Total number of words including references: 8715

#### Towards improvement of grey water footprint assessment: 2 with an illustration for global maize cultivation 3

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**Abstract:** The grey water footprint refers to the volume of water that is required to assimilate polluted 13 14 water. It reflects the intensity of water pollution caused by water use for human activities. This study aims to address some major shortcomings associated with grey water footprint accounting in the 15 literature and discuss several ways towards its improvement. Global maize production is used for 16 illustration. The study specifically tackles three issues: the appropriate water quality standards for grey 17 18 water footprint assessment; grey water footprint for multiple pollutants; and the influence of spatial resolution of the assessment on the level of grey water stress. A biophysical crop model is applied to 19 quantify nitrogen and phosphorus losses to water in maize production on a global scale with a 0.5-20 degree spatial resolution. The study shows that the grey water footprint calculation is highly sensitive 21 22 to the water standards applied. The results also suggest that the grey water footprint relating to nitrogen and phosphorus pollution caused by maize production alone has already exceeded their local 23 water availability in many parts of the world. Grey water stress shows a more critical situation at the 24 grid level than at the watershed level for maize cultivation because the former represents the local 25 concentration whereas the latter gives the average situation of the whole watershed. This study 26 highlights the need for standardizing the setting of water quality standards for a consistent grey water 27 footprint assessment taking into consideration the diverse aquatic ecosystems and ambient water 28 29 quality requirements across regions, as well as the presence of multiple pollutants in water bodies.

- Keywords: nitrogen and phosphorus losses; water quality standards; grey water stress; global
- 31 assessment; PEPIC

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This document is the accepted manuscript version of the following article: Liu, W., Antonelli, M., Liu, X., & Yang, H. (2017). Towards improvement of grey water footprint assessment: with an illustration for global maize cultivation. Journal of Cleaner Production, 147, 1-9. https://doi.org/10.1016/j.jclepro.2017.01.072 32

# 1. Introduction

Nitrogen (N) and phosphorus (P) are key elements to life and are essential for crop and
livestock production. During the period 1960–2010, the application of N and P fertilizers in agriculture
for food production increased nine-fold and three-fold, respectively (Sutton et al., 2013). The use of
fertilizers has, on the one hand, improved agricultural productivity, enabling the feeding of a growing
world population while coping with the dietary shift towards an increased consumption of meat and
dairy products. On the other hand, the use of fertilizers has dramatically increased the amount of N and
P entering the terrestrial biosphere (Bennett et al., 2001; Vitousek et al., 2009). Nutrient losses from
croplands into water bodies have caused major environmental problems, such as water quality
degradation, groundwater contamination, biodiversity loss, fish deaths, and eutrophication (Galloway
and Cowling, 2002; Obersteiner et al., 2013; Vitousek et al., 1997).
The need to account for the impacts of agricultural production in terms of water quantity and
quality led to the development of water footprint indicators in the early 2000s (Hoekstra, 2003). The
water footprint (WF) is a multidimensional indicator of consumptive water use, which accounts for
green (rain) water, blue (surface and underground) water resources, and grey (polluted) water. The
grey water footprint (GWF) was introduced by Hoekstra and Chapagain (2008) as a measure of the
intensity of water pollution caused by water use for human activities. It is defined as the volume of
water that is required to assimilate a load of pollutants to a freshwater body, based on natural
background concentrations and existing ambient water quality standards (Hoekstra et al., 2011). The
idea of measuring water pollution in terms of the amount of water needed to dilute pollutants can be
traced back to Falkenmark and Lindh (1974), who pointed out that the amount of water required to
dilute pollutants to acceptable levels is about 10-50 times the wastewater flow. The GWF indicator
assumes that the gap between a water quality standard and the natural background concentration in a
given water body can be used to dilute the pollution loads to meet the water quality standard. It
expresses water pollution in terms of a water volume needed to dilute contaminated water to a given
quality standard, so that it can be compared with water consumption.

Global GWF assessments have mainly been provided by the Water Footprint Network, e.g.,

A growing number of studies have provided GWF assessments at various geographical levels.

Chapagain et al. (2006), Hoekstra and Mekonnen (2012), and Liu et al. (2012) (Table 1). Other GWF 61 62 studies have been conducted at the national and regional levels (e.g., Cazcarro et al., 2016; Mekonnen et al., 2016); the river basin level (e.g., Miguel Ayala et al., 2016; Vanham and Bidoglio, 2014; Zhi et 63 64 al., 2015); the city level (e.g., Manzardo et al., 2016a; Wang et al., 2013); and with a focus on specific products or crops (e.g., Ene et al., 2013; Lamastra et al., 2014; Suttayakul et al., 2016). GWF 65 assessments have overwhelmingly been focused on N-related loads to freshwater. Only a few 66 considered multiple pollutants, such as N, P, COD (chemical oxygen demand), and NH<sub>4</sub> (ammonium) 67 (Dabrowski et al., 2009; Lu et al., 2016; Pellicer-Martinez and Martinez-Paz, 2016). Most GWF 68 assessments used the drinking water standards (e.g., Bulsink et al., 2010; Chapagain et al., 2006; 69 Mekonnen and Hoekstra, 2011, 2010), with a few exceptions that have used ambient water quality 70 71 standards (e.g., Pellegrini et al., 2016; Pellicer-Martinez and Martinez-Paz, 2016; Zhuo et al., 2016) 72 (Table 1). It has been shown that agriculture, mainly cereal production, accounts for 75% of the global 73 GWF related to anthropogenic N loads, with the highest contribution from Asia (Mekonnen and 74 75 Hoekstra, 2015). In the period 1996–2005, GWF accounted for 15% of the global annual total water 76 footprint (green, blue, and grey), and for 19% of global agricultural and industrial virtual water flows (Hoekstra and Mekonnen, 2012). It has also been shown that about two-thirds of world's major rivers, 77 especially in tropical and sub-tropical areas, have a pollutant load that exceeds the basin's assimilation 78 79 capacity (Liu et al., 2012). A number of studies have shown that large GWF can exacerbate water 80 scarcity conditions (Liu et al., 2012; Mekonnen et al., 2016; Zhuo et al., 2016). Others have pointed 81 out that the international trade of agricultural products has resulted in a globalization of agricultural 82 pollution, which has substantially increased over the past few decades (Galloway et al., 2008; Mekonnen et al., 2016; O'Bannon et al., 2014). 83 84 The GWF indicator has been developed to describe water quality conditions and inform policies related to water pollution. However, the concept has a number of limitations that are mainly 85 related to three aspects. First, there are significant disparities in the water standards for natural nutrient 86 and maximum allowable concentration values used in the GWF assessment. The volume of GWF and 87

associated values are highly sensitive to the standards chosen. This point has not been paid sufficient

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attention in the GWF studies and confusion often occurs. Second, GWF is generally assessed for individual pollutants, while in reality they mostly co-exist. There has been no study that specifically discussed the influences of the integration of GWF of multiple pollutants on GWF by comparing with individual pollutants. Third, the GWF studies have been conducted at different geographical levels, e.g., grid and river basin levels. However, the effects of different spatial resolutions on the GWF assessment have not been addressed. These shortcomings need to be addressed to enhance the usefulness of GWF assessment as a tool for informing sound water management and policy.

The grey water stress (GWS), which is defined as the ratio of GWF to the actual runoff with the same spatiotemporal scale, indicates the status of the assimilative capacity of the corresponding water body under the pressure of GWF, i.e. the water stress caused by GWF. The concept suggests that if the pressure of GWF is higher than the assimilation capacity, then the GWF is environmentally unsustainable (Hoekstra et al., 2011; Mekonnen et al., 2016). It should be noted that Hoekstra et al. (2011) called this indicator the water pollution level (WPL), which was adopted by several studies (e.g., Liu et al., 2012; Mekonnen and Hoekstra, 2015). However, the concept of GWS is considered to be more explicitly in line with the concept of GWF. Therefore, this terminology is used in this study.

