



1 Anthropogenic activities significantly increase annual

greenhouse gas (GHG) fluxes from temperate headwater

3 streams in Germany

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Abstract

Anthropogenic activities increase the contributions of inland waters to global greenhouse gas (GHG;

20 CO₂, CH₄, and N₂O) budgets, yet the mechanisms driving these increases are still not well constrained. In this

21 study, we quantified year-long GHG concentrations and fluxes, as well as water physico-chemical variables from

22 23 streams, 3 ditches, and 2 wastewater inflow sites across five headwater catchments in Germany contrasted by

23 land use. Using mixed-effects models, we determined the overall impact of land use and seasonality on the intra-

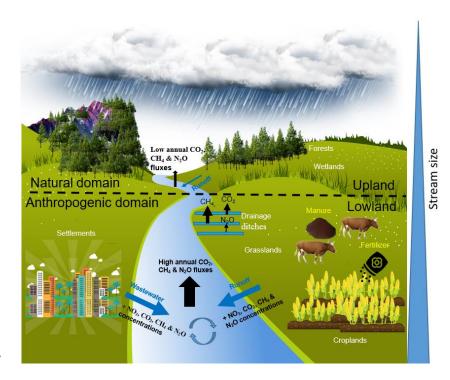
- annual variabilities of these parameters. We found that land use was more significant than seasonality in
- 25 controlling the intra-annual variability of GHG concentrations and fluxes. Agricultural land use and wastewater
- 26 inflows in settlement areas resulted in up to 10 times higher daily riverine CO₂, CH₄, and N₂O emissions than
- 27 forested areas, as substrate inputs by these sources appeared to favor in situ GHG production processes.
- 28 Dissolved GHG inputs directly from agricultural runoff and waste-water inputs also contributed substantially to
- 29 the annual emissions from these sites. Drainage ditches were hotspots for CO2 and CH4 fluxes due to high
- 30 dissolved organic matter concentrations, which appeared to favor in situ production via respiration and
- 31 methanogensis. Overall, the annual emission from anthropogenic-influenced streams in CO₂-equivalents was up
- 32 to 20 times higher (~71 kg CO₂ m⁻² yr⁻¹) than from natural streams (~3 kg CO₂ m⁻² yr⁻¹). Future studies aiming to
- 33 estimate the contribution of lotic ecosystems to GHG emissions should therefore focus on anthropogenically
- 34 perturbed streams, as their GHG emission are much more variable in space and time.





35 Graphical abstract

36







38 1 Introduction

Streams and rivers cover only a small fraction of the earth's land surface (Allen et al., 2018), yet they are significant contributors to global greenhouse gases (CO_2 , CH_4 , and N_2O), emitting approximately 7.6 (6.1–9.1) Pg- CO_2 equivalent into the atmosphere per year. (Li et al., 2021). Several biogeochemical process are responsible for GHG production and consumption within fluvial ecosystems. CO_2 production is attributed to respiration of organic matter (Battin et al., 2008). Production of CH_4 occurs through methanogenesis, with carbon dioxide and acetic acid as substrates under anaerobic conditions (Stanley et al., 2016). Consumption of methane is also possible through methanotrophy in oxygen rich stream waters, producing CO_2 in the process (Shelley et al., 2014). N_2O is mainly a byproduct in nitrification (under aerobic conditions) or an intermediate product in denitrification (under anaerobic conditions), but it can also be reduced to N_2 in organic-rich and nitrate-poor ecosystems (Quick et al., 2019).

Anthropogenic practices such as fertilizer application and construction of drainage ditches to allow agricultural use of former wetlands alter the rates of these processes, thereby influencing in-stream GHG dynamics (Peacock et al., 2021; Wallin et al., 2020; Mwanake et al., 2019). Elevated inorganic nitrogen in streams within fertilized croplands has been shown to favor in situ N₂O (e.g., Beaulieu et al., 2009), CO₂ production (e.g., Bodmer et al., 2016; Borges et al., 2018), and CH₄ production (e.g., Mwanake et al., 2022). While such trends in agricultural streams show similarities across different catchment locations, GHG emissions from streams in predominantly forested catchments with minor influences from croplands and wetlands show more diverse patterns. Some studies indicated that forest streams are hotspots for GHG fluxes (e.g., Wallin et al., 2018; Audet et al., 2019; Herreid et al., 2021), while others found the opposite with much lower fluxes in forests as compare to other land uses (e.g., Bodmer et al., 2016; Mwanake et al., 2022). Drainage ditches, which are characterized by short water residence times, high organic loads, and highly variable O₂ levels, can simultaneously support both aerobic and anaerobic organic carbon mineralization, driving vigorous CH₄ and CO₂ production and subsequent fluxes. In a recent meta-analysis, ditches and canals accounted for up to 3% of the global anthropogenic CH₄ emissions (Peacock et al., 2021). Yet, studies on them are scarce, and thus the main factors making them hotspots of carbon fluxes are still not well-constrained.

In fluvial ecosystems located in settlement areas, inflows of wastewater effluents also act as important drivers of GHG fluxes, by indirectly influencing insitu substrate availability for GHG production and through direct inflows of dissolved GHGs (e.g., Marescaux et al., 2018; Zhang et al., 2021; Wang et al., 2022). For example, in a study of urban-impacted rivers in the Seine basin in France, Marescaux et al. (2018) found elevated CO₂, CH₄, and N₂O concentrations and fluxes downstream of wastewater inflows, which dispropotenately contributed higher basin-wide annual GHG fluxes. Similar findings were also found in urban-impacted rivers in China, were GHG emissions were up to 14 times higher than from other land uses (Zhang et al., 2021). Yet, studies on GHG emissions from urban-impacted fluvial ecosystems are still scarce, and therefore their contributions to riverine annual GHG budgets are not well contrained. Moreover, little is known about the interactive effects of land use and wastewater effluent inflows on riverine GHG fluxes, and whether land use is the overarching controlling factor.





Under temperate climatic conditions, pronounced seasonality regulates the availability of nutrients and to some extent the O₂ in lotic ecosystems, which are both key factors driving *instream* GHG production and gas exchange rates (Borges et al., 2018; Rocher-Ros et al., 2019; Herreid et al., 2021; Aho et al., 2022). Cold winter periods are generally characterized by low *instream* carbon and nitrogen processing, which results in nutrient accumulation (e.g., Herreid et al., 2021), while high *instream* C and N processing are characteristic of warm summer periods (e.g., Borges et al., 2018; Aho et al., 2021, 2022). Seasonality in precipitation regulates discharge, whereby heavy precipitation events or snowmelt during spring result in high discharge events. At the same time, dry summers and winter periods are often characterized by lower discharge (e.g., Aho et al., 2022). Discharge in turn determines the water residence times in streams, thereby influencing rates of carbon and nitrogen processing (e.g., Borges et al., 2018; Mwanake et al., 2022). High discharge events may also increase dissolved GHG supply from upstream terrestrial sources and *instream* GHG production depending on the surrounding land use. For example, studies have found that during high discharge periods, streams draining wetlands show peak CO₂ and CH₄ concentrations (e.g., Aho et al., 2019; Borges et al., 2019) and pronounced N₂O concentrations are found in streams of cropland dominated catchments (e.g., Mwanake et al., 2022).

