

1 Dear editor,

2 Thank you very much for handling our manuscript again. Below we provided a list of all the relevant
3 changes made in the manuscript. The referee responses and a mark-up of the new manuscript version
4 are also included.

5

6 **Referee response**

7 Several textual changes have been made in response to referee comments and suggestions. This includes
8 minor changes to the abstract and conclusions and changes to the results section.

9

10 **Other changes**

- 11 1. Figure 3 caption was wrong and has been adjusted
12 2. Figure 9 outlining was improved
13 3. Code and documentation availability section was updated with new DOI's
14 4. Removed some double spaces

15

16 We hope this list (and attached referee responses and manuscript mark-up) sufficiently describes the
17 manuscript changes made.

18 Sincerely,

19 Bram Droppers on behalf of all co-authors

1 **Referee 1 - Author response**

2 Dear referee,

3

4 Thank you very much for reviewing our paper again. We are pleased that our adjustments the manuscript
5 have been received favourably. Below we address your comments and suggestions (shown in italic),
6 with our responses in blue.

7

8 **Motivations for irrigation efficiency**

9 *“I have just noticed that the authors may have misunderstood one of my comments (...) I wanted to point
10 out that e [irrigation efficiency] would work only if the authors' estimations of C [consumptive water
11 uses] were by chance quite similar to that of Frenken and Gillet (2012). I know this is a question without
12 answer, but at least, if the authors have any concrete idea why they chose Frenken and Gillet (2012) it
13 should be worth mentioned here.”*

14 We agree with the referee that the irrigation efficiency is heavily dependent on the simulated irrigation
15 water consumption (crop evapotranspiration). There have been various other assessments of irrigation
16 efficiency, which all used different methods and models (e.g. Döll and Siebert, 2002; Rohwer et al.,
17 2007; Jägermeyr et al., 2015)

18 We decided to select Frenken and Gillet (2012) since it is a relatively recent and comprehensive study
19 with a high resolution crop modelling basis (5 by 5 arc-minute spatial resolution). Information regarding
20 crop growing areas and seasons were gathered from the same database as the reported crop irrigation
21 water withdrawal, making these values comparable and consistent (to an extent). Therefore, we found
22 this study to be suitable in estimating the actual irrigation efficiency on a aggregated scale.

23 Lines 255-260: “The water loss fraction was based on Frenken and Gillet (2012), who estimated the
24 irrigation efficiency for 22 United Nations sub-regions. Irrigation efficiencies were estimated based on
25 the differences between the calculated crop water requirements (crop evapotranspiration; consumptive
26 water use) and the reported irrigation water withdrawals (including transportation and application
27 losses). Crop water requirements are estimated based on the FAO Irrigation and Drainage paper (Allen
28 et al., 1998). Low irrigation efficiencies can result in irrigation water withdrawals up to four times higher
29 than the crop water requirements in regions such as east- and west Africa.”

30 Will change to: “The water loss fraction was based on Frenken and Gillet (2012), who estimated the
31 aggregated irrigation efficiency for 22 United Nations sub-regions. Irrigation efficiencies were
32 estimated based on the difference between AQUASTAT reported irrigation water withdrawals and

33 calculated irrigation water requirements (Allen et al., 1998), using data on crop information (e.g.
34 growing season, harvest area) from AQUASTAT.”

35

36 We hope the referee agrees with our response, and are open to any further suggestions or comments.

37 Sincerely,

38 Bram Droppers on behalf of all co-authors

39

40 **References**

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42 8-10, 2002.

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47 Sciences*, 19, 2015.

48 Rohwer, J., Gerten, D., and Lucht, W.: Development of functional irrigation types for improved global
49 crop modelling, PIK, 2007.

50

1 **Referee 2 - Author response**

2 Dear referee,

3

4 Thank you very much for reviewing our paper again. We are pleased that our adjustments to the
5 manuscript have been received favourably. Below we address your comments and suggestions (shown
6 in italic), with our responses in blue.

7

8 **Unsupported statement – need clarification**

9 *-L32 and L : "decrease computations times compared to previous versions" – specify the version*
10 *(Haddeland et al. 2006 and pre-VIC5 routing model). This achievement is not a novelty of this*
11 *combination of models only, this is a feature of Hanasaki et al. generic operating rules that does not*
12 *require foresight with respect to Haddeland et al. 2006. The added modules leverage existing*
13 *approaches and their implementation is not computationally more efficient than the other existing*
14 *generic approaches, or at least this is not demonstrated.*

15 *-L524, same as L32 – unsupported statement unless you mention that the previous version is Haddeland*
16 *et al. (2006).*

17 We will more clearly indicate that the decrease in computation time is due to the reduction of the number
18 of model runs needed to simulate human-impacted hydrology, under Haddeland et al. (2006), not a
19 (demonstrated) more efficient approach or implementation. We will also more clearly refer to Haddeland
20 et al. (2006) as the version we are comparing to.

21 We would like to note that the decrease in computation times does not (primarily) relate to the use of
22 the Hanasaki et al. (2006) generic reservoir operation scheme as opposed to the Haddeland et al. (2006)
23 retrospective reservoir optimization scheme. The decrease in computation times is related to the space-
24 before-time processing order of VIC-5. Therefore, multiple successive model simulations are no longer
25 needed to initialize the irrigation water demands (under Haddeland et al. (2006); see further comments
26 below). The following text revisions will be made:

27 Lines 31-32: "The additions presented here make the VIC model more suited for fully-integrated
28 worldwide water-resource assessments and substantially decrease computation times compared to
29 previous versions."

30 Will change to: "The additions presented here make the VIC model more suited for fully-integrated
31 worldwide water-resource assessments."

32 Lines 522-524: “The additions presented here make the VIC model more suited for fully-integrated
33 worldwide water-resource assessments and substantially decrease computation times compared to
34 previous versions.”

35 Will change to: “The additions presented here make the VIC model more suited for fully-integrated
36 worldwide water-resource assessments and substantially decrease computation times compared to
37 Haddeland et al. (2006b).”

38

39 **Most approaches are not new and still need up-front reference.**

40 - L83-84: *important to specify that the Haddeland et al. concept relies on foresight, which is a roadblock*
41 *to full integration in hydrology models, and explain the need for multiple runs. Without the explanation*
42 *that Haddeland representation relies on foresight, the sentence mentioning that multiple runs are needed*
43 *is unclear and does not clearly lead to the conclusion of a lower computation time.*

44 As mentioned above, we think the main roadblock for full hydrological integration is the time-before-
45 space processing order of pre-VIC5 model versions. We will more clearly indicate this distinction in the
46 Introduction (Section 1; see further comments below).

47 Lines 82 – 86: “Lastly, while the model setup of Haddeland et al. (2006b) already included important
48 anthropogenic impact modules (i.e. irrigation and dam operation), these were not fully integrated yet.
49 Therefore multiple successive model runs were required which was computationally expensive,
50 especially for global water resources assessments.”

51 Will change to: “Lastly, while the model setup of Haddeland et al. (2006b) already included important
52 anthropogenic impact modules (i.e. irrigation and dam operation), these were not fully integrated yet.
53 Therefore multiple successive model runs were required (see Section 2.1) which was computationally
54 expensive, especially for global water resources assessments.”

55

56 -L135: *add reference to the new routing model. Note that the new routing model is not motivated by the*
57 *image mode of VIC5, but is associated with the Hanasaki et al. generic release rules and withdrawals*
58 *processes at each time step, which is not compatible with the existing VIC5 Lohman et al. routing model.*
59 *Previous applications of the Hanasaki et al. rules with VIC used a different routing model (MOSART)*
60 *for those exact reasons. Those explanations were provided in Voisin et al. (2013). The existing set up*
61 *extends the Voisin et al. (2013) set up because this VIC-WUR is customized the VIC5 to facilitate 2-way*
62 *coupling (i.e. irrigation water demand and link to groundwater component) the same way that a branch*
63 *of MOSART-WM is not integrated in E3SM to better represent those same processes which are*
64 *differently represented in CLM and VIC. Such framing would better emphasize the novelty in this new*
65 *framework.*

66 We will add further references for the new routing model, which is still based on the equations of
67 Lohmann et al. (1996).

68 The motivation for the new routing module was not primarily motivated by the Hanasaki et al. (2006)
69 generic reservoir operation scheme. The model implementation (including routing) of Haddeland et al.
70 (2006) would have allowed us to run a generic reservoir operation scheme with the Lohmann et al.
71 (1996) post-process routing implementation. However, this would require an irrigation initialization in
72 the form of several additional model runs (Haddeland et al., 2006) or an independent offline irrigation
73 demand dataset (Voisin et al., 2013). Therefore we used VIC-5 as the basis for our model development.
74 This will be made more clear in the Methods (Section 2.1).

75 Lines 132 – 136: “Therefore, water withdrawals could not be taken into account directly and studies
76 using the model setup of Haddeland et al. (2006b) required multiple successive model runs. Since VIC-
77 5 uses the space-before-time processing order, runoff routing could be simulated each timestep. The
78 routing post-process was replaced by our newly developed routing module which simulates routing
79 sequentially (upstream-to-downstream) to facilitate water withdrawals between cells.”

80 Will change to: “In order for reservoirs to account for downstream water demand, an irrigation demand
81 initialization was required. This initialization could either be an independent offline dataset (Voisin et
82 al., 2013) or multiple successive model runs (Haddeland et al. 2006b). Since VIC-5 uses the space-
83 before-time processing order, irrigation water demands and runoff routing could be simulated each
84 timestep. The routing post-process was replaced by our newly developed routing module, which
85 simulates routing sequentially (upstream-to-downstream) based on the Lohmann et al. (1996)
86 equations.”

87

88 *-L138: Most are not “newly developed” modules, rather adapted modules. Please revise, or be more
89 specific.*

90 Lines 138-139: “VIC-WUR extends the existing VIC-5 though the addition of several newly developed
91 anthropogenic-impact modules (Figure 1).”

92 Will change to: “VIC-WUR extends the existing VIC-5 though the addition of several newly
93 implemented anthropogenic-impact modules (Figure 1).”

94

95 *- L233 – sectoral water demands – please add reference to leveraged approaches. References are
96 presently mostly in supplemental material.*

97 We are uncertain which references the referee is referring to here. Approach references for the domestic
98 and industrial sector are present in Section 2.3.2 (Line 269: “(...) estimated similar to Alcamo et al.
99 (2003).”; Lines 272-273: “(...) estimated similar to Flörke et al. (2013) and Voß and Flörke (2010)”.

100 Approach references for the energy and livestock sector are present in Section 2.3.3 (Lines 288-290:
101 “(...) estimated using data from van Vliet et al. (2016). Water use intensity for generation (i.e. the water
102 use per generation unit) was estimated based on the fuel and cooling system type (Goldstein and Smith,
103 2002), (...)”; Lines 296-298: “(...) estimated by combining the Gridded Livestock of the World (GLW3)
104 map (Gilbert et al., 2018) with the livestock water requirement reported by Steinfeld et al. (2006).”). To
105 avoid repeating use of references we have omitted these references from the introduction to Section 2.3
106 (Lines 233-242).

107

108 **Model evaluation**

109 *section 3.2.1 – the differences in sectoral water withdrawals are very large, not only in seasonal*
110 *variations but also in long term trends. What are the next steps for this set up? Given the evaluation,*
111 *what are the recommendations for upcoming energy-water-land analytics?*

112 Our setup was implemented to be able to support future water demand estimations using simulated future
113 population, DGP and GVA. There are several other sectoral water demand setups (e.g. Alcamo et al.,
114 2003; Vassolo and Döll, 2005; Shen et al., 2008; Hanasaki et al., 2013; Wada and Bierkens, 2014),
115 which will have to be compared to our approach. The choice of which approach to take is heavily
116 determined by analysis level (global, regional) and data-availability for calibration. However, as the
117 current model setup allows for user-defined (gridded) inputs of sectoral water withdrawals, withdrawals
118 could be replaced by any other estimation (such as those of Huang et al. (2018) for the historic period).

119 While we will include further recommendations for upcoming for the sectoral water withdrawal setup,
120 the manuscript will mostly refrain from any commenting on upcoming energy-water-land analytics. We
121 think that the model setup as presented currently requires further development for integrated energy-
122 water-land analytics. For example, while the current model setup is able to estimate the effects of
123 anthropogenic water use on water availability (and water stress), this is not translated to actual impacts.
124 This is currently our primary focus.

