance on how to interpret and use results. Bastiaanssen et al. (2007) raised  
62 similar concerns for soil hydrological models. Table 1 gives a short overview of some of the existing tools  
63 based on FAO56. The selection is based on models described in the scientific literature and the provision  
64 of a graphical user interface.

65 The models vary with respect to data integration, with most data being provided by the ICARDA Agro-  
66 Climate tool (Mauget and De Pauw, 2010) and SAPWAT (van Heerden, 2008) for north-west Africa  
67 to central Asia and South Africa, respectively. CROPWAT, SAPWAT and Aquacrop provide climate  
68 data on a global scale via the climate database CLIMWAT, which contains long-term average data from  
69 5000 climate stations (van Heerden, 2008; Smith et al., 1992; Steduto et al., 2009). The data can also  
70 be downloaded and used for the other existing models. Most tools provide default soil profiles and  
71 parameters, but do not use soil maps to increase usability.

72 This study presents the new field scale agricultural water assessment tool Cool Farm Tool Water (CFTW)  
73 which is fully integrated with the already existing greenhouse and biodiversity model Cool Farm Tool  
74 (CFT)(Hillier et al., 2011). The novelty of this tool is that it combines tested algorithms with a database  
75 of climate, soil and crop data on a global scale in an on-line tool and packages them for non-expert use with  
76 limited data availability. In doing so, some of the above documented shortcomings of existing models are  
77 improved. With CFTW, agricultural water assessments can now be performed using local information on  
78 production, climate and management. Growers, companies and non-governmental organizations are thus  
79 no longer dependent on national or regional datasets, own modelling or measurement work to assess their  
80 water use. CFTW provides results on the water footprint (WFP), which describes the water consumed  
81 per unit product as well as irrigation requirements. Furthermore, it provides the possibility to compare  
82 different production sites and systems using the same methodology. Finally, together with the already  
83 existing on-line tool CFT, it enables crop producers and stakeholders to take a more informed and holistic  
84 approach on environmental sustainability in the agricultural sector.

85 In this study we first introduce the existing CFT as the foundation of CFTW (section 2). CFTW is then  
86 presented in detail, describing the model, the database, and the user interface (section 3). To understand  
87 the effect on the accuracy of using global datasets for determining WFPs, the tool is evaluated based  
88 on 16 studies available in the literature in different climatic and soil-plant conditions (section 4). The  
89 study provides also one of the first assessments of different modelled WFPs with observations. Finally,  
90 limitations and future developments are discussed and concluding remarks presented.

## 91 **2 Cool Farm Tool - CFT**

92 The development of the CFT (<https://coolfarmtool.org>) started in 2008 as an on-farm greenhouse gas  
93 (GHG)emission calculator based on a collaboration between the University of Aberdeen, the Sustain-  
94 able Food Lab and Unilever. The GHG tool captures emissions related to crop and livestock production.  
95 Emissions are determined using empirical models and emission factors which consider differences between  
96 production systems, regions and climates (Aryal et al., 2015; Hillier et al., 2011).

97 The interest in the tool from consumer good producers, retailers, non-governmental organisations, fertil-

Table 1: Overview of existing field water assessment tools that deploy the FAO56 approach (Allen et al., 1998). The table provides the level of data integration for climate, soil and crop. Most tools allow the users to update existing soil and crop information.

Name	Source	Climate data	Crop parameters	Soil parameters	Special features
AquaCrop	FAO, Ste-duto et al. (2009)	database with 5000 stations (CLIMWAT)	14 default crops	14 default soil profiles	contains a full crop growth model for yield prediction including different stresses
CRIWAR	Bos et al. (2008)		10 default crops	only needed when determining water requirements	
CROPWAT	FAO, Smith et al. (1992)	database with 5000 stations (CLIMWAT)	36 default crops	3 default soils	
ICARDA Agro-Climate tool	Mauget and De Pauw (2010)	interpolation between 649 climate stations	default crops provided	default soil types provided	only applicable for north-west Africa to central Asia
MABIA-Region	Sahli and Jabloun (2009) & Allani et al. (2012)		> 100 default crops	12 default soil texture classes	GIS based
SALTMED	Ragab (2002)		> 200 default crops	40 default soils	includes advanced soil water model & use of saline water for irrigation
SAPWAT	van Heerden (2008)	database with 5000 stations (CLIMWAT) & South African climate station data	default crops provided	default soils provided	climate station data available for South Africa
SIMDualKc	Rosa et al. (2012)		auxiliary data provided	auxiliary data provided	
SPARE-WATER	Multsch et al. (2013)				GIS based

98 izer producers and small and medium-sized enterprises led to the formation of the Cool Farm Alliance  
99 (CFA) in 2014, which now manages and owns the tool. The CFA currently has over 53 members who are  
100 using and co-developing CFT in collaboration with academics across several research organisations.  
101 The tool was first developed as an Excel spreadsheet and published in 2011 (Hillier et al., 2011). In  
102 2012, CFT on-line was released and has been used by 4900 registered users. Usage requires a one time  
103 registration on <https://coolfarmtool.org> and enables the user to assess up to five crops.  
104 The tool has also been applied in over 30 scientific publications over the last 6 years. The scope of the  
105 different studies ranged from model comparisons (Camargo et al., 2013; Colomb et al., 2013), to product  
106 assessments of, for example wheat, potato and coffee (Aryal et al., 2015; Haverkort et al., 2014; Sapkota  
107 et al., 2014) as well as investigations of mitigation strategies at the global scale (Hillier et al., 2012).  
108 Based on further requests by the different members of the CFA, the tool was extended with the biodiver-  
109 sity module and the water module. The biodiversity module was released in 2016 and is based on the Gaia  
110 biodiversity yardstick (CFA, 2016; CLM, 2017). It provides an evidence-based biodiversity assessment  
111 for the north-west European biome. The water module has been released in 2017 and is described and  
112 assessed in the present study.

### 113 3 Cool Farm Tool Water - CFTW

114 The CFTW is programmed in Python 2.7. It estimates crop water use and the main components of the  
115 soil water balance combining the single crop coefficient approach presented in the "FAO irrigation and  
116 drainage paper No. 56 crop evapotranspiration" (Allen et al., 1998) with global datasets for soil, crops  
117 and climate. Adjustments to crop phenology, soil water balance simulations and management options  
118 have been made to increase accuracy, represent current knowledge or to enhance usability. The adjust-  
119 ments are described in the following section 3.1 and summarised in figure 1. Finally, model and data are  
120 integrated on-line and accessed via a user-friendly interface at <https://coolfarmtool.org> using any internet  
121 browser.

