

Supporting Information for "Economically Efficient and Environmentally Sustainable Irrigation Potentials: a Spatially Explicit Global Assessment"

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Introduction

This Supporting Information provides additional details to the methods applied in the article.

Additionally, country-level results of irrigation potentials (potentially irrigated areas (PIA); potential irrigation water consumption (PIWC); potential irrigation water withdrawals (PIWW)) for 235 countries are provided for two model setups as separate files: (a) under consideration of currently irrigated areas that affect the river flow; (b) not considering currently irrigated areas for irrigation potentials that are purely determined by economic considerations.

1. Detailed Methods: LPJmL Model Description

The Lund-Potsdam-Jena managed Land (LPJmL) model is a spatio-temporally explicit process-based model that simulates the growth and geographical distribution of 11 plant functional types (natural vegetation) and 12 crop functional types (field crops) and additionally pasture as well as (woody and herbaceous) bioenergy crops. It accounts for feedbacks between vegetation, the global terrestrial water, carbon, and nitrogen cycles, and energy fluxes (von Bloh et al., 2018; Schaphoff, von Bloh, et al., 2018; Lutz et al., 2019). The model simulates the terrestrial water balance considering precipitation, snow melt, seepage, interception, plant transpiration and soil evaporation resulting in daily simulations of runoff and discharge and considers its close interactions with plant vegetation in terms of plant growth and productivity that is linked to soil and atmospheric

moisture (Schaphoff, Forkel, et al., 2018). It delivers consistent estimates for spatially explicit irrigated and rainfed potential crop yields, plant water uptake and surface runoff that are the basis for our model. Evaporation of irrigation water during the growing season is calculated based on the fraction of irrigation water in soil moisture and canopy interception (Rost et al., 2008).

The crop types considered in LPJmL are mapped to the crop types considered in our analysis using the following mapping (see table S1). Since LPJmL considers fewer crops than our analysis, LPJmL's groundnut is also the proxycrop for both oilpalm and cotton; maize is also the proxycrop for fodder (forage) and the 'other' crop category including fruits, vegetables and nuts; temperate roots represent both sugar beet and potatoes.

Irrigated and rainfed crop yields as well as consumptive blue water requirements are provided by LPJmL5 with unlimited nitrogen supply (von Bloh et al., 2018). As opposed to previous LPJmL versions (Sitch et al., 2003; Bondeau et al., 2007; Schaphoff, von Bloh, et al., 2018; Schaphoff, Forkel, et al., 2018), LPJmL5 includes an implementation of the global terrestrial nitrogen cycle and consistently accounts for water, grassland and crop management. Since the LPJmL4 and LPJmL5 model version have diverged during the Nitrogen cycle implementation phase, certain natural vegetation dynamics (Forkel et al., 2014) have not yet been included in the newest LPJmL5 version (von Bloh et al., 2018). For this reason, natural vegetation inputs, such as lake evaporation, runoff and monthly discharge are provided by its predecessor LPJmL4 (Schaphoff, von Bloh, et al., 2018; Schaphoff, Forkel, et al., 2018).

2. Detailed Methods: Yield Value Gain Potential

The difference of irrigated and rainfed crop yields as estimated by LPJmL provides the yield gain through irrigation in tons of dry matter. Negative yield gains (irrigated yield < rainfed yield) are technically possible because irrigation may lead to a shift in the growing period in LPJmL resulting in lower irrigated yields. In such cases, the irrigation yield gain is set to 0.

To account for country-specific management effects on yields (e.g. fertilizer and pesticide use; different crop varieties; mechanization; cropping intensity representing multiple cropping or fallow land), LPJmL potential yields are calibrated to meet country-level production as reported by FAO (FAO, 2021) using a multiplicative factor for both rainfed and irrigated yields. Note that both rainfed and irrigated yields are calibrated to FAO country-levels. A potential multiple cropping effect is therefore applied to both irrigated and rainfed yields and cannot capture the effect that irrigation may lead to an additional cropping season and increase yields by one or two additional harvests per year.

Figure S1 shows the potential yield value gain through irrigation in USD per hectare for the historical crop mix in 2010. It represents the areas that would achieve yield gains through irrigation considering irrigated and rainfed potential yields valued at global FAO average prices (in USD per tDM).

3. Reserved Current Agricultural Uses

To derive the grid cell area (in Mha) that was irrigated in the year 2010, we use the irrigated area share provided by the Land-Use Harmonization 2 (LUH2) data set presented in Hurtt et al. (2020) (Hurtt et al., 2019, 2020). LUH2 is based on the HYDE 3.2 data set (Klein Goldewijk et al., 2017) that estimates historically irrigated areas based on Siebert et al. (2015), Portmann, Siebert, and Döll (2010) and FAOSTAT data (FAO,

2021). To obtain grid cell specific crop area for the 19 crop types used in our analysis, we combine the spatially explicit LUH2 cropland map with national crop-type specific data from FAOSTAT that provides country-level harvested areas of crop items.