125 Further efforts are also needed to better represent the energy sector, especially for energy-water-land
126 analysis. Currently only a small part of the energy sector is included, while this is an important sector
127 in the energy-water-land nexus. Lastly, as the irrigation sector is integrated into the hydrological model,
128 which will be part continuous model development.

129 Lines 510-511: “However, note that the model setup of VIC-WUR allows for the evaluation of other
130 sectoral water demand inputs, on various temporal aggregations.”

131 Will change to: “While the current setup to estimate sectoral water demands is well suited for future
132 water withdrawal estimations, there are various other approaches (e.g. Alcamo et al., 2003; Vassolo and
133 Döll, 2005; Shen et al., 2008; Hanasaki et al., 2013; Wada and Bierkens, 2014). As the model setup of

134 VIC-WUR allows for the evaluation of other sectoral water demand inputs (on various temporal
135 aggregations), several different approaches can be used depending on the focus region and data-
136 availability for calibration.”

137 Line 378: “(...) was limited before 1990.”

138 Will change to: “(...) was limited before 1990. Also, data on the disaggregation of industrial sectors (e.g.
139 energy and mining) was limited, which can be important sectors in the water-food-energy nexus.”

140

141 We hope the referee agrees with the changes made, and are open to any further suggestions or comments.

142 Sincerely,

143 Bram Droppers on behalf of all co-authors

144

145 **References**

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173

1 **Referee 3 - Author response**

2 Dear referee,

3

4 Thank you very much for reviewing our paper again. We are very pleased that our adjustments have
5 been received satisfactory.

6

7 Sincerely,

8 Bram Droppers on behalf of all co-authors

1 **Simulating human impacts on global water resources using**
2 **VIC-5**

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10 *Correspondence to:* Bram Droppers (bram.droppers@wur.nl)

11 **Abstract.** Questions related to historical and future water resources and scarcity have been addressed
12 by several macro-scale hydrological models. One of these models is the Variable Infiltration Capacity
13 (VIC) model. However, further model developments were needed to holistically assess anthropogenic
14 impacts on global water resources using VIC.

15 Our study developed VIC-WUR, which extends the VIC model with: (1) integrated routing, (2) surface
16 and groundwater use for various sectors (irrigation, domestic, industrial, energy and livestock), (3)
17 environmental flow requirements for both surface and groundwater systems, and (4) dam operation.
18 Global gridded datasets on sectoral demands were developed separately and used as an input to the VIC-
19 WUR model.

20 Simulated national water withdrawals were in line with reported FAO national annual withdrawals (R^2
21 adjusted > 0.8), both per sector as well as per source. However, trends in time for domestic and industrial
22 water withdrawal were mixed compared to other previous studies. GRACE monthly terrestrial water
23 storage anomalies were well represented (global mean RMSE of 1.9 and 3.5 for annual and interannual
24 anomalies respectively), while groundwater depletion trends were overestimated. The implemented
25 human impact modules increased simulated streamflow performance for 370 out of 462 human-
26 impacted GRDC monitoring stations, mostly due to the effects of reservoir operation. An assessment of
27 environmental flow requirements indicates that global water withdrawals have to be severely limited
28 (by 39 %) to protect aquatic ecosystems, especially for groundwater withdrawals.

29 VIC-WUR has potential for studying impacts of climate change and anthropogenic developments on
30 current and future water resources and sectoral specific-water scarcity. The additions presented here
31 make the VIC model more suited for fully-integrated worldwide water-resource assessments ~~and~~

32 ~~substantially decrease computation times compared to previous versions.~~

Commented [DB1]: In response to referee comments

33 **1 Introduction**

34 Questions related to historical and future water resources and scarcity have been addressed by several
35 macro-scale hydrological models over the last few decades (~~Liang et al., 1994; Alcamo et al., 1997;~~
36 ~~Hagemann and Gates, 2001; Takata et al., 2003; Krinner et al., 2005; Bondeau et al., 2007; Hanasaki et~~
37 ~~al., 2008a; Van Beek and Bierkens, 2008; Best et al., 2011~~)(Liang et al., 1994; Alcamo et al., 1997;
38 ~~Hagemann and Gates, 2001; Takata et al., 2003; Krinner et al., 2005; Bondeau et al., 2007; Hanasaki et~~
39 ~~al., 2008b; Van Beek and Bierkens, 2008; Best et al., 2011~~). Early efforts focussed on the simulation of
40 natural water resources and the impacts of land cover and climate change on water availability (Oki et
41 al., 1995; Nijssen et al., 2001a; Nijssen et al., 2001b). Recently, a larger focus has been on incorporating
42 anthropogenic impacts, such as water withdrawals and dam operations, into water resource assessments
43 (Alcamo et al., 2003; Haddeland et al., 2006b; Biemans et al., 2011; Wada et al., 2011b; Hanasaki et al.,
44 2018).

45 Global water withdrawals increased eight-fold over the last century and are projected to increase further
46 (Shiklomanov, 2000; Wada et al., 2011a). Although water withdrawals are only a small fraction of the
47 total global runoff (Oki and Kanae, 2006), water scarcity can be severe due to the variability of water in
48 both time and space (Postel et al., 1996). Already severe water scarcity is experienced by two-thirds of
49 the global population for at least part of the year (Mekonnen and Hoekstra, 2016). To stabilize water
50 availability for different sectors (e.g. irrigation, hydropower, and domestic uses) dams and reservoirs
51 were built, which are able to strongly affect global river streamflow (Nilsson et al., 2005; Grill et al.,
52 2019). In addition, groundwater resources are being extensively exploited to meet increasing water
53 demands (Rodell et al., 2009; Famiglietti, 2014).

54 One of widely-used macro-scale hydrological models is the Variable Infiltration Capacity (VIC) model.
55 The model was originally developed as a land-surface model (Liang et al., 1994), but has been mostly
56 used as a stand-alone hydrological model (Abdulla et al., 1996; Nijssen et al., 1997) using an offline
57 routing module (Lohmann et al., 1996; Lohmann et al., 1998a, b). Where land-surface models focus on
58 the vertical exchange of water and energy between the land surface and the atmosphere, hydrological
59 models focus on the lateral movement and availability of water. By combining these two approaches,

Commented [DB2]: Updated references

60 VIC simulations are strongly process-based and this, in turn, provides a good basis for climate-impact
61 modelling.

62 VIC has been used extensively in studies ranging from: coupled regional climate model simulations
63 (Zhu et al., 2009; Hamman et al., 2016), combined river streamflow and water-temperature simulations
64 (van Vliet et al., 2016), hydrological sensitivity to climate change (Hamlet and Lettenmaier, 1999;
65 Nijssen et al., 2001a; Chegwidan et al., 2019), global streamflow simulations (Nijssen et al., 2001b),
66 sensitivity in flow regulation and redistribution (Voisin et al., 2018; Zhou et al., 2018), and real-time
67 drought forecasting (Wood and Lettenmaier, 2006; Mo, 2008). Several studies used VIC to simulate the
68 anthropogenic impacts of irrigation and dam operation on water resources (Haddeland et al., 2006a;
69 Haddeland et al., 2006b; Zhou et al., 2015; Zhou et al., 2016) based on the model setup of Haddeland et
70 al. (2006b). However, further developments were needed to holistically assess anthropogenic impacts
71 on global water resources using VIC (Nazemi and Wheeler, 2015a, b; Döll et al., 2016; Pokhrel et al.,
72 2016).

73 Firstly, the VIC model did not yet include groundwater withdrawals or water withdrawals from
74 domestic, manufacturing and energy (thermoelectric) sources. Although these sectors use less water than
75 irrigation (Shiklomanov, 2000; Grobicki et al., 2005; Hejazi et al., 2014) they are locally important
76 actors (Gleick et al., 2013), especially for the water-food-energy nexus (Bazilian et al., 2011). Sufficient
77 water supply and availability are essential for meeting a range of local and global sustainable
78 development goals related to water, food, energy and ecosystems (Bijl et al., 2018). Secondly,
79 environmental flow requirements (EFRs) were often neglected (Pastor et al., 2014), even though they
80 are “necessary to sustain aquatic ecosystems which, in turn, support human cultures, economies,
81 sustainable livelihoods, and well-being” (Brisbane Declaration, 2017). Anthropogenic alterations
82 already strongly affect freshwater ecosystems (Carpenter et al., 2011), with more than a quarter of all
83 global rivers experiencing very high biodiversity threats (Vorosmarty et al., 2010). By neglecting EFRs,
84 sustainable water availability for anthropogenic uses is overestimated (Gerten et al., 2013). Lastly, while
85 the model setup of Haddeland et al. (2006b) already included important anthropogenic impact modules
86 (i.e. irrigation and dam operation), these were not fully integrated yet. Therefore multiple successive

87 model runs were required (see Section 2.1) which was computationally expensive, especially for global
88 water resources assessments.

Commented [DB3]: In response to referee comments

89 Recently version 5 of the VIC model (VIC-5) was released (Hamman et al., 2018), which focussed on
90 improving the VIC model infrastructure. These improvements provide the opportunity to fully integrate
91 human-impacts into the VIC model framework, while reducing computation times. Here the newly
92 developed VIC-WUR model is presented (named after the developing team at Wageningen University
93 and Research). The VIC-WUR model extends the existing VIC-5 model with several modules that
94 simulate the anthropogenic impacts on water resources. These modules will implement previous major
95 works on anthropogenic impact modelling as well as integrate environmental flow requirements into
96 VIC-5. The modules include: (1) integrated routing, (2) surface and groundwater use for various sectors
97 (irrigation, domestic, industrial, energy and livestock), (3) environmental flow requirements for both
98 surface and groundwater systems, and (4) dam operation.

99 The next section first describes the original VIC-5 hydrological model (Section 2.1), which calculates
100 natural water resource availability. Subsequently the integration of the anthropogenic impact modules,
101 which modify the water resource availability, are described (Section 2.2). Global anthropogenic water
102 uses for each sector are also estimated (Section 2.3). To assess the capability of the newly developed
103 modules, the VIC-WUR results were compared with FAO national water withdrawals by sector and by
104 source (FAO, 2016); Huang et al. (2018), Steinfeld et al. (2006), and Shiklomanov (2000) data on water
105 withdrawals by sector; GRACE terrestrial water storage anomalies (NASA, 2002); GRDC streamflow
106 timeseries (GRDC, 2003); and Yassin et al. (2019) and Hanasaki et al. (2006) data on reservoir operation
107 (Section 3.2). VIC-WUR simulations results are also compared with various other state-of-the-art global
108 hydrological models. Lastly, the impacts of adhering to surface and groundwater environmental flow
109 requirements on water availability are assessed (Section 3.3). This assessment is included to indicate the
110 effects of the newly integrated surface and groundwater environmental flow requirements on worldwide
111 water availability.

112 **2 Model development**

113 **2.1 VIC hydrological model**

114 The basis of the VIC-WUR model is the Variable Infiltration Capacity model version 5 (VIC-5) (Liang
115 et al., 1994; Hamman et al., 2018). VIC-5 is an open source macro-scale hydrological model that
116 simulates the full water and energy balance on a (latitude – longitude) grid. Each grid cell accounts for
117 sub-grid variability in land cover and topography, and allows for variable saturation across the grid cell.
118 For each sub-grid the water and energy balance is computed individually (i.e. sub-grid do not exchange
119 water or energy between one another). The methods used to calculate the water and energy balance are
120 summarized in Appendix A, mainly based on the work of Liang et al. (1994). For the description of the
121 global calibration and validation of the water balance one is referred to Nijssen et al. (2001b).

122 VIC version 5 (Hamman et al., 2018) upgrades did not change the model representation of physical
123 processes, but improved the model infrastructure. Improvements include the use of NetCDF for
124 input/output and the implementation of parallelization through Message Passing Interface (MPI). These
125 changes increase computational speed and make VIC-5 better suited for (computationally expensive)
126 global simulations. The most significant modification that enables new model applications is that VIC-
127 5 also changed the processing order of the model. In previous versions all timesteps were processed for
128 a single grid cell before continuing to the next cell (time-before-space). In VIC-5 all grid cells are
129 processed before continuing to the next timestep (space-before-time). This development allows for
130 interaction between grid cells every timestep, which is important for full integration of the anthropogenic
131 impact modules, especially water withdrawals and dam operation.