122

#### 123 3.1 Model

##### 124 3.1.1 Actual evapotranspiration $ET_a$

125 The single crop coefficient approach and - thus CFTW - determines actual evapotranspiration  $ET_a$  (mm  
126  $d^{-1}$ ) based on three distinct steps (Allen et al., 1998). First, the reference evapotranspiration  $ET_0$  (mm  
127  $d^{-1}$ ) is estimated based on the Penman-Monteith equation.  $ET_0$  refers to a short well-watered grass with  
128 an assumed crop height of 0.12 m, a fixed surface resistance of  $70 \text{ s m}^{-1}$  and an albedo of 0.23 (Allen  
129 et al., 1998). Based on these values,  $ET_0$  is determined as follows:

$$ET_0 = \frac{0.408 \cdot \Delta \cdot (R_n - G) + \gamma \cdot \frac{900}{T+273} \cdot u_2 \cdot (VPD)}{\Delta + \gamma \cdot (1 + 0.34 \cdot u_2)} \quad (1)$$

130

131 where  $R_n$  is the net radiation at the crop surface ( $\text{M J m}^{-2} \text{ d}^{-1}$ ),  $G$  is soil heat flux density approximated  
132 as  $0 \text{ M J m}^{-2} \text{ d}^{-1}$  on a daily basis,  $T$  and  $u_2$  are the air temperature ( $^{\circ}\text{C}$ ) and wind speed ( $\text{m s}^{-1}$ ) at 2 m

133 height,  $VPD$  is the vapour pressure deficit of the air (kPa),  $\gamma$  is the psychometric constant (k Pa °C<sup>-1</sup>),  
 134  $\lambda$  is the latent heat of vaporization (MJ kg<sup>-1</sup>) and  $\Delta$  is the slope of the saturation vapour pressure vs.  
 135 air temperature curve (kPa °C<sup>-1</sup>).

136 In the second step,  $ET_0$  is corrected based on the single crop coefficient  $K_c$  to determine the potential  
 137 crop evapotranspiration  $ET_c$  (mm d<sup>-1</sup>) as follows:

$$ET_c = ET_0 \cdot K_c \quad (2)$$

138

139 The single crop coefficient  $K_c$  combines evaporation and crop transpiration into a single coefficient and  
 140 scales the  $ET_0$  so that it resembles a specific crop without any limitation of water and nutrients (Allen  
 141 et al., 1998). The  $K_c$  is not constant over the growing season as shown in the crop growth curve in Figure  
 142 2.

143  $K_c$  is based on adjusted empirical values for various crops and linear interpolation between an initial,  
 144 mid-season and end  $K_c$  over the different crop growing periods. The literature values of  $K_c$  are corrected  
 145 to account for local climate, crop, soil and irrigation management conditions based on the approaches  
 146 presented in Allen et al. (1998).  $ET_c$  at the beginning of the growing period is primarily governed by  
 147 evaporation from the top soil. Therefore,  $K_c$  for the initial phase is defined by the wetting frequency  
 148 of the soil surface, soil texture and the irrigation method. The remaining mid-season and late growing  
 149 period are mostly dependent on crop type and are corrected for humidity and crop height. CFTW does  
 150 not correct for wind speed as this is greatly influenced by field location and its surroundings and uses a  
 151 global average of 2 m s<sup>-1</sup> as recommended by FAO56 (Allen et al., 1998). Even though global wind speed  
 152 data is available, they are often not representative at local scale.

153 Finally, the length of the different growing periods are crop specific and scaled to the length of the total  
 154 growing period defined by the user. In the last step,  $ET_c$  is scaled based on a water stress coefficient  $K_s$   
 155 that accounts for the soil water available for transpiration for the plant and for evaporation and limits  
 156  $ET_c$  to actual evapotranspiration  $ET_a$  (mm d<sup>-1</sup>):

$$ET_a = ET_c \cdot K_s \quad (3)$$

157  $K_s$  ranges between 0 and 1 and is defined by root zone depletion  $D_r$ , which is the water lost from the  
 158 total available water to the plant and described in the section 3.1.2.

$$K_s = \frac{TAW - D_r}{(TAW - RAW)} \quad \text{for } D_r > RAW \quad (4)$$

$$K_s = 1 \quad \text{for } D_r \leq RAW$$

159 where  $TAW$  and  $RAW$  are the total and readily available water (mm) respectively.  $TAW$  represents the  
 160 total storage capacity  $(\theta_{FC} - \theta_{WP}) Z_r$ , where  $\theta_{FC}$  and  $\theta_{WP}$  are soil moisture at field capacity and at  
 161 permanent wilting point, respectively and  $Z_r$  is the rooting depth.  $RAW$  represents the part of  $TAW$   
 162 for which plants do not suffer water stress. In contrast to FAO56, where  $Z_r$  is described as constant,  
 163  $Z_r$  grows from an initial depth to the maximum depth over the initial and developing growth stage in

164 CFTW (Figure 2). This is an important adjustment also made by CROPWAT for example as not all  
 165 soil water within the maximum rooting zone is available to the plant from the beginning of the growing  
 166 period and neglecting this may lead to an underestimation of crop water stress (Bos et al., 2008).

### 167 3.1.2 Soil water balance

168 The soil water balance, as expressed in terms of soil water depletion in the root zone  $D_r$  at time  $i$ , is  
 169 defined by a traditional tipping bucket approach (Allen et al., 1998). The bucket size is defined by field  
 170 capacity and permanent wilting point described by the pedo-transfer function in Saxton and Rawls (2006)  
 171 as well as the maximum  $Z_r$  of the specific crop. The general balance can be written as follows:

$$D_{r,i} = D_{r,i-1} - (P + I - RO - Int_I)_i + ET_{a,i} + DP_i - CR_i \pm LF_i \quad (5)$$

172

173 where  $P$  is precipitation,  $I$  is applied irrigation depth,  $CR$  is capillary rise,  $LF$  is the lateral soil water  
 174 fluxes,  $RO$  is runoff from the soil surface from irrigation and precipitation,  $Int_I$  is the interception loss  
 175 from irrigation and  $DP$  is water loss out of the root zone by deep percolation. All the components are  
 176 expressed in term of time step day  $i$  in  $\text{mm d}^{-1}$ .