The aforementioned interactions between seasonality and land use indicate that temporally sporadic measurements of GHG concentrations and fluxes are limited in revealing intra-annual variations, which are necessary for better estimating annual emissions. Yet, only a handful of studies in temperate streams have assessed the seasonal dynamics of GHG fluxes at sampling points with contrasting land uses (e.g., Marescaux et al., 2018; Borges et al., 2018; Herreid et al., 2021; Galantini et al., 2021). As climate change drives more extreme discharge conditions, and as agricultural intensification and settlement areas continue to increase (Winkler et al., 2021), studies that cover a wide array of land uses, discharge, and temperature conditions are needed to constrain better the effects of land use in controlling intra-annual GHG flux variabilities and to unravel synergistic or antagonistic relationships amongst them.

The main objective of this study was to assess the seasonality-land use relationships of water physicochemical variables and GHG concentration and fluxes by comparing temperate lotic ecosystems of forests and wetlands with those from more human-influenced agricultural and settlement catchments. To do so, we conducted at least tri-weekly measurements covering a full year of observations and mainly focused on headwater streams (stream orders 1–6), which are known hotspots of fluvial emissions, but remain currently underrepresented in global GHG datasets (Drake et al., 2018; Li et al., 2021). We hypothesize that catchment land use is the most important control for stream GHG concentration and fluxes, with higher seasonal variability in human-influenced ecosystems than in natural ones. Moreover, we hypothesized that drainage ditches and headwater streams with wastewater inflow within agricultural and settlement areas are hotspots for GHG emissions, driven by direct dissolved GHG inputs or substrate inputs that favor *in situ* GHG production.

2 Materials and methods

2.1 Study areas and sampling design

Five headwater catchments in central (Schwingbach), southeast (Loisach), and southwest (Ammer, Goldersbach, and Steinlach) Germany were investigated in this study. The catchments covered a wide range of





fluvial ecosystems with different stream orders and land use characteristics (Table 1; Fig. 1). The catchment boundaries for each site were determined based on the most downstream sampling location within each catchment (Fig. 1). Elevation of the Schwingbach catchment (54 km²), located in the central-German state of Hessen, ranges from 176–480 m above sea level (a.s.l). The catchment has a mixed land use of ~41 % mixed forests, 46% croplands, 8 % settlement areas, and 5 % pasturelands (Wangari et al., 2022) (Fig. 1A). The climate is warm and temperate (Cfb, Köppen climate classification), with an annual rainfall of 742 mm (monthly mean min: 51 mm, monthly mean max: 72 mm) (1999–2019) and a mean annual temperature of 9.8 °C (monthly mean min: 1.3 °C, monthly mean max: 18.8 °C) (1991–2021) (Climate-data.org, https://en.climate-data.org/europe/germany/hesse/giessen-151/).

The Upper Loisach catchment (467 km², outlet Eschenlohe town) is located in the mountainous region of the Bavarian Alps, Germany. The catchment is characterized by a pronounced relief and steep slopes, with elevations ranging from 616–2,963 m a.s.l. Land use in the catchment comprises coniferous and deciduous forests interspersed with natural grasslands and rocky surfaces on the mountain slopes (78%). At the valley bottom, the land use is mainly settlement areas (9%), managed grasslands (8%), and wetlands (5%) (Fig. 1B). The climate is cold and temperate (Dfb, Köppen climate classification), with annual precipitation of 1,693 mm (monthly mean min: 87 mm, monthly mean max: 207 mm) (1999–2019) and mean annual temperature of 3.8 °C (monthly mean min: -6.6 °C, monthly mean max: 13.1 °C) (1991–2021) (Climate-data.org, https://en.climate-data.org/europe/germany/free-state-of-bavaria/garmisch-partenkirchen-8762/).

The other three catchments are sub-catchments of the Neckar river (Fig. 1C). The Goldersbach (116 km²), a tributary of the main Ammer stream, is a forested catchment (95%), with elevations ranging from 366–583 m a.s.l. The Steinlach catchment (513 km²) is also dominated by forests (74%), with agricultural lands (croplands and grasslands) and settlement areas occupying 21% and 5% of the landscape, respectively. The elevation range of the hilly area is 321–878 m a.s.l (Fig. 1C). The Ammer catchment (304 km², outlet Pfäffingen) is dominated by agricultural lands (80%), with 11% forests and 9% settlement areas (Fig. 1C). It has moderate slopes with an elevation ranging from 319–610 m a.s.l. The Ammer stream is a gaining stream fed by an extensive groundwater karst system and has significant discharge levels even during the driest periods of the year (Glaser et al., 2020). The climate is warm and temperate (Cfb, Köppen climate classification), with a mean annual rainfall of 923 mm (monthly mean min: 63 mm, monthly mean max: 98 mm) (1999–2019) and a mean annual temperature of 9.3 °C (monthly mean min: 0.2 °C, monthly mean max: 18.6 °C) (1991–2021) (Climate-data.org, https://en.climate-data.org/europe/germany/baden-wuerttemberg/tuebingen-22712/).

Across the five catchments, a total of 28 sites at headwater streams (N=23, orders 1–6, defined after Strahler, 1952), drainage ditches (N=3) and waste water outflows (N=2, Text A1) were sampled every 2–3 weeks for an entire year (Table 1, Fig. 1). The Schwingbach and Loisach catchments were sampled from June 2020 to June 2021 while the Goldersbach, Ammer, and Steinlach catchments, were sampled from April 2021 to April 2022.





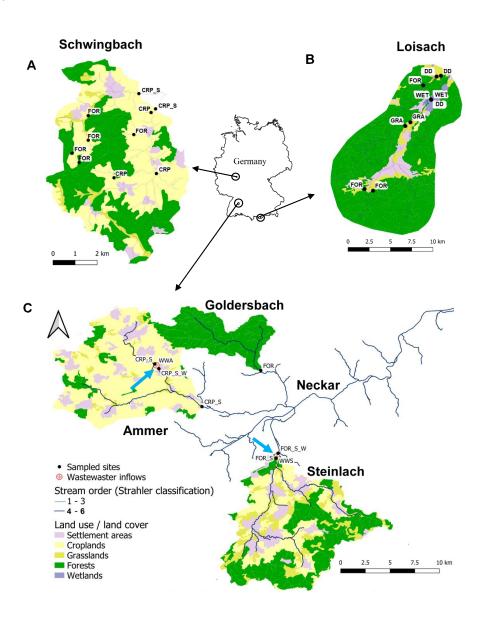


Fig. 1: Land cover maps of the (A) Schwingbach, (B) Loisach, and (C) Neckar sub-catchments (Goldersbach, Ammer, and Steinlach) derived from the Corine Land Cover 2018 inventory with a 25 ha spatial resolution (https://land.copernicus.eu/pan-european/corine-land-cover/clc2018?tab=mapview). Black dots with labels (abbreviations explained in Table 1) represent sampled headwater streams and drainage ditch sampling points. Wastewater inflows sampled are indicated by blue arrows on the maps. Drainage ditches in Loisach catchment were dug in the 1930s to 1960s to lower water levels to improve grassland productivity in areas formerly occupied by wetlands.