To address the above limitations in GWF accounting, a global assessment of the GWF for one of the most important cereal crops in the world, i.e. maize, is conducted for illustration. Globally, maize produces the highest crop production and ranks third on N and P consumption as well as the resulted water pollution, following wheat and rice (West et al., 2014). Considering its wide spatial coverage (Portmann et al., 2010) and relatively less diverse production conditions compared to wheat and rice, maize is selected as an illustration here. A large-scale crop model PEPIC (Liu et al., 2016b), i.e. Environmental Policy Integrated Climate (EPIC) under the Python environment (www.python.org), is applied to quantify the losses of N and P from fertilizer application on a global scale with a 0.5-degree spatial resolution. Total losses of N and P from maize croplands into water bodies are considered in the GWF assessment, as they are the major nutrients applied to crops as agrochemical fertilizers. Water quality standards are clarified and the implications of the disparities are addressed. The GWS is assessed at the grid and river basin levels to highlight the impacts of different geographical scales on the assessment. Given the importance and increasing emphasis of water

management at the basin/watershed level, it is useful to provide the assessment of GWS at this level.

### 2. Methodology and data

#### 2.1 Model description and input data

In this study, the EPIC model (Williams et al., 1984) was used to quantify the N and P losses from maize production. EPIC is a field-scale crop model, which was initially developed to simulate the impacts of soil erosion on soil productivity. It was then extended to simulate the complex processes in the soil—water—climate—management systems (Williams, 1995). PEPIC is a simulation framework, which is able to prepare spatial input data for EPIC, run the EPIC model, extract outputs of the simulation, and finally map the results of desired variables (Liu et al., 2016b). It has been successfully used to investigate the crop-water relations of maize and assess global N losses from major crops (Liu et al., 2016b, 2016c).

The inputs to the PEPIC model include climate, soil, digital elevation model (DEM), slope, and crop management information, e.g., planting and harvesting dates, irrigation, and fertilizer. The fertilizer inputs (including N and P) were downloaded from EarthStat (http://www.earthstat.org/). This dataset was based on Mueller et al. (2012) and West et al. (2014). Information about the other inputs can be found in (Liu et al., 2016b).

The inputs of N considered in this study include chemical fertilizer, manure, precipitation deposition, and crop residue decomposition; while the P inputs include chemical fertilizer, manure, and crop residue decomposition. On the downside, the N outputs include crop uptake, N lost to atmosphere through denitrification and volatilization, and N lost to water bodies through surface runoff, leaching, and bounded with sediments. As for P, the outputting pathway is crop uptake, and losses go into water bodies through surface runoff, leaching, and with sediments. More details of the nutrient dynamics of the EPIC model can be found in Williams (1995) and the calculation of N losses can be found in Liu et al. (2016c). For the calculation of GWF related to N and P, the study focused on the total losses of N and P into the water bodies.

#### 2.2 Grey water footprint and grey water stress

By definition (Hoekstra et al., 2011), GWF is calculated as:

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$$GWF_N = 100 * L_N / (C_{max_N} - C_{nat_N})$$
 (1)

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$$GWF_P = 100 * L_P / (C_{max P} - C_{nat P})$$
 (2)

where GWF<sub>N</sub> [mm] and GWF<sub>P</sub> [mm] are the grey water footprint derived from N and P loads in water bodies, respectively; L<sub>N</sub> [kg N ha<sup>-1</sup>] and L<sub>P</sub> [kg P ha<sup>-1</sup>] are the N and P concentrations, which can be estimated at different levels, e.g., grid and basin levels, as the ratios of total N [kg N] and P [kg P] loads to the whole areas of the corresponding level; C<sub>max\_N</sub> [mg N L<sup>-1</sup>] and C<sub>max\_P</sub> [mg P L<sup>-1</sup>] are the ambient water quality standards for N and P; C<sub>nat\_N</sub> [mg N L<sup>-1</sup>] and C<sub>nat\_P</sub> [mg P L<sup>-1</sup>] are the natural background concentrations of N and P in the receiving water body; 100 is for unit transformation.

In the literature, for multiple pollutants, the GWF was calculated and presented separately for each pollutant, e.g., N, P, COD, and pesticides in some studies (e.g., Dabrowski et al., 2009; Liu et al., 2016a, 2012). For an individual grid cell or a given water body, e.g., river basin, the GWF under the condition of multiple pollutants (hereby referred to as integrated GWF) should be assessed by choosing the highest GWF deriving from single pollutants (Eq. 3). A summation of GWF of individual pollutants would overestimate the total GWF because the volume of water required for diluting one pollutant can simultaneously dilute other pollutants. The highest GWF of a single pollutant can thus represent the integrated GWF for that grid cell or corresponding water body. The integrated grey water footprint is calculated as:

$$161 GWF_{I} = max(GWF_{N}, GWF_{P}) (3)$$

where GWF<sub>I</sub> [mm] is the integrated grey water footprint from considering the N and P loads. The total

GWF is calculated as the product of GWF and the whole area. The GWS is calculated as:

$$164 GWS = GWF/R (4)$$

where R is the runoff with the same spatiotemporal scale as GWF [mm]. Runoff data are produced by the Distributed Biosphere Hydrological (DBH) model forced by the Princeton global meteorological data (Liu et al., 2016d; Tang et al., 2008). For the management of a water body, assessing the consumption of assimilative capacity caused by pollutants is important (Zhi et al., 2015). With this concept, GWS < 1 means that the water body still has excess assimilative capacity to accommodate more pollutants; GWS = 1 means all the assimilative capacity is used up; and GWS > 1 means that the pollution level has exceeded the assimilative capacity and that the water pollution is thus unsustainable.

The higher the GWS, the poorer the water quality. In this study, GWS related to N (GWS<sub>N</sub>), P (GWS<sub>P</sub>), and their integration (GWS<sub>I</sub>) are considered.

#### 2.3 Water quality standards review

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Water standards are the key information required to calculate the GWF and GWS. In different countries, these standards can vary quite substantially. An extensive literature review was conducted on the water standards used for GWF assessments related to N and P (Table 1). It reveals some confusion in water quality standards. From the definition of Hoekstra et al. (2011), the ambient environmental water quality standards should be applied in the GWF calculation. However, in the GWF literature to date, the USA and the EU N standards for drinking water, respectively 10 mg N L<sup>-1</sup> and 11.3 mg N L<sup>-1</sup>, are the most commonly used water quality standards. Among the 40 reviewed studies (which explicitly used the N-related water quality standards) in Table 1, 30 (75%) of them used the N drinking water quality standards. So far, no study has specifically addressed this issue. Some local standards have also been used for regional GWF assessments. It is noted that studies even adopted different standards in the same region/country. For example, in China, three standards for nitrogen were used, i.e. 1 (e.g., Wu et al., 2016), 10 (e.g., Lu et al., 2016), and 12 (Duan et al., 2016) mg N L<sup>-1</sup> in the literature. Obviously, these different standards introduced significant disparities in GWF accounting. The confusion is partly related to the fact that for many countries, the information on ambient water quality standards are not available, either because they do not exist (e.g., for most African countries and less developed countries) or because of the lack of access to the data sources (e.g., for most of other countries). For P, a few papers addressed the GWF with standards varying from  $0.1 \text{ to } 4 \text{ mg P L}^{-1} \text{ (Table 1)}.$ 

Table 1 Literature review of standards used for grey water footprint assessment.

