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	Widespread and increasing violations of environmental flow	
	envelopes	Deleted: : quantifying global, ecosystem-threatening streamflow alterations
5	Vili Virkki ^{1*#} , Elina Alanärä ^{1#} , Miina Porkka ^{1,2} , Lauri Ahopelto ^{1,3} , Tom Gleeson ^{4,5} , Chinchu Mohan ^{4,6} , Lan Wang-Erlandsson ^{7,8} , Martina Flörke ⁹ , Dieter Gerten ^{10,11} , Simon N Gosling ¹² , Naota Hanasaki ¹³ , Hannes Müller Schmied ^{14,15} , <u>Niko Wanders¹⁶</u> , Matti Kummu ^{1*}	
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	¹⁰ Potsdam Institute for Climate Impact Research (PIK), Member of the Leibniz Association, Potsdam, Germany ¹¹ Humboldt-Universität zu Berlin, Geography Department, Berlin, Germany	Deleted: . However,
	¹² School of Geography, University of Nottingham, Nottingham, NG7 2RD, United Kingdom	Deleted: inwever,
20	 ¹³ National Institute for Environmental Studies, Tsukuba, Japan ¹⁴ Institute of Physical Geography, Goethe University Frankfurt, Frankfurt am Main, Germany 	Deleted: method to determine EFs by Environmental Flow Envelopes (
	¹⁵ Senckenberg Leibniz Biodiversity and Climate Research Centre (SBiK-F), Frankfurt am Main, Germany	Deleted:), which
	¹⁶ Utrecht University, Department of Physical Geography, Utrecht, The Netherlands # equal contribution to the article	Deleted: an envelope of variability bounded by discharge limits within which riverine ecosystems are not seriously compromised. The EFE is defined globally in
25	* Correspondence to: Vili Virkki (vili.virkki@aalto.fi), Matti Kummu (matti.kummu@aalto.fi)	Deleted: -
25	Correspondence to: Viii Viikki (Viii. Viikki @aano.11), Matti Kullinu (Illatti. Kullinu @aano.11)	Deleted: considering also
	Abstract. Human actions and climate change have drastically altered river flows across the world, resulting in adverse effects	Deleted: , the EFE introduces
	on riverine ecosystems. Environmental flows (EFs) have emerged as a prominent tool for safeguarding the riverine ecosystems,	Deleted: ,
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	but at the global scale, the assessment of EFs is associated with high uncertainty. Here, we present a novel, in-depth global EF	Deleted:
	assessment using environmental flow envelopes (EFEs). The sub-basin specific EFE is determined for approximately 4,400	Deleted: ,
30	sub _c basins at <u>a</u> monthly time resolution, and its derivation considers the methodological uncertainties related with global-scale	Deleted: commonly done
	EF studies. In addition to a lower bound of discharge, based on existing EF methods, we introduce an upper bound of discharge,	Deleted: , which can be considered as potential threats to rivering ecosystems
	in the EFE. This enables identifying areas where streamflow has substantially increased above natural levels. Further, instead	Formatted: Not Superscript/ Subscript
	of only showing whether EFs are violated as a long-term average, we quantify, for the first time, the frequency, severity, and	Deleted: -
	trends of EFE violations during the recent historical period,	Deleted: and
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20	We use pre-industrial (1801–1860) quasi-natural discharge together with a suite of hydrological EF methods to estimate the	Deleted: and global hydrological models
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	EFEs, applying global hydrological model outputs from the ISIMIP 2b ensemble. We then compare the EFEs to recent	Deleted: data

- 65 historical (1976–2005) discharge to assess the violations of the EFE. These violations most commonly manifest themselves by insufficient streamflow during the loweflow season, with fewer violations during the intermediate flow season, and only a few violations during the high flow season. The EFE violations are widespread, occurring in half of the sub basins of the world during more than 5% of the months between 1976 and 2005, which is double compared to the pre-industrial period. The trends in EFE violations have mainly been increasing, which will likely continue in the future with the projected hydroclimatic
- 70 changes and increases in anthropogenic water use, Indications of increased upper extreme streamflow through EFE upper bound violations are relatively scarce and spatially distributed. While the <u>EFEs provide</u> a quick and globally robust way of determining environmental flow allocations at the sub basin scale, local fine tuning is necessary for practical applications, and further research on the coupling between quantitative discharge and riverine ecosystem responses at the global scale is required.

1 Introduction

- 75 Human exploitation of rivers is a sensitive balance between benefits gained from water use and adverse Earth system responses. While also enabling the development of societies, rivers upkeep two major regulatory Earth system functions: maintaining the hydrological cycle_a and providing habitat for freshwater ecosystems (Gleeson et al., 2020). Nonetheless, they are subject to high anthropogenic pressure e.g. from flow regulation and damming, excessive water withdrawals, pollution, and land use change (Best, 2019; Kummu et al., 2016). Moreover, human induced climate change can increase or decrease the seasonal
- 80 streamflow at different spatial scales (Arnell and Gosling, 2013; Asadieh and Krakauer, 2017; Gudmundsson et al., 2021; Moragoda and Cohen, 2020. The pressure on freshwater ecosystems is only expected to increase in the future due to population growth and projected climate change (Best, 2019; Graham et al., 2020; Thompson et al., 2021).
- Freshwater ecosystems contain nearly 6% of all known species concentrated in 0.8% of Earth's surface (Dudgeon et al., 2006).
 The riverine parts of freshwater ecosystems have been seriously compromised by human actions: rivers containing 65% of the global discharge are classified to be under moderate to high threat in terms of biodiversity (Vörösmarty et al., 2010), 53% of global rivers have experienced marked changes in fish biodiversity (Su et al., 2021), and 48% of global river reaches are impaired by diminished connectivity (Grill et al., 2019). One of the root causes behind this degradation is the anthropogenic alteration of the natural flow regime of a river, i.e. the magnitude, frequency, duration, timing, and rate of change in flow
 (Poff et al., 1997). In addition, human actions impact the intra- and interannual variability, which are often considered as parts
- of the natural flow regime (Richter et al., 2006). These natural streamflow dynamics have already changed in major rivers across the globe (Grill et al., 2015). The flow regime is one of the key factors in defining the integrity of riverine ecosystems, as it maintains their physical habitat as well as their longitudinal and lateral connectivity (Bunn and Arthington, 2002). Furthermore, aquatic species have evolved within and adapted to the natural flow regime, and alterations to it may facilitate

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95 invasive species. Therefore, although riverine ecosystems are extremely complex, the association between flow regime alteration and riverine ecosystem integrity is strong (Poff and Zimmerman, 2010; Rolls et al., 2018).

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To safeguard riverine ecosystems, the concept of environmental flows (hereafter EFs) has emerged during the past three decades (Poff and Matthews, 2013). While multiple definitions of EFs exist, the most comprehensive recent definition comes
140 from the Brisbane Declaration 2018 (Arthington et al., 2018), which states that "environmental flows describe the quantity, timing, and quality of freshwater flows and levels necessary to sustain aquatic ecosystems which, in turn, support human cultures, economies, sustainable livelihoods, and well being." To date, many countries have initiated legislation that would support the establishment of EFs as a concrete means of conserving and restoring riverine ecosystems (Acreman et al., 2014; Arthington et al., 2018; Tickner et al., 2020). In an ideal case, EFs are quantified by assimilating observed hydrological data
145 with local-scale expert knowledge in a collaborative process, resulting in EFs tailored for each unique river (Richter et al., 2006; Poff et al., 2017). Such holistic EF methods include, for example, ELOHA (Poff et al., 2010), DRIFT (King et al., 2003), and PROBFLO (O'Brien et al., 2018).