2.2 Sub-catchment delineation and land use classification

Sub-catchments for each sampling point in the Loisach, Goldersbach, Steinlach, Ammer and Schwingbach catchments were delineated in QGIS from a Digital Elevation Model (DEM) (EU-DEM v1.1) with a 25 m resolution (European Copernicus mission, retrieved August 1, 2021, https://land.copernicus.eu/imagery-in-situ/eu-dem/eu-dem-v1.1). Land use/ land cover percentages of all the delineated sub-catchments were calculated from Corine Land Cover 2018 survey with a 25 ha spatial resolution (retrieved August 1, 2021, https://land.copernicus.eu/pan-european/corine-land-cover/clc2018?tab=mapview). For the purposes of data analysis, we classified sub-catchments according to their dominant land cover (>50% of the total area) into forest (FOR), cropland (CRP), grassland (GRA), and wetland (WET), and further differentiated sub-catchments with influence of settlement areas (S) and wastewater inflows (W). (Table 1). As drainage ditches (DD) in the Loisach catchment were also added as extra land use category, this classification resulted in a total of 9 land use classes (for details see Table 1).

2.3 Hydrological and water physico-chemical characteristics

In the Loisach and Schwingbach catchments, discharge was calculated (Gore, 2007) from stream depth and velocity measurements using an electromagnetic sensor (OTT-MF-Pro, Hydromet, Germany). For streams in the Neckar sub-catchments, velocity was measured using the electromagnetic sensor (OTT-MF-Pro, Hydromet, Germany), and depth and discharge were obtained directly from gauging stations maintained by the water authority of the state of Baden-Württemberg (https://udo.lubw.baden-wuerttemberg.de/public/index.xhtml). The slope of a ~5 m reach at each sampling point was measured using a laser rangefinder with a slope function (Nikon Model: 8381, Japan). The slopes and velocities were used to model the site-specific gas transfer velocities (k in m d⁻¹) for the quantification of daily GHG fluxes per unit stream surface area (mass m⁻² d⁻¹) (see details in the flux calculation section).

Discharge measurements at each sampling location and at every sampling event were complemented by *in situ* measurements of water temperature (°C), electrical conductivity (µS cm⁻¹), dissolved oxygen (DO) (mg L⁻¹), and pH using the Pro DSS multiprobe (YSI Inc., USA). Water samples for nutrient and organic carbon analyses were also collected and filtered on-site through polyethersulfone (PES) filters (0.45 µm pore size, preleached with 60 ml of miliq water). The samples were stored in triplicates in 30 ml acid-washed HDPE sample bottles and transported within 24 h to the laboratories at Karlsruhe Institute of Technology, Campus Alpin, Justus Liebig University Giessen, or the University of Tübingen. On arrival, all samples were immediately frozen for later analysis.

After unfreezing the samples overnight in a 4° C refrigerator, the samples were directly analyzed for dissolved organic carbon (DOC), total dissolved nitrogen (TDN), nitrate (NO₃-N), and ammonium (NH₄-N) concentrations. Dissolved organic nitrogen (DON) concentrations were estimated as the difference between the TDN and dissolved inorganic nitrogen DIN (NO₃-N + NH₄-N) concentrations. DIN concentrations were determined using colorimetric methods, and the absorbance of the samples was measured using a microplate spectrophotometer (Model: Epoch, BioTek Inc., USA). NO₃-N concentrations were analyzed based on reactions with the Griess reagent (Patton & Kryskalla, 2011), and NH₄-N concentrations were analyzed using the indophenol method (Bolleter et al., 1961). The DOC concentrations were measured as non-purgeable organic





carbon (NPOC) using a TOC/TN analyzer (Analytica-Jena; multi N/C 3100, Germany) after pre-treating the sample with 25% HCl acid to remove the dissolved inorganic carbon (DIC). The TDN concentrations were analyzed simultaneously with the same instrument (Analytica-Jena; multi N/C 3100, Germany).

2.4 Gas sampling, analysis, and calculations of annual areal fluxes

GHG samples of stream, ditch and waste water were collected in triplicates simultaneously with the water physico-chemical samples using the headspace equilibration technique (Raymond et al., 1997). In brief, 80 ml of background water was equilibrated with 20ml of atmospheric air in a syringe at *in situ* water temperatures, and the headspace gas sample transferred into 10ml glass vials for GHG concentration analysis in the laboratory of the Karlsruhe Institute of Technology, Campus Alpin (see full sampling details in Mwanake et al., 2022). Atmospheric air samples were taken twice (morning and afternoon) on each sampling day to correct for background atmospheric GHG concentrations. GHG concentrations from the headspace were analyzed using an SRI gas chromatograph (8610C, Germany) with an electron capture detector (ECD) for N₂O and a flame ionization detector (FID) with an upstream methanizer for simultaneous measurements of CH₄ and CO₂ concentrations. Dissolved GHG concentrations in the stream water were calculated from post-equilibration gas concentrations in the headspace after correcting for atmospheric (ambient) GHG concentrations (e.g., Aho et al., 2019; Mwanake et al., 2022).

Daily diffusive fluxes (F) (moles m⁻² d⁻¹) of the GHGs were estimated using Fick's Law of gas diffusion, where the F is the product of the gas exchange velocity (k) (m d⁻¹) and the difference between the stream water (C_{aq}) (moles m⁻³) and the ambient atmospheric gas concentration in water assuming equilibrium with the atmosphere (C_{sat}) (moles m⁻³) (Equation 1). GHG concentrations and fluxes were expressed in mass units by multiplying by the respective molar masses.

$$F = k \left(C_{aa} - C_{sat} \right) \tag{1}$$

The temperature-specific gas transfer velocities (k) for each of the gases were calculated from normalized gas transfer velocities (k_{600}) (m d⁻¹) (corresponding to the k of CO₂ at 20° C with a Schmidt number of 600) and temperature-dependent Schmidt numbers (Sc) (unit-less) of the respective gases (Equation 2).

$$220 k = k_{600} \times (600/s_c)^{0.5} (2)$$

The k_{600} was modeled using Equation 3 (drawn from equation 4 in Table 2 of Raymond et al. (2012)), which was calibrated from headwater streams of similar characteristic of our study sites, where V is stream velocity (m s⁻¹) and S is the slope (m m⁻¹).

