Authors' Response

Dear Prof. Harrie-Jan Hendricks Franssen,

Thank you for your suggestions and comments on our manuscript entitled "Benchmark levels for the consumptive water footprint of crop production for different environmental conditions: a case study for winter wheat in China" (ID: hess-2016-47). We appreciate the opportunity to revise the paper. Based on the suggested amendments and comments from the Editor and two Referees, we carefully revised the manuscript. The revised parts are in RED color in the updated manuscript.

In response to Referee#1 comment (1) and Referee#2 comment (2), in Section 1 Introduction, we elaborate on the practical relevance of the current study. When water is scarce, raising production per unit of water (i.e. increasing water productivity in terms of t m^{-3} or reducing the WF in $m^3 t^{-1}$) is a key challenge in order to save water and achieve sustainable water use at catchment level. We will need WF benchmarks as reference of what water use per unit of crop is reasonable. The current paper explores what variables need to be considered when setting WF benchmarks (and finds that climate is the most important factor when setting WF benchmarks). In Section 3.6 we add a reflection on the significance of saving water and WF benchmarking in the wet South China.

In response to Referee #2 comment (1), we add the information in Section 2.2 explaining why we did not distinguish green and blue WF benchmarks in the current study. We do, however, of course, distinguish between green and blue ET in the simulations, as elaborated now in more detail in Section 2.

As suggested by both Referees, in the revised manuscript, we split Section 2 Method and Data into two subparts: 2.1 Estimating consumptive WF of growing a crop and 2.2 Benchmarking consumptive WF of growing a crop. We added detailed information on the AquaCrop model, parameterization, calibration and assumptions in Section 2.1 as suggested by both Referees. In 3.7 Discussion, we add sentences on results validation and the effects on core results and conclusions from the uncertainties in absolute magnitude of simulated WFs.

In response to Referee#1 comment (3) we add in Section 3.7 Discussion a discussion of the need for using different models to identify uncertainties around the WF benchmark values presented.

We thank again all the efforts of Editor and two Referees on our manuscript.

Sincerely,

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Benchmark levels for the consumptive water footprint of crop production for different environmental conditions: a case study for winter wheat in China

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1 Abstract.

2 Meeting growing food demands while simultaneously shrinking the water footprint (WF) of agricultural production is one of 3 the greatest societal challenges. Benchmarks for the WF of crop production can serve as a reference and be helpful in setting WF reduction targets. The consumptive WF of crops, the consumption of rainwater stored in the soil (green WF) and the 4 5 consumption of irrigation water (blue WF) over the crop growing period, varies spatially and temporally depending on 6 environmental factors like climate and soil. The study explores which environmental factors should be distinguished when 7 determining benchmark levels for the consumptive WF of crops. Hereto we determine benchmark levels for the consumptive 8 WF of winter wheat production in China for all separate years in the period 1961-2008, for rain-fed versus irrigated croplands, 9 for wet versus dry years, for warm versus cold years, for four different soil classes and for two different climate zones. We 10 simulate consumptive WFs of winter wheat production with the crop water productivity model AquaCrop at a 5 by 5 arc min 11 resolution, accounting for water stress only. The results show that (i) benchmark levels determined for individual years for the 12 country as a whole remain within a range of $\pm 20\%$ around long-term mean levels over 1961-2008; (ii) the WF benchmarks for 13 irrigated winter wheat are 8-10% larger than those for rain-fed winter wheat; (iii) WF benchmarks for wet years are 1-3% 14 smaller than for dry years, (iv) WF benchmarks for warm years are 7-8% smaller than for cold years, (v) WF benchmarks 15 differ by about 10-12% across different soil texture classes; and (vi) WF benchmarks for the humid zone are 26-31% smaller 16 than for the arid zone, which has relatively higher reference evapotranspiration in general and lower yields in rain-fed fields. 17 We conclude that when determining benchmark levels for the consumptive WF of a crop, it is useful to primarily distinguish 18 between different climate zones. If actual consumptive WFs of winter wheat throughout China were reduced to the benchmark 19 levels set by the best 25% of Chinese winter wheat production (1224 m³ t⁻¹ for arid areas and 841 m³ t⁻¹ for humid areas), the 20 water saving in an average year would be 53% of the current water consumption at winter wheat fields in China. The majority 21 of the yield increase and associated improvement in water productivity can be achieved in southern China.

1 1 Introduction

2 Half of the large river basins in the world face severe blue water scarcity for at least one month a year (Hoekstra et al., 2012). 3 Agriculture is the largest consumer of water in the world and therefore responsible for a large part of the water scarcity in the 4 world. Still, global food demand continues to increase, due to growing populations and changing diets. Meeting growing food 5 demands and simultaneously reducing the water footprint (WF) of agricultural production is therefore one of the greatest 6 societal challenges of our time (Foley et al., 2011; Hoekstra and Wiedmann, 2014). In crop production, individual farmers 7 generally aim to maximize their economic return through raising their productivity per unit of input such as capital, labour, 8 land, and fertilizer. When water is scarce, raising production per unit of water (i.e. increasing water productivity in terms of t 9 m^{-3} or reducing the WF in $m^{3}t^{-1}$) is a key challenge in order to save water and achieve sustainable water use at catchment level. Even when water is not scarce, it makes sense to have a reasonable level of water productivity, i.e. a good amount of crop per 10 11 drop. Farmers, however, generally lack incentives for saving water, since they pay little for their water use compared to other input factors, even under conditions of high water scarcity. In order to provide producers with an incentive to reduce the WF 12 13 of their products to reasonable levels. Hoekstra (2014, 2013) has proposed to develop WF benchmarks, which can be used by 14 governments, farmers and customers (crop traders and retailers) for setting WF reduction targets. Setting WF benchmarks for 15 different products, particularly water-intensive products like crops, is fundamental for wise water allocation and fair sharing 16 of water resources among different sectors and users (Hoekstra, 2013). WF benchmarks of crop production could be global, 17 but would preferably be context-specific, given the fact that the WF of growing a crop varies as a function of environmental factors such as climate and soil (Mekonnen and Hoekstra, 2011; Siebert and Döll, 2010; Tuninetti et al., 2015). 18