132 For example, surface and subsurface runoff routing to produce river streamflow was typically done as a
133 post-process operation (Lohmann et al., 1996; Hamman et al., 2017). ~~due to the time before space~~
134 ~~processing order of previous versions. Therefore, water withdrawals could not be taken into account~~
135 ~~directly and studies using the model setup of Haddeland et al. (2006b) required multiple successive~~
136 ~~model runs. Since VIC 5 uses the space before time processing order, runoff routing could be simulated~~
137 ~~each timestep. The routing post process was replaced by our newly developed routing module which~~
138 ~~simulates routing sequentially (upstream to downstream) to facilitate water withdrawals between cells.~~

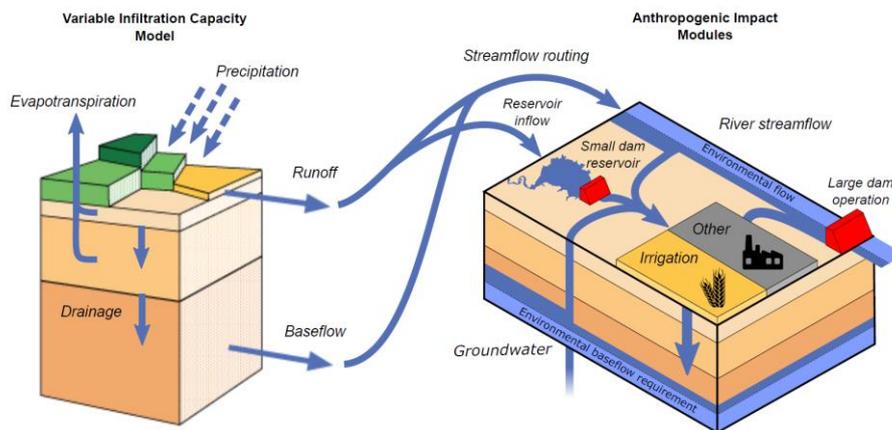
139 due to the time-before-space processing order of previous versions. In order for reservoirs to account
 140 for downstream water demand, an irrigation demand initialization was required. This initialization could
 141 either be an independent offline dataset (Voisin et al., 2013a) or multiple successive model runs
 142 (Haddeland et al., 2006b). Since VIC-5 uses the space-before-time processing order, irrigation water
 143 demands and runoff routing could be simulated each timestep. The routing post-process was replaced
 144 by our newly developed routing module, which simulates routing sequentially (upstream-to-
 145 downstream) based on the Lohmann et al. (1996) equations.

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146 2.2 Anthropogenic-impact modules

147 VIC-WUR extends the existing VIC-5 though the addition of several newly developed/implemented
 148 anthropogenic-impact modules (Figure 1). These modules include sector-specific water withdrawal and
 149 consumption, environmental flow requirements for both surface and groundwater systems and dam
 150 operation for large and small (within-grid) dams.

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151
 152 **Figure 1:** Schematic overview of the VIC-WUR model that includes the VIC-5 model and several anthropogenic impact
 153 modules. Water from river streamflow, groundwater and small (within-grid) reservoirs are available for withdrawal.
 154 Surface and groundwater withdrawals are constrained by environmental flow requirements. Withdrawn water is
 155 available for irrigation, domestic, industrial, energy and livestock use. Unconsumed irrigation water is returned to the
 156 soil column of the hydrological model. Unconsumed water for the other sectors is returned to the river streamflow.
 157 Small reservoirs fill using surface runoff from the cell they are located, while large dam reservoirs operate solely on
 158 rivers streamflow.

159 **2.2.1 Water withdrawal and consumption**

160 In VIC-WUR, sectoral water demands need to be specified for each grid cell (Section 2.3). To meet
161 water demands, water can be withdrawn from river streamflow, small (within-grid) reservoirs, and
162 groundwater resources. Streamflow withdrawals are abstracted from the grid cell discharge (as
163 generated by the routing module) and reservoir withdrawals are abstracted from small dam reservoirs
164 (located in the cell). Groundwater withdrawals are abstracted from the third layer soil moisture and an
165 (unlimited) aquifer below the soil column. Aquifer abstractions represent renewable and non-renewable
166 abstractions from deep groundwater resources. Subsurface runoff is used to fill the aquifer if there is a
167 deficit.

168 The partitioning of water withdrawals between surface and ground water resources is data driven
169 (similar to e.g. Döll et al., 2012; Voisin et al., 2017; Hanasaki et al., 2018). Partitioning was based on
170 the study of Döll et al. (2012), who estimated groundwater withdrawal fractions for each sector in around
171 15.000 national and sub-national administrative units. These groundwater fractions were based mainly
172 on information from the International Groundwater Resources Assessment Centre (IGRAC; un-
173 igrac.org) database. Surface water withdrawals were partitioned between river streamflow and small
174 reservoirs relative to water availability. Groundwater withdrawals were first withdrawn from the third
175 soil layer, second from the (remaining) river streamflow resources and lastly from the groundwater
176 aquifer. This order was implemented to avoid overestimation of non-renewable groundwater
177 withdrawals as a result of errors in the partitioning data. Aquifer withdrawals are additionally limited
178 by the pumping capacity from Sutanudjaja et al. (2018), who estimated regional pumping capacities
179 based on information from IGRAC.

180 Water can also be withdrawn from the river streamflow of other ‘remote’ cells in delta areas. Since
181 rivers cannot split in the routing module, the model is unable to simulate the redistribution of water
182 resources in dendritic deltas. Therefore, streamflow at the river mouth is available for use in delta areas
183 (partitioned based on demand) to simulate the actual water availability. Delta areas were delineated by
184 the global delta map of Tessler et al. (2015).

185 In terms of water allocation, under conditions where water demands cannot be met, water withdrawals
186 are allocated to the domestic, energy, manufacturing, livestock and irrigation sector in that order.
187 Withdrawn water is partly consumed, meaning the water evaporates and does not return to the
188 hydrological model. Consumption rates were set at 0.15 for the domestic and 0.10 for the industrial
189 sector, based on the data of Shiklomanov (2000). The water consumption in the energy sector was based
190 on Goldstein and Smith (2002) and varies per thermoelectric plant based on the fuel type and cooling
191 system. For the livestock sector the assumption was made that all withdrawn water is consumed.
192 Unconsumed water withdrawals for these sectors are returned as river streamflow. For the irrigation
193 sector, consumption was determined by the calculated evapotranspiration. Unconsumed irrigation water
194 remains in the soil column and eventually returns as subsurface runoff.

195 **2.2.2 Environmental flow requirements**

196 Water withdrawals can be constrained by environmental flow requirements (EFRs). These EFRs specify
197 the timing and quantity of water needed to support terrestrial river ecosystems (Smakhtin et al., 2004;
198 Pastor et al., 2019). Surface and groundwater withdrawals are constrained separately in VIC-WUR,
199 based on the EFRs for streamflow and baseflow respectively. EFRs for streamflow specify the minimum
200 river streamflow requirements while EFRs for baseflow specify the minimum subsurface runoff
201 requirements (from groundwater to surface water). Since baseflow is a function groundwater
202 availability, baseflow requirements are used to constrain groundwater (including aquifer) withdrawals.

203 Various EFR methods are available (Smakhtin et al., 2004; Richter et al., 2012; Pastor et al., 2014). Our
204 study used the Variable Monthly Flow (VMF) method (Pastor et al., 2014) to calculate the EFRs for
205 streamflows. VMF calculates the required streamflow as a fraction of the natural flow during high (30
206 %), intermediate (45 %) and low (60 %) flow periods, as described in Appendix B. The VMF method
207 performed favourably compared to other hydrological methods, in 11 case studies where EFRs were
208 calculated locally (Pastor et al., 2014). The advantage of the VMF method is that the method accounts
209 for the natural flow variability, which is essential to support freshwater ecosystems (Poff et al., 2010).

210 EFR methods for baseflow have been rather underdeveloped compared to EFR methods for streamflow.
211 However, a presumptive standard of 90 % of the natural subsurface runoff through time was proposed
212 by Gleeson and Richter (2018), as described in Appendix B. This standard should provide high levels
213 of ecological protection, especially for groundwater dependent ecosystems.

214 Note that part of the EFRs for baseflow are already captured in the EFRs for streamflow, especially
215 during low-flow periods that are usually dominated by baseflows. However, the EFRs for baseflow
216 specifically limit local groundwater withdrawals while EFRs for streamflow include the accumulated
217 runoff from upstream areas. Also, the chemical composition of groundwater derived flows is inherently
218 different, making them a non-substitutable water flow for environmental purposes (Gleeson and Richter,
219 2018).

220 **2.2.3 Dam operation**

221 Due to the lack of globally available information on local dam operations, several generic dam operation
222 schemes were developed for macro-scale hydrological models to reproduce the effect of dams on natural
223 streamflow (Haddeland et al., 2006a; Hanasaki et al., 2006; Zhao et al., 2016; Rougé et al., 2019; Yassin
224 et al., 2019). In VIC-WUR a distinction is made between ‘small’ dam reservoirs (with an upstream area
225 smaller than the cell area) and ‘large’ dam reservoirs, similar to Hanasaki et al. (2018), Wisser et al.
226 (2010a) and Döll et al. (2009). Small dam reservoirs act as buckets that fill using surface runoff of the
227 grid-cell they are located in and reservoirs storage can be used for water withdrawals in the same cell.
228 Large dam reservoirs are located in the main river and used the operation scheme of Hanasaki et al.
229 (2006), as described in Appendix C.

230 The scheme distinguishes between two dam types: (1) dams that do not account for water demands
231 downstream (e.g. hydropower dams or flood protection dams) and (2) dams that do account for water
232 demand downstream (e.g. irrigation dams). For dams that do not account for demands, dam release is
233 aimed at reducing annual fluctuations in discharge. For dams that do account for demands, dam release
234 is additionally adjusted to provide more water during periods of high demand. The operation scheme
235 was validated by Hanasaki et al. (2006) for 28 reservoirs and was used in various other studies (Hanasaki

236 ~~et al., 2008a; Döll et al., 2009; Pokhrel et al., 2012b; Voisin et al., 2013; Hanasaki et al., 2018).~~ (Hanasaki
237 ~~et al., 2008b; Döll et al., 2009; Pokhrel et al., 2012b; Voisin et al., 2013b; Hanasaki et al., 2018).~~ Here,

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238 the scheme was adjusted slightly to account for monthly varying EFRs and to reduce overflow releases,
239 which is described in Appendix C.

240 The Global Reservoir and Dam (GRanD) database (Lehner et al., 2011) was used to specify location,
241 capacity, function (purpose), and construction year of each dam. The capacity of multiple (small- and
242 large) dams located in the same cell were combined.

243 **2.3 Sectoral water demands**

244 VIC-WUR water withdrawals are based on the irrigation, domestic, industry, energy and livestock water
245 demand in each grid-cell. Water demands represent the potential water withdrawal, which is reduced
246 when insufficient water is available. Irrigation demands were estimated based on the hydrological model
247 while water demands for other sectors are provided to the model as an input. Domestic and industrial
248 were estimated based on several socioeconomic predictors, while energy and livestock water demands
249 were derived from power plant and livestock distribution data. Due to data limitations the energy sector
250 was incomplete, and energy water demands were partly included in the industrial water demands (which
251 combined the remaining energy and manufacturing water demands). For more details concerning
252 sectoral water demand calculations the reader is referred to Appendix D.

253 **2.3.1 Irrigation demands**

254 Irrigation demands were set to increase soil moisture in the root zone so that water availability is not
255 limiting crop evapotranspiration and growth. The exception is paddy rice irrigation (Brouwer et al.,
256 1989), where irrigation was also supplied to keep the upper soil layer saturated. Water demands for
257 paddy irrigation practices are relatively high compared to conventional irrigation practices due to
258 increased evaporation and percolation. Therefore, the crop irrigation demands for these two irrigation
259 practices were calculated and applied separately (i.e. in different sub-grids). Note that multiple cropping
260 seasons are included based on the MIRCA2000 land-use dataset (Portmann et al., 2010) (see Section
261 3.1 for more details).