		Nitrogen		Phosphorus		
Num.	Study area	$C_{max}$	C <sub>nat</sub>	$C_{max}$	$C_{nat}$	References
		$(mg N L^{-1})$	$(mg N L^{-1})$	$(mg P L^{-1})$	$(mg P L^{-1})$	
1	Argentina <sup>R</sup>	10 (USA)*				(Rodriguez et al., 2015)
2	$Brazil^{G+W}$			0.1 (Brazil)	0	(Miguel Ayala et al.,
						2016)
3	Brazil <sup>R</sup>	10 (Brazil)*	0.8			(Scarpare et al., 2016)
4	China <sup>R</sup>	12 (China)*				(Duan et al., 2016)
5	China <sup>W</sup>	1 (China)#	0			(Liu et al., 2016a)
6	China <sup>R</sup>	$10 (USA)^*$				(Huang et al., 2012)
7	China <sup>R</sup>	10 (China)*	0			(Lu et al., 2016)
8	China <sup>R</sup>	1 (China)#		0.2 (China)		(Wang et al., 2013)

9	China <sup>R</sup>	1 (China)#	0	0.2 (China)	0	(Wu et al., 2016)
10	China <sup>R</sup>	$10 (USA)^*$				(Xu et al., 2015)
11	China <sup>W</sup>	10 (China)*	0			(Zeng et al., 2013)
12	China <sup>G+R</sup>	1 (China)#	0.2	0.2 (China)	0.02	(Zhuo et al., 2016)
13	England <sup>R</sup>	12.86 (England and Wales)#	6.38	0.25 (England and Wales)	0.01	(Zhang et al., 2014)
14	Europe <sup>G+R</sup>	3.1 (Liu et al.,	1.5	0.95 (Liu et al.,	0.52	(Mekonnen et al., 2016)
15	Europe <sup>R</sup>	2012) 10 (USA)*		2012)		(Thaler et al., 2012)
16	Europe <sup>W</sup>	10 (USA)*	0			(Vanham and Bidoglio,
	•	, ,	U			2014)
17	France <sup>G+R</sup>	10 (USA)*				(Ercin et al., 2013)
18	Global <sup>R</sup>	10 (USA)*	0			(Chapagain et al., 2006)
19	Global <sup>R</sup>	11.3 (EU)*				(Chapagain and Hoekstra, 2011)
20	Global <sup>G+R</sup>	10 (USA)*	0			(Hoekstra and Mekonnen, 2012)
21	Global <sup>W</sup>	3.1 (Estimated)	1.5	0.95 (Estimated)	0.52	(Liu et al., 2012)
22	Global <sup>G+W</sup>	2.9 (Canada)#	0.4			(Mekonnen and Hoekstra, 2015)
23	Global <sup>G+R</sup>	10 (USA)*	0			(Mekonnen and Hoekstra, 2010)
24	Global <sup>G+R</sup>	10 (USA)*	0			(Mekonnen and Hoekstra, 2011)
25	Global <sup>G+R</sup>	10 (USA)*	0			(Mekonnen and Hoekstra,
26	Indonesia <sup>R</sup>	10 (USA)*				2014) (Bulsink et al., 2010)
27	Italy <sup>G+R</sup>	10 (USA)*	0			(Aldaya and Hoekstra,
	•	, ,				2010)
28	Italy <sup>R</sup>	15 (Italy)#	0			(Pellegrini et al., 2016)
29	Kenya <sup>G+W</sup>	10 (USA)*	0			(Mekonnen et al., 2012)
30	Latin America and the Caribbean <sup>G+R</sup>	10 (USA)*				(Mekonnen et al., 2015)
31	New Zealand <sup>R</sup>	11.3 (New Zealand)*	0-1.3			(Deurer et al., 2011)
32	New Zealand <sup>R</sup>	11.3 (New Zealand)*	0			(Herath et al., 2013)
33	Morocco <sup>R</sup>	10 (USA)*				(Schyns and Hoekstra, 2014)
34	Romania <sup>R</sup>	10 (USA)*				(Ene et al., 2013)
35	South Africa <sup>R</sup>	4 (South Africa)#	0.62	0.13 (South Africa)	0.06	(Dabrowski et al., 2009)
36	South Korea <sup>R</sup>	40 (South Korea) <sup>#</sup>		4 (South Korea)		(Yoo et al., 2014)
37	Spain <sup>R</sup>	11.3 (EU)*				Cazcarro et al., 2016)
38	Spain <sup>R</sup>	11.3 (EU)*				(Chapagain and Orr, 2009)
39	Spain <sup>W</sup>	11.3 (EU)*				(Chico et al., 2013)
40	Spain <sup>R</sup>	5.6 (Spain)#		0.13 (Spain)		(Pellicer-Martinez and
10	Spani	J.0 (Spain)		o.io (Spain)		Martinez-Paz, 2016)
41	Taiwan, China <sup>R</sup>	10 (Taiwan)*	0			(Su et al., 2015)
42	Tunisia <sup>R</sup>	10 (USA)*	Ö			(Chouchane et al., 2015)
43	USA <sup>R</sup>	10 (USA)*	0			(Manzardo et al., 2016b)
		- 0 (0 221)	<u> </u>			(

R: Regional level; W: Watershed level; G+R: Grid and regional levels; G+W: Grid and watershed levels; \*: Nitrogen standard for drinking water; #: Nitrogen standard for surface water; --: no information.

Several national N and P standards for drinking water and ambient surface water were reviewed (Table 2). They show that the allowed N concentrations for surface water are generally lower than that for drinking water. This means that the estimated GWF tends to be underestimated by adopting drinking water standards, which was the case for many previous studies (as shown in Table

1). As P is not directly toxic for humans, drinking water standards have not been set for this pollutant. Some points to note are that in some cases the national N and P standards are not unique values but region specific. For example, the USA was divided into 13 ecoregions and there is one standard for each ecoregion (https://www.epa.gov/nutrient-policy-data/ecoregional-criteria). This situation also holds for the Netherlands (Table 2). The way to determine surface water N standards also varies. For example, the N surface standard in Canada (3 mg N L<sup>-1</sup>) was derived from toxicity tests of aquatic species exposed to environments with different levels of nitrate (http://ceqg-rcqe.ccme.ca/download/en/197?redir=1465564929). In the USA, however, it was determined by the 25th of measured nitrate concentrations from surface water (https://www.epa.gov/nutrient-policy-data/ecoregional-criteria).

After careful assessment and discussions with experts in ecosystem sciences, the N standard adopted in this study for surface water is 3 mg N L $^{-1}$ . This standard is also recommended by Franke et al. (2013). For comparison, the most commonly used N standard for drinking water (10 mg N L $^{-1}$ ) was also used for GWF calculation related to N. For P, the standard used here is 0.15 mg P L $^{-1}$ . This standard is recommended by Germany and the EU (Table 2). Many previous studies assumed the natural pollutant concentration to be 0 due to a lack of data (e.g., Chapagain et al., 2006; Liu et al., 2016a; Zeng et al., 2013). However, the natural concentrations of N and P are generally higher than 0, thus such an assumption leads to an underestimation of GWF. In this study, the natural concentration was set to be 0.4 mg N L $^{-1}$  and 0.01 mg P L $^{-1}$  for N and P, respectively. The natural concentration for N of 0.4 mg N L $^{-1}$  is recommended by Hoekstra et al. (2011) and Franke et al. (2013), and is also used by Mekonnen and Hoekstra (2015). As for the natural concentration of P, 0.01 mg P L $^{-1}$ , is recommended by Franke et al. (2013). The natural concentration was also considered to be 0 for illustrating the impacts of using different natural concentrations.