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20 The WF of a crop is determined by both environmental conditions (e.g. climate, soil texture, CO_2 concentration in the air) that 21 cannot be controlled by humans and managerial factors (e.g. application of fertilizers and pesticides, irrigation technology and 22 strategy, mulching practice) (Zwart et al., 2010; Mekonnen and Hoekstra, 2011; Brauman et al., 2013). Benchmarks for the 23 WF of growing a crop can, for example, be set by looking at what WF level is not exceeded by the best 20-25% of the total 24 production in an area. Alternatively, benchmarks can be determined by estimating the WF associated with best-available 25 technology and management practice (Hoekstra, 2014, 2013). Mekonnen and Hoekstra (2014) followed the first approach and 26 developed global benchmarks for both the consumptive (green plus blue) WF and the degradative (grey) WF for a large number 27 of crops, based on estimated WF values for 1996-2005 at a spatial resolution of 5 by 5 arc minute. Chukalla et al. (2015) 28 followed the second approach and explored reduction potentials of consumptive WFs for a few crops by applying different 29 alternative irrigation techniques and strategies and different alternative mulching practices. They found that the highest 30 reduction (~29%) in the consumptive WF of a crop could be achieved when applying drip/subsurface drip irrigation in 31 combination with deficit irrigation and synthetic mulching.

1 Research in developing benchmark levels for the consumptive WF of crop production is still in its infancy. An important 2 question that has been insufficiently addressed is which environmental factors should play a role when developing WF 3 benchmarks. It is nice to have one global benchmark for the consumptive WF per crop, as a global reference, like the ones developed by Mekonnen and Hoekstra (2014), but it remains unclear whether it is reasonable to expect the same water 4 5 productivity under different environmental conditions. In their global analysis, Mekonnen and Hoekstra (2014) found that a 6 crop in a temperate climate generally has a smaller WF than the same crop in a tropical climate, but this can still be due to 7 other factors (e.g. better management practices in temperate climates), so that this is not a sufficient finding to diversify 8 benchmark levels based on the distinction between temperate and tropical. Besides, even though Mekonnen and Hoekstra 9 (2014) found a difference between different climates, for each crop considered it was found that the 10% best global production 10 (e.g. with smallest WFs) were always at least partly in the tropics as well. In other words, a WF benchmark developed in the 11 temperate part of the world still offers a reference value that can be achieved in the tropics as well. Next to climate also soil 12 affects evapotranspiration and yield and thus the WF of a crop. Tolk and Howell (2012), for example, analyse the variation of consumptive WFs of sunflower in relation to different types of soils. There has not been yet, though, a systematic study looking 13 14 at how environmental factors influence the consumptive WFs of crops and to which extent it makes sense to diversify WF benchmark levels based on specific environmental factors. 15

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17 The current study aims to contribute to this discussion through an explorative study for winter wheat in China. We explore 18 which environmental factors should be distinguished when determining benchmark levels for the consumptive WF of crops. 19 We subsequently determine benchmark levels for the consumptive WF of winter wheat production in China for all separate 20 years in the period 1961-2008, for rain-fed versus irrigated croplands, for wet versus dry years, for warm versus cold years, 21 for four different soil classes and for two different climate zones. Winter wheat in China accounts for 95% of total wheat 22 production in China, which is the world biggest wheat producer (FAO, 2014). Winter wheat covers 96% of China's harvested 23 wheat area and occurs across China's different climate zones (NBSC, 2013). In order to avoid interference from managerial factors that cause differences in evapotranspiration and yield, we simulate WFs by means of FAO's water productivity model 24 25 AquaCrop (Hsiao et al., 2009; Raes et al., 2009; Steduto et al., 2009), at a resolution of 5 by 5 arc minute, considering only 26 water stress and not taking into account other stresses such as from soil fertility, salinity, frost, or pest and diseases.

27 2 Method and data

28 2.1 Estimating consumptive WF of growing a crop

The consumptive (green and blue) WF of growing a crop ($m^3 t^{-1}$) equals the total actual evapotranspiration (ET, $m^3 ha^{-1}$) over the cropping period divided by the crop yield (Y, t ha⁻¹). In current study, the ET and Y of growing winter wheat in China were simulated at daily basis, at 5 by 5 arc min resolution, with FAO's crop water productivity model AquaCrop (Hsiao et al., 2009;

32 Raes et al., 2009; Steduto et al., 2009), run for the whole period 1961-2008. Compared to other crop growth models, AquaCrop

has a significantly smaller number of parameters and better balances between simplicity, accuracy and robustness (Steduto et al., 2007; Confalonieri et al., 2016). The model performance on simulating crop growth and water use has been well tested for
a variety of crop types under diverse environmental conditions (e.g. Kumar et al., 2014; Jin et al., 2014; Abedinpour et al., 2012; Mkhabela and Bullock, 2012; Andarzian et al., 2011; Stricevic et al., 2011; Heng et al., 2009; Farahani et al., 2009;
García-vila et al., 2009). AquaCrop has been applied in WF accounting at field (Chukalla et al., 2015), river basin (Zhuo et al., 2016a) and national level (Zhuo et al., 2016b) at high spatial resolution.