262 Total irrigation demands also included transportation and application losses. Note that transportation
263 and application losses are not ‘lost’ but rather returned to the soil column without being used by the
264 crop. The water loss fraction was based on Frenken and Gillet (2012), who estimated the aggregated
265 irrigation efficiency for 22 United Nations sub-regions. Irrigation efficiencies were estimated based on
266 the differencesdifference between the calculated crop water requirements (crop evapotranspiration;
267 consumptive water use) and theAQUASTAT reported irrigation water withdrawals (including
268 transportation and application losses). Crop water requirements are estimated based on the FAO
269 irrigation and Drainage paper (Allen et al., 1998). Low and calculated irrigation efficiencies can result
270 in irrigation water withdrawals up to four times higher than the crop water requirements in regions such
271 as east and west Africawater requirements (Allen et al., 1998), using data on crop information (e.g.
272 growing season, harvest area) from AQUASTAT.

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273 2.3.2 Domestic and industrial demands

274 Domestic and industrial water withdrawals were estimated based on Gross Domestic Product (GDP) per
275 capita and Gross Value Added (GVA) by industries respectively (from Bolt et al. (2018), Feenstra et al.
276 (2015) and World bank (2010); see Appendix D for more details). These drivers do not fully capture the
277 multitude of socioeconomic factors that influence water demands (Babel et al., 2007). However, the
278 wide availability of data allows for extrapolation of water demands to data-scarce regions and future
279 scenarios (using studies such as Chateau et al. (2014)).

280 Domestic water demands per capita (used for drinking, sanitation, hygiene and amenity uses) were
281 estimated similar to Alcamo et al. (2003). Demands increased non-linearly with GDP per capita due to
282 the acquisition of water using appliances as household become richer. A minimum water supply is
283 needed for survival, and the saturation of water using appliances sets a maximum on domestic water
284 demands. Industrial water demands (used for cooling, transportation and manufacturing) were estimated
285 similar to Flörke et al. (2013) and Voß and Flörke (2010). Industrial demands increased linearly with
286 GVA (as an indicator of industrial production). Since industrial water intensities (i.e. the water use per
287 production unit) vary widely between different industries (Flörke and Alcamo, 2004 ; Vassolo and Döll,
288 2005; Voß and Flörke, 2010), the average water intensity was estimated for each country. Both domestic

289 and industrial water demands were also influenced by technological developments that increase water-
290 use efficiency over time, as in Flörke et al. (2013).

291 Domestic water demands varied monthly based on air temperature variability as in Huang et al. (2018)
292 (based on Wada et al. (2011b)). Using this approach, water demands were higher in summer than in
293 winter, especially for counties with strong seasonal temperature differences. Domestic water demand
294 per capita were downscaled using the HYDE3.2 gridded population maps (Goldewijk et al., 2017).
295 Industrial water demands were kept constant throughout the year. Industrial demands were downscaled
296 from national to grid cell values using the NASA Black Marble night-time light intensity map (Roman
297 et al., 2018). National industrial water demands were allocated based on the relative light intensity per
298 grid cell for each country.

299 **2.3.3 Energy and livestock demands**

300 Energy water demands (used for cooling of thermoelectric plants) were estimated using data from van
301 Vliet et al. (2016). Water use intensity for generation (i.e. the water use per generation unit) was
302 estimated based on the fuel and cooling system type (Goldstein and Smith, 2002), which was combined
303 with the generation capacity. Note that the data only covered a selection of the total number of
304 thermoelectric power plants worldwide. Around 27 % of the total (non-renewable) global installed
305 capacity between 1980 and 2011 was included in the dataset due to lack of information on cooling
306 system types for the majority of thermoelectric plants. To avoid double counting, energy water demands
307 were subtracted from the industrial water demands.

308 Livestock water demands (used for drinking and animal servicing) were estimated by combining the
309 Gridded Livestock of the World (GLW3) map (Gilbert et al., 2018) with the livestock water requirement
310 reported by Steinfeld et al. (2006). Eight varieties of livestock were considered: cattle, buffaloes, horses,
311 sheep, goats, pigs, chicken and ducks. Drinking water demands varied monthly based on temperature as
312 described by Steinfeld et al. (2006), whereby drinking water requirements were higher during higher
313 temperatures.

314 **3 Model application**

315 **3.1 Setup**

316 VIC-WUR results were generated between 1979 and 2016, excluding a spin-up period of one year
317 (analysis period from 1980 to 2016). The model used a daily timestep (with a 6-hourly timestep for snow
318 processes) and simulations were executed on a 0.5° by 0.5° grid (around 55 km at the equator) with three
319 soil layers per grid cell. Soil and (natural) vegetation parameters were the same as in Nijssen et al.
320 (2001c) (disaggregated to 0.5°), who used various sources to determine the soil (Cosby et al., 1984;
321 Carter and Scholes, 1999) and vegetation parameters (Calder, 1993; Ducoudre et al., 1993; Sellers et al.,
322 1994; Myneni et al., 1997).

323 Nijssen et al. (2001c) used the Advanced Very High Resolution Radiometer vegetation type database
324 (Hansen et al., 2000) to spatially distinguish 13 land cover types. The land cover type ‘cropland’ in the
325 original land-cover dataset was replaced by cropland extents from the MIRCA2000 cropland dataset
326 (Portmann et al., 2010). MIRCA2000 distinguishes the monthly growing area(s) and season(s) of 26
327 irrigated and rain-fed crop types around the year 2000. Crop types were aggregated into three land cover
328 types: rain-fed, irrigated and paddy rice cropland. The natural vegetation was proportionally rescaled to
329 make up discrepancies between the natural vegetation and cropland extents.

330 Cropland coverage (the cropland area actually growing crops) varied monthly based on the crop growing
331 areas of MIRCA2000. The remainder was treated as bare soil. Cropland vegetation parameters (e.g. Leaf
332 Area Index (LAI), displacement, vegetation roughness and albedo) vary monthly based on the crop
333 growing seasons and the development-stage crop coefficients of the Food and Agricultural Organisation
334 (Allen et al., 1998).

335 The latest WATCH forcing data Era Interim (aggregated to 6 hourly), developed by the EU Water and
336 Global Change (WATCH; Harding et al., 2011) project, was used as climate forcing (WFDEI; Weedon
337 et al., 2014). The dataset provides gridded historical climatic variables of minimum and maximum air
338 temperature, precipitation (as the sum of snowfall and rainfall, GPCP bias-corrected), relative humidity,
339 pressure and incoming shortwave and longwave radiation.

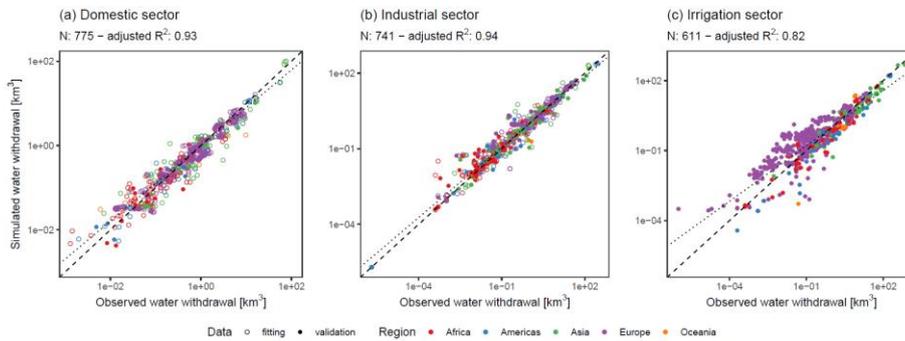
340 For naturalized simulations only the routing module was used. For the human-impact simulations the
341 sectoral water withdrawals and dam operation modules were turned on in the model simulations. For
342 the EFR-limited simulations water withdrawals and dam operations were constrained as described.

343 **3.2 Validation and evaluation**

344 In order to validate the VIC-WUR human-impact modules, water withdrawal, terrestrial total water
345 storage anomalies, streamflow and reservoir operation simulations were compared with observations.
346 The validation specifically focused on the effects of the newly included human-impact modules,
347 meaning that streamflow and total-water storage anomaly results are shown for river basins that are
348 strongly influenced by human activities. A general validation for streamflow and terrestrial total water
349 storage anomalies (including basins with limited human activities) is shown in Appendix E.

350 **3.2.1 Sectoral water withdrawals**

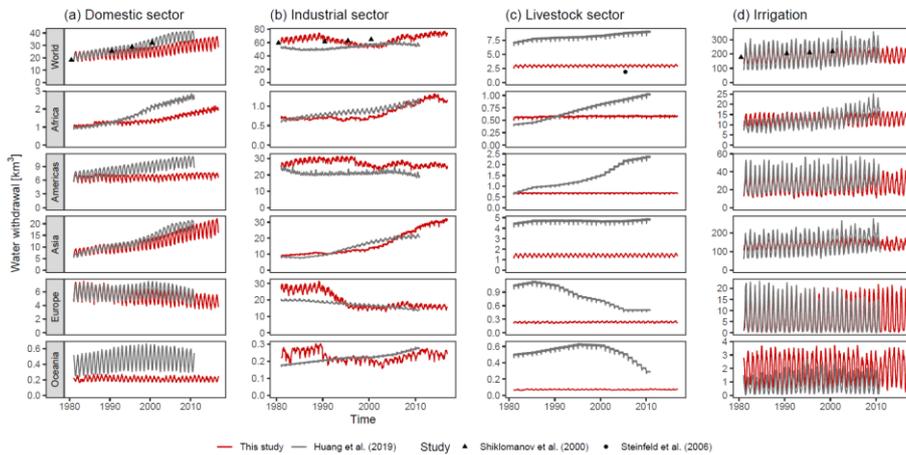
351 Simulated global domestic, industrial, livestock and irrigation mean water withdrawals were 310, 771,
352 36 and 2202 km³ year⁻¹ respectively for the period of 1980 to 2016. Sectoral water withdrawals were
353 compared with FAO national annual water withdrawals (FAO, 2016), monthly withdrawal data from
354 Huang et al. (2018) and annual withdrawal data from Shiklomanov (2000) and Steinfeld et al. (2006).
355 For the latter studies, water withdrawals were aggregated by region (world, Africa, Asia, Americas,
356 Europe and Oceania). Note that Huang et al. (2018) irrigation water withdrawals integrate results of four
357 other macro-scale hydrological models (WaterGAP, H08, LPJmL, PCR-GLOBWB), using the same
358 land-use and climate setup as our study. Results from individual macro-scale hydrological models are
359 also shown.



360

361 **Figure 2: Comparison between simulated and reported national annual water withdrawals for the (a) domestic, (b)**
 362 **industrial and (c) irrigation sector. Colours distinguish between regions. Open circles were also used in the calibration**
 363 **of the water withdrawal demands. The dashed line indicates the 1:1 ratio and the spotted line indicates the simulated**
 364 **best linear fit. Note the log-log axis which is used to display the wide range of water withdrawals. The R^2 adjusted**
 365 **is also based on the log values.**

366 Simulated domestic, industrial and irrigation water withdrawals correlated well to reported national
 367 water withdrawals, with adjusted R^2 of 0.93, 0.94 and 0.82 for domestic, industrial and irrigation water
 368 withdrawal respectively (Figure 2a-c). Generally, smaller water withdrawals were overestimated and
 369 larger water withdrawals were underestimated. Differences for the domestic and industrial sector were
 370 small and probably related to the fact that smaller countries were poorly delineated on a 0.5° by 0.5°
 371 grid. However, irrigation differences were larger with overestimations of irrigation water withdrawals
 372 in (mostly) Europe. Since irrigation water demands are the results of the simulated water balance,
 373 overestimations would indicate a regional underestimation of water availability for Europe or
 374 differences in irrigation efficiency.