Table 2 National water standards for nitrogen (N) and phosphorus (P).

	1	N (mg N L <sup>-1</sup> )	TD ( D I = 1)
	Drinking water	Surface water	$ TP (mg P L^{-1}) $
Canada	10 (NO <sub>3</sub> -N) <sup>R1</sup>	3 (NO <sub>3</sub> -N) <sup>R2</sup>	$0.02^{R3}$
China	$10 (NO_3-N)^{R4}$	1 (TN, lakes) <sup>R5</sup>	0.05 (lakes) <sup>R5</sup> 0.2 (rivers) <sup>R5</sup>
EU	$11.3 (NO_3-N)^{R6}$	$5.6 (NO_3-N)^{R7}$	$0.07 - 0.15^{R8}$
Germany	$11.3 (NO_3-N)^{R9}$	$3 (TN)^{R10}$	$0.15^{R11}$

		2.5 (NO <sub>3</sub> -N) <sup>R10</sup>	
Switzerland	5.6 (NO <sub>3</sub> -N) <sup>R12</sup>	$7 (TN)^{R13}$	$0.07^{R13}$
		5.6 (NO <sub>3</sub> -N) <sup>R13</sup>	
The Netherlands	$11.3 (NO_3-N)^{R14}$	0.12-18.05 (TN)*, R15	0.01-2.5*, R15
USA	$10 (NO_3-N)^{R16}$	0.1-1.27 (TN, lakes) <sup>R17</sup>	0.008-0.038 (lakes) <sup>R17</sup>
	` /	0.12-2.18 (TN, rivers) <sup>R17</sup>	$0.01-0.076  (rivers)^{R17}$

- TN: total nitrogen; TP: total phosphorus. \*: the ranges are quite high because they incorporate specific goals from more than 225 226 227 500 water bodies.
- R1: http://www.hc-sc.gc.ca/ewh-semt/pubs/water-eau/sum\_guide-res\_recom/index-eng.php;
- 228 R2: http://cegg-rcqe.ccme.ca/download/en/197?redir=1465564929;
- 229 R3: http://ceqg-rcqe.ccme.ca/download/en/205?redir=1465564939;
- 230 R4: http://www.moh.gov.cn/zwgkzt/pgw/201212/33644.shtml;
- R5: http://kjs.mep.gov.cn/hjbhbz/bzwb/shjbh/shjzlbz/200206/t20020601 66497.shtml;
- 231 232 R6: http://eur-lex.europa.eu/LexUriServ/LexUriServ.do?uri=OJ:L:1998:330:0032:0054:EN:PDF;
- 233 R7: http://eur-lex.europa.eu/legal-content/EN/TXT/PDF/?uri=CELEX:31991L0676&from=en;
- 234 R8: http://eur-lex.europa.eu/LexUriServ/LexUriServ.do?uri=OJ:L:2006:264:0020:0031:EN:PDF;
- 235 R9: http://www.gesetze-im-internet.de/bundesrecht/trinkwv\_2001/gesamt.pdf;
- 236 R10: http://www.vsr-gewaesserschutz.de/33.html;
- 237 R11: http://www.vsr-gewaesserschutz.de/20.html;
- 238 R12: https://www.admin.ch/opc/en/classified-compilation/19983281/index.html#app4;
- 239 R13: http://www.modul-stufen-konzept.ch/download/ChemieD\_Juni2010.pdf;
- 240 R14: http://wetten.overheid.nl/BWBR0027061/2016-01-01/0/#BijlageIII;
- 241 R15: http://www.helpdeskwater.nl/onderwerpen/wetgeving-beleid/kaderrichtlijn-water/2016-2021/;
- R16: https://www.epa.gov/wqc/national-recommended-water-quality-criteria-human-health-criteria-table; 242
- 243 R17: https://www.epa.gov/nutrient-policy-data/ecoregional-criteria.

#### 3. Results

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#### 3.1 Grey water footprint assessment

Global GWF for maize cultivation using 3 mg N L<sup>-1</sup> and 0.15 mg P L<sup>-1</sup> as standards with the natural concentrations of 0.4 mg N L<sup>-1</sup> and 0.01 mg P L<sup>-1</sup> is presented in Fig. 1. It can be seen that GWF<sub>P</sub> is much higher than GWF<sub>N</sub> in many regions. The GWF<sub>N</sub> and GWF<sub>P</sub> also present different spatial patterns. The North China Plain, northeastern parts of China, and north central parts of the USA are the major regions with a high level of GWF<sub>N</sub> (> 500 mm) (Fig. 1a). While for GWF<sub>P</sub>, not only the above mentioned regions, but also southwest parts of China, southern Mexico, and southern Brazil, as well as some parts of central Europe show a high level of GWF<sub>P</sub> (Fig. 1b). GWF<sub>I</sub> presents quite a similar pattern to GWF<sub>P</sub> (Fig. 1c), implying that GWF<sub>I</sub> is more related to GWF<sub>P</sub> for maize cultivation.

Fig. 1 about here.

The global total GWF<sub>N</sub> for maize is 706 km<sup>3</sup> using the drinking water standard (10 mg N L<sup>-1</sup>) and increases to 2607 km<sup>3</sup> using the ambient water standard (3 mg N L<sup>-1</sup>) (Table 3). The conversion of GWF between different standards is straightforward. For example, the GWF<sub>N</sub> with the ambient water quality standard can be obtained by multiplying 3.69, i.e. the ratio of (10 - 0.4) to (3 - 0.4), by the

GWF<sub>N</sub> calculated using the drinking water quality standard. The global GWF<sub>P</sub> is about 2.7 times that of GWF<sub>N</sub>. Although globally N loads are generally higher than P loads, the higher GWF<sub>P</sub> is mainly due to the much lower value of maximum allowable concentration for P than for N. The GWF<sub>I</sub> is only a little higher than GWF<sub>P</sub> (4%). This indicates that GWF relating to maize production is mainly determined by P. Asia and North America contribute the largest proportion of global GWF both for N and P, while South America also contributes a high proportion of global GWF<sub>P</sub>. At the country level, China and the USA account for the highest proportion of GWF<sub>N</sub>, while the USA, China, and Brazil contribute the most GWF<sub>P</sub>. For the top ten watersheds, the Mississippi plays the dominant role in contributing to GWF<sub>N</sub>, while the Mississippi and La Plata show similar roles for GWF<sub>P</sub>. Disparities of GWF at continental, country, and river basin levels are noticeable. For example, the GWF<sub>P</sub> (P0.15\_0.01, with 0.15 mg P L<sup>-1</sup> and 0.01 mg P L<sup>-1</sup> as standard and natural concentration) is about 2.2 times that of GWF<sub>N</sub> (N3\_0.4, with 3 mg N L<sup>-1</sup> and 0.4 mg N L<sup>-1</sup> as standard and natural concentration) in Asia, however it decreases to 1.4 times in China and only accounts for 55% of the GWF<sub>N</sub> of the Haihe river basin.