177 Initial soil water depletion is provided by the user and then simulated daily using the daily water balance.  
 178 As in Allen et al. (1998) and Bos et al. (2008), CFTW assumes that  $LF$  and  $CR$  are negligible and, for  
 179 this reason, not simulated. Therefore, CFTW is currently only applicable when these terms are small  
 180 and do not influence the soil water balance significantly. Precipitation and irrigation are provided by a  
 181 global data base and users, respectively. Net soil water infiltration is defined by net precipitation and  
 182 net irrigation, which considers interception loss, surface run-off, and deep percolation.  $Int_I$  is based on  
 183 Hoyningen-Huene (1983) and Braden (1985) (Kroes et al., 2008).

$$Int_{P+I,i} = a * LAI_i * \left( 1 - \frac{1}{1 + \frac{b * P_{tot,i}}{a * LAI_i}} \right) \quad (6)$$

184 where  $Int_{P+I}$  ( $\text{mm d}^{-1}$ ) is intercepted precipitation and irrigation on day  $i$ ,  $a$  is an empirical coefficient  
 185 of  $0.025 \text{ cm d}^{-1}$ ,  $b$  is the soil cover fraction approximated as  $b = \frac{LAI}{3}$  and  $P_{tot}$  ( $\text{mm d}^{-1}$ ) is the total  
 186 precipitation including above canopy irrigation on day  $i$ .  $LAI$  is derived from global average crop specific  
 187 values (Breuer et al., 2003; Scurlock et al., 2001), which are reached after a linear increase from  $0 \text{ m}^2/\text{m}^2$  to  
 188  $0.1 \text{ m}^2/\text{m}^2$  over the initial and a further linear increase to the maximum average  $LAI$  over the developing  
 189 growth stage similar to the crop coefficient (Figure 2, Table 2).

190 If above canopy irrigation (e.g. sprinkler irrigation) and precipitation occur on the same day, interception  
 191 is attributed based on their relative fractions.

$$Int_{P,i} = Int_{P+I,i} \cdot \frac{P_i}{P_{tot,i}} \quad (7)$$

192

$$Int_{I,i} = Int_{P+I,i} \cdot \frac{I_i}{P_{tot,i}} \quad (8)$$

193 Runoff  $RO_i$  is determined using the approach of Jägermeyr et al. (2015) for the global dynamic vegetation  
 194 model LPJmL.

$$RO_i = 1 - P_{tot,i} \cdot \sqrt{1 - \frac{\theta_i}{\theta_{SAT} - \theta_{WP}}} \quad (9)$$



195 where  $P_{tot}$  refers to the sum of precipitation and all irrigation,  $\theta$  is soil water content,  $\theta_{WP}$  is soil water  
 196 content at the permanent wilting point and  $\theta_{SAT}$  is soil water content at saturation. After larger rain  
 197 or irrigation events, soil water content may exceed field capacity and therefore water holding capacity of  
 198 the soil and trigger deep percolation  $DP$ . FAO56 and CFTW work with the simplified assumption that  
 199 all excess water above  $\theta_{FC}$  drains into deeper soil layers the same day (Allen et al., 1998).

### 200 3.1.3 Effect of management practices

201 The crop water use is controlled by many factors some of which cannot be altered nor managed by the  
 202 farmer. Soil texture and climate including precipitation are defined by the field location. However, water  
 203 usage is in some respects influenced by the farmer, and these are reflected in CFTW.

204 First and foremost the choice of crop has a big influence on total  $ET_a$ . User can select 24 different crops,  
 205 which vary with respect to growing period and length,  $K_c$ , stress tolerance (e.g. via  $RAW$ ), crop height,  
 206 rooting depth as well as  $LAI$ .

207 Organic matter content in the soil is important for determining the total water holding capacity and can  
 208 be influenced by the crop producer for e.g. by reduced tillage or applying organic mulch as described in  
 209 Cannell and Hawes (1994) and Mulumba and Lal (2008), respectively. This is implemented in CFTW by  
 210 using the pedo-transfer function of Saxton and Rawls (2006), where field capacity and permanent wilting  
 211 point is determined based on sand and clay content as well as soil organic matter. A higher organic  
 212 matter content thus may reduce  $DP$  and increase resilience against water stress.

213 More options to impact  $ET_a$  arise when irrigation is applied to the field. The model considers four different  
 214 methods for irrigating: pivot, rain gun, flooding and drip irrigation. The methods vary with respect to  
 215 their application efficiency as the model considers interception loss and runoff, with only infiltrating water  
 216 being utilized by the crop. Irrigation also affects the initial crop factor  $K_{c,ini}$  in two ways; firstly,  $K_{c,ini}$  is  
 217 determined by wetting interval as evaporation requires frequent wetting and, secondly, different irrigation  
 218 practices wet different soil fractions (Allen et al., 1998). A smaller irrigated soil fraction, as for example  
 219 when applying drip irrigation, where only 35% is wetted, implies lower evaporation. The wetted soil  
 220 fraction for flood, pivot and rain gun irrigation, on the other hand, is 100% (Allen et al., 1998).

### 221 3.1.4 Model outputs

222 The model determines the components of the soil water balance as discussed above. These results are  
 223 used to estimate the crop irrigation requirements  $I_{req}$  as follows:

$$I_{req} = \sum(ET_0 \cdot K_c) - \sum P_{net} \quad (10)$$

224  $P_{net}$  is the sum of net precipitation and net irrigation. The tool provides the green and blue WFP  
 225 (Hoekstra et al., 2011). The green water footprint  $WFP_{green}$  reflects the total precipitation water used  
 226 by the production of a crop, whereas the blue water footprint  $WFP_{blue}$  reflects the used surface and  
 227 groundwater via irrigation. Both WFPs are determined in accordance with the water footprint network  
 228 as follows (Hoekstra et al., 2011):

$$WFP_{blue} = \frac{\min(I_{req}, I_{net})}{Y} \quad (11)$$

229 where  $I_{net}$  is the part of the applied irrigation not lost via interception, surface runoff or deep percolation.  
230 The model does not consider losses related to water transport (conveyance efficiency).  $Y$  is the harvested  
231 yield in  $\text{kg ha}^{-1}$ .

$$WFP_{green} = \frac{\sum ET_a}{Y} - WFP_{blue} \quad (12)$$

232  $\sum ET_a$  describes the cumulative  $ET_a$  of the entire growing season in l. The quotient of  $\sum ET_a$  and  $Y$  is  
233 the total WFP of a crop. Water stored in the final harvested product is neglected because this generally  
234 consists of less than 1% of the total WFP and, in fact, is commonly in the order of 0.1% (Hoekstra et al.,  
235 2011).