While the holistic methods available to quantify EFs are comprehensive, the data required to implement them are unavailable
 at a global scale. Hence, in global studies, the concept of EFs is typically quantified by computing environmental flow requirements (EFRs) based on hydrological EF methods (Pastor et al., 2014). These methods assume that not transgressing EFRs will retain a fair state of riverine ecosystems. Although this proxy relationship is uncertain and varies across spatial and temporal scales (Bunn and Arthington, 2002; Poff and Zimmerman, 2010; Rolls et al., 2018), the hydrological EF methods are often used in global studies as presumptive standards of sustaining riverine ecosystems (Gerten et al., 2020, 2013; Hanasaki

- 155 et al., 2008; Hoekstra and Mekonnen, 2011; Hogeboom et al., 2020; Jägermeyr et al., 2017; Liu et al., 2021; Pastor et al., 2014, 2019; Steffen et al., 2015). In addition to ecological uncertainty, discharge data used for determining EFRs in global studies / are uncertain; runoff and discharge estimated by global hydrological models (GHMs) that are forced with modelled climate from general circulation models (GCMs) tend to be highly dispersed between different GHMs and GCMs (Gädeke et al., 2020; Hattermann et al., 2018; Müller Schmied et al., 2016; Schewe et al., 2014; Veldkamp et al., 2018; Zaherpour et al., 2019). As
- 160 the <u>underlying hydrological data</u> are generally uncertain, determining EFRs based on <u>them</u> and hydrological <u>EF</u> methods is equally uncertain. Moreover, hydrological <u>EF</u> methods often <u>only set</u> a minimum discharge boundary, disregarding the potentially adverse effects of flows increasing significantly above natural levels <u>–</u> especially in floodplain ecosystems (Hayes et al., 2018; Junk et al., 1989; Schneider et al., 2017; Talbot et al., 2018). Although reviews of EFs have recognised <u>the threat</u> of <u>increased upper extreme</u> flows (Acreman et al., 2014; Poff and Zimmerman, 2010; Richter, 2010), <u>and limiting upper</u> extreme flows has been conceptually proposed (Richter et al., 2012), <u>a</u> global scale methodology to quantify <u>this does not yet</u>.

exist.

Existing global studies are also limited in their EF violation assessment. Commonly, EFs are treated in global studies as simple, long-term monthly or annual limits that are either violated or not, lacking quantification of how frequently or how severely these violations manifest themselves (Pastor et al., 2014; Steffen et al., 2015). Some, more detailed studies incorporate

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additional factors, such as the magnitude by which EFs are violated, but do not account for the seasonality of streamflow (Hogeboom et al., 2020; Jägermeyr et al., 2017). Given that particularly low flows are often the most impacted by anthropogenic actions, such as water withdrawals and flow regulation by damming (Döll et al., 2009; Schneider et al., 2017), EF assessments should be able to separate violations during different flow seasons. Finally, while recent studies have shown that river flows have changed considerably due to direct human actions (Graham et al., 2020; Müller Schmied et al., 2016) and climate change (Gudmundsson et al., 2021; Moragoda and Cohen, 2020) during the past decades, no study has yet assessed the past trends in EF violations. Therefore, new knowledge is required to compose a combined and comprehensive outlook on

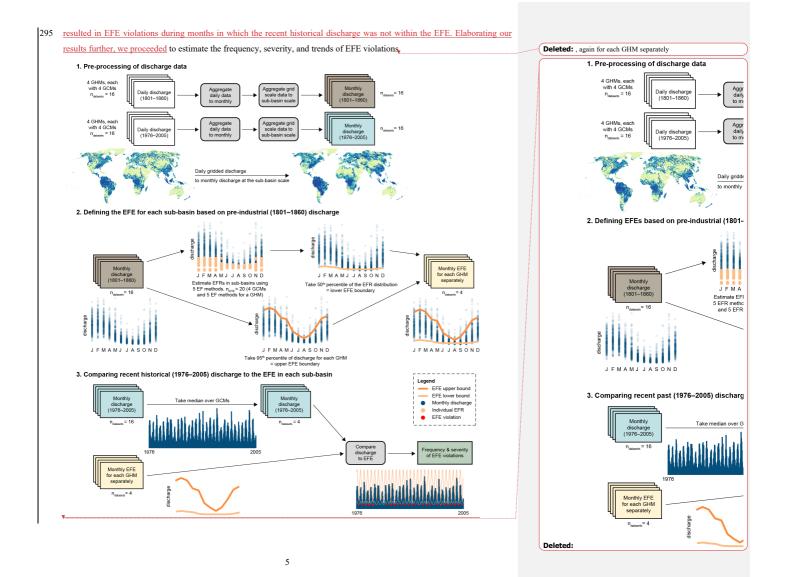
250 the past trends in EP violations. Therefore, new knowledge is required to compose a combined and comprehensive outlook these three aspects of EF violation.

Here, we present an in-depth global EF assessment by applying a robust, global scale methodology of environmental flow / envelopes (EFEs). Defined at the sub basin scale in monthly time resolution, the EFE is an envelope of discharge variability, which advances the existing methods in two main ways. First, in order to reduce uncertainties in global EF assessments, the EFE is composed of a number of hydrological EF methods applied to an ensemble of GHM outputs simulated using multiple. GCMs. Secondly, we include a preliminary upper bound in the EFE, aiding in identifying areas where streamflow has increased substantially above the presumed natural levels. In addition to the methodological advances, we present a novel quantification of the seasonal frequency, severity, and trends of EFE violations by comparing recent historical (1976–2005), anthropogenically influenced discharge to pre-industrial (1801–1860) state EFEs.

2 Methods and data

Estimating EFE violations was divided into three parts, which are outlined in Fig. 1 and detailed in the following sections. Our method is based on discharge data, which are simulated by four global hydrological models (GHMs). Simulating discharge with the GHMs involves modelling the global terrestrial hydrological cycle through mechanistic equations, as well as forcing the models with observed or modelled climate. For this study, we used modelled climate from four different general circulation models (GCMs). Combining each of the four GHMs with four GCMs provided us with 16 distinct data sets of gridded global-scale daily discharge. First, for each discharge data set in this 16-member ensemble, we transformed the gridded daily discharge to monthly discharge at the sub₂basin scale according to HydroBASINS sub₂basin division, (Lehner and Grill, 2013). This was done separately for the pre-industrial period (1801–1860) and the recent historical period (1976–2005). Second, we took the
pre-industrial monthly discharge for each GHM at the sub-basin scale and estimated EFRs using five hydrological EF methods for four discharge data sets from different GCMs. This totalled 20 EFRs (5 EF methods x 4 GCMs) for each GHM and subbasin. From this EFR distribution, we drew the median as the GHM-specific EFE lower bound for each sub-basin. Further, we determined the EFE upper bound from the pre-industrial discharge – again, separately for each GHM and sub-basin. Finally, we took the monthly discharge at the sub-basin scale from the recent historical period and compared it to the EFEs. This

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300 Figure 1. The methodological outline of this study: defining the environmental flow envelope (EFE) and estimating the frequency and severity of EFE violations in each sub-basin. GHM stands for global hydrological model, GCM for general circulation model, and EFR for environmental flow requirement.