375

376 **Figure 3: Comparison between simulated and compiled monthly and annual regional water withdrawals for the (a)**
 377 **domestic sector, (b) industrial sector, (c) livestock sector, and (d) irrigation. Colours and shapes distinguish between**
 378 **studies. Note that the jitter in livestock withdrawals is due to the different days per month.**

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379 When domestic, industrial and livestock water withdrawals were compared to other studies, results were
 380 mixed (figure 3a-c). Simulated domestic withdrawals followed a similar trend in time. However,
 381 simulated domestic water withdrawals trends were overall somewhat underestimated with a mean bias
 382 of $54 \text{ km}^3 \text{ year}^{-1}$ compared to Huang et al. (2018). Asia is the main contributor to the global
 383 underestimation, but results are similar in most regions. Simulated industrial water withdrawal were
 384 (mostly) higher in our study with a mean bias of $107 \text{ km}^3 \text{ year}^{-1}$ compared to Huang et al. (2018) but
 385 only a mean bias of $5 \text{ km}^3 \text{ year}^{-1}$ compared to Shiklomanov (2000). Also, industrial water withdrawal
 386 trends in time were less consistent.

387 Withdrawal differences for the domestic and industrial sector are probably due to the limited data
 388 availability. Our approach to compute water demands was data-driven and sensitive to data gaps (as
 389 opposed to Huang et al. (2018) who also combined model results). For example, domestic withdrawal
 390 data for China was not available before 2007 and industrial withdrawal data was limited before 1990.

391 Also, data on the disaggregation of industrial sectors (e.g. energy and mining) was limited, which can
 392 be important sectors in the water-food-energy nexus.

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393 For livestock water withdrawals there is a large discrepancy between the Huang et al. (2018) and
 394 Steinfeld et al. (2006). Both studies used similar livestock maps, but there was large differences in

395 livestock water intensity [litre animal⁻¹ year⁻¹]. Since our study used Steinfeld et al. (2006) to estimate
 396 livestock water intensity, our results were closer to their values (slightly higher due to the inclusion of
 397 buffaloes, horses and ducks). Note that Huang et al. (2018) shows trends in livestock water withdrawals
 398 while our study used static livestock maps.

399 **Table 1: Global irrigation water withdrawals as calculated by several global hydrological models. **Includes livestock**
 400 **withdrawals.**

Model	Irrigation withdrawal [km ³ year ⁻¹]	Representative years	Reference
VIC-WUR	2202 (± 60)	1980-2016	Our study
H08	(a) 2810 (b) 2544 (± 75)	(a) 1995 (b) 1984 - 2013	(a) Hanasaki et al. (2008b) (a) Hanasaki et al. (2008a) (b) Hanasaki et al. (2018)
MATSIRO	(a) 2158 (± 134) (b) 3028 (± 171)	(a) 1983 - 2007 (b) 1998 - 2002	(a) Pokhrel et al. (2012a) (b) Pokhrel et al. (2015)
LPJmL	2555	1971 - 2000	Rost et al. (2008)
PCR-GLOB	(a) 2644 (b) 2309 **	(a) 2010 (b) 2000 - 2015	(a) Wada and Bierkens (2014) (b) Sutanudjaja et al. (2018)
WaterGAP	(a) 3185 (b) 2400	(a) 1998-2002 (b) 2003 - 2009	(a) Döll et al. (2012) (b) Döll et al. (2014)
WBM	2997	2002	Wisser et al. (2010b)

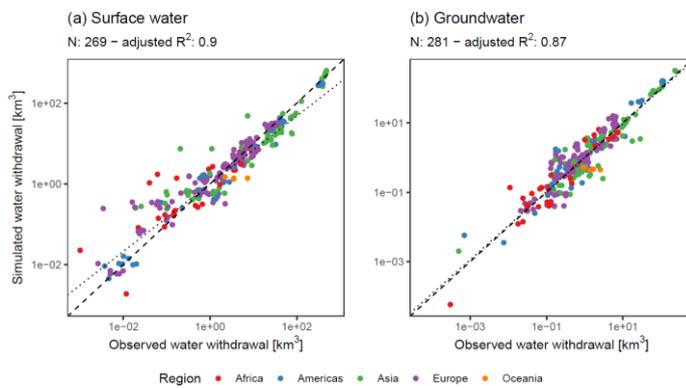
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401 Simulated irrigation water withdrawals were within range of other macro-scale hydrological model
 402 estimates (Table 1). Simulated monthly variability in irrigation water withdrawals is reduced compared
 403 to the compiled results of Huang et al. (2018) (Figure 3d), especially in Asia. Also, trends in time are
 404 less pronounced as can be seen in Africa. These differences may indicate a relative low weather/climate
 405 sensitivity of evapotranspiration in VIC-WUR, as annual and interannual weather changes affect
 406 irrigation water demands to a lesser degree.

407 3.2.2 Groundwater withdrawals and depletion

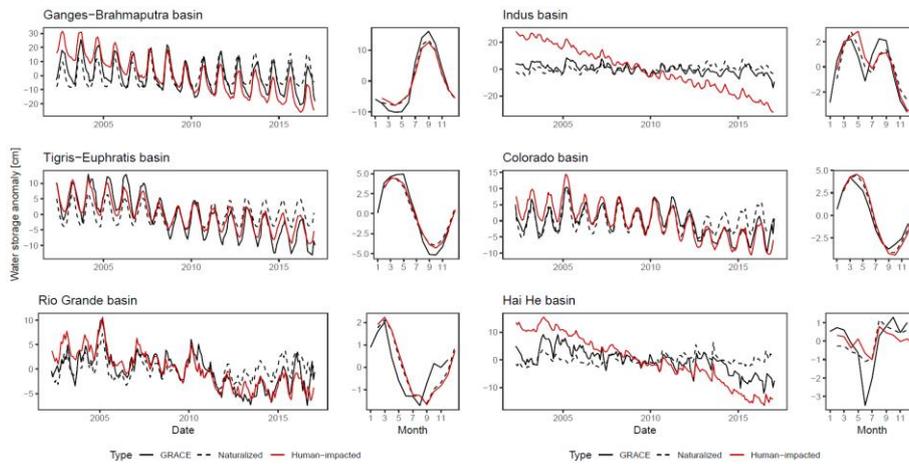
408 Simulated global mean withdrawals were 2327 and 992 km³ year⁻¹ for surface and groundwater
 409 respectively for the period of 1980 to 2016. Of the global groundwater withdrawals, 334 km³ year⁻¹
 410 contributed to groundwater depletion. Simulated ground and surface water withdrawals and terrestrial
 411 total water storage anomalies were compared FAO national annual water withdrawals (FAO, 2016) and

412 monthly storage anomaly data from the GRACE satellite (NASA, 2002). GRACE satellite total water
413 storage anomalies were used to validate total water storage dynamics as well as groundwater exploitation
414 contributing to downward trends in total water storage. Groundwater depletion results from other macro-
415 scale hydrological models are shown as well. In order to compare the simulation results to the GRACE
416 dataset, a 300km gaussian filter was applied to the simulated data (similar to Long et al. (2015)).



417
418 **Figure 4: Comparison between simulated and reported national annual water withdrawals from (a) surface water and**
419 **(b) groundwater. Colours distinguish between regions. The dashed line indicates the 1:1 ratio and the spotted line**
420 **indicates the simulated best linear fit. Note the log-log axis which is used to display the wide range of water withdrawals.**
421 **The R^2 adjusted is also based on the log values.**

422 Simulated surface and groundwater withdrawals correlated well to the reported national water
423 withdrawals, with adjusted R^2 of 0.90 and 0.87 for surface and groundwater respectively (Figure 4a-b).
424 Surface water withdrawals were overestimated for low withdrawals and underestimated for large
425 withdrawals. There is a weak correlation (-0.35) between the underestimations in surface water
426 withdrawals and the overestimation in groundwater withdrawals, meaning water withdrawal differences
427 could be related to the partitioning between surface and groundwater resources. Also, it is likely that
428 low water demands are overestimated (as discussed in Section 3.2.1), resulting in an overestimation of
429 low surface water withdrawals.



430

431 **Figure 5: Comparison between simulated and observed monthly terrestrial total water storage anomalies. Figures**
 432 **indicate timeseries and multi-year mean average for naturalized simulations (dashed), human-impacted simulations**
 433 **(red) and observed (black) terrestrial total water storage anomalies.**

434 Simulated monthly terrestrial water storage anomalies correlated well to the GRACE observations, with
 435 mean annual and inter-annual Root Mean Squared Error (RMSE) of 1.9 mm and 3.5 mm respectively.
 436 The difference between annual and inter annual performance was primarily due to the groundwater
 437 depletion process (Figure 5). Simulated groundwater depletion was (mostly) overestimated (e.g. Indus
 438 and Hai He basins), with higher declining trends in terrestrial total water storage for most basins.
 439 However, compared to other macro-scale hydrological models, simulated groundwater withdrawal and
 440 exploitation was within range (Table 2), even though total groundwater withdrawals were relatively
 441 high.

442 As with the FAO comparison, these results seems to indicate that withdrawal partitioning towards
 443 groundwater is overestimated. However, conclusions regarding groundwater depletion are limited by
 444 the relatively simplistic approach to groundwater used in our study (as discussed by Konikow (2011)
 445 and de Graaf et al. (2017)). For example, processes such as wetland recharge and groundwater flows
 446 between cells are not simulated, even though these could decrease groundwater depletion.

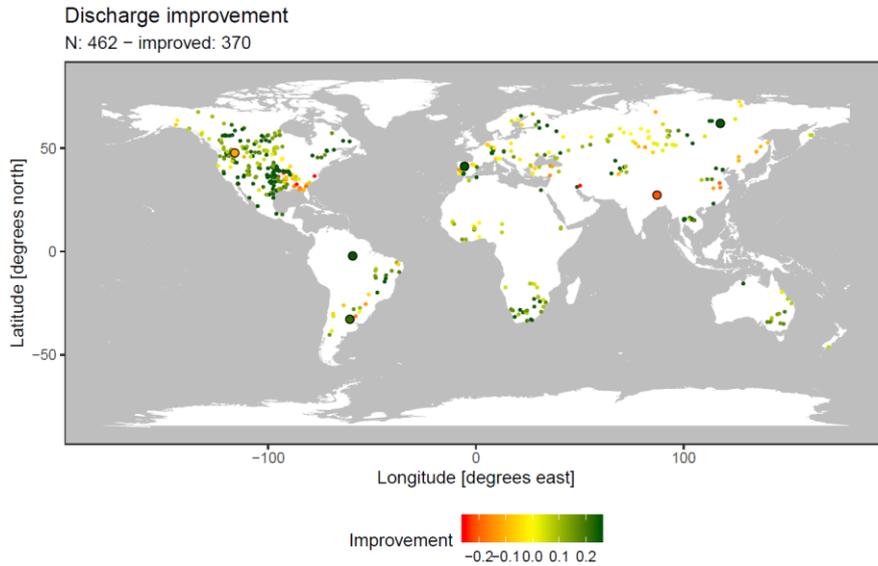
447 **Table 2: Global groundwater withdrawals and depletion as calculated by several global hydrological models.**

Model	Groundwater withdrawal [km ³ year ⁻¹]	Groundwater depletion [km ³ year ⁻¹]	Representative years	Reference
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VIC-WUR	992 (\pm 51)	316 (\pm 63)	1980 - 2016	Our study
H08	789 (\pm 30)	182 (\pm 26)	1984 - 2013	Hanasaki et al. (2018)
MATSIRO	570 (\pm 61)	330	1998 - 2002	Pokhrel et al. (2015)
GCAM		(a) 600 (b) 550	(a) 2005 (b) 2000	(a) Kim et al. (2016) (b) Turner et al. (2019)
PCR-GLOB	(a) 952 (b) 632	(a) 304 (b) 171	(a) 2010 (b) 2000 - 2015	(a) Wada and Bierkens (2014) (b) Sutanudjaja et al. (2018)
WaterGAP	(a) 1519 (b) 888	(a) 250 (b) 113	(a) 1998-2002 (b) 2000 - 2009	(a) Döll et al. (2012) (b) Döll et al. (2014)

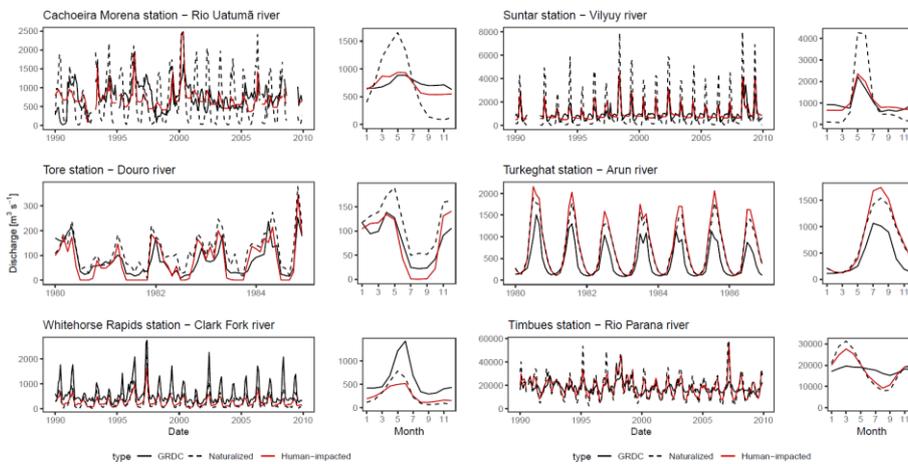
448 **3.2.3 Discharge modification**

449 Simulated discharge was compared to GRDC station data (GRDC, 2003) for various human-impacted
450 rivers. Stations were selected if the upstream area was larger than 20,000 km², matched the simulated
451 upstream area at the station location, and the available data spanned more than 2 years. Subsequently,
452 stations where the human-impact modules did not sufficiently impacted discharge were omitted. In order
453 validate the reservoir operation more thoroughly, simulated reservoir inflow, storage and release was
454 compared with operation data from Hanasaki et al. (2006) and Yassin et al. (2019). Reservoirs were
455 included if the simulated storage capacity (which is the combined storage capacity of all large dams in
456 a grid) was similar to observed storage capacity.