### 236 3.2 Data

237 Table 2 and figure 1 provide an overview of data requirements. All data which is not required from the  
238 user is stored in a PostgreSQL database. The datasets include the Harmonized World Soil Database  
239 (HWSD), the ERA-Interim climate data, the FAO56 crop and soil parameters as well as a dataset of crop  
240 specific leaf area index (LAI) values.

241 ERA-Interim is a climate reanalysis dataset developed by the European Centre for Medium-Range  
242 Weather Forecasts (ECMWF) and it provides precipitation and meteorological variables for determining  
243 reference  $ET_0$  and  $ET_a$  according to the FAO56 and as described in the previous section (Dee et al.,  
244 2011). The three-hourly values available in the ECMWF database were adjusted for time zone and ag-  
245 gregated to daily values. The database provides climate data since the year 2004 and is updated every  
246 three months.

247 HWSD is the assimilation of multiple national and multinational soil databases (FAO et al., 2010) and  
248 is used to determine soil texture defined by sand, silt and clay content and organic mater content if the  
249 user does not provide this information. The pedo-transfer function of Saxton and Rawls (2006) are used  
250 to estimate field capacity and permanent wilting point based on this information.

251 The model includes crop factors, length of growing stages and other crop parameters for 25 different  
252 annual crops as well as perennial grass (See crop section in table 2). Default values can be derived from  
253 Allen et al. (1998). Average LAI values are primarily based on two publications by Breuer et al. (2003)  
254 and Scurlock et al. (2001).

### 255 3.3 User interface

256 The CFTW user interface is fully integrated into the CFT to avoid redundancies of input variables  
257 between the GHG calculator and the water tool. For example, some inputs, such as crop and growing  
258 area, are required for both metrics of the CFT. Several questions presented to the user, such as intensity  
259 and average temperature are, however, only relevant for the GHG metrics and have no influence on the  
260 water results (see figure 3). The input and output user interfaces are designed to make water assessments  
261 easily accessible via an interface that is quick and self-explanatory as displayed in figure 3 and 4. All  
262 relevant user inputs for the water component are presented in table 2.

263 Input fields required only for CFTW concern field location, growing period, initial soil moisture and  
264 irrigation management and are highlighted using a droplet icon. For all inputs which require a unit

Table 2: Data requirements for the CFTW and data-sources. ✓ indicates that data input is only optional. ✓ shows that data input is mandatory. The column D. or C. indicates if the parameter is a constant (C) for the entire season or varies daily (D). CFTGHG input and CFTW input shows if the variable is new to CFT for the water module or has been part of the GHG model already.

	Variable	D. or C.	CFTGHG input	CFTW input	FAO56	ERA-Interim	HWSD	Others	Comments
crop	field location	C		✓					
	crop type	C	✓						
	crop yield	C	✓						
	growing area	C	✓						
	planting and harvesting date	C		✓					
	length growth stages	C			✓				
	default crop factors	C			✓				
	rooting depth	D			✓				adapted from Smith et al. (1992), see figure 2
	crop height	D			✓				adapted from Smith et al. (1992), see figure 2
	leaf area index	D						✓	adapted from Allen et al. (1998) with data from Breuer et al. (2003) and Scurlock et al. (2001), see figure 2
soil	soil texture	C		✓				✓	
	soil organic matter	C		✓				✓	
	initial soil water content	C		✓					
	clay fraction	C			✓			✓	
	sand fraction	C			✓			✓	
	readily evaporable water	C			✓				
	readily available water	D			✓			✓	
	porosity	C						✓	
climate	min. and max. temperature	D				✓			
	dew point temperature	D				✓			
	net radiation	D				✓			
	surface pressure	D				✓			
	minimum relative humidity	D				✓			
	precipitation	D				✓			
	wind speed	D							global average of 2 m s <sup>-1</sup> used
management	irrigation amount	C	✓						
	period irrigated	C		✓					
	number of irrigation events	C	✓						
	fraction irrigated	C			✓				
	irrigation method	C	✓						

265 the on-line tool provides a selections of units from which to choose. The location is entered via "Farm  
266 settings", where the user can select the field location by providing longitude and latitude or by tagging  
267 the location on a map (Figure 3). Growing and irrigation period are not provided as dates, but as  
268 early, middle or late in a given month and year representing the 5<sup>th</sup>, 15<sup>st</sup> and 25<sup>th</sup> day of each month  
269 respectively. The total irrigation amount is then distributed equally between both dates and the total  
270 number of irrigation events. Initial soil moisture content is a required input and can be entered as high  
271 (soil moisture at field capacity), mid (2/3 of available water capacity filled) and low (1/3 of available water  
272 capacity filled). The use of approximate dates and classes were identified to be a suitable compromise  
273 between the accuracy in the input and model usability since specific values are not always available.  
274 The blue and green WFP and the irrigation requirements based on the assessment of the entire growing  
275 period are displayed in the results section. Furthermore, additional information about the results are  
276 provided after pressing the info icons on the results page.

## 277 4 Assessment of CFTW

278 The assessment of CFTW was done in three ways: First we compare the CFTW estimates with field  
279 observations, which represents a very time consuming, but reliable approach to assess total water foot-  
280 prints. This is done by using different field trials of water footprints in the scientific literature as well as  
281 different eddy covariance measurement sites (see appendix B). Secondly, model results are compared with  
282 available estimates by the water footprint network (WFPN), which are quickly accessible and represent  
283 state level averages. In a final step we analyse the usability of the tool based on feedback we have received  
284 by members of the CFA.

285 The goal of this study is not to analyse the quality of each input dataset individually, but to provide a  
286 functional evaluation of the model results based on all input data used and test if CFTW responds to  
287 differences in management, climate, soil and crop.