2.1 Data

 We used the Inter-Sectoral Impact Model Intercomparison Project (ISIMIP) simulation round 2b outputs of global daily
 discharge (Frieler et al., 2017; available at: https://esg.pik-potsdam.de). ISIMIP is a communitycdriven climate_impacts modelling initiative that collects and harmonises global model outputs (The Inter-Sectoral Impact Model Intercomparison Project, 2021). To guarantee cross-model consistency regarding the parameterisation of both human and climate factors, the ISIMIP 2b experiments are directed by a protocol (Frieler et al., 2017). Due to the protocol, the outputs of ISIMIP 2b are comparable between the pre-industrial and recent historical period model runs, as well as between different models.
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I	To decrease the uncertainties related to using single GHMs with single or few GCMs, we chose to use discharge estimates
1	from four different GHMs, namely H08 (Hanasaki et al., 2018), LPJmL (Schaphoff et al., 2018), PCR-GLOBWB (Sutanudjaja
	et al., 2018), and WaterGAP2 (Müller Schmied et al., 2016). In ISIMIP 2b, each of the GHMs is forced with modelled climate
	from four GCMs, namely GFDL-ESM2M, (Dunne et al., 2012), HadGEM2-ES, (Collins et al., 2011), IPSL-CM5A-LR,
315	(Dufresne et al., 2013), and MIROC5, (Watanabe et al., 2010). All of these GCMs were included in our discharge ensemble.
	The selected GHMs have undergone extensive intercomparison and validation with observed data (see e.g. Gädeke et al., 2020;
	Zaherpour et al., 2018). Furthermore, the hydrologically best fitting model varies largely based on sub-basin characteristics
	(Zaherpour et al., 2018), which is why we chose to estimate EFEs for many GHMs instead of pursuing the hydrological best
	fit and estimating EFEs for one GHM only. This ensemble approach decreases uncertainty stemming from two separate
320	sources: (1) using more than one GCM within one GHM decreases the GHM parameterisation uncertainty, and (2) using a
•	number of GHMs in an analysis decreases the uncertainty of modelling the hydrological cycle within a single GHM (Döll et
1	al., 2016; Schewe et al., 2014; Sood and Smakhtin, 2015).
•	
	The discharge data (for both periods 1801-1860 and 1976-2005) were first temporally aggregated from daily to monthly

325 discharge by <u>calculating the monthly mean of daily values</u>, <u>followed by a spatial aggregation</u> at the sub basin scale according to the HydroBASINS level 5. HydroBASINS is a global polygon layer series, which divides the world into <u>sub basins at</u> different <u>scale levels from the lowest detailed level 1 to the highest detailed level 12</u> (Lehner and Grill, 2013). Within each <u>level</u>, the geographical areas of sub-basins are relatively equal. Level 5 was selected since it is the highest level of detail that <u>can be rasterized into a 0.5</u> degree resolution grid without an excessive loss of sub basins that are smaller than a grid cell. In <u>330</u> total, 352 out of 4,734 sub basins were excluded due to being smaller than a grid cell, corresponding to less than 1 million <u>1</u>

km². or 1% of the global land area. The average size of the remaining sub_x basins was 30,700 km² and median size 19,600 km².
 Minor additional exclusions of five to six sub_x basins per GHM were caused by non_x overlapping discharge data grids. To aggregate the discharge at the sub_x basin scale, we selected the maximum discharge value within the borders of each sub_x basin,

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	Simple metrics, such as the ensemble mean or median, ide globally decent estimates when compared to observed
discharge (see e.g.
	(2015) and Huang et al., (2017)), although individual
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assuming that the sub_k basin drains out from that <u>location</u>. Hence, we consider this <u>location</u> – and any violation in it – as representative of the whole sub_k basin, though the situation may vary in different parts of the sub_k basin.

380 2.2 Defining EFEs

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We defined the EFEs based on the pre-industrial period (1801–1860). As no significant human modification (e.g. large dams and extensive irrigation schemes) had occurred by 1860, we presumed that this time period is quasi-natural – i.e. near the natural flow regime. Furthermore, reaching back to the pre-industrial time period enabled us to quantify the joint effect of both direct anthropogenic flow alteration and long-term anthropocentric climate change, although explicit separation between these two drivers is not possible with the ISIMIP 2b data used in this study.

Following Pastor et al. (2014), we selected five <u>hydrological EF</u> methods to accommodate for the differences in the methods' definitions of ecosystem water needs. The selected methods include Smakhtin's method (Smakhtin et al., 2004), Tennant's method (Tennant, 1976), Tessmann's method (Tessmann, 1980), the Q90-Q50-method (Pastor et al., 2014), and the variable
 monthly flow (VMF) method (Pastor et al., 2014). These methods have been validated against *in situ* EFR estimates by Pastor

et al. (2014) and Jägermeyr et al. (2017). We opted to use these relatively simplistic hydrological EF methods because more advanced EF quantifications, such as ELOHA-based methods, are limited in global scale applicability due to the high data and resource requirements (Richter et al., 2012). The selected methods are based on simple flow metrics, such as mean annual or monthly flow, and they determine EFRs according to hydrological seasons. All methods distinguish between low_eflow and high_eflow months, while the Tessmann and VMF methods supplement this with a third class for intermediate_eflow months.

The equations to compute EFRs according to the selected <u>EF</u> methods are presented in Table 1.

 Table 1. Descriptions of hydrological EF methods used to calculate environmental flow requirements (EFRs) in this study (adapted from Pastor et al. (2014)). MMF refers to mean monthly flow of each month, MAF to mean annual flow (the mean monthly flow of all months within a year), Q50 and Q90 to flow exceeding 50% and 90% of the flows during the period of interest respectively, and coefHF to high-flow coefficient used in Smakhtin's method.

Hydrological	Smakhtin	Tennant	Q90-Q50	Tessmann	Variable monthly flow
season	(2004)	(1976)	(Pastor et al 2014)	(1980)	(Pastor et al., 2014)
Low	$MMF \le MAF$	$MMF \le MAF$	$MMF \le MAF$	$MMF \le 0.4 \text{ x } MAF$	$MMF \le 0.4 \text{ x } MAF$
month definition					
EFR of low	090	0.2 x MAF	090	MMF	0.6 x MMF
flow month	~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~		~		
High flow	MMF > MAF	MMF > MAF	MMF > MAF	MMF > 0.4 x MAF	MMF > 0.8 x MAF
month				and	
definition				$0.4 \times MMF > 0.4 \times MAF$	

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precindustrial monthly discharge – simulated using modelled climate from all four GCMs – can thereby be considered as a remarkable signal of increased flows, although the underlying drivers vary. Though the mechanism of ecosystem degradation
 caused by increased flows is known to exist, no hydro-ecologically grounded quantitative methods have been introduced. Therefore, we used the 95th percentile as the first step and inspiration towards future methodological advances.

For illustration, a conceptual definition of the EFE is presented in Fig. A1, a comparison between monthly pre<u>r</u>industrial discharge and the EFE lower bound is presented in Fig. A2, and a comparison between EFEs and recent <u>historical discharge</u> in sub_z basins in variable flow regimes across the world is presented in Fig. A3.

2.3 Evaluating EFE violations

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Finally, we compared the recent <u>historical (1976–2005)</u> discharge to the <u>EFE in each subpasin</u>. The end date of the recent <u>historical period was</u> limited to 2005 by the ISIMIP 2b simulation protocol owing to a lack of reliable estimates regarding, for example, irrigation extent for the years thereafter (Siebert et al., 2015). For each GHM, we calculated a monthly violation ratio
between the median discharge over four GCMs and the GHM-specific EFE (Table 2). The violation ratio yields a value between 0 and 100 if discharge is within the EFE a negative value if discharge is help with the EFE a negative value if discharge is help with the EFE and and a value.

between 0 and 100 if discharge is within the EFE, a negative value if discharge is below the EFE lower bound, and a value over 100 if discharge is above the EFE upper bound. In the few cases where the EFE was unavailable due to no flow during the precindustrial period, we considered the violation ratio to be zero__i.e. no violation.

495 Table 2. Computing the EFE violation ratio. Q stands for monthly discharge between 1976 and 2005; EFE_{lower} for the <u>EFE</u> lower <u>bound</u>, and EFE_{upper} for the <u>EFE</u> upper <u>bound</u>.