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8 AquaCrop simulates water-driven crop water productivity with a dynamic daily soil water balance:

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10
$$S_{[t]} = S_{[t-1]} + PR_{[t]} + IRR_{[t]} + CR_{[t]} - ET_{[t]} - RO_{[t]} - DP_{[t]}$$
 (1)

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where $S_{[t]}$ (mm) refers to the soil water content at the end of day t, $PR_{[t]}$ (mm) the precipitation on day t, $IRR_{[t]}$ (mm) the rigidion water applied on day t, $CR_{[t]}$ (mm) the capillary rise from groundwater, $RO_{[t]}$ (mm) daily surface runoff and $DP_{[t]}$ (mm) deep percolation. $CR_{[t]}$ is assumed to be zero because the ground water depth is considered to be much larger than 1m (Allen et al., 1998).

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The green and blue WFs are determined by green and blue ET over the cropping period, respectively, divided by Y. Following
Chukalla et al. (2015) and Zhuo et al. (2016a, b), the daily green and blue ET (mm) were separated by tracking the daily
incoming and outgoing green and blue water fluxes at the boundaries of the root zone:

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$$\begin{cases} S_{\text{green}[t]} = S_{\text{green}[t-1]} + \left(PR_{[t]} + IRR_{[t]} - RO_{[t]}\right) \times \frac{PR_{[t]}}{\left(PR_{[t]} + IRR_{[t]}\right)} - \left(DP_{[t]} + ET_{[t]}\right) \times \frac{S_{\text{green}[t-1]}}{S_{[t-1]}} \\ S_{\text{blue}[t]} = S_{\text{blue}[t-1]} + \left(PR_{[t]} + IRR_{[t]} - RO_{[t]}\right) \times \frac{IRR_{[t]}}{\left(PR_{[t]} + IRR_{[t]}\right)} - \left(DP_{[t]} + ET_{[t]}\right) \times \frac{S_{\text{blue}[t-1]}}{S_{[t-1]}} \end{cases}$$
(2)

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where S_{green} and S_{blue} refer to the green and blue soil water content, respectively. The initial soil water moisture at the start of the growing period is assumed as green water. The contribution of precipitation (green water) and irrigation (blue water) to surface runoff was calculated based on the respective magnitudes of precipitation and irrigation to the total green plus blue water inflow. The green and blue components in DP and ET were calculated per day based on the fractions of green and blue water in the total soil water content at the end of the previous day.

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29 Y was determined by multiplying the above-ground biomass (B) and the harvest index (HI, %). HI was adjusted to water and

30 temperature stress depending on timing and extent of the stress by an adjustment factor (f_H) from the reference harvest index

31 (HI₀) (Raes et al., 2011):

- 1
- $2 \quad HI = f_{HI} \times HI_0 \qquad (3)$
- 3

4 Only water stress is considered in modelling, which is determined by the water availability in the root zone, thus leaving out 5 the effects of non-environmental factors (e.g. technology, fertilization) on crop growth. For irrigated fields, we assume that 6 the applied irrigation volumes are equal to the net irrigation requirement. We used the same input crop parameters, including 7 a fixed crop calendar, reference harvested index and maximum root depth as calibrated for China's winter wheat, as Zhuo et 8 al. (2016b). We simulated winter wheat production per grid cell over the years based on the irrigated and rain-fed harvested 9 areas of around the year 2000, as obtained from Portmann et al. (2010) (Fig. 1) in order to avoid in the simulations the effects 10 of changes in where how much wheat is grown.

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Data on monthly precipitation, reference evapotranspiration (ET_0) and temperature at 30 arc min resolution were taken from the CRU-TS 3.10 dataset (Harris et al., 2014). Soil texture data were obtained from Dijkshoorn et al. (2008). For hydraulic characteristics for each type of soil, the indicative values provided by AquaCrop were used. Data on total soil water capacity were obtained from Batjes (2012).

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17 2.2 Benchmarking consumptive WF of growing a crop

Following Mekonnen and Hoekstra (2014), benchmark levels for the consumptive WF of crop production were determined by ranking the grid-level WF values from the smallest to the largest against the corresponding cumulative percentage of total crop production. As in the earlier study, we did not distinguish between green and blue WF benchmarks for two reasons. Firstly, the ratio of green to blue WF of a crop heavily depends on local green water resources availability, which is defined by the climate of certain time in a certain location. Location-specific blue WF benchmarks can be developed as a function of the overall consumptive WF benchmarks and local green water availability (Mekonnen and Hoekstra, 2014). Secondly, the purpose of the current study is to find out to which environmental factor the consumptive WF benchmark is most sensitive.

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In order to analyse differences in consumptive WFs in relatively dry versus relatively wet years, we evenly group the fortyeight considered years (1961-2008) into relative dry, average and relatively wet years. We ranked the years based on the annual precipitation over the cropping area of winter wheat in China (Fig. 2a) and classified the sixteen years with the lowest precipitation into the group of dry years and the sixteen years with the highest precipitation into the group of wet years, with the other sixteen years remaining for the group of average years. The average annual precipitation levels of the relatively dry, average and relatively wet years are 760, 799 and 850 mm y⁻¹, respectively.

We also grouped the years considered into relatively cold, average and relatively warm years based on annual mean temperature (Fig. 2b) and into years with relatively low, average and high ET_0 (Fig. 2c). The average annual mean temperatures of the relative cold, average and warm years are 10.7, 11.2 and 11.8 °C, respectively. The average annual ET_0 values in the three categories of years are 874, 896 and 927 mm y⁻¹.