### 288 4.1 Experimental tests

#### 289 4.1.1 Case studies

290 CFTW was tested using observations from 16 published crop water productivity studies for potato,  
291 wheat and maize as those represent the most commonly used crops in CFT. Each of the studies explored  
292 between 1 to 18 different trials. The studies were selected to represent different irrigation management  
293 practices, climates, soils and potential yields in order to investigate the response to these important  
294 drivers. The selected studies are presented in table 3. The evaluation is only based on studies that use  
295 site observations as soil water balance measurements, lysimeter studies or eddy covariance measurements.  
296 Modelling studies were not used in order to avoid interdependencies in modelling results. Model runs were  
297 performed using all available information about growing period, irrigation design and soil. In contrast  
298 to the on-line model interface this study uses exact dates for sowing and harvesting as well as beginning  
299 and end of irrigation. Furthermore, the trials also include a fifth irrigation method representing furrow  
300 irrigation with a soil wetting fraction of 50% which is not yet available on-line.

301 Furthermore, state level WFP data published by the WFPN and CFTW WFPs are compared to the

Table 3: The table provides information on the 16 studies used for validating CFTW. Most studies are using the soil water balance method (SWB) to determine  $ET_a$ . Other methods used are lysimeter studies (LM), eddy covariance measurements (EC) and the Bowen ratio (BR).

study	crop	location	country	method	study aim (assessing impact of)
Aksic et al. (2014)	potato	43.3 N, 21.9 E	Serbia	SWB	irrigation amount
Ati et al. (2012)	potato	33.3 N, 44.2 E	Iraq	SWB	irrigation method & amount, fertilizer rates
Bandyopadhyay and Mallick (2003)	wheat	23.0 N, 88.1 E	India	SWB	irrigation amount
Corbeels et al. (1998)	wheat	33.9 N, 5.6 W	Morocco	SWB	irrigation amount, fertilizer rates
Cossani et al. (2012)	wheat	41.2 N, 1.1 E	Spain	SWB	irrigation amount, fertilizer rates
Erdem et al. (2006)	potato	41.0 N, 27.5 E	Turkey	SWB	irrigation method, amount & interval
Fengrui et al. (2000)	potato, maize	35.7 N, 107.9 E	China	SWB	crop rotations
Hernández et al. (2015)	maize	37.8 S, 58.3 W	Argentina	SWB & Micro-LM	irrigation amount, fertilizer rates
Igbadun et al. (2008)	maize	8.6 S, 33.9 E	Tanzania	SWB	irrigation interval & periode
Jabro et al. (2012)	potato	48.2 N, 103.1 W	USA	SWB	irrigation interval
Jia et al. (2014)	wheat	36.2 N, 117.2 E	China	SWB	irrigation amount & method
Kang et al. (2000)	maize	38.0 N, 103.1 E	China	SWB	irrigation amount & method
López-Urrea et al. (2009)	wheat	39.2 N, -2.1 W	Spain	LM	- (only single wheat crop)
Parent and Anctil (2012)	potato	46.8 N, 72.3 W	Canada	EC	- (only single potato crop)
Suyker and Verma (2009)	maize	41.2 N, 96.5 W	USA	EC	irrigation amount
Young et al. (2008)	wheat	31.7 S, 150.5 E	Australia	SWB & BR	-

302 observed total WFPs of the case studies. The WFN values are part of a global modelling study for  
303 various crops using a grid based water balance model also derived from Allen et al. (1998) using global  
304 datasets for crop distribution, precipitation, long-term monthly  $ET_0$  and soil properties (Mekonnen and  
305 Hoekstra, 2011). Soil and climate data used for WFN estimates differ from datasets used for CFTW and  
306 are further described in Mekonnen and Hoekstra (2011). Results have been aggregated for administrative  
307 units and results are representative for the years 1996 until 2005 (Mekonnen and Hoekstra, 2011). The  
308 comparison highlights the potential differences between state level averages and local agricultural practices  
309 for a specific year.

#### 310 4.1.2 Result and discussion of case studies

311 CFTW explained more than 50% of all the variance in all 16 studies of observed  $ET_a$  ( $R^2 = 0.53$ , p-value  
312  $< 0.05$ ). The best results are obtained for potato ( $R^2 = 0.63$ , p-value  $< 0.05$ ), followed by wheat ( $R^2 =$   
313  $0.61$ , p-value  $< 0.05$ ) and maize ( $R^2 = 0.57$ , p-value  $< 0.05$ ). The Root Mean Square Error (RMSE) of  
314 all studies combined is 103 mm and ranges between 28 mm for Aksic et al. (2014) and 190 mm for Ati  
315 et al. (2012) (Figure 5).

316 The median relative error between simulated and observed  $ET_a$  is 1.3% with an inter-quartile range of  
317 -20.2% and 15.5% and thus shows no clear bias towards over- or underestimation of  $ET_a$ . The model  
318 reproduced a significant positive correlation for 10 out of 13 studies with more than two trials (Figure  
319 5). Only results for Corbeels et al. (1998) show a significant negative correlation. This shows that, based  
320 on 13 studies, CFTW correctly identifies water management improvements. The magnitude of change in  
321  $ET_a$  is however underestimated for 8 and overestimated for 2 of the 10 studies - with significant positive  
322 correlation. In cases where underestimation of the change in  $ET_a$  occurs, this may in fact also result from  
323 the method of measurement of  $ET_a$ . Most studies used in this work are based on soil water balances  
324 which tend to underestimate  $ET_a$ , in particular for high precipitation or irrigation (Sadras and Angus,  
325 2006) due to the fact that these studies often neglect runoff and deep percolation.

326 The biggest positive relative error between model results and measurements is from a trial of Cossani  
327 et al. (2012).  $ET_a$  is overestimated by over 54.4% and 65.4 mm. The trial shows a control trial of the  
328 study without any irrigation and only little precipitation which means  $ET_a$  is highly sensitive to initial  
329 soil moisture. CFTW currently permits only three levels of initial soil moisture with the lowest being  
330 one-third of the available water capacity. This may therefore lead to an overestimation of the available  
331 water in the soil, when actual soil moisture is below this at the time of sowing as in this trial.