$$k_{600} = VS^{0.76} \times 951.5 \tag{3}$$

Before chosing the aforementioned equation for modeling the k_{600} values, we compared the k_{600} values calculated from all seven empirical models from Raymond et al 2012. The predicted k_{600} values from models 3,4,5 and 6 in Table 2 of Raymond et al. (2012), which all use velocity and slope as input parameters, were mostly similar for the three discharge periods and across all stream orders 1–6 (ANOVA; p>0.05). In contrast, the calculated k_{600} values from equation 1, 2 and 7, which use a stream depth parameter, were higher (ANOVA; p<0.05), particularly from the higher stream orders (5–6). This finding is inconsistent with the energy dissipation

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model of turbulent streams where k_{600} is predicted to decrease with stream order. We therefore interpreted this to indicate a breakdown of these models for higher stream orders. This is also in agreement with recommendations from Raymond et al. 2012, and we therefore choose not to use models 1, 2 and 7 for this study. Out of the remaining equations 3, 4, 5 and 6, we used equation 4, which calculates k_{600} based on the slope and velocity parameters, and was also in line with several previous studies spanning a wide range of stream orders similar to our study. (See, Aho et al., 2019; Borges et al., 2019; Mwanake et al., 2019; Hall & Ulseth, 2020; Aho et al., 2021; Mwanake et al., 2022). The uncertainties in the modelled gas transfer velocities were reduced in this study by parametrization of the velocities and slopes based on actual field measurements of both variables. Equation 3 was also used to estimate the gas transfer velocities in the drainage ditches that also had a measurable flow velocity and slope.

Water-to-atmosphere fluxes for all three GHGs across all land use classes in each subcatchment were calculated from the mean daily CO₂, CH₄ and N₂O fluxes during different discharge conditions. Total GHG fluxes were expressed as CO₂ equivalents emissions (mg CO₂-eq m⁻² d⁻¹) computed from global warming potentials (GWP₁₀₀) using 28 for CH₄ and 298 for N₂O (IPCC, 2014). In order to scale tri-weekly measurements to annual flux estimates, we followed the procedure developed in Mwanake et al. (2022). Briefly, we classified each sampling date of every location into low, medium, or high discharge conditions according to whether normalized discharge fell in the 0–33% percentile (low), 34–66% (medium), or 67-100% (high) days.

Normalized discharge for each site was determined by dividing each absolute discharge measurement for every site visit during the year with the maximum measured discharge. The number of days in each discharge period were estimated as the ratio of the number of observations in each discharge period to the total number of flux observations in individual land use classes in each catchment. CO₂ equivalents fluxes were then calculated for the three different discharge periods in each land use class by multiplying the daily mean CO₂ equivalents flux measured during each period and the number of days within each period. Annual fluxes were finally estimated by summing up the emissions of the low, medium, and high discharge periods for the individual land use classes in each catchment.





2.5 Statistical analysis

Linear mixed-effects models were used to investigate the effect of seasonality and land use on water physico-chemical variables, GHG concentrations, and fluxes ("Ime4" package in R version 4.1.1). Fixed effects in the models consisted of land use classes in each catchment (Table 1) and seasons: summer June 1–August 31, autumn September 1–November 30, winter December 1–February 28, and spring March 1–31th May. Random effects accounting for repeated measures were also included in the models. Model performance was assessed based on the distribution of residuals (i.e residuals should be normally distributed with a mean close to zero) and conditional r^2 values calculated from significant models (p-value <0.05) ("MuMln" package in R). A Tukey post-hoc test (p-value <0.05) of least-square means was used on the mixed models to identify individual differences within each categorical fixed effect. GHG concentration and flux data and other water physico-chemical variables were transformed using the natural logarithm to meet the assumption of normality. Because we quantified occasional negative fluxes in some of our sites, constant flux values of 50 mg m⁻² d⁻¹ for CO₂-C, 0.5 mg m⁻² d⁻¹ for CH₄-C, and 10 μ g m⁻² d⁻¹ for N₂O-N were added to the fluxes to enable the natural logarithm transformations.

Path analysis from structural equation models (SEMs, "lavaan" package in R version 4.1.1) were used to determine how environmental factors linked to seasonality and landuse, directly or indirectly influenced *in stream* GHG production and consumption processes as well as external GHG sources, i.e., dissolved GHGs inputs to the streams originating from either wastewater inflows or terrestrial landscapes which were not produced *in situ*. In brief, these SEMs were constructed on the basis of causal relationships between exogenous variables (interpreted as ultimate drivers of GHG concentrations) and endogenous variables, which are affected by the exogenous variables and also act as immediate drivers that affect GHG concentrations. Endogenous variables in the models, which are known to influence *in situ* biogeochemical GHG production and consumption processes directly, included dissolved oxygen DO (% saturation), DOC (mg L⁻¹), NH₄-N (mg L⁻¹), and NO₃-N (mg L⁻¹) concentrations (Battin et al., 2008; Stanley et al., 2016; Quick et al., 2019). The exogenous variables in the models, which influence *in situ* GHG concentrations either directly by facilitating dissolved GHG inputs or indirectly by controlling the endogenous variables, were water temperature (°C) (a proxy for different seasons), stream velocity V (m s⁻¹), % upstream agricultural area for each sampling point (AGR: grassland + cropland area) and wastewater inflows (WW: Boolean numbers, i.e., 1 for the presence of wastewater inflow and 0 for absence).

The hypothesized relationships between the endogenous and exogenous drivers of instream GHG concentrations were assessed in the overall theoretical SEM, which is made up of several multivariate regression equations shown in Equations 4-8. To get the best-fit SEM, removal of parts of the theoretical SEM was done manually until the model with the highest parsimony fit index (PNFI), and a root mean squared error of approximation (RMSEA) of <=0.05 was found (Schumacker and Lomax, 2016). Graphical representations of the significant relationship pathways from the best-fit model, including standardized slope parameter estimates, were done using the "semPlot" package in R software.

 Log_e GHG concentration = DO + DOC + stream velocity + water temperature + Log_eNO_3 + 293 Log_e NH_4 + wastewater inflow + agricultural area

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295 296	$\label{eq:logenomega} Log_eNO_3 = DO + Log_eNH_4 + DOC + was tewaterinflow + agriculturalarea$ stream velocity (5)	+
297 298	$Log_e NH_4 = DO + DOC + was tewater in flow + a gricultural area + stream velocity $ (6)	
299	DOC = wastewater inflow + agricultural area + stream velocity	(7)
300	DO = DOC + wastewater inflow + agricultural area + stream velocity	(8)





Steinlach/Neckar Wastewater effluent

(Fig. 1). The land use (%) was calculated for the site-specific upstream sub-catchments based on the Corine Land Cover 2018 survey of Europe (See main Table 1: Summary descriptions of sampling sites located in the Schwingbach, Loisach, and Neckar sub-catchments (Goldersbach, Ammer and Steinlach) text for details).