Condition	Equation for violation ratio		Violation ratio	
Q < EFE _{lower}	$\frac{Q - EFE_{lower}}{EFE_{lower}} \ge 100$	(1)	< 0	
$EFE_{lower} \leq Q \leq EFE_{upper}$	$\frac{Q - EFE_{lower}}{EFE_{upper} - EFE_{lower}} \ge 100$	(2)	0 – 100 (no EFE violation)	
$Q > EFE_{upper}$	$\left(rac{Q-EFE_{upper}}{EFE_{upper}}+1 ight)x$ 100	(3)	> 100	

Throughout the analysis, we excluded time periods during which the EFE is violated for less than three consecutive months. This emphasises long_tterm flow alterations that are likely to threaten riverine ecosystems beyond individual species (Biggs et al., 2005). Simultaneously, potential one_xmonth outliers in recent<u>historical</u> discharge are eliminated and do not therefore cause bias to violation metrics. On the other hand, flow alteration events lasting less than three months, such as rapid floods and short-term water withdrawals, are inevitably masked from the results. Therefore, the EFE violations shown in this study present persistent and long-term flow alterations. In addition to results presented in the following section with a minimum three_xmonth

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sequence of violations, we repeated the analysis with other minimum lengths of the violation streak. The results of this sensitivity analysis are presented in the supplementary material (Fig. S1–S3). Finally, although we defined the EFE for all subbasins, we excluded sub basins with extremely low flow from further analysis. If at least three out of four GHMs estimated mean annual flow (the mean monthly flow of all months; MAF) to be less than 10 m³ s⁻¹ at the sub-basin outlet, the sub-basin was excluded. During the recent historical period, sub-basins covering 6.5% of the global land area were excluded due to this criterion.

- 540 Using equations in Table 2, we determined the violation ratio in each sub basin for each month in 1976–2005. Considering the four GHMs, this resulted in a total of 1,440 violation ratios for each sub basin (4 GHMs x 30 years x 12 months). We treated the violation ratios from different GHMs as independent observations of violation since the EFE was defined and evaluated strictly GHM_c wise. We then defined two metrics: violation frequency and violation severity. The violation frequency is defined as the fraction of violated months out of all 1,440 months. The violation severity is defined as the unweighted mean of violation
- 545 <u>ratios</u> during the violated months, the count of which may vary. These metrics were computed separately for the lower and upper EFE bounds. A numerical example is provided in Fig. A1. In addition to the results presented in the following section, we conducted the analysis for individual GHMs, the results for which are shown in the supplementary material (Fig. <u>\$4-\$11</u>).

Elaborating the EFE violation patterns further, we analysed the violations with respect to flow seasons. For this, we classified each month into low (Q < 0.4MAF), intermediate (0.4MAF $\leq Q \leq MAF$), and high (Q > MAF) flow classes, in which MAF stands for mean annual flow. This classification was based on the flow season limits in the EF methods selected for this study (Table 1), and it aims to illustrate the dependency between the amount of discharge and EFE violations. For each GHM, we computed the flow season of each month from median discharge across all GCMs. MAF was computed from the respective year of each month, so that individual months could be classified into different seasons during different years, thus 555 accommodating for drier and wetter years,

Further, we conducted a seasonal trend analysis on the recent historical EFE violation frequency and severity. For the trend analysis, we computed the frequency and severity of violations according to the definitions above, but instead of all years (1976–2005), we applied fiver year moving windows starting from the first window 1976–1980 and ending in the last window 2001–2005. Each of the moving windows was computed over four GHMs and consisted of 240 violation ratios (4 GHMs x 5 years x 12 months). Then, for each sub basin and separately for frequency and severity, we computed the Kendall rank correlation coefficient and fitted a linear regression model into the moving window series (n = 26). We eliminated statistically non significant (p > 0.05) trends using the Kendall rank correlation test (Hollander and Wolfe, 1973) and the linear regression slope t–test (Chambers et al., 1990).

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Finally, we performed a fuzzy c-means clustering (Bezdek, 1981) for each flow season separately. The four clustering variables constituted violation frequency, violation severity, and the linear trend slopes associated with each. We chose to create six
 clusters, of which the most likely one was selected for each sub-basin. If no cluster was selected with over 30% likelihood for a sub-basin, that sub-basin was left unassigned. The output of the cluster analysis is a set of sub-basin clusters, within which intra-cluster similarity and inter-cluster dissimilarity are maximised. Therefore, in sub-basins belonging to one cluster, the EFE violation characteristics and trends are similar.

3 Results

610 3.1 Recent historical and pre-industrial EFE violations

Our findings show that (1) EFE violations are widespread around the world, (2) that lower bound violations are more common than upper bound violations, and (3) that the most impacted regions are located mainly in the arid and dry temperate climate zones (Fig. 2a-c). All of the results presented in this section include only violations in a minimum of three-month streaks, which emphasises long-term flow alterations and masks short-term variation (see Sect. 2.3). The EFE is violated in 49.8% of

- 615 the total 3,860 sub_basins during more than 5.0% of the total 1,440 months across all GHMs (Fig. 2a). The EFE lower bound is violated in 43.2% of sub-basins during more than 5.0% of all months (Fig. 2b), whereas the respective figure for the EFE upper bound is 9.6% (Fig. 2c). Regional patterns are more visible in the EFE lower bound violations than in the EFE upper bound violations, as sub-basins showing lower bound violations are more commonly grouped together. Notable EFE violation patterns emerge in areas with high anthropogenic pressure, such as the Middle East, India, Eastern Asia, and Central America.
- 620 As the violation frequency shown in Fig. 2a-c is computed over all 1,440 months between 1976 and 2005 (4 GHMs x 360 months), it corresponds to the unweighted ensemble mean of the four GHMs. The most impacted regions also remain comparable for individual GHMs, of which PCR-GLOBWB shows the least frequent and LPJmL shows the most frequent violations, with H08 and WaterGAP2 falling in between these (Fig. S4).
- 625 For a comparison between the pre-industrial and recent historical periods, we computed the change in violation frequency between them (Fig. 2d-f; Fig. A4g-i). During the pre-industrial period, the EFE is violated in 24.0% of all sub-basins during more than 5.0% of all months (Fig. A4a). The majority of this consists of EFE lower bound violations, as no sub-basins have more than 5.0% of months violated solely due to upper bound violations (Fig. A4c). Since EFE violations prevail in certain regions also during the pre-industrial period, some of the violations can be assumed to be caused by natural variability.
- 630 However, the EFE violation frequency has widely increased since the pre-industrial period (Fig. 2e), which indicates remarkable changes in discharge. These changes are highlighted when counting sub-basins with more than 10.0% of all months violated (32.7% recent historical; 9.6% pre-industrial) or 25.0% of all months violated (9.5% recent historical; 0.08% pre-industrial). While the EFE lower bound violation frequency has been considerable especially in the driest mid-latitudes and Australia during the pre-industrial period (Fig. A4b) many of these regions have also experienced the largest increases in

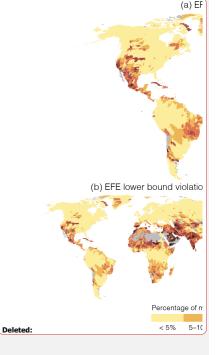
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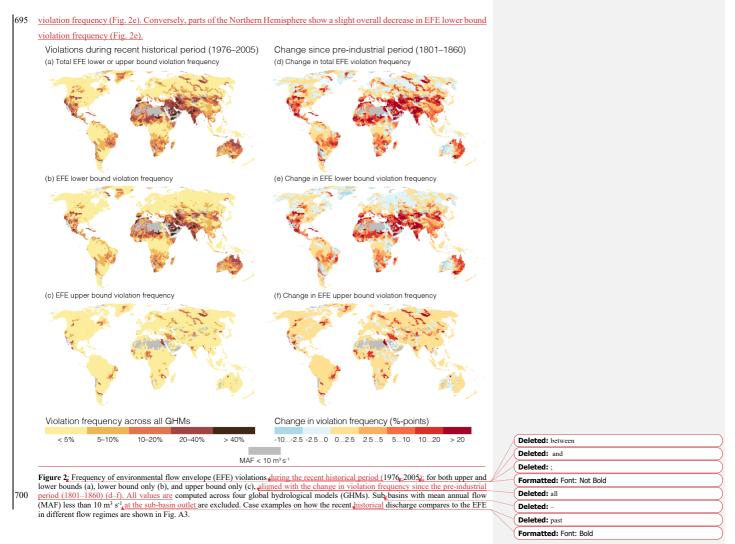
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3.2 Seasonal characterisation of recent historical EFE violations