457

458 **Figure 6: Discharge improvement from naturalized to human-impacted simulations (as a fraction of the naturalized**
 459 **RMSE). Circled larger stations are shown in Figure 7.**



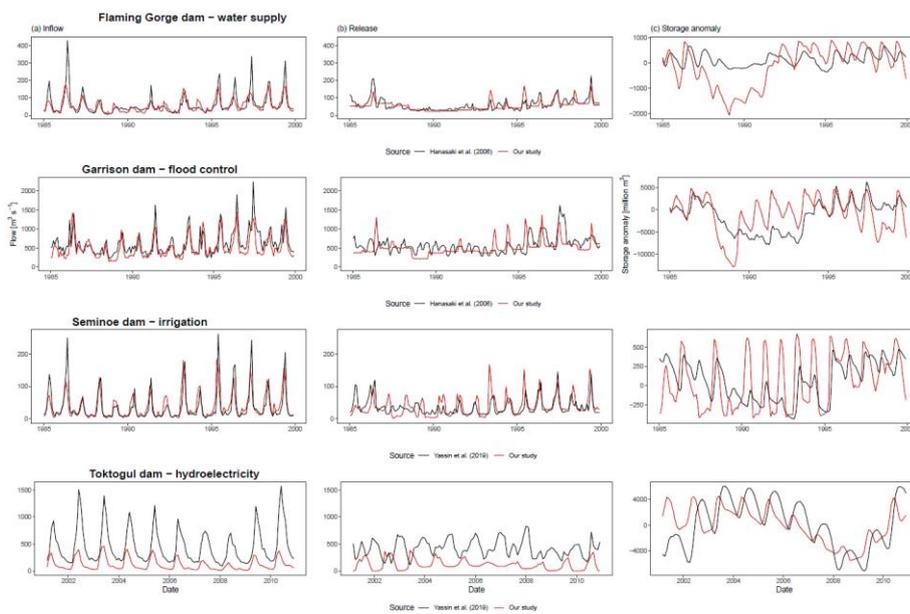
460

461 **Figure 7: Comparison between simulated and observed discharge. Figures indicate timeseries and multi-year average**
 462 **of for naturalized simulations (dashed), human-impacted simulations (red) and observed (black) discharge.**

463 The inclusion of the human-impact modules improved discharge performance, measured in RMSE, for
 464 370 out of 462 stations (80 %; Figure 6 and 7). Improvements were mainly due to the effects of reservoir
 465 operation on discharges (e.g. Cachoeira Morena and Suntar stations), but also due to withdrawal

466 reductions (e.g. Tore station). Reservoir effects on discharge were sometimes underestimated however
467 (e.g. Timbues station).

468 Decreased performance was mostly related to under or overestimations of (calibrated) natural
469 streamflow which was subsequently exacerbated by reservoir operation and water withdrawals. For
470 example, the Clark Fork river naturalized streamflow was underestimated, which was subsequently
471 further underestimated by the human-impact modules (Whitehorse Rapids station). Also, increases in
472 discharge due to groundwater withdrawals could increase naturalized streamflow (e.g. Turkeghat
473 station). Further improvements to discharge performance would most likely require either a recalibration
474 of the VIC model parameters.



475
476 **Figure 8: Comparison between simulated and observed reservoir operation. Figures indicate timeseries and multi-year**
477 **averages of (a) inflow, (b) release and (c) storage anomalies for human-impacted simulations (red) and observations**
478 **(black).**

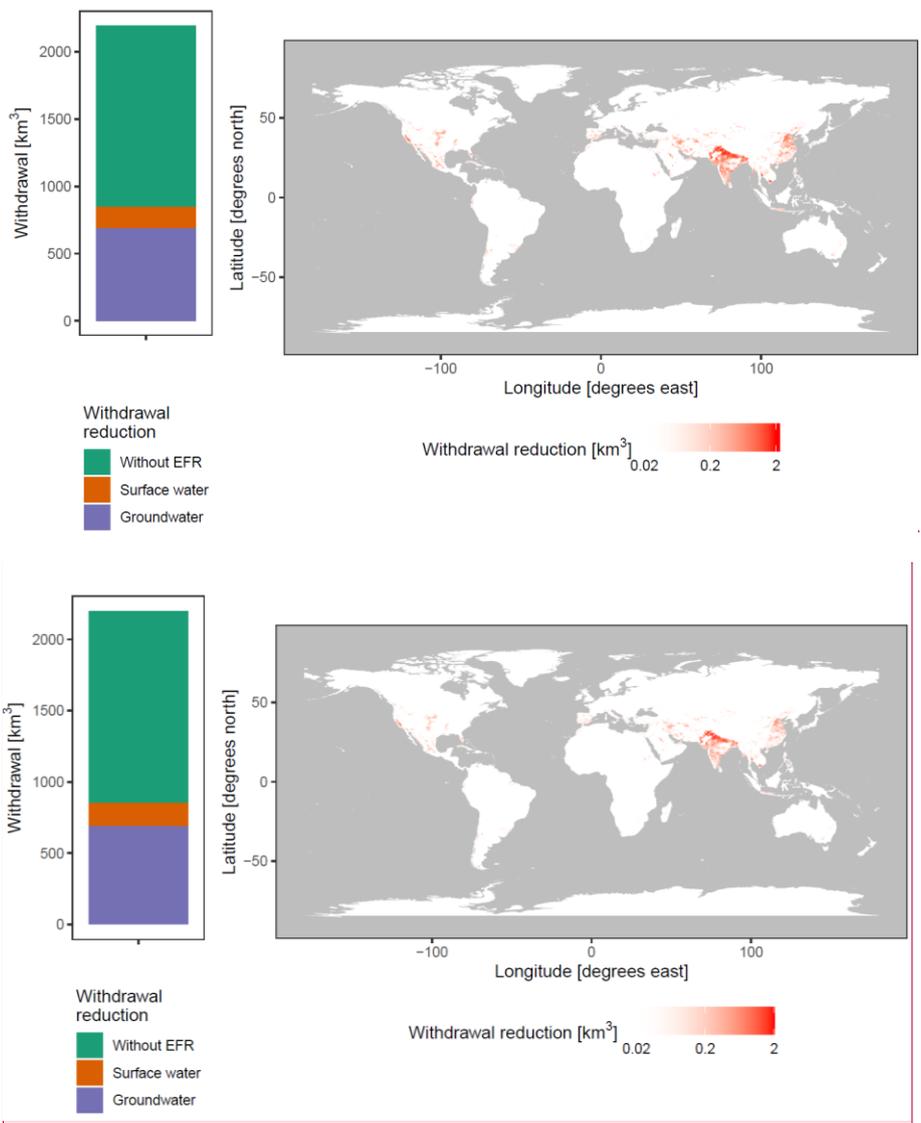
479 For individual reservoirs, operation characteristics were generally well simulated (Figure 8), with
480 reductions in annual discharge variations (e.g. Flaming Gorge and Garrison dams) and increased water
481 release for irrigation (e.g. Seminoe dam). However, due to changes in locally simulated and actual
482 inflow, dam operation can take on different characteristics (e.g. Toktogul dam). Also, peak discharge

483 events caused by reservoir overflow (as also described by ~~Masaki et al. (2018a)~~ Masaki et al. (2018))
484 were not always sufficiently represented in the observations (e.g. Garisson dam). These differences
485 indicate locally varying reservoir operation strategies. Several studies have developed reservoir
486 operation schemes that can be calibrated to the local situation (Rougé et al., 2019; Yassin et al., 2019).
487 However, worldwide implementations of these operation schemes remains limited by data availability.

488 3.3 Integrated environmental flow requirements

489 In order to assess the impact and capabilities of the newly integrated environmental flow requirements
490 (EFRs) module, simulated water withdrawals with and without adhering to EFRs were compared.

Commented [DB11]: Updated references



491

492

493 **Figure 9:** Average annual irrigation water withdrawal reductions when adhering to EFRs as (left) global gross total and
 494 (right) spatially distributed. Global gross totals are separated into withdrawals without any reduction (green), surface
 495 water withdrawal reductions (orange) and groundwater withdrawal reductions (purple). Note the log axis for the
 496 spatially distributed withdrawal reductions to better display the spatial distribution of the reductions.

497 If water-use would be limited to EFRs, irrigation withdrawals would need to be reduced by about 39 %

498 (851 km³ year⁻¹) (Figure 9a). Under the strict requirements used in our study, 81 % (693 km³ year⁻¹) of

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499 the reduction could be attributed to limitations imposed on groundwater withdrawals. Subsequently, the
500 impact of the environmental flow requirements (if adhered to) would be largest in groundwater
501 dependent regions (Figure 9b). Note that, due to the full integration of EFRs, downstream surface water
502 withdrawals increased by $98 \text{ km}^3 \text{ year}^{-1}$ when limiting groundwater withdrawals on top of limiting
503 surface water withdrawals, due to increase subsurface runoff.

504 Reductions due to EFRs were similar to Jägermeyr et al. (2017), who calculated irrigation withdrawal
505 reductions of 41 % ($997 \text{ km}^3 \text{ year}^{-1}$) assuming only surface water abstractions. In our study, surface
506 water reductions were smaller since the strict groundwater requirements increases subsurface runoff to
507 surface waters. It can be discussed to what extent the EFRs for baseflow were too constricting, since
508 they were based on the relatively stringent EFR for streamflow of Richter et al. (2012) (10 % of the
509 natural streamflow). However, in the absence of any other standards, this baseflow standard remains the
510 best available. Note that, even when accounting for EFRs for baseflow on a grid scale, withdrawals
511 could still have local and long-term impacts that are not captured by the model. The timing, location and
512 depth of groundwater withdrawals are important factors due to their interactions with the local
513 geohydrology, as discussed by Gleeson and Richter (2018).

514 **4 Conclusion**

515 The VIC-WUR model introduced in this paper aims to provide new opportunities for global water
516 resource assessments using the VIC model. Accordingly, several anthropogenic impact modules, based
517 on previous major works, were integrated into the VIC-5 macro-scale hydrological model: domestic,
518 industrial, energy, livestock and irrigation water withdrawals from both surface water and groundwater
519 as well as an integrated environmental flow requirement module and dam operation module. Global
520 gridded datasets on domestic, industrial, energy and livestock demand were developed separately and
521 used to force the VIC-WUR model.