		Stream	Coordinates (d	Coordinates (decimal degrees) Subcatchment	Subcatchment	Elevation at	Sub-c	atchment	Sub-catchment Landuse / landcover (%) Was tewater	andcover	A (%)	/as tewater		Main land use
Main Catchment	Site		Latitude	Longitude	area (Ha)	sampling point	Forest	Wetland (Forest Wetland Grassland Cropland Urban	Cropland	Urban	inflow	Main sub-catchment landuse class	Abreviations
Loisach	Stream	-	47.5694	11.1554	4	651	40	09	0	0	0	No	Wetland	WET
Loisach	Stream	2	47.5689	11.1556	10	645	22	78	0	0	0	No	Wetland	WET
Loisach	Stream	_	47.5440	11.1193	11	099	0	0	100	0	0	No	Grassland	GRA
Loisach	Stream	-	47.5399	11.1105	13	663	19	0	81	0	0	So	Grassland	GRA
Loisach	Stream	-	47.4670	11.0537	40	750	98	0	14	0	0	No	Forest	FOR
Loisach	Stream	2	47.4691	11.0394	75	756	66	0	0	0	_	No	Forest	FOR
Loisach	Stream	2	47.5858	11.1429	102	719	100	0	0	0	0	No	Forest	FOR
Loisach	Drainage ditch		47.5963	11.1730	11	630	27	0	73	0	0	No	Drainage ditch	DD
Loisach	Drainage ditch		47.5953	11.1657	11	645	43	57	0	0	0	So	Drainage ditch	DD
Loisach	Drainage ditch		47.5696	11.1550	17	630	47	0	53	0	0	No	Drainage ditch	DD
Schwingbach	Stream	-	50.5051	8.6127	41	297	%	0	0	4	0	No	Forest	FOR
Schwingbach	Stream	-	50.4695	8.6179	09	187	0	0	0	100	0	No	Cropland	CRP
Schwingbach	Stream	2	50.4811	8.5407	62	241	86	0	2	0	0	No	Forest	FOR
Schwingbach	Stream	-	50.4756	8.5472	29	334	98	0	0	14	0	So	Forest	FOR
Schwingbach	Stream	2	50.4922	8.5971	220	260	47	0	0	53	0	No	Cropland	CRP
Schwingbach	Stream	2	50.5032	8.5553	220	272	65	0	0	35	0	No	Forest	FOR
Schwingbach	Stream	2	50.4887	8.5555	268	204	83	0	0	17	0	No	Forest	FOR
Schwingbach	Stream	-	50.4669	8.5792	355	207	14	0	0	8	2	No	Cropland	CRP
Schwingbach	Stream	33	50.5050	8.6148	2337	183	37	0	9	84	6	No	Cropland+settlement	CRP_S
Schwingbach	Stream	33	50.5166	8.5992	5345	189	4	0	4	45	7	No	Cropland+settlement	CRP_S
Goldersbach (Neckar)	Stream	5	48.5588	9.0591	11623	367	76	0	0	3	0	No	Forest	FOR
Ammer (Neckar)	Stream	5	48.5649	8.8986	26157	379	Π	0	-	8	4	No	Cropland+settlement	CRP_S
Ammer (Neckar)	Stream	9	48.5640	8.8997	26361	377	Π	0	-	83	5	Yes	Cropland+settlement+was tewater	CRP_S_W
Ammer (Neckar)	Stream	9	48.5271	8.9615	30441	348	4	0	2	11	∞	No	Cropland+settlement	CRP_S
Steinlach/Neckar	Stream	9	48.4796	9.0634	51332	348	74	0	10	=	4	No	Fores t+s ett lement	FOR_S
Steinlach/Neckar	Stream	9	48.4812	9.0639	51332	345	75	0	10	Ξ	4	Yes	Forest+settlement+wastewater	FOR_S_W
Ammer/Neckar	Wastewater effluen	ıt.	48.5644	8.8993									Wastewater	WWA

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302 3 Results

3.1 Hydrological variables

Across all sampling points and seasons, tri-weekly sampled stream velocity measurements (annual mean \pm SE) were two-folds higher for streams (0.19 \pm 0.009 m s⁻¹,range: 0.01- 1.17) than ditches (0.05 \pm 0.06 m s⁻¹, range: 0.01–0.23) (Fig A1). Seasonality had an overall significant effect on stream velocities across all sampling points, with higher stream velocities observed in spring (0.24 \pm 0.02 m s⁻¹) than in autumn (0.12 \pm 0.01 m s⁻¹) (Table 2; Table B2). Discharge in streams (3.9–18,500 L s⁻¹) and in ditches (0.1–37 L s⁻¹) was highly variable, reflecting differing stream sizes and seasonal variability (Fig. A1). The Neckar sub-catchments, dominated by streams (orders 5 - 6), had an order of magnitude higher mean annual discharge (874.7 \pm 178 L s⁻¹) than the streams in the other catchments (Loisach: 50.5 \pm 6 L s⁻¹ and Schwingbach: 26.7 \pm 4 L s⁻¹). The average discharge at the stream and ditch sampling points in all our study catchments were 3 to 5-fold higher in spring and summer (384.1 \pm 96 and 526.4 \pm 171 L s⁻¹, respectively) than in autumn and winter (86.25 \pm 13.07 and 157.3 \pm 31.58, respectively; Table 2; Table B2).





Table 2: Results of multiple linear mixed-effects models predicting the effect of seasonality (summer, autumn, winter, and spring) and sub-catchment land use (Table 1) on stream velocity, discharge, water physico-chemical variables, GHG concentration, gas-transfer velocity, and GHG flux. The model performance was assessed based on conditional r² and the distribution of residuals, including the variances explained by fixed effects and repeated mea sures' random effects.