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6 For determining WF benchmarks for different soil texture classes, the soil types in the USDA (U.S. Department of Agriculture) 7 soil texture triangles were grouped into four soil classes (Raes et al., 2011): sandy soils, loamy soils, sandy clayey soils, and 8 silty clayey soils. Each soil class has different ranges of field capacity, permanent wilting point and saturated water content 9 (Table 1). The difference between soil water content and permanent wilting point defines the total available soil water content 10 in the root zone. Given certain soil water content, a soil with a higher field capacity has less deep percolation. With the same 11 water input from precipitation or irrigation and the same soil water content, soils with a smaller saturated soil water content will generate more surface runoff (Raes et al., 2011). Figure 3 shows the spatial distribution of the four soil classes across 12 mainland China. 13

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For determining WF benchmarks for different climate zones, we classify climate based on UNEP's aridity index (AI) (Middleton and Thomas, 1997, 1992). The AI is an indicator of dryness, defined as the ratio of precipitation to reference evapotranspiration, with five levels of aridity: hyper-arid (AI< 0.05), arid (0.05 < AI < 0.2), semi-arid (0.2 < AI < 0.5), dry sub humid (0.5 < AI < 0.65), and humid (AI > 0.65). To determine the geographic spread of the five climate zones in China we used the data on annual precipitation and ET₀ averaged over the period 1961-2008 at 30 by 30 arc min resolution (Harris et al., 2014) (Fig. 4). In the current study, we group the five climate zones into two broad zones: the arid-semi-arid (Arid) zone (AI < 0.5) and the humid-semi-humid (Humid) zone (AI > 0.5).

22 3 Result

3.1 Benchmark levels for the consumptive WF as determined for different years and for rain-fed and irrigated croplands separately

25 We calculated the benchmark levels at different production percentiles for the consumptive WF of winter wheat $(m^3 t^{-1})$ for the 26 country as a whole, year by year, for the period 1961-2008. The results are summarized in Fig. 5. The benchmarks, determined per year and per production percentile, generally vary within $\pm 20\%$ of the long-term mean value over the period 1961-2008. 27 28 We find that the best 10% of winter wheat production in China (with smallest WFs) has a maximum long-term average consumptive WF of 777 m³ t⁻¹, which is larger than the maximum consumptive WF of the best 10% of wheat production 29 globally (592 m³ t⁻¹) that was reported by Mekonnen and Hoekstra (2014). We note here that the figures are not fully 30 comparable, because Mekonnen and Hoekstra (2014) consider total wheat (both spring and winter wheat), use another model 31 32 and consider another period. We find that the best 20% of winter wheat production in China has a maximum long-term average 1 consumptive WF of 825 m³ t⁻¹, which is *smaller* than the reported maximum consumptive WF of the best 20% of wheat 2 production globally (992 m³ t⁻¹). Finally, we find that the best 25% of winter wheat production in China has a maximum long-3 term average consumptive WF of 849 m³ t⁻¹, which is again *smaller* than the maximum consumptive WF of the best 25% of 4 wheat production globally (1069 m³ t⁻¹).

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6 The national average consumptive WF of rain-fed winter wheat $(1120 \text{ m}^3 \text{ t}^{-1})$ is larger than the national average consumptive 7 WF of irrigated winter wheat $(1075 \text{ m}^3 \text{ t}^{-1})$. However, the benchmark levels determined by the best 10%, 20% and 25% of 8 production for rain-fed winter wheat are lower than for irrigated winter wheat. The reason is that the yields in rain-fed 9 production are generally higher than the yields in irrigated production at the same benchmark percentile. The highest rain-fed 10 yields occur in the southern wet area with sufficient precipitation over the cropping period, so that little water stress results in 11 high rain-fed yields. The WF benchmarks for irrigated winter wheat are 8% (for the 10th production percentile) to 10% (for the 25th production percentile) higher than for rain-fed winter wheat.

13 3.2 Benchmark levels for the consumptive WF for dry versus wet years

In a relatively dry or wet year, when considering winter wheat areas in China as a whole, we do not find typically different consumptive WFs in winter wheat production (Table 2). The WF benchmarks are consistently higher in dry than in wet years (1-3%), but the differences between benchmark levels for the consumptive WF for dry versus wet years are small compared to the variations within the dry and wet year categories (±11-14%).

18 **3.3 Benchmark levels for the consumptive WF for warm versus cold years**

Overall, considering irrigated and rain-fed croplands together, WF benchmarks for relatively warm years are 7-8% smaller than for relatively cold years, which is not much when seen in the context of fluctuations in the WFs within the three temperature categories (Table 3). In irrigated areas, WF benchmarks for warm years are 11% smaller, on average, than for cold years. In rain-fed areas, WF benchmarks for warm years are smaller than for cold years as well, but WF benchmarks in average years are not in between the WF benchmarks found for cold and warm years but higher than both. The lower values in cold years relate to lower ET, while the lower values in warm years relate to higher yields.

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The findings when considering different ET_0 classes are similar when looking at the different temperature classes (Table 4). Overall, considering irrigated and rain-fed croplands together, WF benchmarks for years with high ET_0 are on average 5% smaller than for years with average ET_0 and only 2% smaller than for years with low ET_0 . Again, differences between consumptive WFs for years with relatively low or high ET_0 are small when seen in the context of fluctuations in the WFs within the three ET_0 categories (±3-6%).