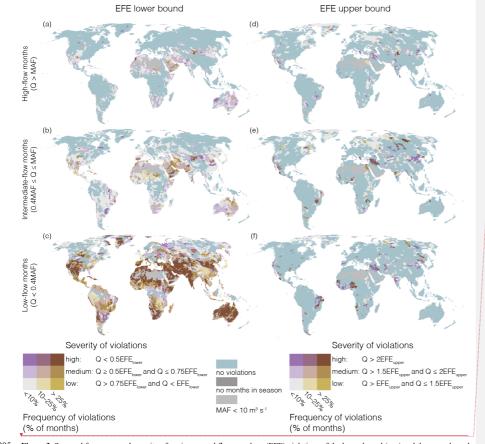
- 710 The low_eflow season is clearly the most impacted in terms of EFE lower bound violations, while the violations decrease gradually from low₋ to intermediate₋ and intermediate₋ to high_eflow seasons (Fig. 3a-c). The distinction between flow seasons is stronger for the frequency than the severity of violations. Between 1976 and 2005, the EFE has been violated in 83.5%, 59.0%, and 28.6% of sub_ebasins during low_e, intermediate_e, and high_eflow seasons for at least one three-month streak (frequency > 0). The medians of EFE lower bound violation severities for low_e, intermediate_e, and high_eflow seasons are -
- 715 37.1%, -19.0%, and_c24.7%, respectively. These figures mean that the typical EFE lower bound violation is caused by discharge falling 19–37% below the EFE lower bound. Although <u>violation</u> severity appears to be less dependent on flow season <u>compared</u> to the dependency of violation frequency on flow season, the low-flow season remains the most impacted overall. This is supported by the spatial coverage of sub_cbasins in the class of the most frequent (> 25%) and the most severe (Q < 0.5EFE_{lower}) violations, which reaches <u>across</u> all continents during low_cflow season (Fig. 3c) and decreases in prevalence during
- 720 intermediate_ and high_flow seasons (Fig. 3a-b). While the spread in EFE lower bound violation frequency and severity is notable between GHMs (Fig. S6), especially the distinction between flow seasons is clearly visible in all single-GHM results (Fig. S5).
- EFE upper bound violations are less dependent than EFE lower bound violations on flow season and exhibit less consistent spatial patterns of frequency and severity (Fig. 3d-f). The shares of sub-basins within which the EFE upper bound is violated for at least one three-month streak between 1976 and 2005 are 15.5%, 24.6%, and 18.9% for low_{ex} intermediate, and high_e flow seasons respectively. The medians of EFE upper bound violation severities during low_{ex} intermediate_{ex} and high_e flow seasons are 153%, 121%, and 123%. Although the summarised statistics would suggest typical EFE upper bound violations to be caused by discharge exceeding the EFE upper bound by 21–53%, many of the sub-basins experiencing EFE upper bound
- violations fall into the high-severity categories, within which discharge exceeds the EFE upper bound at least twofold (Fig. 3d-f). These extremes often cover a small number of sub-basins at a time (Fig. 3e-f; e.g. Tigris-Euphrates, river system, northern China, Niger River), whereas larger scale patterns covering more sub-basins show less frequent and less severe EFE upper bound violations (Fig. 3d-e; e.g. north-eastern Europe, Central Asia). For individual GHMs, the spread in EFE upper bound violations (Fig. 3d-e; e.g. north-eastern Europe, Central Asia). For individual GHMs, the spread in EFE upper bound violations (Fig. S8). Most of the EFE upper bound violations originate from other models except PCR-GLOBWB, but the three other models show fair agreement.

in identifying the sub-basins with major EFE upper bound violations (Fig. S7).

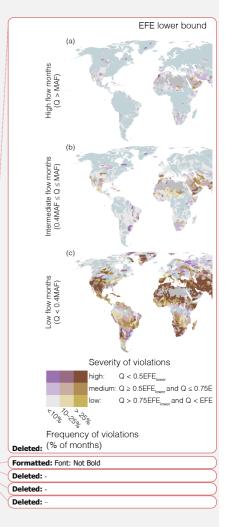
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905 Figure 3: Seasonal frequency and severity of environmental flow envelope (EFE) violations of the lower bound (acc) and the upper bound (dcf). Q stands for monthly discharge and MAF for mean annual flow. For each subbasin in each season, violation frequency corresponds to the fraction of violated months out of all months in that season, and violation frequency to the mean violation ratio during those violated months. If there are no months between 1976 and 2005 during which discharge would fall below the low-flow season limit, the respective sub-basin is classified as "no months in season".



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3.3 Trends in recent historical EFE violations

- 915 Between 1976 and 2005, the frequency and severity of EFE violations have often increased or decreased together. Although we are unable to analytically determine the main drivers of this, we show that more subrabasins have experienced amplifying rather than attenuating EFE violation trends. For the EFE lower bound violations, a statistically significant violation trend is observed for 15.0–51.9% of all subrabasins depending on flow season (Fig. 4a–c). This violation trend consists of a statistically significant trend in violation frequency, severity, or both. Many of the trends (41.0–64.8% of all detected trends depending on
- 920 flow season) consist of a frequency and a severity trend in the same direction, <u>- i.e. both violation frequency and severity are increasing or decreasing. For the EFE upper bound violations</u> 10.3-16.6% of all sub basins show statistically significant violation trends, and most of the trends (68.4-72.1%) consist of changes in the same direction (Fig. 4d-f). Conflicting violation trends are rare; trends consisting of an increase in one variable and a decrease in the other cover 0.5-5.4% of all detected trends across both EFE bounds and all three flow seasons.

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The agreement between the direction of EFE violation frequency and severity trends highlights that the violation trends coe develop rather than conflict. This also holds when computing trends for individual GHMs. Though a statistically significant trend is identified by all individual GHMs in relatively few sub-basins, cases in which some GHMs would detect increasing and some decreasing trends are rare (Fig. <u>\$11</u>). Since increasing violation frequency combined with increasing violation

- 930 severity is the single most common trend for both EFE lower and upper bound violations (28,4% and 53,1% of all detected trends across all flow seasons), the general trend of EFE violations has been towards the intensifying direction during the recent historical period.
- In most of the world, the trends of EFE lower and upper bound violations are independent, but signs of EFE violation trends 935 shifting from the lower bound to the upper bound can be identified __especially in the Northern Hemisphere and the Pan-Arctic areas. Trends in which the EFE lower bound violation frequency and severity are decreasing prevail, for example, in parts of Russia and northern Canada (Fig. 4c), yet the same regions show increasing trends in EFE upper bound violations (Fig. 4e). Therefore, while the EFE lower bound violations can be alleviated by increasing discharge, the EFE upper bound violations _ may be amplified at the same time. The net EFE violations may then balance out, or the EFE upper bound violations may in 940 turn dominate and increase the overall violations. For most of the world, however, this shifting of violations is not visible, and trends – as well as the EFE violations overall – concentrate on one boundary of the envelope only.

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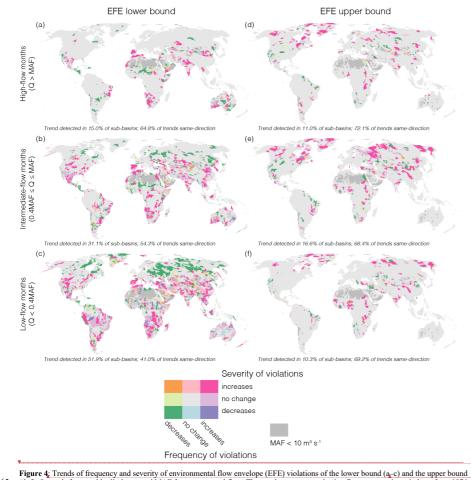
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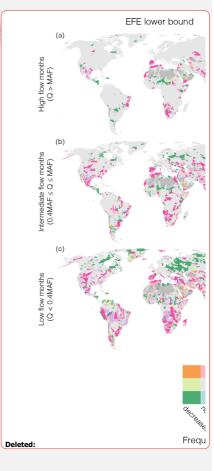


Figure 4: Trends of frequency and severity of environmental flow envelope (EFE) violations of the lower bound (a_x-c) and the upper bound (d_x-f). Q stands for monthly discharge and MAF for mean annual flow. The trends are computed using five-year moving windows from 1976 to 2005; only statistically significant trends are shown. <u>Same-direction trends are defined as having both frequency and severity trends in the same direction, i.e. both violation frequency and severity are increasing or decreasing.</u> The steepness of the trend slope is not considered here but the trends are classified only by increasing or decreasing direction.