332 The greatest underestimations occurred at individual trials from Suyker and Verma (2009), Ati et al.  
333 (2012), Young et al. (2008) and Jia et al. (2014). Suyker and Verma (2009) shows a substantial underes-  
334 timation of precipitation in ERA-Interim during summer months in 2003 and 2005, which triggers water  
335 stress and an underestimation of rain-fed trials. Other reasons may also contribute to the discrepancies,  
336 since rainfall is not underestimated in 2001 while simulated  $ET_a$  is 150 mm lower than the observed.  
337 Nevertheless, the rain-fed trial in 2001 shows the smallest error of all trials without irrigation in Suyker  
338 and Verma (2009).

339 Data points from Ati et al. (2012) represent furrow and drip irrigated potato grown between September  
340 and January with different fertilizer application levels.  $ET_a$  from all trials in this study are underes-

Table 4: The table displays the RMSE for estimated WFPs using CFTW or WFN state level values representative for 1996 to 2005.

crop	RMSE for WFN WFP [ $\text{m}^3 \text{ kg}^{-1}$ ]	RMSE for CFTW [ $\text{m}^3 \text{ kg}^{-1}$ ]
potato	0.174	0.048
wheat	1.442	0.404
maize	0.555	0.154
all	0.941	0.264

341 timated by CFTW. The model results do not show crop water stress and therefore indicates that  $ET_c$   
342 is underestimated.  $ET_0$  is high in the beginning, but decreases towards the end of the growing season  
343 when crop factors are higher. Therefore the underestimation is because  $ET_0$  is underestimated during  
344 the winter months or the  $K_c$  in the early month of the growing season is too low, possibly linked to an  
345 underestimation of soil wetting fraction. In addition, different levels of fertilization - and thus different  
346 crop growth curves per crop - are currently not implemented in CFTW which assumes optimal nutrient  
347 levels for all crops. Hence the model results show no variance in  $ET_a$  associated with different fertilization  
348 rates. The strong offset of the Ati et al. (2012)  $ET_a$  estimates results in a mean relative error of -17% for  
349 all potato studies.

350 For Jia et al. (2014) the reasons are more complex and the underestimation of  $ET_a$  cannot be clearly  
351 attributed. The initial crop factor is very low, which leads to a low water use in the beginning, which  
352 again triggers great deep-percolation. Moreover, the results show high crop water stress in April and  
353 June at the end of the growing season.

354 The observed total WFPs of all studies range between  $0.083 \text{ m}^3 \text{ kg}^{-1}$  and  $8.686 \text{ m}^3 \text{ kg}^{-1}$  (Figure 6). The  
355 WFPs above  $1.500 \text{ m}^3 \text{ kg}^{-1}$  belong to trials with little irrigation or precipitation causing low yields and  
356 thus often served as control trials. Only four trials of Corbeels et al. (1998) and Igbadun et al. (2008)  
357 exceed a total WFP of over  $1.500 \text{ m}^3 \text{ kg}^{-1}$ . The underestimation of potato  $ET_a$  also leads to an underes-  
358 timation of potato water footprints. Nevertheless, CFTW accounts for 92.6% to 99.1% of the variance of  
359 all WFPs for the specific crop (potato:  $R^2 = 0.926$ , p-value  $< 0.05$ ; wheat:  $R^2 = 0.991$ , p-value  $< 0.05$ ).

360 In contrast, WFPs estimated based on the WFN state level are much more diverse. Using CFTW with  
361 limited user input and global datasets reduces the RMSE of the WFP by over 70% for all crops in  
362 comparison of WFN estimates (Table 4, Figure 6). This shows the differences between local WFPs and  
363 average state level WFPs presented by WFN, but also the benefit of using field level yield data as well as  
364 more local climate, soil and management information. Therefore, WFN state level WFP data cannot be  
365 used as approximation for individual fields within one state as it does not reveal the variability of WFPs  
366 on state and even individual field level.

367

## 368 4.2 Usability trials

369 The tool was tested with users via presentations to candidate user groups in the form of a webinar, a  
370 workshop and individual trials organized by the CFA between February and June 2017. Here the impor-  
371 tance of science based methods for voluntary assessments in the agricultural and cooperate sector was

372 emphasized in order to assess, improve and communicate the sustainability of crop production as well  
373 as global supply chains. During these trials, it was recognized how CFTW has provided a practical tool  
374 for the assessment of agricultural water use and increased the usability of FAO56 by, (i), limiting the  
375 user input to basic questions that constitute common farmer knowledge and, (ii), integrating a climate,  
376 soil and crop database. Users, however, acknowledged that the use of gridded climate data, default crop  
377 parameters and soil data may not capture the spatial or crop specific variability in those domains. There-  
378 fore, results should be interpreted with caution when these values do not well represent local conditions.  
379 In addition, the tool tried to minimize user input by integrating the GHG and water user interface.  
380 While this reduced the redundancy of questions in CFT significantly, it may also lead to questions when  
381 the input is not explicitly allocated to the GHG or water metrics. Therefore, defining user pathways  
382 depending on user interest is desirable and should be considered in further developments.  
383 CFTW enables users to assess their green and blue WFP considering local meteorology, soil and harvest,  
384 capturing different growing seasons and the annual variability of the weather. Although Hoekstra et al.  
385 (2011) also addresses farmers and gives recommendations on how to reduce the WFP of crop production,  
386 uptake by farmers has been low, whereas a focus on irrigation requirements, water productivity and crop  
387 water stress appears to have greater meaning with farmers.  
388 The discussion with members of the CFA also revealed difference of opinion and uncertainty on how to  
389 best assess the environmental impact of water consumption and how to define reduction targets. While  
390 some showed a strong interest in the WFP, others target a reduction of abstracted blue water, an increase  
391 in irrigation efficiency or avoiding water scarcity and risk. The discussion observed here is also present  
392 in the scientific literature where recommendations vary to the point of the reduction of the WFP to an  
393 approach that focuses on water scarcity (Boulay et al., 2015; Hoekstra et al., 2011; ISO, 2014; Ridoutt  
394 and Pfister, 2010).