1 3.4 Benchmark levels for the consumptive WF for different soil classes

2 Tables 5 shows the consumptive WFs of winter wheat at different production percentiles in four soil classes in China. The 3 simulated winter wheat production in sandy clayey soils accounts for 60% of national total, followed by the production in 4 sandy soils (24%), silty clayey soils (8%) and loamy soils (8%) in average over the studied period. No consistent trends can 5 be observed when we compare the benchmarks across the different soil classes. Overall, when we take irrigated and rain-fed 6 fields together, the WF benchmarks for sandy soils are 10-12% lower than the WF benchmarks for loamy soils. More 7 specifically, we find that the WF benchmarks for irrigated winter wheat in sandy soils are about 15% smaller than the WF 8 benchmarks for the other three soil classes, due to relatively low ET. Without water stress, as is the case in the irrigated 9 croplands, soil evaporation from sandy soils is less than from the other soil types because of the fast percolation of water below 10 the root zone in the sandy soils, causing lower ET over the cropping period (Asseng et al., 2001). At rain-fed fields with limited 11 water availability, crop yields are mainly affected by the soil water holding capacity. Therefore, consumptive WFs in sandy 12 soils are larger than in the other three soils, due to the smaller crop yield in case of poorer water holding capacity. The observed 13 differences in WFs of winter wheat in different soil classes agree with the experimental observations by Tolk and Howell 14 (2012) for the case of irrigated sunflower in a semiarid environment as well as with the fieldwork-based simulations by Asseng 15 et al. (2001) for irrigated and rain-fed wheat in the Mediterranean climatic region of Western Australia.

16 **3.5 Benchmark levels for the consumptive WF for different climate zones**

Consumptive WFs of winter wheat at different production percentiles in arid and humid zones in China are shown in Table 6. Significant differences between the benchmarks for different climate zones can be observed. Overall, considering irrigated and rain-fed croplands together, WF benchmarks for the humid zone are 26% (for the 10th production percentile) to 31% (for the 25th production percentile) smaller than for the arid zone. The WF benchmarks for winter wheat in China as a whole (when we take the arid and humid zones together) are close to the benchmarks for the humid zone, caused by the fact that most (96% in average over the study period) of the simulated winter wheat production in China occurs in the humid zone.

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In the irrigated areas, WF benchmarks for the humid zone are 26-30% smaller than for the arid zone; in the rain-fed areas, they are 29-43% smaller. The relatively large WFs in rain-fed fields in the arid zone logically follow from the water stress and resultant low yields. For the irrigated fields, the larger WFs in the arid zone are caused by the relatively high ET_0 and ET. The results confirm the findings from previous studies that the WF of crops, especially rain-fed crops, is negatively correlated with precipitation and positively correlated with ET_0 (Zwart et al., 2010;Zhuo et al., 2014). The differences between the WF benchmarks for irrigated and rain-fed winter wheat are 7-9% in the humid zone and 3-11% in the arid zone.

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Figure 6 shows, for both the humid and arid part of China, for the various winter wheat production areas whether they contribute to the best 10% of national winter wheat production in that climate zone (in the sense of having smallest WFs), to

1 the next best 10%, to the best 5% after that, or to the worst 75% (with WFs beyond the 25th percentile benchmark). Within the 2 arid zone, consumptive WFs below the 25th percentile benchmark level were mostly located in Xinjiang province, with 3 relatively high irrigation density (~98% of the harvested area). In the humid zone, consumptive WFs below the 25th percentile 4 benchmark level were gathered in the southwest, where ET_0 is smaller than in other places (Fig. 4b).

5 **3.6** Water saving potential by reducing WFs to selected benchmark levels

6 The WF benchmarks for different climate zones differ much more significantly (26-31%) than for different soils (10-12%). WF benchmarks differ even less if we compare irrigated versus rain-fed fields (8-10%), warm versus cold years (7-8%), or 7 8 wet versus dry years (1-3%). Therefore, when determining benchmark levels for the consumptive WF of a crop, it seems most 9 useful to primarily distinguish between different climate zones, at least in the case of winter wheat in China. In this section, 10 we analyse the potential water saving if actual consumptive WFs of winter wheat throughout China were reduced to the climate-specific benchmark levels set by the best 10% of Chinese winter wheat production (1042 m³ t⁻¹ for arid areas and 776 11 $m^3 t^{-1}$ for humid areas), the best 20% of Chinese winter wheat production (1170 m³ t⁻¹ for arid areas and 819 m³ t⁻¹ for humid 12 areas), or the best 25% of Chinese winter wheat production (1224 $m^3 t^{-1}$ for arid areas and 841 $m^3 t^{-1}$ for humid areas). 13

14 Taking the estimated actual consumptive WFs of winter wheat in 2005, an average climatic year, as calibrated by the provincial 15 statistics on yield of winter wheat (NBSC, 2013), we find that consumptive WFs in 75% of the planted grids in arid zones and 16 in 96% of the planted grids in humid zones are over the 25th percentile benchmarks. This is largely due to low actual versus 17 potential yields. Figure 7 shows differences between actual provincial yields of winter wheat and the simulated yield potentials 18 from the current study (assuming no crops stresses except water stress in rain-fed areas). The largest yield gaps occur in the 19 southern provinces in the humid zone. The largest yield gap was observed in Fujian province. South China has 81% of national 20 blue water resources (Jiang et al., 2015). However, the risk of water shortage is increasing in the wet South with the operation 21 of the South-to-North Water Transfer Project and the increasing competition for water resources between different sectors 22 (Barnnet et al., 2015). Therefore, water saving and benchmarking WF for the South China are as equally important as for the 23 drier North.

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Table 7 shows the (green plus blue) water saving that would be achieved if actual consumptive WFs of winter wheat everywhere in China were reduced to the climate-differentiated WF benchmark levels set by the 10th, 20th and 25th percentiles of production, in an average year (2005). We find that if in both the arid and humid zone the actual consumptive WFs were reduced to the respective 25^{th} percentile benchmark level, the water saving in an average year would be 53% of the current water consumption at winter wheat fields in China, which is 201 billion m³ y⁻¹ in absolute terms. We further find that the water saving potential in the arid zone is substantially higher than in the humid zone.