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3.4 Categorisation of sub_ebasins by <u>recent historical</u> EFE lower bound violations and trends

The arid mid-latitudes along with parts of tropical South America and subtropical Africa and Asia emerge as the most impacted regions in terms of EFE lower bound violations when the frequency, severity, and trends associated with both are combined in a seasonal cluster analysis. In the relative paucity of sub basins experiencing EFE upper bound violations, we performed 075 the cluster analysis for the EFE lower bound violations only (for further details, see Sect. 2.3). In Fig. 5, the seasonal clusters are grouped together into cluster groups and named according to the characteristics of EFE lower bound violations, within each

group. The blue cluster group A encompasses areas with very few violations, while the first EFE violations appear during lowflow season in the turquoise cluster group B. EFE violation frequency and severity increase in the purple and yellow cluster groups C-D compared to A-B. The orange cluster group E consists of sub-basins with highly variable EFE violation 080 characteristics, and the red cluster group F corresponds to the areas with the highest frequency and severity of EFE lower

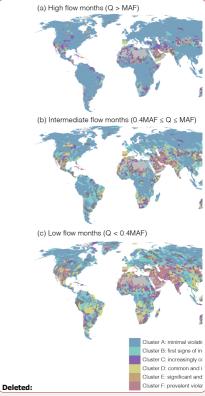
bound violations within each season.

The cluster groups A-C represent sub-basins with minor or stable EFE violations, while EFE violations in sub-basins belonging to cluster groups D-F can be considered the most remarkable throughout all flow seasons. Adding to the nearly non-existent 085 EFE violations in the cluster group A, the cluster group B shows minor violations with decreasing low-flow season trends. Although the cluster group C shows relatively common violations, sub-basins in this group are not experiencing amplifying trends. The regions previously identified as the most impacted (Sect. 3.1-3.2) are mainly covered by cluster groups D-F. Specifically, sub-basins within the vellow cluster group D currently experience moderate violations but show the steepest increasing trends in both violation frequency and severity during low-flow season, and increasing trends during other seasons. 090 If past trends continue, these sub-basins - which include densely populated regions in Asia as well as regions in South America



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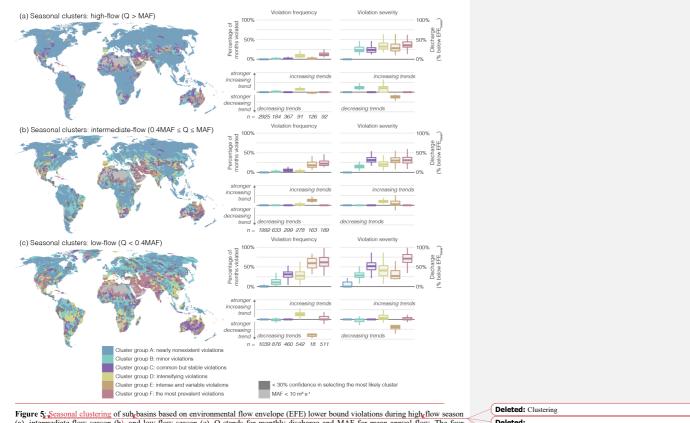


Figure 5; Seasonal clustering of sub basins based on environmental flow envelope (EFE) lower bound violations during high-flow season (a), intermediate-flow season (b), and low-flow season (c). Q stands for monthly discharge and MAF for mean annual flow. The four 230 clustering variables constitute violation frequency, violation severity, and linear trend slopes associated with both variables. In the trend slope box plots, the position of each box is proportional to the linear trend slope; boxes further away from the middle line indicate steeper trend slopes and thereby more rapid change in violation metrics. Sub basins with mean annual flow (MAF) less than 10 m³ s⁻¹ at the subbasin outlet were excluded from the clustering.

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4 Discussion

In this work,— which is the first to quantitatively address frequency, severity, and trends of EFE violations combined – we show that the EFE is widely violated across the globe between 1976 and 2005 (Fig. 2a). Given that the change from the pre-245 industrial period is substantial (Fig. 2d) and all considered violations last three or more months (Sect. 2.3), the EFE violations represent both long-term and recent flow alterations. The EFE lower bound violation patterns are strongly seasonal, with the low flow season being the most affected in terms of both frequency and severity of violations (Fig. 3a–c). The EFE upper bound violation patterns are more dispersed and harder to characterise (Fig. 3d–f). Some sub-basins have experienced drastic flow alteration, with discharge either falling to half of the EFE lower bound or increasing to more than double the EFE upper

250 <u>bound (Fig. 3).</u> Further, recent historical trends in EFE violations have been amplifying rather than attenuating showing increases in both violation frequency and severity (Fig. 4). <u>Combined</u>, our results show that many subvasins in the most <u>densely populated</u> and ecologically diverse areas which as East Asia, South Asia, and parts of South America, are already experiencing considerable EFE lower bound violations, and these can be expected to intensify based on past trends (Fig. 5).

4.1 Comparison to existing studies

- 255 EFE violations, in the arid mid_clatitudes, India, Eastern Asia, and the west coast of North America, compare well to the EF violations estimated by Jägermeyr et al. (2017). These are also in line with the areas requiring the largest reductions in water withdrawals to meet EFRs (Droppers et al., 2020). Our EFE violations are, however, more widespread in large parts of Australia, South America, and Southern Africa (Fig. 2–3) than those reported by Jägermeyr et al. (2017). Central Europe and parts of North America show minor EFE violations (Fig. 2–3; Jägermeyr et al., 2017), although rivers in these regions are
- 260 highly fragmented, regulated, and threatened or even degraded in terms of biodiversity (Grill et al., 2015, 2019; Grizzetti et al., 2017; Vörösmarty et al., 2010). This may indicate limitations in the ecological representativeness of global studies that apply hydrological EF methods that are based on discharge only. Regarding EF violation magnitude, Jägermeyr et al. (2017) report discharge deficits mainly under 10%, whereas our results show substantially higher violation severities (Fig. 3a-c). However, Jägermeyr et al. (2017) determine EFRs based on pristine discharge simulation between 1980 and 2009 (i.e. not
- 265 <u>taking potential climate change impacts into account as we do</u> and report annual averages, which is different from our seasonal analysis based on the pregindustrial period.

The key benefit of our ensemble approach (see Sect. 2.2) is that the ensemble metrics can be considered globally feasible (Arsenault et al., 2015; Huang et al., 2017). Since using the ensemble counters the hydrological uncertainty always embedded
 in GHMs (Telteu et al., 2021), our results could be assumed to be more robust compared to studies using single GHMs or EF methods (e.g., Gerten et al., 2020; Hockstra and Mekonnen, 2011; Pastor et al., 2019; Steffen et al., 2015). Although the ISIMIP 2b data are representative of historical anthropogenic drivers including dams and reservoirs (Frieler et al., 2017), the inclusion and parameterisation of human impacts invokes substantial uncertainty in GHM outputs, particularly in terms of flooding and

Deleted:- which is the first to quantitatively address frequency, severity, and trends of EFE violations combined – we show that recent past discharge in nearly half of the sub-basins of ...he world violates ...FE is widely violated across the EFE – a safe envelope of discharge variability – for extensive and recurrent periods (...[14])

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525 dam operation (Masaki et al., 2017; Veldkamp et al., 2018). Regarding the Pan-Arctic areas in particular, GHMs have recently been shown to perform relatively poorly (Gädeke et al., 2020), which calls for cautious interpretation of our results. As the individual GHMs show notable differences in EFE violations (Fig. S4–S8) and the spread in ensemble EFRs is often substantial (Fig. A2; Hogeboom et al., 2020; Jägermeyr et al., 2017; Liu et al., 2021; Pastor et al., 2014), selecting an array of GHMs and EF methods would be highly desirable in global EF assessments to understand and address the related uncertainties.