Africa 33.5 123.7 563.2 58 Asia 284.1 1049.1 2297.2 24 Europe 75.1 277.2 666.9 68 N. America 236.1 871.9 2056.5 21 Oceania 0.3 1.0 6.6 6.6 S. America 76.8 283.6 1395.1 13 Global 706.0 2606.7 6985.5 72  Top 10 maize-producing countries*  N10_0.4* N3_0.4* P0.15_0.01* GV USA 160.4 592.4 1233.4 12 China 208.5 769.9 1097.3 12 Brazil 54.9 202.6 1088.2 10 Mexico 46.9 173.3 547.4 54 India 18.2 67.2 365.7 36 Russia 2.8 10.4 64.0 64 Nigeria 4.1 15.0 116.5 11 Argentina 8.2 30.4 125.6 12 Ukraine 1.8 6.6 35.5 35 France 25.6 94.5 123.3 13  Top 10 maize-producing watersheds*  N10_0.4* N3_0.4* P0.15_0.01* GV Ukraine 1.8 6.6 35.5 35 France 25.6 94.5 123.3 13  Top 10 maize-producing watersheds*  N10_0.4* N3_0.4* P0.15_0.01* GV Mississippi 132.6 489.8 988.0 10 La Plata 44.6 164.5 843.2 84 Danube 11.5 42.3 144.0 14 Amur 37.1 137.0 113.4 15 Haihe 35.2 129.8 71.4 13 Yangtze 37.7 139.3 368.3 368.3			level	Continental		
Asia 284.1 1049.1 2297.2 24 Europe 75.1 277.2 666.9 68 N. America 236.1 871.9 2056.5 21 Oceania 0.3 1.0 6.6 S. America 76.8 283.6 1395.1 13 Global 706.0 2606.7 6985.5 72  Top 10 maize-producing countries*  N10_0.4 <sup>a</sup> N3_0.4 <sup>b</sup> P0.15_0.01 <sup>c</sup> GV USA 160.4 592.4 1233.4 12 China 208.5 769.9 1097.3 12 Brazil 54.9 202.6 1088.2 10 Mexico 46.9 173.3 547.4 54 India 18.2 67.2 365.7 36 Russia 2.8 10.4 64.0 64 Nigeria 4.1 15.0 116.5 11 Argentina 8.2 30.4 125.6 12 Ukraine 1.8 6.6 35.5 35 France 25.6 94.5 123.3 13  Top 10 maize-producing watersheds*  N10_0.4 <sup>a</sup> N3_0.4 <sup>b</sup> P0.15_0.01 <sup>c</sup> GV  Top 10 maize-producing watersheds*  N10_0.4 <sup>a</sup> N3_0.4 <sup>b</sup> P0.15_0.01 <sup>c</sup> GV  Mississippi 132.6 489.8 988.0 10  La Plata 44.6 164.5 843.2 84  Danube 11.5 42.3 144.0 14  Amur 37.1 137.0 113.4 15  Haihe 35.2 129.8 71.4 13  Yangtze 37.7 139.3 368.3 368.3	WF <sub>I</sub> <sup>d</sup>	G	P0.15_0.01°	N3_0.4 <sup>b</sup>	N10_0.4a	
Europe 75.1 277.2 666.9 68 N. America 236.1 871.9 2056.5 21 Oceania 0.3 1.0 6.6 6.6 S. America 76.8 283.6 1395.1 13 Global 706.0 2606.7 6985.5 72  Top 10 maize-producing countries*  N10_0.4a N3_0.4b P0.15_0.01c GV USA 160.4 592.4 1233.4 12 China 208.5 769.9 1097.3 12 Brazil 54.9 202.6 1088.2 10 Mexico 46.9 173.3 547.4 54 India 18.2 67.2 365.7 36 Russia 2.8 10.4 64.0 64 Nigeria 4.1 15.0 116.5 11 Argentina 8.2 30.4 125.6 12 Ukraine 1.8 6.6 35.5 35 France 25.6 94.5 123.3 13  Top 10 maize-producing watersheds*  N10_0.4a N3_0.4b P0.15_0.01c GV Ukraine 1.8 6.6 35.5 35 France 25.6 94.5 123.3 13  Mississippi 132.6 489.8 988.0 10 La Plata 44.6 164.5 843.2 84 Danube 11.5 42.3 144.0 14 Amur 37.1 137.0 113.4 15 Haihe 35.2 129.8 71.4 13 Yangtze 37.7 139.3 368.3 368.3	36.9	58	563.2	123.7	33.5	Africa
N. America 236.1 871.9 2056.5 21 Oceania 0.3 1.0 6.6 6.6 S. America 76.8 283.6 1395.1 13 Global 706.0 2606.7 6985.5 72  Top 10 maize-producing countries*    N10_0.4a	159.0	24	2297.2	1049.1	284.1	Asia
Oceania         0.3         1.0         6.6         6.6           S. America         76.8         283.6         1395.1         13           Global         706.0         2606.7         6985.5         72           Top 10 maize-producing countries*           USA         160.4         592.4         1233.4         12           China         208.5         769.9         1097.3         12           Brazil         54.9         202.6         1088.2         10           Mexico         46.9         173.3         547.4         54           India         18.2         67.2         365.7         36           Russia         2.8         10.4         64.0         64           Nigeria         4.1         15.0         116.5         11           Argentina         8.2         30.4         125.6         12           Ukraine         1.8         6.6         35.5         35           France         25.6         94.5         123.3         13           Top 10 maize-producing watersheds*           Mississippi         132.6         489.8         988.0         10 <t< td=""><td>31.0</td><td>68</td><td>666.9</td><td>277.2</td><td>75.1</td><td>Europe</td></t<>	31.0	68	666.9	277.2	75.1	Europe
S. America       76.8       283.6       1395.1       13         Global       706.0       2606.7       6985.5       72         Top 10 maize-producing countries*         N10_0.4a       N3_0.4b       P0.15_0.01c       GV         USA       160.4       592.4       1233.4       12         China       208.5       769.9       1097.3       12         Brazil       54.9       202.6       1088.2       10         Mexico       46.9       173.3       547.4       54         India       18.2       67.2       365.7       36         Russia       2.8       10.4       64.0       64         Nigeria       4.1       15.0       116.5       11         Argentina       8.2       30.4       125.6       12         Ukraine       1.8       6.6       35.5       35         France       25.6       94.5       123.3       13         Top 10 maize-producing watersheds*         Mississippi       132.6       489.8       988.0       10         La Plata       44.6       164.5       843	102.7	2	2056.5	871.9	236.1	N. America
Global         706.0         2606.7         6985.5         72           Top 10 maize-producing countries*           N10_0.4a         N3_0.4b         P0.15_0.01c         GV           USA         160.4         592.4         1233.4         12           China         208.5         769.9         1097.3         12           Brazil         54.9         202.6         1088.2         10           Mexico         46.9         173.3         547.4         54           India         18.2         67.2         365.7         36           Russia         2.8         10.4         64.0         64           Nigeria         4.1         15.0         116.5         11           Argentina         8.2         30.4         125.6         12           Ukraine         1.8         6.6         35.5         35           France         25.6         94.5         123.3         13           Top 10 maize-producing watersheds*           Mississippi         132.6         489.8         988.0         10           La Plata         44.6         164.5	6	6.	6.6	1.0	0.3	Oceania
Top 10 maize-producing countries*   N10_0.4a   N3_0.4b   P0.15_0.01c   GV	397.5	13	1395.1	283.6	76.8	S. America
N10_0.4a   N3_0.4b   P0.15_0.01c   GV	233.7	72	6985.5	2606.7	706.0	Global
USA 160.4 592.4 1233.4 12 China 208.5 769.9 1097.3 12 Brazil 54.9 202.6 1088.2 10 Mexico 46.9 173.3 547.4 54 India 18.2 67.2 365.7 36 Russia 2.8 10.4 64.0 64 Nigeria 4.1 15.0 116.5 11 Argentina 8.2 30.4 125.6 12 Ukraine 1.8 6.6 35.5 35 France 25.6 94.5 123.3 13  Top 10 maize-producing watersheds*  N10_0.4a N3_0.4b P0.15_0.01c GV Mississippi 132.6 489.8 988.0 10 La Plata 44.6 164.5 843.2 84 Danube 11.5 42.3 144.0 14 Amur 37.1 137.0 113.4 15 Haihe 35.2 129.8 71.4 13 Yangtze 37.7 139.3 368.3 36			ing countries*	Top 10 maize-produc		
China       208.5       769.9       1097.3       12         Brazil       54.9       202.6       1088.2       10         Mexico       46.9       173.3       547.4       54         India       18.2       67.2       365.7       36         Russia       2.8       10.4       64.0       64         Nigeria       4.1       15.0       116.5       11         Argentina       8.2       30.4       125.6       12         Ukraine       1.8       6.6       35.5       35         France       25.6       94.5       123.3       13         Top 10 maize-producing watersheds*         N10_0.4a       N3_0.4b       P0.15_0.01c       GV         Mississisppi       132.6       489.8       988.0       10         La Plata       44.6       164.5       843.2       84         Danube       11.5       42.3       144.0       14         Amur       37.1       137.0       113.4       15         Haihe       35.2       129.8       71.4       13         Yangtze       37.7       139.3       368.3       36	$WF_{I}^{d}$	G	P0.15_0.01°	N3_0.4 <sup>b</sup>	N10_0.4a	