522 Simulated national water withdrawals were in line with reported national annual withdrawals (R^2
523 adjusted > 0.8 ; both per sector as per source). However, the data-oriented methodology used to derive
524 sectoral water demands resulted in different withdrawal trends over time compared to other studies

525 (Shiklomanov, 2000; Huang et al., 2018). ~~However, note that~~ While the current setup to estimate sectoral
526 water demands is well suited for future water withdrawal estimations, there are various other approaches
527 (e.g. Alcamo et al., 2003; Vassolo and Döll, 2005; Shen et al., 2008; Hanasaki et al., 2013; Wada and
528 Bierkens, 2014). As the model setup of VIC-WUR allows for the evaluation of other sectoral water
529 demand inputs (on various temporal aggregations), several different approaches can be used depending
530 on the focus region and data-availability for calibration. Terrestrial water storage anomaly trends were
531 well simulated (mean annual and inter-annual RMSE of 1.9 mm and 3.6 mm respectively), while
532 groundwater exploitation was overestimated. Overestimated groundwater depletion rates are likely
533 related to an over-partitioning of water withdrawals to groundwater. The implemented human impact
534 modules increased simulated discharge performance (370 out of 462 stations), mostly due to the effects
535 of reservoir operation.

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536 An assessment of the effect of EFRs shows that, when one would adhere to these requirements, global
537 water withdrawals would be severely limited (39 %). This limitation is especially the case for
538 groundwater withdrawals, which, under the strict requirements used in our study, need to be reduced by
539 81 %.

540 VIC-WUR has potential for studying impacts of climate change and anthropogenic developments on
541 current and future water resources and sectoral specific-water scarcity. The additions presented here
542 make the VIC model more suited for fully-integrated worldwide water-resource assessments and
543 substantially decrease computation times compared to ~~previous versions~~ Haddeland et al. (2006a).

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544 5 Code availability

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545 All code for the VIC-WUR model is freely available at github.com/wur-wsg/VIC (tag VIC-
546 WUR.2.01.0; DOI 10.5281/zenodo.33994503934325) under the GNU General Public License, version
547 2 (GPL-2.0). VIC-WUR documentation can be found at vicwur.readthedocs.io. The original VIC model
548 is freely available at github.com/UW-Hydro/VIC (tag VIC.5.0.1; DOI 10.5281/zenodo.267178) under
549 the GNU General Public License, version 2 (GPL-2.0). VIC documentation can be found at
550 vic.readthedocs.io. Documentation and scripts concerning ~~inputs, configurations and analysis~~ input data

551 used in our study is freely available at github.com/bramdr/VIC-WUR_support (tag VIC-WUR.2.01.0;
552 DOI 10.5281/zenodo.3401413934363) under the GNU General Public License, version 3 (GPL-3.0).

553 **6 Appendix**

554 **6.1 Appendix A: VIC water and energy balance**

555 In VIC each sub-grid computes the water and energy balance individually (i.e. sub-grid do not exchange
556 water or energy between one another). For the water balance, incoming precipitation is partitioned
557 between evapotranspiration, surface and subsurface runoff, and soil water storage. Potential
558 evapotranspiration is based on the Penman-Monteith equation without the canopy resistance
559 (Shuttleworth, 1993). The actual evapotranspiration is calculated by two methods, based on whether the
560 land cover is vegetated or not (bare soil). Evapotranspiration of vegetation is constrained by stomatal,
561 architectural and aerodynamic resistances and is partitioned between canopy evaporation and
562 transpiration based on the intercepted water content of the canopy (Deardorff, 1978; Ducoudre et al.,
563 1993). Bare soil evaporation is constrained by the saturated area of the upper soil layer. The saturated
564 area is variable within the grid since (as the model name implies) the infiltration capacity of the soil is
565 assumed heterogeneous (Franchini and Pacciani, 1991). Saturated areas evaporate at the potential
566 evaporation rate while in unsaturated areas evaporation is limited. Surface runoff is produced by
567 precipitation over saturated areas. Precipitation over unsaturated areas infiltrates into the upper soil layer
568 and drains through the soil layers based on the gravitational hydraulic conductivity equations of Brooks
569 and Corey (1964). In the first and second layer water is available for transpiration, while the third layer
570 is assumed to be below the root zone. From the third layer baseflow is generated based on the non-linear
571 Arno conceptualization (Franchini and Pacciani, 1991). Baseflow increases linearly with soil moisture
572 content when the moisture content is low. At higher soil moisture contents the relation is non-linear,
573 representing subsurface storm-flows.

574 For the energy balance, incoming net radiation is partitioned between sensible, latent, and ground heat
575 fluxes and energy storage in the air below the canopy. The energy storage below the canopy is omitted
576 if it is considered negligible (e.g. the canopy surface is open or close to the ground). The latent heat flux

577 is determined by the evapotranspiration as calculated in the water balance. The sensible heat flux is
578 calculated based on the difference between the air and surface temperature and the ground heat flux is
579 calculated based on the difference between the soil and surface temperature. Since the incoming net
580 radiation is also a function of the surface temperature (specifically the outgoing longwave radiation),
581 the surface temperature is solved iteratively. Subsurface ground heat fluxes are calculated assuming an
582 exponential temperature profile between the surface and the bottom of the soil column, where the bottom
583 temperature is assumed constant. Later model developments included options for finite difference
584 solutions of the ground temperature profile (Cherkauer and Lettenmaier, 1999), spatial distribution of
585 soil temperatures (Cherkauer and Lettenmaier, 2003), a quasi-2-layer snow-pack snow model
586 (Andreadis et al., 2009), and blowing snow sublimation (Bowling et al., 2004).

587 **6.2 Appendix B: EFRs for surface and groundwater**

588 VIC-WUR used the Variable Monthly Flow (VMF) method (Pastor et al., 2014) to limit surface water
589 withdrawals. The VMF method (Pastor et al., 2014) calculates the EFRs for streamflow as a fraction of
590 the natural flow during high (Eq. A.1), intermediate (Eq. A.2) and low (Eq. A.3) flow periods. The
591 presumptive standard Gleeson and Richter (2018) is used to limit groundwater withdrawals -(including
592 aquifer groundwater withdrawals). This standard calculates the EFRs for baseflow as 90 % of the natural
593 subsurface runoff through time (Eq. A.4). Here, daily instead of monthly EFRs were used to better
594 capture the monthly flow variability.

$$595 \quad EFR_{s,d} = 0.6 \cdot NF_{s,d} \quad \text{Eq. (A.1)}$$

$$596 \quad \text{where } NF_{s,d} \leq 0.4 \cdot NF_{s,y}$$

$$597 \quad EFR_{s,d} = 0.45 \cdot NF_{s,d} \quad \text{Eq. (A.2)}$$

$$598 \quad \text{where } 0.4 \cdot NF_{s,y} < NF_{s,d} \leq 0.8 \cdot NF_{s,y}$$

$$599 \quad EFR_{s,d} = 0.3 \cdot NF_{s,d} \quad \text{Eq. (A.3)}$$

$$600 \quad \text{where } NF_{s,d} > 0.8 \cdot NF_{s,y}$$

$$601 \quad EFR_{b,d} = 0.9 \cdot NF_{b,d} \quad \text{Eq. (A.4)}$$

602 Where $EFR_{s,d}$ is the daily EFRs for streamflow [$\text{m}^3 \text{s}^{-1}$], $EFR_{b,d}$ the daily EFRs for baseflow [$\text{m}^3 \text{s}^{-1}$],
603 $NF_{s,d}$ is the average natural daily streamflow [$\text{m}^3 \text{s}^{-1}$], and $NF_{s,y}$ is the average natural yearly streamflow
604 [$\text{m}^3 \text{s}^{-1}$], and $NF_{b,d}$ is the average natural daily baseflow [$\text{m}^3 \text{s}^{-1}$].

605 EFRs for streamflow and baseflow were based on VIC-WUR naturalized simulations between 1980 and
606 2010. Average natural daily flows were calculated as the interpolated multi-year monthly average flow
607 over the simulation period.

608 **6.3 Appendix C: Dam operation scheme**

609 VIC-WUR used a dam operation scheme based on Hanasaki et al. (2006). Target release (i.e. the
610 estimated optimal release) was calculated at the start of the operational year. The operational year starts
611 at the month where the inflow drops below the average annual inflow, and thus the storage should be at
612 its desired maximum. The scheme distinguished between two dam types: (1) dams that did not account
613 for water demands downstream (e.g. hydropower dams or flood control) and (2) dams that did account
614 for water demands downstream (e.g. irrigation dams). The original scheme of Hanasaki et al. (2006)
615 also accounts for EFRs, which were fixed at half the annual mean inflow. Other studies lowered the
616 requirements to a tenth of the mean annual inflow, increasing irrigation availability and preventing
617 excessive releases (Biemans et al., 2011; Voisin et al., 2013b). In our study the original dam operation
618 scheme was adapted slightly to account for monthly varying EFRs.

619 For dams that did not account for demands, the initial release was set at the mean annual inflow corrected
620 by the variable EFRs (Eq. A.5). For dams that did account for demands, the initial release was increased
621 during periods of higher water demand. If demands were relatively high compared to the annual inflow,
622 the release was corrected by the demand relative to the mean demand (Eq. A.6). If demands were
623 relatively low compared to the annual inflow, release was corrected based on the actual water demand
624 (Eq. A.7).

625

$$626 \quad R'_m = EFR_{s,m} + (I_y - EFR_{s,y}) \quad \text{Eq. (A.5)}$$

627 *where* $D_y = 0$

628 $R'_m = EFR_{s,m} + (I_y - EFR_{s,y}) * \frac{D_m}{D_y}$ Eq. (A.6)

629 where $D_y > 0$ and $D_y > (I_y - EFR_{s,y})$

630 $R'_m = EFR_{s,m} + (I_y - EFR_{s,y}) - D_y + D_m$ Eq. (A.7)

631 where $D_y > 0$ and $D_y \leq (I_y - EFR_{s,y})$

632 Where R'_m is the initial monthly target release [$m^3 s^{-1}$], $EFR_{s,m}$ is the average monthly EFR for
 633 streamflow demand [$m^3 s^{-1}$], I_y is the average yearly inflow [$m^3 s^{-1}$], $EFR_{s,y}$ is the average yearly EFR
 634 for streamflow [$m^3 s^{-1}$], D_m is the average monthly water demand [$m^3 s^{-1}$] and D_y is the average yearly
 635 water demand [$m^3 s^{-1}$].

636 As in Hanasaki et al. (2006), the initial target release was adjusted based on storage and capacity. Target
 637 release was adjusted to compensate differences between the current storage and the desired maximum
 638 storage (Eq. A.8). Target release was additionally adjusted if the storage capacity is relatively low
 639 compared to the annual inflow, and unable to store large portions of the inflow for later release (Eq.
 640 A.9).

641 $R_m = k \cdot R'_m$ Eq. (A.8)

642 where $c \geq 0.5$

643 $R_m = \left(\frac{c}{0.5}\right)^2 \cdot k \cdot R'_m + \left\{1 - \left(\frac{c}{0.5}\right)^2\right\} \cdot I_m$ Eq. (A.9)

644 where $0 \leq c \leq 0.5$

645 Where I_y is the average monthly inflow [$m^3 s^{-1}$], c the capacity parameter [-] calculated as the storage
 646 capacity divided by the mean annual inflow and k the storage parameter [-] calculated as current storage
 647 divided by the desired maximum storage. The desired maximum storage was set at 85 % of the storage
 648 capacity as recommended by Hanasaki et al. (2006).

649 Water inflow, demand and EFRs were estimated based on the average of the past five years. Water
 650 demands were based on the water demands of downstream cells. Only a fraction of water demands were
 651 taken into account, based on the fraction of discharge the dam controlled. For example: if a dam

652 controlled 70 % of the discharge of a downstream cell, than 70 % of its demands were taken into account.

653 Fractions smaller than 25 % were ignored.

654 The original dam operation scheme of Hanasaki et al. (2006) was shown to produce excessively high

655 discharge events due to overflow releases ~~(Masaki et al., 2018b)~~(Masaki et al., 2018). These overflow

656 releases occurred due to a mismatch between the expected and actual inflow. In our study, dam release

657 was increased during high-storage events to prevent overflow and accompanying high discharge events.

658 If dam storage was above the desired maximum storage, target dam release was increased to negate the

659 difference (Eq. A.10). If dam storage was below the desired minimum storage, release is decreased (Eq.