		Type 2 ANOVA table			
		Season (df=3)	Land use (df=11)		
Dependent variables	Conditional r ²	F-statistic/significance	F-statistic/significance		
Water physico-chemical and hydrological variable	es				
Temperature (° C)	0.87	66.3***	9.1***		
pH	0.80	3.1*	97.8***		
DO (mg L ⁻¹)	0.83	20.1***	143.7***		
Electrical Conductivity (µs cm ⁻¹)	0.83	4.9**	86.1***		
NO_3 -N (mg L^{-1}) ^a	0.80	4.9**	141***		
NH_4 - $N (mg L^{-1})^a$	0.60	ns	32.3***		
TDN (mg L ⁻¹) ^a	0.79	5.6**	93.8***		
DON (mg L ⁻¹) ^a	0.55	ns	13.9***		
$DOC (mg L^{-1})^a$	0.59	ns	47.3***		
DOC:DIN	0.84	3.2*	133.2***		
DOC:DON	0.63	ns	15.1***		
Velocity ^a	0.59	3.7*	34.5***		
Discharge ^a	0.86	4.6**	96.9***		
$k_{600,}$ Gas concentration and flux					
CO ₂ -C concentration (µg L ⁻¹) ^a	0.86	25.6***	219.3***		
CH ₄ -C concentration (µg L ⁻¹) ^a	0.89	ns	273.1***		
N ₂ O-N concentration (ng L ⁻¹) ^a	0.75	3.3*	69***		
k_{600} (m d^{-1}) ^a	0.57	ns	31.2 ***		
CO ₂ -C flux (mg m ⁻² d ⁻¹) a	0.57	ns	50.2***		
CH_4 - C flux (mg m ⁻² d ⁻¹) a	0.79	ns	113***		
N_2 O-N flux ($\mu g \ m^{-2} \ d^{-1}$) ^a	0.70	3.9*	75.6***		
Total fluxes CO ₂ -eq (g m ⁻² d ⁻¹) ^a	0.67	ns	67***		
Level of significance (p-value) * <0.05	^a Natural loga	rithim transformation			
** <0.01					
*** <0.001	Conditional r ²	= Variance explained by fixed a	nd random effects of sampling da		
ns >0.05	df= degrees of	f freedom			

3.2 Water physico-chemical variables

3.2.1 Seasonal variation

Water temperature, DO, and pH ranged from 0.9–24° C, 1.1–15.7 mg O_2 L^{-1,} and 6.7–9.0, respectively. Streams in the mountainous Loisach catchment had a mean annual (\pm SE) water temperature of 9.0 \pm 0.2 °C, which was ~1 °C colder than streams of the Schwingbach catchment (10.0 \pm 0.4 °C) and 3 degrees colder than





streams in the Neckar sub-catchments (12.0 ± 0.3 °C). The annual ranges of NH₄-N, NO₃-N, DON, TDN, and DOC concentrations across all catchments were 0.05–1.0 mg L⁻¹, 0.5–14.8 mg L⁻¹, 0.05–10.9 mg L⁻¹, 0.6–17.0 $mg~L^{-1}$, and $0.9-16.0~mg~C~L^{-1}$, respectively. DO, NO_3 , and TDN concentrations showed significant seasonal variability (Table 2, Table B2). DO was higher in winter and spring than in summer and autumn. NO₃-N and TDN concentrations were highest in winter and lowest in autumn and summer, while NH₄-N, DOC and DON showed no significant seasonal variation (Table 2; Table B2). We additionally calculated DOC:DIN and DOC:DON molar ratios, which had interquartile ranges from 0.9-4.9 and 4.1-29.0, respectively. DOC:DIN ratios showed significant seasonal variability, with higher values in summer and spring than in winter, while no seasonal variability was found for DOC:DON ratios (Table 2: Table B2).

3.2.2 Land use variation

Catchment land use was more significant than seasonality in explaning the variability of most water physico-chemical variables (Table 2). In the Loisach catchment, ditches had up to 2.6 times lower DO and up to 8 times lower NO₃-N concentrations than the streams across all land use types (Fig. 2; Table B3). In contrast, NH₄-N and DOC concentrations, as well as the DOC:DIN ratio were 6-10 times higher in the ditches than in the streams (Fig. 2; Table B3). In the Neckar sub-catchments, forested streams had 1-2 times higher DO and DOC concentrations than cropland, settlement, and wastewater-influenced streams. The opposite was true for NO₃-N and DON concentrations, which were an order of magnitude higher in the cropland, settlement, and wastewater-influenced streams than in the forested streams (Fig. 2; Table B3). As a result, DOC:DIN and DOC:DON ratios in the Neckar sub-catchments were therefore higher in forested streams than in cropland, settlement, and wastewater-influenced streams (Table B3).

In addition, cropland streams directly receiving wastewater inflows also had significantly lower DO and higher DOC than cropland streams without wastewater inflows (Fig. 2; Table B3). While NO₃-N and DON concentrations were not significantly different in cropland streams with or without wastewater inflows, the concentrations of both variables was slightly higher in cropland streams with wastewater inflows (Table B3). In streams of the Schwingbach catchment, surrounding croplands and settlement areas also influenced NO₃-N concentrations, which were up to 3-fold higher than in the forested streams. Across all the three catchments, DO concentrations, DOC:DIN and DOC:DON ratios were higher in the forested streams and decreased in streams of sub-catchments with predominant agricultural land uses or settlement areas, while the opposite was found for NO₃-N and DON concentrations (Table B3). Additionally, forested streams in the Loisach catchment had an order of magnitude higher DOC:DON ratios than forested streams in the Neckar and Schwingbach catchments (Table B3).





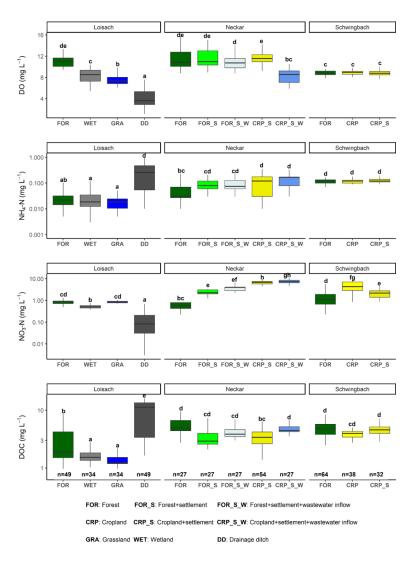


Fig. 2: Boxplots of DO, NH_4 -N, NO_3 -N, and DOC concentrations in stream and ditch waters in the three catchments grouped by dominating land uses (see Table 1 methods). Letters on top of the boxplots represent significant differences (p<0.05) among land use classes across the three catchments based on Tukey post-hoc analyses from the linear mixed-effects model results (Table 2).

3.3 GHG concentrations and fluxes

3.3.1 Seasonal variation

In all headwater streams, CH₄ and N₂O concentrations varied greatly, spanning three orders of magnitude, i.e., from 0.03–58 μ g-C L⁻¹ (pCH₄ 1.3–2,145 μ atm) for CH₄ and from 20–18,717 ng-N L⁻¹ (pN₂O





21–15,813 natm) for N₂O. In contrast, CO₂ concentrations varied less, spanning only one order of magnitude
 from 219–4,868 μg-C L⁻¹ (pCO₂ 369–7,979 μatm). GHG concentrations in ditches also varied widely, with CH₄.
 N₂O and CO₂ concentrations spanning 1-2 orders of magnitude ranging from 27–831 μg-C L⁻¹ (pCH₄ 1,469–34,482 μatm), 56–1,540 ng-N L⁻¹ (pN₂O 35–1,512 natm), and 1,722–9,746 μg-C L⁻¹ (pCO₂ 2,888–13,400 μatm), respectively (Fig. A2–A5).