Brazil       54.9       202.6       1088.2       10         Mexico       46.9       173.3       547.4       54         India       18.2       67.2       365.7       36         Russia       2.8       10.4       64.0       64         Nigeria       4.1       15.0       116.5       11         Argentina       8.2       30.4       125.6       12         Ukraine       1.8       6.6       35.5       35         France       25.6       94.5       123.3       13         Top 10 maize-producing watersheds*         N10_0.4a       N3_0.4b       P0.15_0.01c       GV         Mississisppi       132.6       489.8       988.0       10         La Plata       44.6       164.5       843.2       84         Danube       11.5       42.3       144.0       14         Amur       37.1       137.0       113.4       15         Haihe       35.2       129.8       71.4       13         Yangtze       37.7       139.3       368.3       36	275.3	12	1233.4	592.4	160.4	USA
Mexico       46.9       173.3       547.4       54         India       18.2       67.2       365.7       36         Russia       2.8       10.4       64.0       64         Nigeria       4.1       15.0       116.5       11         Argentina       8.2       30.4       125.6       12         Ukraine       1.8       6.6       35.5       35         France       25.6       94.5       123.3       13         Top 10 maize-producing watersheds*         Nississippi       132.6       489.8       988.0       10         La Plata       44.6       164.5       843.2       84         Danube       11.5       42.3       144.0       14         Amur       37.1       137.0       113.4       15         Haihe       35.2       129.8       71.4       13         Yangtze       37.7       139.3       368.3       36	252.2	12	1097.3	769.9	208.5	China
India       18.2       67.2       365.7       36         Russia       2.8       10.4       64.0       64         Nigeria       4.1       15.0       116.5       11         Argentina       8.2       30.4       125.6       12         Ukraine       1.8       6.6       35.5       35         France       25.6       94.5       123.3       13         Top 10 maize-producing watersheds*         Top 10 maize-producing watersheds*         Mississippi       132.6       489.8       988.0       10         La Plata       44.6       164.5       843.2       84         Danube       11.5       42.3       144.0       14         Amur       37.1       137.0       113.4       15         Haihe       35.2       129.8       71.4       13         Yangtze       37.7       139.3       368.3       36	089.2	10	1088.2	202.6	54.9	Brazil
Russia       2.8       10.4       64.0       64         Nigeria       4.1       15.0       116.5       11         Argentina       8.2       30.4       125.6       12         Ukraine       1.8       6.6       35.5       35         France       25.6       94.5       123.3       13         Top 10 maize-producing watersheds*         Mississisppi       132.6       489.8       988.0       10         La Plata       44.6       164.5       843.2       84         Danube       11.5       42.3       144.0       14         Amur       37.1       137.0       113.4       15         Haihe       35.2       129.8       71.4       13         Yangtze       37.7       139.3       368.3       36	18.5	54	547.4	173.3	46.9	Mexico
Nigeria       4.1       15.0       116.5       11         Argentina       8.2       30.4       125.6       12         Ukraine       1.8       6.6       35.5       35         France       25.6       94.5       123.3       13         Top 10 maize-producing watersheds*         N10_0.4a       N3_0.4b       P0.15_0.01c       GV         Mississisppi       132.6       489.8       988.0       10         La Plata       44.6       164.5       843.2       84         Danube       11.5       42.3       144.0       14         Amur       37.1       137.0       113.4       15         Haihe       35.2       129.8       71.4       13         Yangtze       37.7       139.3       368.3       36	55.7	30	365.7	67.2	18.2	India
Argentina       8.2       30.4       125.6       12         Ukraine       1.8       6.6       35.5       35         France       25.6       94.5       123.3       13         Top 10 maize-producing watersheds*         Mississippi       132.6       489.8       988.0       10         La Plata       44.6       164.5       843.2       84         Danube       11.5       42.3       144.0       14         Amur       37.1       137.0       113.4       15         Haihe       35.2       129.8       71.4       13         Yangtze       37.7       139.3       368.3       36	ł.1	64	64.0	10.4	2.8	Russia
Ukraine         1.8         6.6         35.5         35           France         25.6         94.5         123.3         13           Top 10 maize-producing watersheds*           N10_0.4a         N3_0.4b         P0.15_0.01c         GV           Mississisppi         132.6         489.8         988.0         10           La Plata         44.6         164.5         843.2         84           Danube         11.5         42.3         144.0         14           Amur         37.1         137.0         113.4         15           Haihe         35.2         129.8         71.4         13           Yangtze         37.7         139.3         368.3         36	16.6	1	116.5	15.0	4.1	Nigeria
France         25.6         94.5         123.3         13           Top 10 maize-producing watersheds*           N10_0.4a         N3_0.4b         P0.15_0.01c         GV           Mississippi         132.6         489.8         988.0         10           La Plata         44.6         164.5         843.2         84           Danube         11.5         42.3         144.0         14           Amur         37.1         137.0         113.4         15           Haihe         35.2         129.8         71.4         13           Yangtze         37.7         139.3         368.3         36	25.6	12	125.6	30.4	8.2	Argentina
Top 10 maize-producing watersheds*    N10_0.4a	5.5	35	35.5	6.6	1.8	Ukraine
N10_0.4a         N3_0.4b         P0.15_0.01c         GV           Mississippi         132.6         489.8         988.0         10           La Plata         44.6         164.5         843.2         84           Danube         11.5         42.3         144.0         14           Amur         37.1         137.0         113.4         15           Haihe         35.2         129.8         71.4         13           Yangtze         37.7         139.3         368.3         36	31.8	13	123.3	94.5	25.6	France
Mississippi     132.6     489.8     988.0     10       La Plata     44.6     164.5     843.2     84       Danube     11.5     42.3     144.0     14       Amur     37.1     137.0     113.4     15       Haihe     35.2     129.8     71.4     13       Yangtze     37.7     139.3     368.3     36			ng watersheds*	op 10 maize-produc		
La Plata       44.6       164.5       843.2       84         Danube       11.5       42.3       144.0       14         Amur       37.1       137.0       113.4       15         Haihe       35.2       129.8       71.4       13         Yangtze       37.7       139.3       368.3       36	$WF_I{}^d$	<u>G</u>	P0.15_0.01°	N3_0.4 <sup>b</sup>	N10_0.4a	
Danube       11.5       42.3       144.0       14         Amur       37.1       137.0       113.4       15         Haihe       35.2       129.8       71.4       13         Yangtze       37.7       139.3       368.3       36	)23.9	10	988.0	489.8	132.6	Mississippi
Amur       37.1       137.0       113.4       15         Haihe       35.2       129.8       71.4       13         Yangtze       37.7       139.3       368.3       36	13.2	84	843.2	164.5	44.6	La Plata
Haihe     35.2     129.8     71.4     13       Yangtze     37.7     139.3     368.3     36	14.7	14	144.0	42.3	11.5	Danube
Yangtze 37.7 139.3 368.3 36	58.4	15	113.4	137.0	37.1	Amur
	36.1	13	71.4	129.8	35.2	Haihe
St Lawrence 21.4 79.1 148.1 14	58.6	30	368.3	139.3	37.7	Yangtze
	19.7	14	148.1	79.1	21.4	St Lawrence
China Coast 47.3 174.7 281.4 28	37.6	28	281.4	174.7	47.3	China Coast
Ganges- 12.0 44.3 248.0 24 Bramaputra	18.0	24	248.0	44.3	12.0	C
Yellow 22.8 84.3 63.6 91	4	91	63.6	84.3	22.8	-