660 A.11). Dam release was adjusted exponentially based on the relative storage difference: small storage

661 differences were only corrected slightly, but if the dam was close to overflowing or emptying, the

662 difference was corrected strongly.

$$663 \quad R_a = R_m + \frac{(S-C\alpha)}{\gamma} \cdot \left(\frac{S}{C-\alpha} \right)^b \quad \text{Eq. (A.10)}$$

664 *where* $S > C\alpha$

$$665 \quad R_a = R_m + \frac{(S-C(1-\alpha))}{\gamma} \cdot \left(\frac{(1-\alpha)-\frac{S}{C}}{1-\alpha} \right)^b \quad \text{Eq. (A.11)}$$

666 *where* $S < C(1-\alpha)$

667 Where R_a is the actual dam release [$\text{m}^3 \text{s}^{-1}$], S the dam storage capacity [m^3], α the fraction of the capacity

668 that is the desired maximum [-], β the exponent determining the correction increase [-] and γ the

669 parameter determining the period when the release is corrected [s^{-1}]. In testing the exponent and period

670 were tuned to 0.6 and 5 days respectively.

671 **6.4 Appendix D: Water demand**

672 **6.4.1 Fitting and validation data**

673 Data on irrigation, domestic and industrial water withdrawals were based on the AQUASTAT database

674 (FAO, 2016), EUROSTAT database (EC, 2019) and United Nations World Water Development Report

675 (Connor, 2015). Data on GDP per capita and GVA was abstracted from the Maddison Project Database

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676 2018 (Bolt et al., 2018), Penn World Table 9.0 (Feenstra et al., 2015) and World Bank Development
677 Indicators (World bank, 2010).

678 Available data for domestic and industrial withdrawals were divided into a dataset used for parameter
679 fitting (80 %) and a dataset used for validation (20 %). Domestic water demands were estimated for
680 each United Nations sub-region, and thus data was divided per sub-region to ensure a good global
681 coverage of data. In the same manner industrial water demand were divided per country. In case there
682 is only a single data point, the data was added to both the fitting and validation data.

683 6.4.2 Irrigation sector

684 Conventional irrigation demands were calculated when soil moisture contents drop below the critical
685 threshold where evapotranspiration will be limited. Demands were set to relieve water stress (Eq. A.12).

686 Paddy irrigation demands were set to always keep the soil moisture content of the upper soil layer
687 saturated (Eq. A.13), similar to ~~Hanasaki et al. (2008a)~~Hanasaki et al. (2008b) and Wada et al. (2014).

688 For paddy irrigation, the saturated hydraulic conductivity of the upper soil layer was reduced by its
689 cubed root to simulate puddling practices, as recommended by the CROPWAT model (Smith, 1996).

690 Total irrigation demands were adjusted by the irrigation efficiency (Eq. A.14). Paddy irrigation used an
691 irrigation efficiency of 1 since the water losses were already incorporated in the water demand
692 calculation.

$$693 ID'_{conventional} = (W_{cr,1} + W_{cr,2}) - (W_1 + W_2) \quad \text{Eq. (A.12)}$$

$$694 \quad \text{where } W_1 + W_2 < W_{cr,1} + W_{cr,2}$$

$$695 ID'_{paddy} = W_{max,1} - W_1 \quad \text{Eq. (A.13)}$$

$$696 \quad \text{where } W_1 < W_{max,1}$$

$$697 ID = ID' * IE \quad \text{Eq. (A.14)}$$

698 Where $ID'_{conventional}$ is the conventional crop irrigation demand [mm], ID'_{paddy} is the paddy crop irrigation
699 demand [mm], ID is the total irrigation demand [mm], W_1 and W_2 are the soil moisture contents of the

Commented [DB17]: Updated references

700 first and second soil layer respectively [mm], W_{cr} is the critical soil moisture content [mm], W_{max} the
701 maximum soil moisture content [mm], and IE is the irrigation efficiency [mm mm⁻¹].

702 6.4.3 Domestic sector

703 Domestic water demands were represented by using a sigmoid curve for the calculation of structural
704 domestic water demands (Eq.A.15) and a efficiency rate for the calculation of water-use efficiency
705 increases (Eq. A.16). These equations differ slightly from Alcamo et al. (2003) since our study used the
706 base 10 logarithms of GDP and water withdrawals per capita as they provided a better fit.

$$707 \quad DSW_y = DSW_{min} + (DSW_{max} - DSW_{min}) * \frac{1}{1+e^{-f(GDP_y-o)}} \quad \text{Eq. (A.15)}$$

$$708 \quad DW_y = 10^{DSW_y} \cdot TE^{y-y_{base}} \quad \text{Eq. (A.16)}$$

709 Where DSW is the yearly structural domestic withdrawal [log10 m³ cap⁻¹], DW the yearly domestic
710 withdrawal [m³ cap⁻¹], DW_{min} the minimum structural domestic withdrawal [log10 m³ cap⁻¹], DW_{max} the
711 maximum structural domestic withdrawal (without technological improvement) [log10 m³ cap⁻¹], GDP
712 the yearly gross domestic product [log10 USD_{equivalent} cap⁻¹], f [-] and o [log10 USD_{equivalent}] the
713 parameters that determine the range and steepness of the sigmoid curve, y the year index, TE the
714 technological efficiency rate [-], and y_{base} the base year (taken to be 1980).

715 DW_{min} was set at 7.5 l cap⁻¹ d⁻¹ based on the World Health Organisation standard (Reed and Reed, 2013),
716 DW_{max} was estimated at around 450 l cap⁻¹ y⁻¹ based on a global curve fit, and TE was set at 0.995, 0.99,
717 and 0.98 for developing, transition and developed countries respectively (United Nations development
718 status classification) based on Flörke et al. (2013). Curve parameters f and o were estimated for the 23
719 United Nations sub-regions based on the GDP per capita and domestic water withdrawal data. In case
720 insufficient data was available to calculate parameters values, regional (4 sub-regions) or global (4 sub-
721 regions) parameter estimates were used.

722 **6.4.4 Industrial sector**

723 Industrial water demands were represented by using a linear formula for the calculation of structural
724 industrial water demands (Eq. A.17) and a efficiency rate for the calculation of water-use efficiency
725 increases (Eq. A.18).

726 $ISW_y = ISW_{int} \cdot GVA_y$ Eq. (A.17)

727 $IW_y = ISW_y \cdot TE^{y-y_{base}}$ Eq. (A.18)

728 Where ISW is the yearly structural industrial withdrawal [m^3], IW_{int} the country specific industrial water
729 intensity [$m \text{ USD}_{equivalent}^{-1}$], IW the yearly industrial withdrawal [m^3], GVA the yearly gross value added
730 by industry [$\text{USD}_{equivalent}$], y the year index, y_{base} the base year (taken to be the year when the industrial
731 water intensity is determined), and TE the technological efficiency rate [-].

732 TE was set at 0.976 and 1 for OECD and non-OECD countries respectively before the year 1980, 0.976
733 between the years 1980 and 2000 and 0.99 after the year 2000 based on Flörke et al. (2013). Industrial
734 water intensities were estimated for the 246 United Nations countries based on the GVA and industrial
735 water withdrawal data. In case insufficient data was available to calculate the industrial water intensities,
736 either sub-regional (56 countries), regional (17 countries) or global (9 countries) intensities estimates
737 were used.

738 **6.4.5 Energy sector**

739 For each thermoelectric power plant the water intensity was combined with their generation to calculate
740 the water demands (Eq. A.19). Actual generation is estimated by adjusting the installed generation
741 capacity by 46 % for fossil, 72 % for nuclear and 56 % for biomass power plants (based on EIA national
742 annual generation data (EIA, 2013))

743 $EW_y = EW_{int} \cdot G_y$ Eq. (A.19)

744 Where EW is the yearly energy withdrawal [m^3], EW_{int} the energy water intensity [$m^3 \text{ MWh}^{-1}$], G the
745 yearly generation for each plant [MWh], and y the year index.

746 The energy water demands were subtracted from the industrial water demands at the location of each
747 power plant. In cases where the grid cell industrial water demand was less than the energy water demand,
748 national industrial water demands were lowered. In cases where even the national industrial water
749 demands were less than the national energy water demand (3 countries), the energy water demands were
750 lowered instead. Energy demands were lowered until 10 % of the national industrial water demand
751 remains, to ensure some spatial coverage of industrial and energy water demands.

752 **6.4.6 Livestock sector**

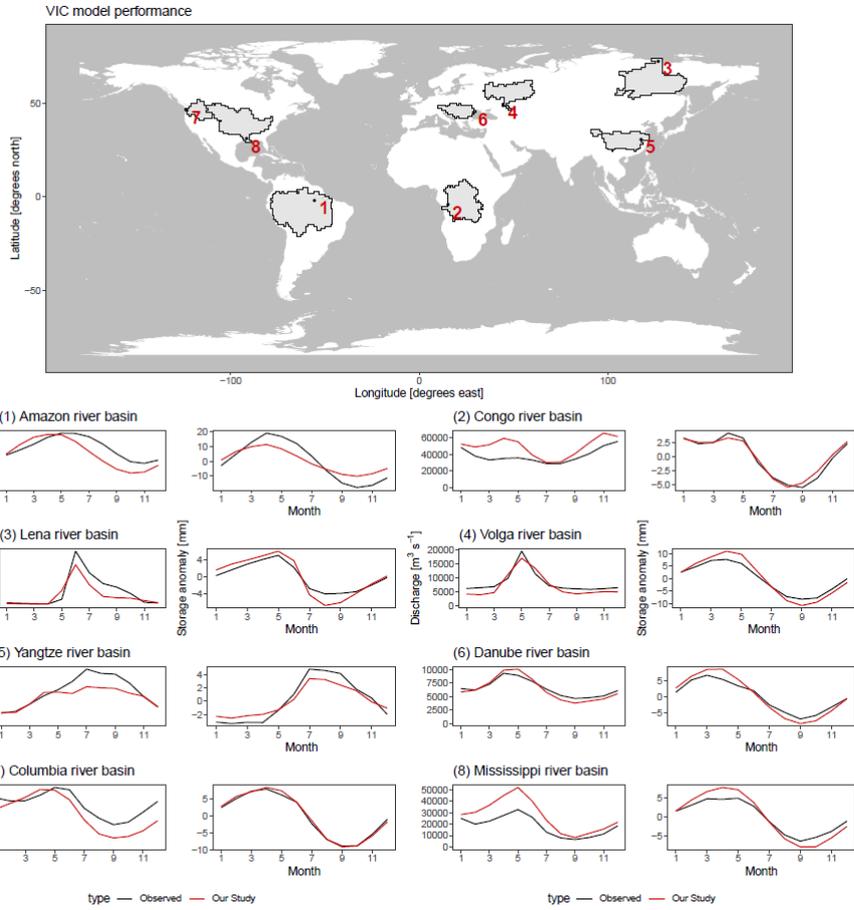
753 Livestock water demands were estimated by combining the livestock population with the water
754 requirements for each livestock variety (Eq. A.20).

$$755 \quad LW_y = LW_{int} \cdot L \quad \text{Eq. (A.20)}$$

756 Where LW is the yearly livestock withdrawal [m^3], LW_{int} the livestock water intensity [$\text{m}^3 \text{livestock}^{-1}$],
757 L the livestock number for each variety [livestock].

758 **6.5 Appendix E: General performance**

759 VIC-WUR monthly discharge and monthly terrestrial total water storage anomalies were compared with
760 observations from the GRDC dataset (GRDC, 2003) and GRACE satellite dataset (NASA, 2002) for
761 eight major river basins (not included in the main text). Discharge stations were selected if the upstream
762 area was larger than 10000 m^2 , matched the simulated upstream area at the station location,): Amazon,
763 Congo, Lena, Volga, Yangtze, Danube, Columbia and Mississippi river basins. A 300km gaussian filter
764 has been applied to the total water storage simulation data (similar to Long et al. (2015)).



765

766 **Figure A1: Comparison between simulated and observed discharge and terrestrial total water storage anomalies.**

767 **Figures indicate multi-year averages of human-impacted simulations (red) and observations (black).**

768 **7 Author contribution**

769 Bram Droppers and Wietse H.P. Franssen developed and tested the model additions introduced in VIC-

770 WUR. Bram Droppers generated and analysed the results. Michelle T.H. van Vliet, Bart Nijssen and

771 Fulco Ludwig provided overall oversight and guidance. Bram Droppers prepared the manuscript with

772 contributions from all co-authors.

773 **8 Competing interests**

774 The authors declare that they have no conflict of interest.

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