1 3.7 Discussion

The consumptive WF of a crop in $m^3 t^{-1}$ most strongly depends on the crop yield in t ha⁻¹ and much less on the 2 3 evapotranspiration from the crop over the growing period in m³ ha⁻¹ (Tuninetti et al., 2015;Mekonnen and Hoekstra, 2011). 4 The simulated consumptive WFs of winter wheat in China have been based on modelling under a hypothetical condition 5 without effects of managerial factors on crop growth. For evaluating our simulations of crop growth, we compared the 6 simulated averaged yields of winter wheat of Chinese provinces for 1961-1990 to the corresponding agro-climatic attainable 7 yields at different agricultural input levels in the GAEZ database (FAO/IIASA, 2011) (Fig. 8). The GAEZ agro-climatic 8 attainable yields account for different levels of yield constraints from four factors in addition to water stress: (i) pest, diseases 9 and weed damages on plant growth, (ii) direct and indirect climatic damages on quality of produce, (iii) efficiency of farming 10 operations, and (iv) frost hazards. Current simulated yields of irrigated winter wheat are closest to the agro-climatically 11 attainable yields with intermediate input levels and the yields of rain-fed winter wheat are closest to the agro-climatically attainable yields with high input levels. The simulated national average yield in the current study (6.5 t ha⁻¹) is 23% higher 12 13 than the attainable wheat yield for China in the year 2000 (5.3 t ha^{-1}) estimated by Mueller et al. (2012).

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15 The study shows that climate is the primary factor to be considered when setting consumptive WF benchmarks. This finding 16 is probably little sensitive to the model used; the precise WF benchmark figures found per climate zone, however, will be more 17 sensitive to the model used. Subsequent studies, comparing WF benchmark estimates per climate zone using different models, 18 are necessary to quantify the uncertainty in the WF benchmarks presented in this study.

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Further research could also explore whether crop varieties used should play a role when developing WF benchmarks, given the fact that some crop varieties may inherently be more productive than others. On the other hand, one could also consider that choosing a productive crop variety is part of the managerial choices. Since crop variety is not a given environmental condition but a choice, one could argue that accepting a less strict WF reference level for a less productive crop variety cannot be justified.

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An important remaining research question is also how combinations of specific techniques and practices can actually lead to the WF reductions that will be necessary in different locations if Chinese government would adopt certain WF benchmarks as targets to achieve greater water productivity. Suppose, for example, that two WF benchmarks for winter wheat were adopted in China: 1224 m³ t⁻¹ for arid areas and 841 m³ t⁻¹ for humid areas. Although the simulations suggest that these levels are feasible throughout the arid and humid zone, respectively, whatever is the soil, whether fields are rain-fed or irrigated, whether it is a cold or warm year, and whether it is a dry or wet year, in some places it will be harder and more would need to be done than in other places.

We studied benchmarks for combined green and blue WFs and did not look at each colour separately. For rain-fed lands, the benchmark levels presented in this study are obviously green WF benchmarks. For irrigated lands, the presented benchmark levels for overall consumptive WFs would need further specification into green and blue. Further research would need to be done to translate a certain benchmark level for the overall consumptive WF of a crop into a specific blue WF benchmark level per specific location as a function of the amount of rain per location, recognizing that the blue ratio in the WF will need to be larger if less green water is available.

7 4 Conclusions

8 Based on the case of winter wheat in China we find that (i) benchmark levels for the consumptive WF determined for individual 9 years for the country as a whole remain within a range of $\pm 20\%$ around long-term mean levels over 1961-2008; (ii) the WF 10 benchmarks for irrigated winter wheat are 8-10% larger than those for rain-fed winter wheat; (iii) WF benchmarks for wet 11 years are on average 1-3% smaller than for dry years, (iv) WF benchmarks for warm years are on average 7-8% smaller than 12 for cold years, (v) WF benchmarks differ by about 10-12% across different soil texture classes; and (vi) WF benchmarks for 13 the humid zone are 26-31% smaller than for the arid zone, which has relatively higher ET_0 in general and lower yields in rain-14 fed fields. Therefore, we conclude that when determining benchmark levels for the consumptive WF of a crop, it is useful to 15 primarily distinguish between different climate zones. We estimated that when in both the arid and humid zone the actual consumptive WFs are reduced to climate-specific benchmark levels set by the 25th percentile of production, the water saving 16 17 in an average year would be 53% of the current water consumption at winter wheat fields in China, with greatest relative 18 savings in the arid zone.

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21 Author contribution

A. Y., L. Z. and M. M. M. designed the study. L. Z. carried it out. L. Z. prepared the manuscript with contributions from all
 co-authors.

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44 Tables

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Table 1. Soil classes.

		Soil water content (vol %)			
Soil class	Soil types	Field capacity	Permanent wilting point	Saturation	
Sandy	Sand, loamy sand, sandy loam	9 - 28	4 - 15	32 - 51	
Loamy	Loam, silt loam, silt	23 - 42	6 - 20	42 - 55	
Sandy clayey	Sandy clay, sandy clay loam, clay loam	25 - 45	16 - 34	40 - 53	
Silty clayey	Silty clay loam, silty clay, clay	40 - 58	20 - 42	49 - 58	

Source: Raes et al. (2011).