395

### 396 **4.3 Limitations and possible improvements**

397 In this study and in trialling of the tool we have identified several specific areas where there is scope for  
398 further development and opportunities for improvement. This could mean an improvement of default  
399 data provided in the tool, enhancing the model itself or an advancement of the user interface.  
400 Pereira et al. (2015) and Allen et al. (1998) emphasize the importance of accurate measurements of mete-  
401 orological variables to reduce the uncertainty for  $ET_0$ . CFTW uses global gridded data to determine  $ET_a$   
402 based on FAO56 and is therefore taking a similar approach as Siebert and Döll (2010) and Mekonnen  
403 and Hoekstra (2011). The average climate in the grid cell may not represent the meteorology at the field  
404 location for various reasons, as for example, topography. In particular ERA Interim precipitation data is  
405 mainly linked to uncertainties due to the spatial variability of rainfall. Furthermore, some studies show  
406 an underestimation of ERA Interim precipitation (de Leeuw et al., 2015; Szczypta et al., 2011), similar  
407 to what was observed for the trials of Suyker and Verma (2009). A refined analysis of using local mete-  
408 orological data versus ERA Interim data with respect to CFTW using 10 eddy covariance measurement  
409 sites confirmed that results only show a small improvement (see appendix B). Furthermore, the outline  
410 of the  $0.75^\circ$  grid cell that is used under "Farm settings" enables the user to assess how representative



411 weather is across this area based on own local knowledge. Still, future versions of the model may consider  
412 using meteorological data with a higher spatial resolutions, such as the newly released ERA5 dataset  
413 (ECMWF, 2018) or allow for the replacement of individual climate variables with local meteorological  
414 data if available.

415 The tool might also make better use of the data available in the HWSO, by displaying the different soil  
416 textures available in the HWSO for a given location and offering the user to select the most representa-  
417 tive soil from these options rather than defaulting to the most abundant soil texture as is done currently.  
418 Moreover, the quality of the HWSO varies strongly across different world regions and countries (Avellan  
419 et al., 2012; FAO et al., 2010). It is possible that alternatives such as the newly available high resolution  
420 soil map SOLIDGRIDS could replace HWSO in the future (Hengl et al., 2017) and address this known  
421 issue. Nevertheless, the use of local soil data whenever available is likely to be the most reliable option.  
422 Furthermore, default crop coefficients, rooting depth, crop height and LAI are currently set internally  
423 and not by the user. Since global averages may not be representative at farm level (Allen et al., 1998),  
424 we propose to improve the on-line tool in the future to enable users to overwrite this crop data where  
425 desirable. Still, CFTW is different to most other models as it determines initial  $K_c$  based on wetting  
426 frequency and soil texture and adjusts the remaining  $K_c$ s automatically and therefore eliminates a great  
427 source of error present in most tools (Pereira et al., 2015).

428 The model itself can be improved by enhancing current model components or increasing its scope. For  
429 example, the water balance estimated in CFTW is similar to CROPWAT (Smith et al., 1992) and does  
430 not account for any impermeable layers or capillary rise from groundwater layers. Raes et al. (2012) have  
431 overcome this limitation in AquaCrop by enhancing the input requirements for the tool. In addition,  
432 the pedo-transfer function used for determining water holding capacity for the soil profile in CFTW was  
433 calibrated for top soils and is applied here for the full rooting depth. This could be replaced by a different  
434 pedo-transfer function to reduce uncertainties for water holding capacity.

435 CFTW represents water stress and assumes that crop growth is not limited by other factors as nutrients,  
436 temperature or salt stress. Inclusion of these features could increase the scope of the tool as has also been  
437 done in AquaCrop (Raes et al., 2012). In addition, more management interventions such as mulching,  
438 contour ploughing, fertilizing or further irrigation methods to reduce runoff and decrease evaporation as  
439 well as transpiration might be included. Future developments can show the benefits of such practices on  
440 water management and therefore would give greater relevance to the tool and encourage adoption and  
441 reporting of these practices. Mulching, for example, is estimated to reduce  $ET_a$  by 10% per area covered  
442 (Allen et al., 1998; Chukalla et al., 2015).

443 Currently the tool only reflects the water use element of the WFP estimation while yield is defined by  
444 the user. Coupling CFTW with a crop growth model could help to show the co-benefits of different  
445 management for water and yield and therefore show the full reduction potential of the WFP.

446 Finally, the WFP has been criticised in the past for not being easily comparable and not reflecting local  
447 water scarcity (Ridoutt and Pfister, 2010). In practice it is more important to manage water efficiently  
448 when a river basin suffers from water scarcity and the WFP does not convey the importance of this  
449 context. The tool (and any reporting of WFP) would therefore benefit from provision of information on  
450 local water scarcity or availability.

451 These examples show how existing model components can be further developed in the future. Most of the  
452 changes discussed above, would imply an adjustment to the user interface and higher user input. Thus, all  
453 changes have to be thoroughly evaluated based on the added accuracy and functionality, while remaining  
454 a user friendly tool, which is easy to use. This can only be done and decided in close collaboration with  
455 targeted user groups, as conducted so far. This also holds true for more precise user input for already  
456 existing input parameters, as for example, irrigation scheduling and initial soil moisture.

457 While we show that CFTW provides reliable estimates based on 16 field studies, further testing is es-  
458 sential to consider a wider range of management interventions, climates, crops and soils. The estimation  
459 and communication of uncertainties within the tool remains an important task in terms of model eval-  
460 uation and usability. Using various environmental conditions and management during testing showed  
461 that CFTW is sensitive to those changes and that ERA Interim is sufficiently accurate. A sensitivity  
462 assessment of CFTW considering the uncertainties for crop, soil and climate input using information  
463 from additional field sites is foreseen.

464

## 465 5 Conclusion

466 The CFTW is to our knowledge the first on-line water tool for farmers, suppliers, NGOs and consumer  
467 goods producers that provides WFP results, irrigation requirements and the main components of soil  
468 water balance using gridded climate data, global soil maps and local management information. It over-  
469 comes some of the main constraints with current models as it provides default input data where users find  
470 provision of such data difficult, uses terminology known to the farmer and has an on-line user interface.  
471 The strong collaboration with the Cool Farm Alliance helped us to shape the tool based on demand and  
472 enabled us to make scientific models and datasets available to end-users.

473 The validation of CFTW using 16 studies for potato, wheat and maize in 12 different countries with a  
474 total of 106 observations showed that the CFTW was effective in modelling  $ET_a$  and total WFP and is  
475 able to indicate the correct direction of change in water use for management interventions or location  
476 changes for most studies investigated.