The LUH2 cropland map is subdivided into only five crop functional types (C3 annuals; C4 annuals; C3 perennials; C4 perennials; C3 nitrogen fixers). These five functional types are further disaggregated into crop groups using relative shares of area harvested on country level from FAOSTAT. Because rice plays a special role in terms of irrigation as well as greenhouse gas accounting, the spatial distribution of rice areas is especially important. We therefore determine the distribution of physical rice areas by assigning the country's rice production first to flooded areas provided at cellular level by LUH2. Upland (aerobic) rice is accounted by distributing country-level FAO rice areas beyond country-aggregated LUH2 flooded area (i.e. where FAO reports higher country-level rice areas than there are LUH2 flooded areas in the respective country) equally across the remaining country's cropland area. Note that flooded areas are not accounted as irrigated areas. For one, because flooded rice production is often only partially irrigated with blue water and often just retains the rainwater in paddies (Klein Goldewijk et al., 2017; Hurtt et al., 2020) and also because flooding fulfills a special management purpose in terms of pest control (Ampong-Nyarko & De Datta, 1991).

Given the area irrigated and the crop pattern in 2010 derived from LUH2 and FAOSTAT, the volume of current cellular irrigation water use ($U_{c,w}$) is calculated (see equation 1).

$$U_{c,w} = \sum_k V_{c,k,w} \cdot A_{c,k} \quad (1)$$

where $V_{c,k,w}$ refer to the crop water requirements per crop type (k) and grid cell (c) for the two water use types (w = consumption and withdrawal), $A_{c,k}$ is the irrigated area per grid cell and crop.

4. Environmental Flow Requirements

The share of yearly discharge to be reserved for EFR per grid cell is calculated over a long-term reference period (1985-2015) based on monthly discharge provided by LPJmL4 (Schaphoff, von Bloh, et al., 2018). For a functioning freshwater ecosystem, a certain base flow (low flow requirements, LFR) is necessary to avoid aquatic species loss. Additionally, flooding plays an important role for riverine vegetation and wetlands. It can be accounted for by high flow requirements (HFR) (Smakhtin et al., 2004). We follow the Variable Monthly Flow (VMF) method introduced by Pastor, Ludwig, Biemans, Hoff, and Kabat (2014). It determines EFR using the flow variation throughout a year with different requirements for low-, intermediate- and high-flow months parametrized to a ‘fair’ ecosystem preservation status. In low-flow months (i.e. months in which mean monthly flow is smaller or equal to 40 % of the mean annual flow), 60 % of mean monthly flows are reserved for the environment; in intermediate-flow months (i.e. months in which mean monthly flow is greater than 40 %, but smaller than 80 % of the mean annual flow) 45 %; and in high-flow months (i.e. months in which mean monthly flow is greater than 80 % of the mean annual flow) 30 % of mean monthly flows is reserved (Pastor et al., 2014). We adopted this method by splitting EFR into LFR and HFR-equivalents. Discharge reserved in low-flow months is attributed to LFR, discharge reserved in high-flow months is attributed to HFR, and half of intermediate-flow requirements are attributed to LFRs and

the other half to HFRs to appropriately consider the interaction of EFR and inaccessible discharge.

Not all water on Earth can easily be brought into productive use (Postel et al., 1996; de Fraiture et al., 2001). Especially highly variable flows are difficult to access for humans and could only be used for irrigation with appropriate storage infrastructure (reservoirs), which are costly to install. To account for such inaccessible (or hardly accessible) discharge, we use the coefficient of variation (CV) of monthly discharge over a reference period of 30 years (here: 1980-2010) assuming a functional relationship that leads to a decrease in accessibility with increasing long-term seasonal variability of discharge (see equation 2).

$$a_c = 2^{\frac{\sigma_c}{\mu_c}} \quad (2)$$

where a_c is the share of discharge in cell c that can be accessed, σ is the standard deviation of long-term monthly discharge in cell c and μ is the mean discharge of cell c over the same long-term period. The CV is the ratio of the two ($\frac{\sigma_c}{\mu_c}$). With the monthly discharge time series provided by LPJmL4, the CV ranges between 0 and 19.11 resulting in a functional form as displayed in figure S2b). The bulk of the data lies between 0 and 3.61 with the 25th percentile at 1.08 and the 75th percentile at 2.09 (see figure S2a). Only a few grid cells show discharge variability that results in complete inaccessibility.