		Consumptive WF (m ³ t ⁻¹) at different production percentiles*			
Crop		10th	20th	25th	Average
	Dry years	787±69	837±70	858±71	1103±82
Winter wheat	Average years	763±107	826±72	849±74	1073±97
	Wet years	770±68	813±60	838±50	1048 ± 77
T • 4 1 • 4	Dry years	822±118	862±110	876±112	1095±110
Irrigated winter wheat	Average years	814±97	856±97	881±98	1078±93
	Wet years	799±97	850±100	870±96	1052±96
	Dry years	757±44	802±57	812±56	1121±97
Rain-fed winter wheat	Average years	736±62	771±70	783±70	1074±133
	Wet years	755±96	784±103	794±104	1164±561

Table 2. Benchmark levels for the consumptive water footprint (WF) benchmarks $(m^3 t^{-1})$ of winter wheat for relative dry, average and wet years in China.

		Consumptive WF $(m^3 t^{-1})$ at different production percentiles*			
Crop		10th	20th	25th	Average
	Cold years	795±101	848±63	870±67	1103±96
Winter wheat	Average years	794±79	840±66	864±58	1087 ± 82
	Warm years	732±42	788±58	811±57	1033±70
	Cold years	862±86	902±87	924±87	1121±86
Irrigated winter wheat	Average years	810±107	863±102	878±96	1083±93
wiicat	Warm years	763±96	804±93	824±96	1022±98
	Cold years	760±59	791±68	798±69	1088±144
Rain-fed winter wheat	Average years	772±95	821±99	831±100	1218±553
	Warm years	716±31	744±40	761±44	1053±63

Table 3. National consumptive water footprint (WF) benchmarks $(m^3 t^{-1})$ of winter wheat for relative cold, warm and average years in China.

		Consumptive WF $(m^3 t^{-1})$ at different production percentiles*			
Crop		10th	20th	25th	Average
	Low-ET ₀ years	774±99	822±64	841±62	1065±82
Winter wheat	Average years	806±80	846±73	866±76	1095 ± 107
	High-ET ₀ years	741±51	808±62	839±58	1065 ± 70
	Low-ET ₀ years	831±111	874±108	892±106	1089±98
Irrigated winter wheat	Average years	820±105	868±96	887±96	1073±103
	High-ET ₀ years	784±93	827±97	847±97	1064±102
Rain-fed winter wheat	Low-ET ₀ years	749±55	774±56	781±54	1038±100
	Average years	784±90	828±98	841±98	1249±550
	High-ET ₀ years	716±72	755±59	767±58	1072 ± 78

Table 4. National consumptive water footprint (WF) benchmarks $(m^3 t^{-1})$ of winter wheat for relative low-, high- and average-ET₀ years in China.

		Consumptive WF (m ³ t ⁻¹) at different production percentiles*			
Crop	Soil class	10th	20th	25th	Average
	Sandy	748±143	814±115	834±116	1017±125
Winter wheat	Loamy	846±53	912±77	928±73	1108 ± 74
whiter wheat	Sandy clayey	788±76	848±61	881±66	1071±48
	Silty clayey	822±48	895±43	912±46	963±22
	Sandy	767±158	782±177	846±128	1000±126
Imigated winter wheat	Loamy	931±91	937±93	996±70	1189±107
Irrigated winter wheat	Sandy clayey	879±98	932±98	969±102	1164 ± 100
	Silty clayey	920±68	942±72	958±66	1070±52
	Sandy	785±58	834±88	850±96	1151±272
Rain-fed winter	Loamy	757±77	822±73	843±73	1040±160
wheat	Sandy clayey	764±66	799±68	818±70	1096±129
	Silty clayey	769±62	814±60	837±60	931±103

Table 5. Benchmark levels for the consumptive water footprint (WF) $(m^3 t^{-1})$ of winter wheat for different soil classes in China.

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		Consumptive WF (m ³ t ⁻¹) at different production percentile*			
Crop	Climate zones	10th	20th	25th	Average
	Arid	1042±100	1170±130	1224±125	1757±200
Winter wheat	Humid	776±70	819±66	841±66	1044±83
	Overall	777±72	825±67	849±65	1075±87
	Arid	1088±66	1205±73	1245±84	1399±163
Irrigated winter wheat	Humid	807±104	853±100	872±99	1055±97
	Overall	812±103	856±100	875±100	1075±99
	Arid	1058±310	1311±406	1399±415	2919±1004
Rain-fed winter wheat	Humid	749±70	784±78	795±79	1076±338
	Overall	750±70	785±78	796±78	1120±332

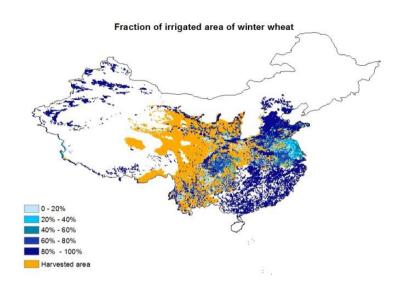
Table 6. Benchmarks for the consumptive water footprint (WF) $(m^3 t^{-1})$ of winter wheat for different climate zones in China.

Table 7. Water saving if actual consumptive water footprint (WF) of winter wheat everywhere in China were reduced to the climate-differentiated WF benchmark levels set by the 10^{th} , 20^{th} and 25^{th} percentiles of production, in an average year (2005).