Streams and drainage ditches across all seasons were predominantly sources of atmospheric CH₄, N₂O, and CO₂, as indicated by concentrations mostly above atmospheric background and the positive flux values displayed in Figure 3. CO₂ fluxes from streams ranged from -0.05–179 g C m⁻² d⁻¹ (mean 19 g C m⁻² d⁻¹), CH₄ fluxes ranged from -0.40–325 mg C m⁻² d⁻¹ (mean 30 mg C m⁻² d⁻¹), and N₂O fluxes ranged from -9.2–199.5 mg N m⁻² d⁻¹ (mean 12 mg N m⁻² d⁻¹). CO₂ and CH₄ fluxes from the ditches varied between 2–63 g C m⁻² d⁻¹ (mean 13.7 g C m⁻² d⁻¹) and from 117–7,933 mg C m⁻² d⁻¹ (mean 1,532 mg C m⁻² d⁻¹), respectively, while N₂O fluxes ranged from -0.8–7.1 mg N m⁻² d⁻¹ (mean 1.2 mg N m⁻² d⁻¹).

Seasonal variation in GHG concentrations and fluxes were GHG dependent and varied across the different land uses within each catchment (Fig. 3; Fig. A2, A3, and A4). In the Loisach catchment, there was a decline in *instream* CO_2 concentrations in the summer followed by a subsequent increase in autumn, particularly at non-forested sampling points (Fig. A2). Similar *instream* CO_2 concentration trends with lower values in the summer season and increasing values in autumn, were also found for non-forested streams of the Neckar subcatchments (Fig. A3). However, non-forested streams of the Schwingbach catchments showed slightly different trends, with a decline of CO_2 concentrations in spring and an increase of CO_2 concentrations in the late summer season. (Fig. A4). Considering all data over all catchments, seasonality had an overall significant effect (p<0.05) with CO_2 concentrations in summer being 1.6 times lower than in autumn, while CO_2 fluxes showed no significant seasonal variability (Table 2; Table B2).

In contrast to CO_2 , N_2O concentrations in the Loisach and Schwingbach catchments decreased from summer to autumn but increased again towards the beginning of the winter season (Fig. A2, A4). In autumn, N_2O concentrations at first and second order forested streams in the Loisach and Schwingbach catchments were often below atmospheric concentrations (Fig. A2, A4), characterizing these sites as N_2O sinks (Fig. 2). A similar autumn decline in N_2O concentrations was not observed in the streams of the Neckar sub-catchments, but rather, N_2O concentrations increased from autumn to winter (Fig. A3). Across all catchments and sampling points, N_2O concentrations were 2.4 times higher in winter than in the other seasons (Table B2). N_2O fluxes were up to 1.6 times higher in summer and winter than in autumn and spring (Fig. 3; Table B2), which represented periods of either high N_2O concentrations and moderate gas transfer velocities (winter) or moderate N_2O concentrations and high gas transfer velocities (summer) (Table B2).

CH₄ concentrations showed a seasonal pattern only in the Schwingbach catchment (Fig. A4), which showed a decline from summer through autumn and winter. This trend was not observed for the other catchments (Fig. A2, A3) and resulted in non-significant seasonal effect on both concentrations and fluxes when all data from all catchments were considered together (Table 2; Table B2). Overall, strong seasonal trends of GHG fluxes throughout the year were mostly found in human-influenced land use classes such as streams and





ditches in grasslands, croplands, and settlement areas, but not at streams whose sub-catchments were dominated by forests or wetlands (Fig. 3).

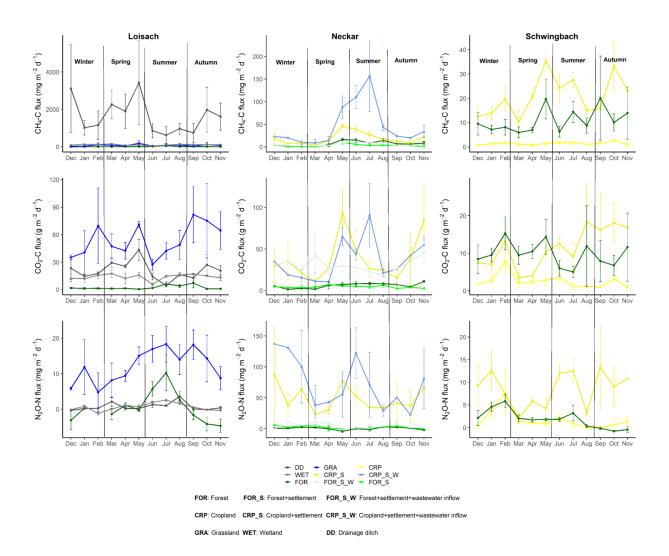


Fig. 3: Monthly mean \pm SE of CO₂, CH₄, and N₂O fluxes across all 26 sampled streams and ditches in the Loisach, Neckar, and Schwingbach catchments (see Table 1 methods). The colors of the lines and labels on the graph indicate the nine dominant land use classes.

408 3.3.2 Land use variation

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Similar to water physico-chemical variables, the variability in GHG concentrations and fluxes was more strongly linked to catchment land use than seasonality (Table 2). In the Loisach catchment, CO₂ concentrations and fluxes were an order of magnitude higher for the ditch and stream sites that were dominated





by grassland land uses than forested-dominated sites (Fig. 3; Fig. 4; Table B3). N₂O concentrations and fluxes in streams were also an order of magnitude higher in the grassland streams compared to the wetland and forested ones, with the latter functioning as occasional sinks for atmospheric N₂O (Fig. 3; Fig. 4; Table B3). Wetland streams had higher CH₄ fluxes than the other stream sites (Fig. 3; Fig. 4; Table B3). Overall, ditches showed up to 14 times higher CO₂ and up to 850 folds higher CH₄ concentrations than the streams of Loisach catchment (Fig. A5; Table B3). In contrast, N₂O concentrations in the ditches were highly variable, with both higher and lower than atmospheric concentrations over the sampling year (Fig. A2). CH₄ fluxes were two orders of magnitude higher in ditches than in streams (Fig. 3; Fig. 4; Table B3). Interestingly, the ditches were even more often N₂O sinks than forests, which resulted in overall lowest N₂O fluxes e.g. 10 times lower than the ones of grassland-dominated streams (Fig. 3; Table B3)

In the Neckar sub-catchments, CO_2 , CH_4 , and N_2O concentrations and fluxes were 1-10 times higher in the streams located in cropland and settlement areas as compared to streams in forested areas (Fig. 3; Fig. 4; Fig. A5; Table B3). Generally, GHG concentrations and fluxes of streams in cropland and settlement areas further increased if sampling points were affected by wastewater inflows (Fig. 3; Fig. 4; Fig. A5; Table B3). For the latter, it is noteworthy that pronounced differences in wastewater characteristics existed in our study, even though the treatment procedures and the number of served households (80000) were comparable for the two wastewater treatment plants. Overall, the wastewater outflow in the Ammer catchment had higher TDN, DOC, CH_4 and N_2O concentrations than the one in the Steinlach catchment (Table B1).