We assume that seasonally highly variable flows are difficult to access by humans, but may serve an ecosystem function similar to HFRs. The baseflow or LFR, on the other hand, cannot be served by such variable flows and must be left untouched by human intervention when the environmental flow protection is considered. For this reason, we split discharge reserved for EFR into HFRs and LFRs. When discharge is constrained

based on the accessibility constraint, HFRs count towards the inaccessible discharge, while LFRs are excluded from human access in addition to inaccessible discharge.

Natural Land Protection

The following map shows areas of ecological importance following the Half-Earth protection approach based on the data provided by Kok et al. (2020). It includes currently protected areas based on the World Database of Protected Areas (WDPA), biodiversity hotspots (Mittermeier et al., 2005) and intact forest landscapes (Potapov et al., 2017). On top of these areas, at least 50 % of the land surface of each ‘ecoregion’ as described in Dinerstein et al. (2019) is protected.

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Crop types considered in LPJmL	Crop types considered in this study
Temperate cereals	Temperate cereals
Tropical cereals	Tropical cereals
Maize	Maize; Others (fruits, vegetable, nuts); Forage
Rice	Rice
Oil crops (soybean)	Soybean
Oil crops (rapeseed)	Other oil crops (including rapeseed)
Oil crops (groundnut)	Groundnuts; Oilpalms; Cotton
Oil crops (sunflower)	Sunflower
Pulses	Pulses
Temperate roots	Potatoes; Sugar beet
Tropical roots	Tropical roots (including cassava)
Sugar cane	Sugar cane
Biomass grass	Short rotation grasses
Biomass trees	Short rotation trees

Table S1. Mapping of LPJmL crop types to crop types considered in our analysis.

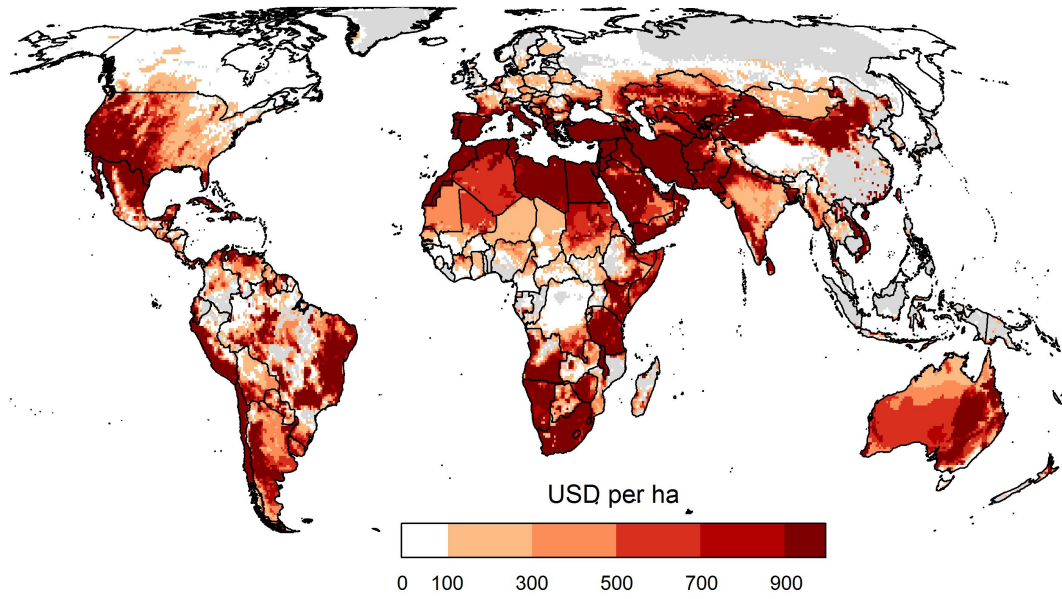


Figure S1. Potential yield value gain through irrigation in USD ha⁻¹. Areas in grey have a yield value gain of 0.