477 In contrast, the long-term and spatially averaged results provided by the WFN were not able to represent  
478 local conditions. By that, it is shown how CFTW helps crop producers to identify adaptation strategies  
479 relevant for the specific local conditions. Finally, by integrating this water assessment tool with the  
480 already existing on-line CFT developed for the assessment of green-house gases emissions (Hillier et al.,  
481 2011) and the biodiversity module (CFA, 2016), it provides a unique platform to engage farmers and  
482 users towards a holistic assessment of the agricultural sector.

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## 746 Figure Captions

Figure 1: Schematic representation of CFTW model components and related publications. The figure also shows where CFTW makes adjustments to FAO56, by introducing different or new model components. A more detailed visual description of the model is presented in the appendix.

Figure 2: The schematic plot shows the crop phenology in CFTW as represented by the crop growth curve showing the crop coefficient  $K_c$ , rooting depth  $Z_r$  and the leaf area index  $LAI$ .

Figure 3: CFTW on-line input user interface. The figures show part of the user interface for a potato crop grown in England in 2014, which is irrigated between May and early July using a rain-gun system.

Figure 4: CFTW on-line results for a potato crop in England as described in figure 3.

Figure 5: Simulated  $ET_a$  using CFTW versus observed  $ET_a$  for 16 different studies for wheat, potato and maize. The dashed lines indicate an offset between simulated and observed  $ET_a$  of more than 25%.

Figure 6: Comparison of state level total WFP estimates by WFN for 1996 to 2005 and CFTW total WFPs to observed total WFPs of the 16 case studies. Two points are removed from the wheat plot where observed water footprints exceed  $3.000 \text{ m}^3 \text{ kg}^{-1}$ , to enhance visibility of the remaining studies.

## 747 Appendix

### 748 A. Schematic representation of CFTW model

Figure A1: Schematic representation of CFTW model. The dashed boxes represent datasets used in the model and stored in the PostgreSQL database. The boxes represent sub-models or related publications employed in CFTW.

### 749 B. Testing model performance based on climate data input

750 CFTW was further tested for climate data uncertainty. The model is based on the global climate dataset  
751 ERA Interim, which has been evaluated independently in several studies (Gao, 2013; Thiemig et al., 2012;  
752 Zhang et al., 2016). This section provides an analysis of model results dependent in field data input and  
753 ERA Interim input specifically tailored to cropped fields.

754 Field data was used from 10 eddy covariance measurement sites of the AmeriFlux network and European  
755 Fluxes Database Cluster using local information on daily maximum and minimum temperature, net  
756 radiation, wind speed, atmospheric pressure and precipitation if available (Table B1). This data was  
757 used as it provides field relevant meteorological data, agronomic information as well as observed  $ET_a$ .  
758 As field data contained gaps we allowed for an up to 10% gap filling using ERA Interim data, except for  
759 atmospheric pressure where we used ERA Interim if this was not observed at the eddy covariance site.  
760 Furthermore, CFTW required dewpoint temperature, which was not observed at any of the sites and  
761 therefore was replaced with daily minimum temperature as recommend by FAO56 (Allen et al., 1998).  
762 Observed  $ET_a$  was gap filled using linear interpolation and only for a maximum of 5% of the growing  
763 season.

764 Our comparison based on 10 eddy covariance sites and a total of 60 individual observations indicated that  
765 CFTW performs only slightly better when using local meteorological input ( $R^2 = 0.59$ , RMSE = 97mm)  
766 opposed to results based on ERA Interim ( $R^2 = 0.53$ , RMSE = 100mm)(Figure B1). Furthermore, CFTW  
767  $ET_a$  results driven by local and global climate data showed a very good agreement ( $R^2 = 0.78$ )(Figure  
768 B2). However, we also found that CFTW results based on ERA Interim were on average 12.5% lower  
769 compared to model runs based on local data. Still, this was driven by 3 sites (US-Ne1, US-Ne2 and  
770 US-Ne3) located in the same ERA Interim grid cell, due to an average underestimation of precipitation of  
771 41.3% to 56.7% (see also 4.1). The bias was significantly reduced (-4.9%) when these sites were excluded  
772 from the analysis.

773 Therefore, we conclude that ERA Interim is sufficiently accurate for the purpose of CFTW, but may  
774 benefit from improved precipitation input.

Figure B1: CFTW model results for  $ET_a$  based on local (FLUX data) and global (ERA Interim) climate data input compared to observations based on eddy covariance measurements.

Table B1: Eddy covariance sites from AmeriFlux and European Carbon Flux that were used for testing model performance based on climate input data. Sites were used for minimum and maximum temperature, precipitation, net radiation, surface pressure and wind speed to drive CFTW.

site	crops	years	location	country	reference
CH-Oe2	Wheat, Potato, Bar- ley	2005 - 2007	47.3N, 7.7E	Switzerland	Dietiker et al. (2010) and Emmel et al. (2018)
DE-Kli	Maize, Barley, Maize	2007, 2008, 2011, 2012	50.9N, 13.5E	Germany	Prescher et al. (2010)
FR-Gri	Maize, Bar- ley, Wheat, Triticale	2005 - 2011	48.8N, 2.0E	France	Loubet et al. (2011)
IT-BCi	Maize	2004, 2005	40.5N, 15.0E	Italy	Vitale et al. (2007, 2009)
US-ARM	Maize	2004 - 2006, 2009	36.6N, 97.5W	USA	Fischer et al. (2007)
US-Bo1	Maize, Soy- bean	2001 - 2006	40.0N, 88.3W	USA	Meyers (2004) and Wilson and Meyers (2007)
US-IB1	Maize, Soy- bean	2006, 2008	41.9N, 88.2W	USA	Matamala et al. (2008)
US-Ne1	Maize	2002 - 2012	41.2N, 96.5W	USA	Suyker and Verma (2009)
US-Ne2	Maize, Soy- bean	2002, 2004 - 2012	41.2N, 96.5W	USA	Suyker and Verma (2009)
US-Ne3	Maize, Soy- bean	2002, 2004 - 2012	41.2N, 96.4W	USA	Suyker and Verma (2009)

Figure B2: CFTW model results for  $ET_a$  based on local (FLUX data) and global (ERA Interim) climate data input.