530 4.2 Key drivers of <u>recent historical</u> EFE violations

Three key drivers of EFE violations can be identified from existing global research: increasing water use (Graham et al., 2020; Müller Schmied et al., 2016), flow regulation, especially by dam operation (Döll et al., 2009; Schneider et al., 2017), and the indirect impact of climate change on streamflow (Arnell and Gosling, 2013; Asadieh and Krakauer, 2017; Gudmundsson et al., 2021; Moragoda and Cohen, 2020; Thompson et al., 2021; Wanders et al., 2015). Major EFE violations prevail in the

- 535 densely populated mid-latitudes, in which anthropogenic impacts often dominate long term streamflow alterations (Müller Schmied et al., 2016). The anthropogenic impacts on flow alteration are also reflected in the projected increase of water stress (use to availability ratio) that is driven primarily by increasing water use (Graham et al., 2020). Factors beyond the sub-basin, scale, such as water use in upstream sub-basins (Munia et al., 2020) or land use induced changes in atmospheric moisture recycling (Wang-Erlandsson et al., 2018), can further affect the net anthropogenic flow alteration within a sub-basin. However, some EFE lower bound violations are caused by natural streamflow variability rather than anthropogenic flow alteration, since
- the EFE lower bound is violated in some areas also during the pre-industrial period (Fig. A4b).

In the subtropical Southern Hemisphere, increasing EFE lower bound violation trends can be expected to co-occur with the projected trends of increasing droughts (Asadieh and Krakauer, 2017; Wanders et al., 2015), as both indicate abnormally low amounts of water in a system. On the other hand, the decreasing EFE lower bound violation trends and the increasing, EFE upper bound violation trends in high-latitude Europe and Siberia (Fig. 4b-e) may link to climate change-induced changes in discharge (Arnell and Gosling, 2013; Asadieh and Krakauer, 2017). Though our 30-year recent historical period is relatively short for identifying climate trends, Gudmundsson et al. (2021) report decadal trends already within this period. However, climatic changes are not the sole cause of EFE upper bound violation trends; for example, dam operation, can increase discharge

550 especially <u>during low-flow</u> season and therefore result in EFE upper bound violations (Döll et al., 2009; Poff et al., <u>2017</u>; see also Fig. A3b). This can be seen in EFE upper bound violations located above Boguchany and Krasnovarsk mega-dams in Siberia (Fig. 2c; Lehner et al., 2011).

4.3 Relationship between EFEs and riverine ecosystem integrity

Our method – and EFs in general – assumes that violating or respecting the EFE is associated with degrading or preserving /
 riverine ecosystems. However, this might not hold for simplified hydrological EF methods (Poff and Zimmerman, 2010; /
 Richter, 2010; Richter et al., 2012; Mohan et al., 2021), and practical discharge allocation based on insufficient EF methods /

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 ecosystem well-being
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1775 has even been argued to potentially cause further degradation of riverine ecosystems (Arthington et al., 2006; Shenton et al., 2012). This is because the ecosystem response to altered flow regimes varies across spatial and temporal scales, as well as between different species (Biggs et al., 2005; Poff et al., 1997; Rolls et al., 2018). Moreover, quantitative water flows have been shown to be less important than water quality and invasive species for assessing rivers' ecological status, determining fish biodiversity, and driving fish habitat loss (Barbarossa et al., 2021; Grizzetti et al., 2017; Su et al., 2021). Though fish make

780 up only a part of a riverine ecosystem, these studies support incorporating water quality-related factors in a comprehensive EF definition. Holistic EF methods that include these factors correlate much better with ecosystem states, but require in situ data, ancillary variables, and local expert knowledge (Poff et al., 2017; Tharme, 2003) that are not available at the glob Therefore, the impacts of anthropogenic flow alteration at the global scale shown here need to be complemented by local studies using holistic EF methods for practical implications, and analyses of sub-basin scale riverine ecosystem integrity

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Our selection of the 95th percentile of pre-industrial discharge as the EFE upper bound is only a first step towards a more informed choice (Sect 2.2). The link between EFE upper bound violations and cosystems exists, since, for example flo ecosystems in monsoon flood pulse systems require distinct dry and wet periods (Hayes et al., 2018; Junk et al., 1989; Schneider et al., 2017). However, case studies that would quantify this link are scarce. Here, we intentionally set the EFE upper provide the second seco

from this, the EFE upper bound is exceeded very rarely during the pre-industrial period (Fig. A4c). Hence, EFE upper bound violations are strong signals of increased upper extreme flows, although it cannot be inferred from this study whether these are detrimental to riverine ecosystems outside monsoon regions. We stress that our proposal is not meant to be established as a global presumptive standard, but rather as a first trial to inspire methodological advances grounded in hydro-ecology.

795 4.4 Limitations and way forward

The EFEs have their specific limitations related to temporal and spatial scales. Temporally, we aggregated daily discharge to monthly discharge, incurring a loss of temporal detail - especially regarding extreme high and low flows. However, we consider this necessary to assess long-term EFE violations, which is also the rationale behind enforcing the three-month consecutive violation streak rule (Sect. 2.3). Spatially, we consider the sub-basin outlet location as representative for the whole

- 800 upstream area, which simplifies the sub-basin into one hydrological unit (Sect 2.1). On one hand, this masks the highest upstream variability, but on the other hand, it also masks local EFE violations that may vary within the sub-basin itself. This is a notable limitation particularly in the case of temporary rivers, which have recently been shown to be globally important (Messager et al., 2021). In addition, HydroSHEDS level 5 lumps small coastal sub-basins with many estuaries together, which in our analysis results in many small catchments to be represented by the outlet location of the largest lumped catchment 805 Therefore, our results are the most robust in relatively large sub-basins with a single outlet channel. Applications at the scale
- of smaller catchments (< 10,000 km², corresponding to approximately four 0.5-degree grid cells) should rather resort to highdetail observed data instead of global data simulated in a coarse grid,

altered now regimes including variation...aries across spatial and temporal scales and distinct parts of the riverine ecosystem, ... as well as well as the adaptation of ... etween different species to flow regime changes over long timespans ...Biggs et al., 2005; Poff et al., 1997; Rolls et al., 2018). Moreover, a recent study on global fish biodiversity has shown that several other...018). Moreover, quantitative water flows have been shown to be less important than Though fish make up only a part of a riverine ecosystem, these Hough is in lance up only a part of a fiveline ecosystem, uses studies support incorporating water quality-related factors, such as water quality and the presence of invasive species, may be more important in maintaining riverine ecosystems than quantitative flow (Su et al., 2021). Despite their flaws, hydrological EFR methods have remained as the primary option for global scale studies since direct recommends of givening. assessments of riverine ... in a comprehensive EF definition. Holistic EF methods that include these factors correlate much better with system well-being or more advanced EFR methods (.... [22]) Formatted: Font: Italic

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Moved up [7]: 2015; Huang et al.,

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Deletee: The underlying hydrological data partially restricts the conclusions that can be made based on our results. First (.... [26]) Deleted: incurs

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operation (Masaki et al., 2017; Veldkamp et al., Deleted: 2018). The between-GHM uncertainty is illustrated in our

nsitivity ...ydroSHEDS level 5 lumps small coastal sub-ba [28] Moved up [6]: S11).