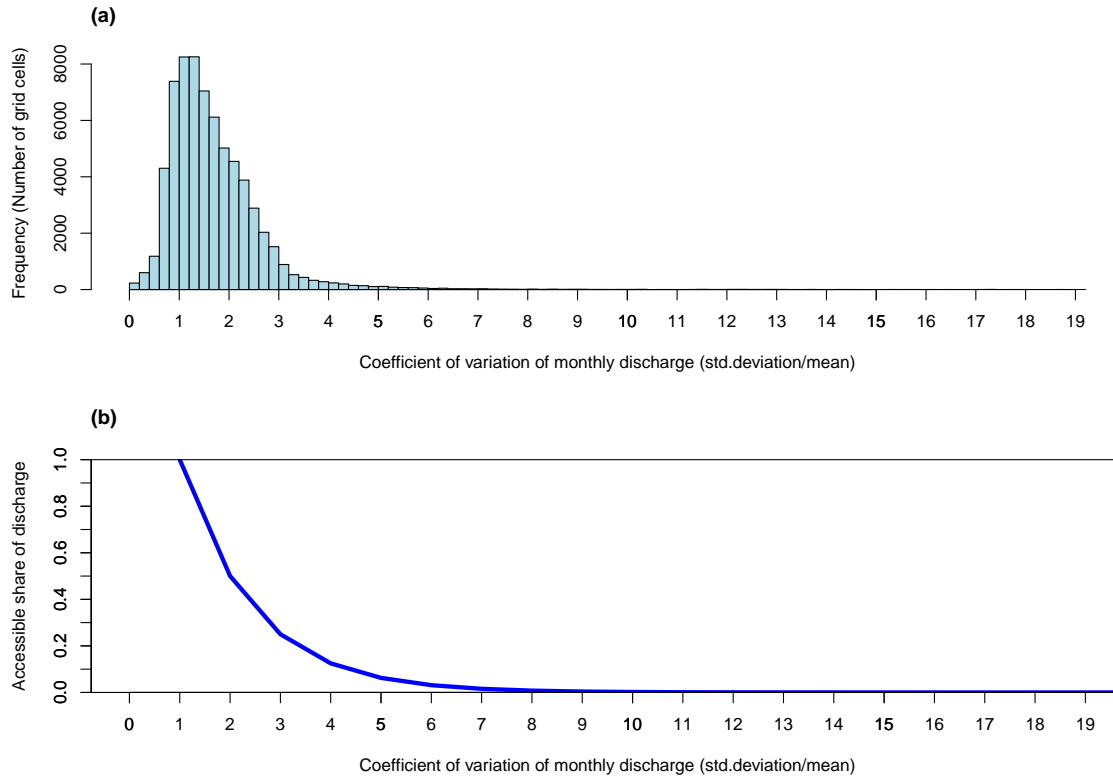


Figure S2. Frequency Coefficient of Variation of monthly discharge (a) and functional relationship between discharge accessibility share and coefficient of variation of discharge for time series of monthly discharge over the period from 1980 to 2010.

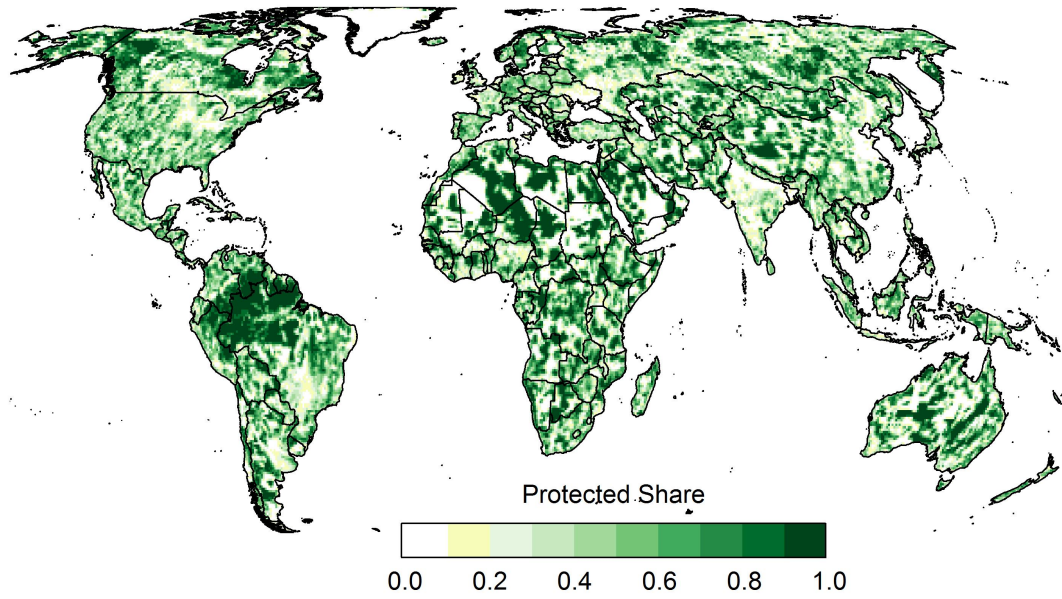


Figure S3. Share of grid cell that would be protected according to the Half-Earth protection approach.