<sup>\*</sup>The top 10 maize-producing countries and top 10 maize-producing watersheds are based on the maize cultivated areas; a:  $GWF_N$  with 10 mg N  $L^{-1}$  and 0.4 mg N  $L^{-1}$  as standard and natural concentration; b:  $GWF_N$  with 3 mg N  $L^{-1}$  and 0.4 mg N  $L^{-1}$  as standard and natural concentration; c:  $GWF_P$  with 0.15 mg P  $L^{-1}$  and 0.01 mg P  $L^{-1}$  as standard and natural concentration; d:  $GWF_I$  with 3 mg N  $L^{-1}$  and 0.15 mg P  $L^{-1}$  as the standards, and 0.4 mg N  $L^{-1}$  and 0.01 mg P  $L^{-1}$  as natural concentrations for nitrogen and phosphorus.

#### 3.2 Grey water stress assessment

At the grid level, the GWS shows quite similar patterns to GWF (Figs. 1–2). In the high GWF regions, GWS is generally higher than 1. For example, the GWS<sub>N</sub> is higher than 1 in the northeastern parts of China and northeastern parts of the USA (Fig. 2a). The high GWS in these regions implies

that the assimilative capacity of the local water resource has been fully consumed and that a process of water environmental degradation is ongoing. In contrast, the GWS presents an optimistic view at the watershed level, as the average values computed for a watershed may hide the extreme values of GWS at a grid level. Most river basins present GWS lower than 0.3, except GWF<sub>N</sub> and GWS<sub>I</sub> in the Haihe river basin of China.

Fig. 2 about here.

To demonstrate the disparities of GWS due to different water quality standards and natural concentrations, different pollutants, and the integration of different pollutants, the fractions of grid number with GWS > 1 in the total grid number for maize cultivation under different conditions were compared (Fig. 3). The fraction for GWS<sub>N</sub> using the ambient water quality standard (3 mg N L<sup>-1</sup>) is much higher (about 5 times) than that using the drinking water quality standard (10 mg N L<sup>-1</sup>), although the difference in water quality standards is only 3.3 times. Different natural concentrations also contribute to the disparities for both GWS<sub>N</sub> and GWS<sub>P</sub>. The fractions derived from P loads are much higher than those derived from N loads. For example, the fraction is about 2.5% for GWS<sub>N</sub> with 3 mg N L<sup>-1</sup> and 0.4 mg N L<sup>-1</sup> as standard and natural concentration, while it increases to 5.1% for GWS<sub>P</sub> with 0.15 mg P L<sup>-1</sup> and 0.4 mg P L<sup>-1</sup> as standard and natural concentration. When considering the integration of N and P, about 5.8% of total maize cultivation grids show GWS > 1.

Fig. 3 about here.

#### 4. Discussion

The key point addressed in this study is that there are multi-dimensional disparities of GWF assessments that prevent the indicator from providing a consistent value for water quality and being a tool for sound policy supporting in the water sector. It is here argued that improving the robustness of GWF assessments is necessary to gain a more comprehensive understanding of the impacts of human activities on water quality. To the best of our knowledge, this study is the first attempt to comprehensively address the main shortcomings of GWF accounting and to propose ways of improving it.

First, the influence of water quality standards and natural concentration levels on GWF assessment was addressed. The GWF is generally estimated using water quality standards for the receiving freshwater bodies, which can be formulated at the national (e.g., for Switzerland) or regional level (e.g., the EU Water Framework Directive). However, as these standards do not exist for all pollutants and for all countries, it is difficult to make consistent analyses in some circumstances. The choice of whether to use environmental or drinking water quality standards in GWF assessments is also a matter of great importance (Table 3). It is highly recommended that the standards used are verified when performing a comparison with results from different studies. Ambient water quality standards also vary considerably from one substance to another and from one country or region to another. This point needs to be addressed, as the choice of inputs and parameter values determines the major disparities in GWF assessments, as also highlighted by Mekonnen and Hoekstra (2015).

It should be highlighted here that for N, ambient water standards are consistently higher than drinking water standards (Table 2). A number of water scientists from the Swiss Federal Institute for Aquatic Science and Technology (Eawag) have been consulted and a review of the relevant literature for the explanation of these differences has been conducted. It has emerged that fish, shellfish, and smaller organisms that live in water are more sensitive to N than humans are. Their small bodies make them less tolerant to the concentration of the pollutants. Therefore, the use of drinking water standards underestimates GWF (Table 2). The P pollution in freshwater bodies is not directly toxic to humans and animals. This is the main reason that there is generally no drinking water standard for P. It has, however, an indirect toxicity as it causes proliferation of toxic algal blooms and results in the depletion of oxygen, which kills fish (Carpenter et al., 1998). Nitrate loads into water, in contrast, pose a direct threat for the health of humans and other mammals. High concentration of nitrate has been linked to cancers and methemoglobinemia in infants, as well as to toxic effects on livestock (e.g., abortions in cattle).

The assumptions on natural concentration also largely influence the results of GWF analyses. As shown by Liu et al. (2012), changes in natural concentration levels can result in large disparities in GWS assessment. It is also noteworthy to say that it is not easy to account for natural concentration in

a receiving body. Some studies have assumed the natural concentration to be zero. In reality, this cannot be the case, as nutrients are transported through river systems and interact with sediments. As different ecosystems respond differently to nutrient loads, to produce meaningful results, GWF assessments should be based on basin-specific values for both maximum allowable concentration and natural concentration for different nutrients. Experimental studies are needed to support more robust GWF assessments.

The second point raised in the present study is that the co-existence of multiple pollutants in water bodies (e.g. N and P, pesticides, herbicides, heavy metals, and other harmful chemical compounds) needs to be considered in GWF assessments. The majority of GWF assessments only considers anthropogenic N loads to water (e.g., Hoekstra and Mekonnen, 2012; Mekonnen and Hoekstra, 2015; O'Bannon et al., 2014) (Table 1), while neglecting the existence of other pollutants. In this study, an integrated GWF by considering the highest GWF of individual pollutants was explicitly illustrated using N and P losses in global maize cultivation as example. The result demonstrates that it is important to consider P-related loads into water, as the dilution of this input is much more water intensive than the dilution of N-loads. This is because eutrophication in water bodies is more likely caused by P (Schindler et al., 2008). Therefore, the P-related water quality standards are far more stringent than those for N. It can also be argued that, as the pollutant that has the highest GWF is often unknown or cannot be determined in the water body, an aggregated water quality indicator, which takes all possible pollutants into account with a weighting factor, may be more appropriate for GWF accounting. But how to determine such an indicator deserves further investigation.