Climate zones	Water saving when actual consumptive WF of winter wheat everywhere in China were to be reduced to a certain percentile benchmark level				
	10th	20th	25th		
Arid	83%	81%	80%		
Humid	49%	46%	45%		
Overall	56%	54%	53%		

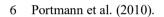
1 Figures

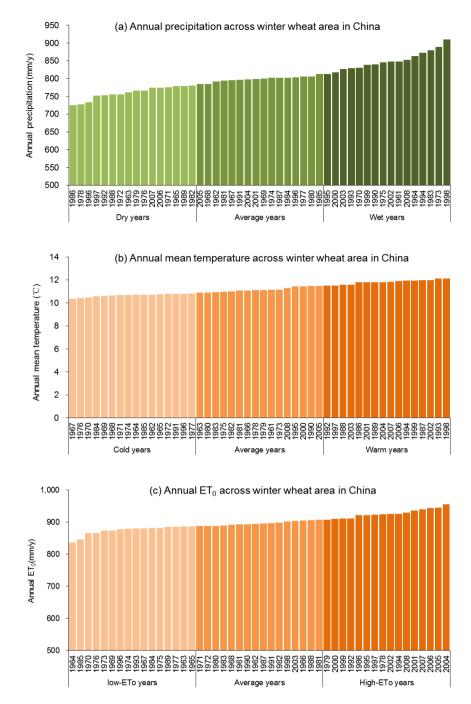




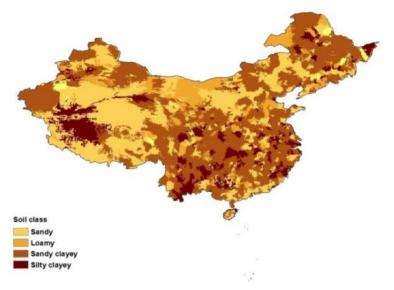


5 Figure 1. Harvested winter wheat areas in China in the year 2000 and fractions of the harvested areas irrigated. Data source:



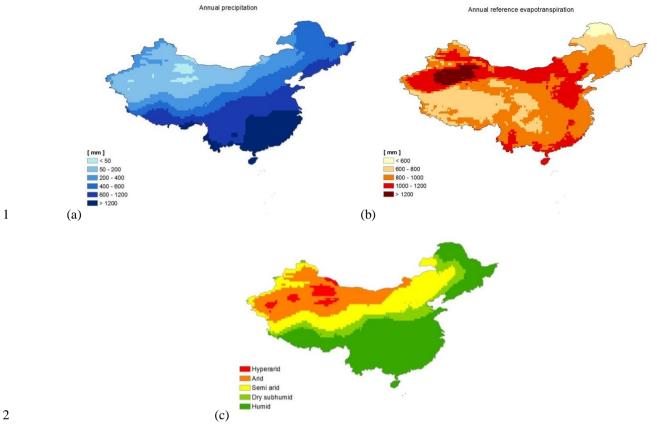


4 Figure 2. Annual precipitation (a), mean temperature (b), and $ET_0(c)$ over the cropping area of winter wheat in China for the 5 years in the period 1961-2008, ranked from lowest to highest values. Data source: Harris et al. (2014).



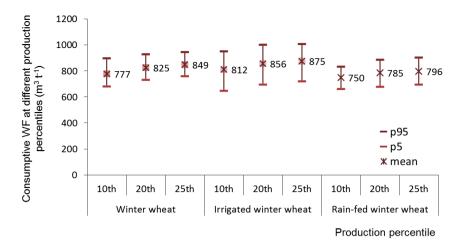
2 Figure 3. Soil classes across mainland China, generated from the ISRIC Soil and Terrain database for China. Data source:

³ Dijkshoorn et al. (2008).



4 Figure 4. Zoning of annual precipitation (a), annual reference evapotranspiration (b), and aridity (c) in China (1961-2008).

5 Data source: Harris et al. (2014).



2 Figure 5. Benchmark levels for the consumptive water footprint (WF) of winter wheat in China at different production
3 percentiles, considering all separate years in the period 1961-2008. Cross marks refer to the mean values; ranges refer to the
4 5% - 95% of accumulative frequencies.

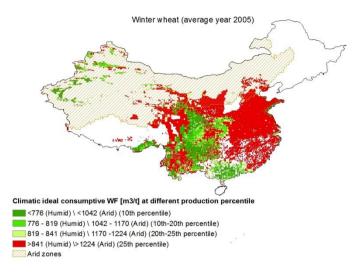




Figure 6. Simulated consumptive water footprints (WFs) of winter wheat, categorized into four classes (the best 10% of
production, the next best 10%, the second next best 5% and the worst 75% of production), accounting for different benchmark
levels for humid versus arid part of China, for the year 2005 (climatic average year).

Actual yield vs. climatic ideal yield of winter wheat in China (2005)

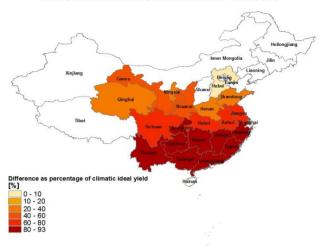


Figure 7. Differences between actual provincial yields of winter wheat in China in 2005 (NBSC, 2013) and simulated yields
from the current study (assuming no crop stress except for water stress in rain-fed areas), expressed as percentage of the
simulated yield.

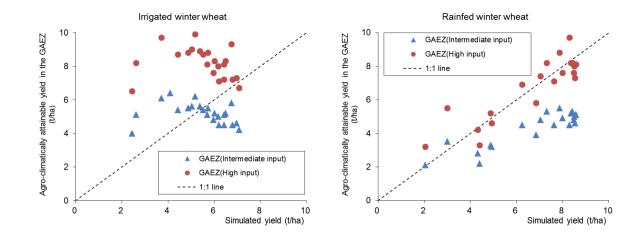


Figure 8. Comparison between the simulated yield of winter wheat and the agro-climatically attainable yield according to
(FAO/IIASA, 2011) at provincial level in China. Averaged over the period 1961-1990.