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In the future, the EFEs should be developed by complementing our global analysis with more advanced EF methods and more detailed hydrological data that better correlate with riverine ecosystem status. Furthermore, case studies quantifying riverine ecosystem responses to prolonged and increased upper extreme flows would benefit the development of the EFE upper bound. as the mechanism is known (Hayes et al., 2018; Junk et al., 1989; Schneider et al., 2017) but quantitative methods are lacking. 185 While the main drivers of EFE violations are recognised here, a systematic analysis on the couplings between them and EFE violations would provide more insights into our results. In addition, separating natural and anthropogenic flow alterations

could prove useful in estimating how major violations - and in which regions - should be deemed as the most serious. Our EFE methodology is lightly parameterised and applicable with open global data sets,- the availability and quality of which is constantly increasing. While fine tuning is required for local contexts, the EFEs provide a quick and globally robust way of 190 assessing the extent and degree of flow alteration.

5 Conclusion

Direct and indirect anthropogenic flow alterations have drastically changed flow regimes across the world and are likely to threaten riverine ecosystem integrity. In this study, we have conducted a global, in-depth analysis of the flow alterations using environmental flow envelopes (EFEs). The widespread and long-standing flow alterations found can be expected to be

195 amplified in response to projected future increases in human water use, the building of new dams, and climate change. On one hand, operationalising our results at the basin scale requires more detailed data, assimilation of cross-scale information, and interdisciplinary knowledge to more fully portray the ecological and hydrological conditions of each unique river. On the other hand, our results highlight the need to consider environmental flows in both global research and policies on water resources management as major anthropogenic flow alterations prevail across wide areas.

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In the future, developing and applying the EFE methodology presented in this study should concentrate on validating the correspondence between the estimated EFEs and riverine ecosystem responses. Although derived from a robust ensemble, the EFE is still based on rule-of-thumb style EFR methods, which must be augmented with local knowledge for practical applications Deleted: quantification of the

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Deleted: novel methodology of Environmental Flow Envelopes CeFEs) to quantify both the frequency and severity of these threats. Comparing recent past discharge with the EFEs based on pre-industrial conditions shows that a significant part of

Deleted: sub-basins is experiencing long-standing flow alteration. These EFE violations most commonly manifest themselves insuffi

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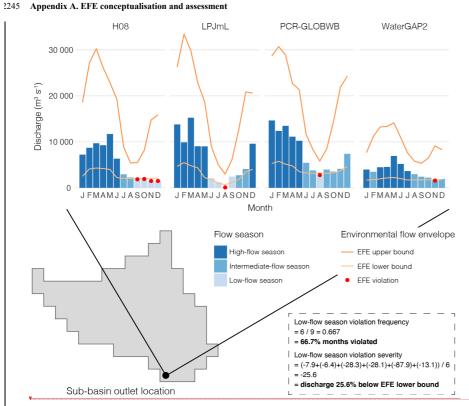
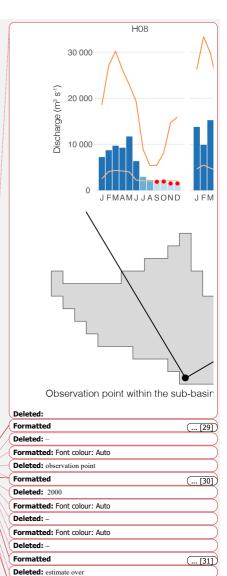


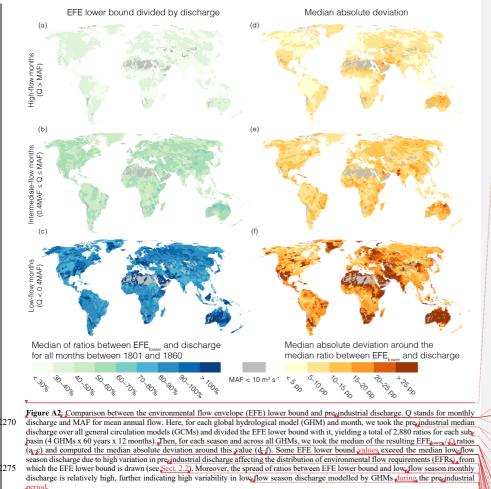
Figure A1: Case example on the conceptual definition of the environmental flow envelope (EFE) and the assessment of EFE violations. The example subpasin is a part of the Rio Paraguay basin: the outlet is located a little upstream from Asunción, Paraguay. For simplicity, we show discharge and assess EFE violations only for the <u>EFE</u> lower bound and <u>one year</u>. In addition, we do not enforce the 3-month violation streak rule (see Sect. 2.3) in this example but count all individual violated months. If the 3-month rule was enforced, violations from the H08 model only would be counted. For each global hydrological model (GHM; H08, LPJmL, PCR-GLOBWB, and WaterGAP2), the discharge is the median of four general circulation models (GCMs). The EFE violation frequency and severity are computed according to definitions in Sect. 2.3



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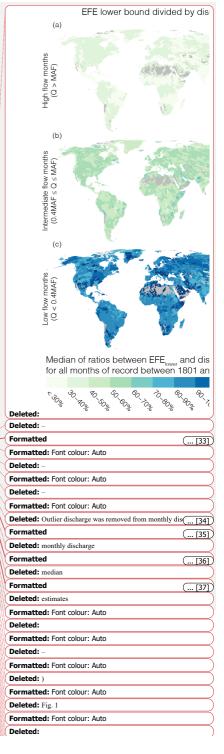
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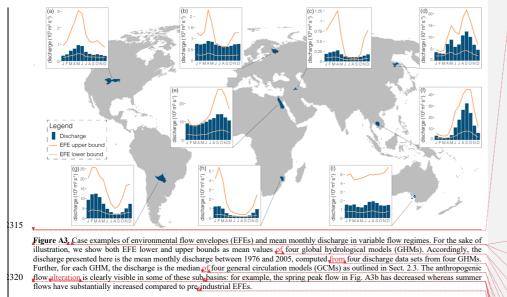


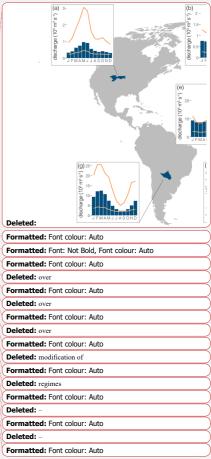
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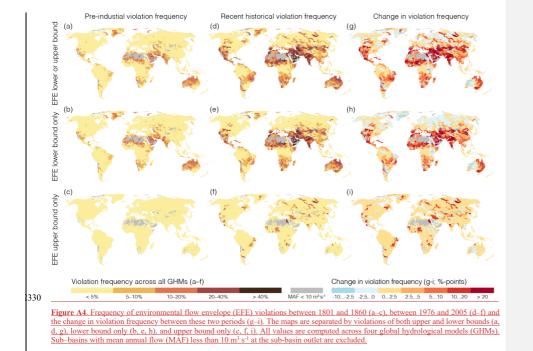
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2335 Code and data availability

The code and data used in producing the results shown in this research article will be released in an open repository upon publication.

Author contribution

MK, EA, and VV conceptualised the study with input from MP, LA, TG, CM, LWE, and DG. DG, MF, NH, HMS, and NW 2340 performed the ISIMIP simulations, which were coordinated by SNG and HMS. EA processed the raw data, wrote the implementation of the EFE methodology, and conceived the initial analysis with help from VV, MP, LA, and MK. VV revised and performed the final analysis and produced the results and visualisation shown in the study, discussing together with MK, MP, LA, TG, CM, and LWE. VV wrote the manuscript based on EA's work with contributions from all authors.

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2345 Competing interests

The authors declare that they have no conflict of interest.

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research and innovation programme (grant agreement No. <u>\$19202), and by the Erling-Persson Family Foundation. LWE</u> acknowledges financial support from the European Research Council through the "Earth Resilience in the Anthropocene" project (no. ERC-2016-ADG 743080).

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