Third, this study addressed the disparities arising from the use of different spatial scales. The analysis has shown that the good condition of GWS at the basin level hides the potential negative impacts at the grid level. This is because the grid level assessment provides information on local concentration of pollution, whereas the basin level assessment gives the average concentration in the whole basin which can smooth out the high concentration in many local areas. Hence, the river basin level GWF may provide wrong information on the criticality of water pollution situation, impairing sound policy making for water resource management. Though watersheds hold the complete

hydrological processes, it seems that water pollution intensity should be assessed by treating sources of pollution which are located in different parts of a basin. When the water pollution happens in the upstream of a basin, the situation may be different from if it happens in the down steam. For example, the GWS of N at Mississippi is only 0.04–0.08 when taking it as a whole area (Fig. 2). However, a large number of grids in its northeastern parts present GSW larger than 1. It is difficult to use the downstream water resources to deal with the upstream water pollution. Therefore, assessing the GWF and GWS at the grid level should have a higher priority. The grid level was also recommended by Mekonnen and Hoekstra (2014), where they compared the green and blue water footprint at the grid, provincial, and country levels. For this reason, the current study focused on the grid level for evaluating the disparities of GWS as shown in Fig. 3.

Finally, it should be mentioned that this study did not consider accumulation and degradation of the lost N and P in the receiving water bodies. Also, the effects of different forms of lost N and P on the aquatic ecosystem were not considered. This is because it is difficult to determine exact loads of individual forms in which the lost N and P present in water bodies due to the complex chemical and biological processes involved. For this reason, studies in grey water footprint so far have used a certain element as a proxy for N and P loads, e.g., total losses to water regardless forms and fates of the pollutant (such as in this study and many other studies, including Hoekstra and Mekonnen (2012) and Mekonnen and Hoekstra (2015)). Only few have used dissolved inorganic N and P, dissolved organic N and P, and total N and P (Liu et al., 2012). It is clear that using different forms of a certain pollution element will lead to different results of GWF. However, a detailed analysis of the fate of the lost N and P and the involved chemical and biological processes is beyond the scope of this study, but deserves further investigation in the future.

#### 5. Conclusions

This study has discussed some major limitations of GWF assessment in the literature and proposed several ways towards its improvement. It provides a first exploration on how the GWF assessment can be enhanced from a methodological point of view, with an illustration for N and P loads into water relating to global maize cultivation. The insights provided can be used to investigate GWF from

multiple crops and pollutants and water quality parameters. The approach taken in this study can also be used to provide a scientifically sound frame of reference for evaluating the trade-offs of importing rather than producing different crops in different locations from both a water quality and water quantity perspective. Finally, improving the scientific soundness of GWF assessments can better inform national and local governments about the pressure on water systems from agrochemical pollutants and serve as a tool for setting water pollution reduction targets based on the consumption of assimilation capacity of water bodies.

### Acknowledgements

The authors would like to thank Dr. Nele Schuwirth, Dr. Kai Ubert, and Dr. Marion Junghans from Eawag, and Dr. Els Smit from the Centre for Safety of Substances and Products, National Institute for Public Health and the Environment of the Netherlands for the scientific support, particularly for clarification of drinking and surface water quality standards and the knowledge underlying these standards. We also thank four anonymous reviewers' constructive comments for improving the quality of the initial manuscript.

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## Figure captions

Fig. 1 Grey water footprint (GWF) related to nitrogen (GWF<sub>N</sub>), phosphorus (GWF<sub>P</sub>), and their integration (GWF<sub>I</sub>) in maize production by using 3 mg N  $L^{-1}$  and 0.15 mg P  $L^{-1}$  as the standards, and 0.4 mg N  $L^{-1}$  and 0.01 mg P  $L^{-1}$  as natural concentrations for nitrogen and phosphorus.

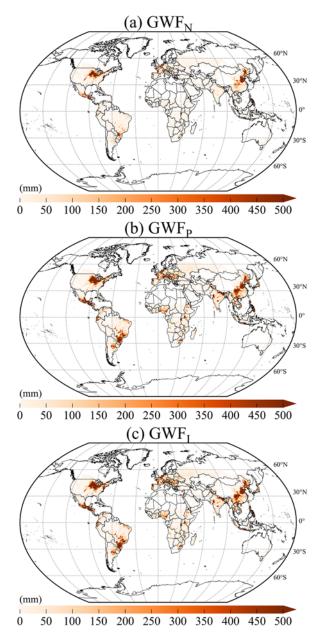


Fig. 2 Grey water stress (GWS) related to nitrogen (GWS<sub>N</sub>), phosphorus (GWS<sub>P</sub>), and their integration (GWS<sub>I</sub>) in maize production at grid and watershed levels by using 3 mg N  $L^{-1}$  and 0.15 mg P  $L^{-1}$  as standards, and 0.4 mg N  $L^{-1}$  and 0.01 mg P  $L^{-1}$  as natural concentrations for nitrogen and phosphorus.

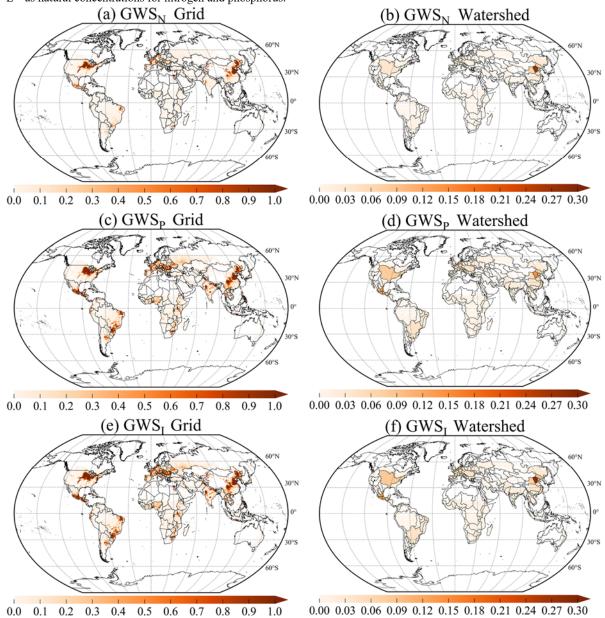


Fig. 3 Fraction of grid number with grey water stress (GWS) higher than 1 in total grid number of maize cultivation. For X axis, N10\_0.4 refers to GWS related to N (GWS<sub>N</sub>) with 10 mg N L<sup>-1</sup> and 0.4 mg N L<sup>-1</sup> as standard and natural concentration; N3\_0.4 refers to GWS<sub>N</sub> with 3 mg N L<sup>-1</sup> and 0.4 mg N L<sup>-1</sup> as standard and natural concentration; N3\_0 refers to GWS<sub>N</sub> with 10 mg N L<sup>-1</sup> and 0 mg N L<sup>-1</sup> as standard and natural concentration; P0.15\_0.01 refers to GWS related to P (GWS<sub>P</sub>) with 0.15 mg P L<sup>-1</sup> and 0.01 mg P L<sup>-1</sup> as standard and natural concentration; P0.15\_0 refers to GWS<sub>P</sub> with 0.15 mg P L<sup>-1</sup> and 0 mg P L<sup>-1</sup> as standard and natural concentration; integrated GWS (GWS<sub>I</sub>) with 3 mg N L<sup>-1</sup> and 0.15 mg P L<sup>-1</sup> as the standards, and 0.4 mg N L<sup>-1</sup> and 0.01 mg P L<sup>-1</sup> as natural concentrations for N and P.