In contrast to the other two catchments, forested streams in the Schwingbach catchment had CO_2 and CH_4 concentrations and fluxes comparable to cropland and settlement-influenced streams within the catchment (Fig. 3; Fig. 4; Fig. A5; Table B3). However, N_2O concentrations and fluxes were higher in streams with cropland and settlement influences than in forested streams (Fig. 3; Fig. 4; Fig. A5; Table B3).





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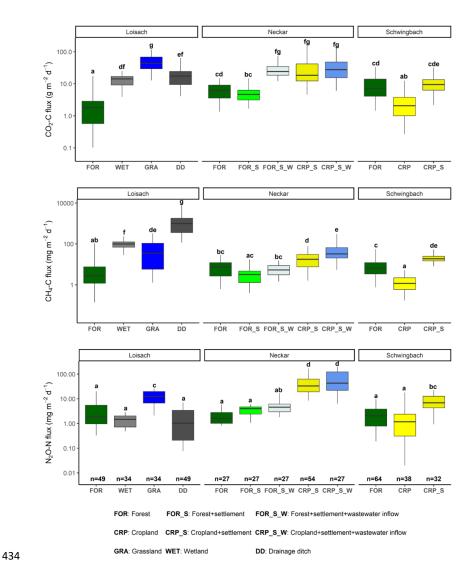


Fig. 4: Boxplots of CO_2 , CH_4 , and N_2O fluxes in stream and ditch waters in the three catchments grouped by land uses (see Table 1 methods). Letters on top of the boxplots represent significant differences (p<0.05) amongst the land use classes across the three catchments based on Tukey post-hoc analyses from the linear mixed-effects models' results (Table 2).





3.4 Direct and indirect drivers of greenhouse gas concentrations

We used path analyses from SEMs based on all our dataset in order to explain how indirect factors such as upstream agricultural area, wastewater inflow and stream velocity controlled the spatial-temporal dynamics of GHG concentrations that drove the fluxes. The slopes parameter estimates from the SEMs revealed significant (p<0.05) interactions between the aforementioned indirect drivers and DO (% saturation), DOC mg L⁻¹ and NO₃-N mg L⁻¹, i.e. drivers that directly control *in situ* GHG concentrations (Fig. 5, Table B4). An increase in upstream agricultural area resulted in a ~45% increase in *in situ* NO₃-N concentrations. Wastewater inputs resulted in a ~22% increase in *in situ* NO₃ concentrations, while DOC concentrations were not significantly affected. DO decreased with increasing DOC concentrations, while NO₃-N concentrations followed an opposite pattern and increased with increasing DO concentrations (Fig. 5).

CO₂ and CH₄ concentrations had a negative relationship with DO (Fig 5A-B), but N₂O concentrations were not significantly related to DO (Fig 5C). Besides DO, CO₂ concentrations decreased by 17% with stream velocity, and increased by 18% with wastewater inflows and by 24% with increasing upstream agricultural area (Fig 5A). CH₄ concentrations also decreased by 16% with increasing stream velocity. However, the effect of wastewater inflows (+5%) or increased share of agricultural areas (+12%) on CH₄ concentrations was lower than for CO₂. Additionally, CH₄ concentrations also decreased by 31% with increasing NO₃-N concentrations (Fig. 5B). In contrast to CO₂ and CH₄, N₂O concentrations increased by 40% with increasing NO₃-N concentrations, while the effect of stream velocity was of minor importance (-8%). Compared to CH₄ and CO₂, N₂O concentrations in stream and river waters showed similar or stronger relationships to wastewater inflows (+18%) and upstream agricultural area (+34%) (Fig 5C). Overall, the best-fit SEMs explained 60, 66, and 46 % of the observed variances in CO₂, CH₄, and N₂O concentrations, respectively (Table B4)





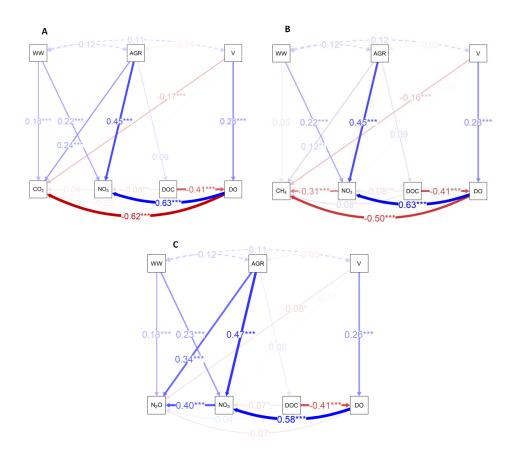


Fig. 5: Regression pathways predicting A) Loge CO_2 concentration μ g-C L⁻¹, B) Loge CH_4 concentration μ g-C L⁻¹ and C) Loge N_2O concentration ng-N L⁻¹ across all sampling points and seasons from best-fit SEMs consisting of endogenous (DO, DOC, and NO_3 -N) and exogenous variables (stream velocity (V), percentage agricultural area (AGR; grassland+cropland areas), and wastewater inflows (WW). The numbers on the lines represent standardized slope parameters, with significant relationships indicated by *. Solid lines represent actual fitted relationships, while dashed lines represent co-variances in the exogenous variables. Blue lines represent positive relationships and red represents negative relationships, with width and color intensity represting the strength of the relationships.

3.5 Annual areal fluxes

Based on global warming potential calculations, CO_2 dominated the annual GHG emissions across all headwater streams, with contributions ranging from 56 %–100%. The non- CO_2 gasses' contributions were much lower and ranged from 0–43% for CH_4 and 0–21% for N_2O (Fig. 6). The highest contribution of CH_4 (43%) was found at ditch sampling points in the Loisach, while the highest N_2O contributions (up to 21%) were observed at the cropland-influenced streams fed by wastewater inflows in the Neckar sub-catchments (Fig. 6). Overall, the

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 annual CO_2 -equivalent emissions from anthropogenic-influenced streams were up to 20 times higher than from natural forested and wetland streams (~71 kg CO_2 m⁻² yr¹ vs. ~3 kg CO_2 m⁻² yr¹ respectively; Fig. 6). Its also noteworthy that the total annual GHG emission from oligotrophic forested streams in the Loisach catchment were significantly lower than other forested catchments in the more human influenced Schwingbach and Neckar sub-catchments (Fig. 6).

Regarding different discharge periods, high and medium discharge periods contributed up to 91 % to total GHG emissions in anthropogenic-influenced streams, but only 4% in forested streams (Fig. 6). Overall, the high and medium discharge periods contributed the most to the annual fluxes quantified in lower-order streams (Strahler 1-2) and ditch sampling points, which were prevalent in the Loisach and Schwingbach sub-catchments (Fig. 6B, C). The opposite was true for larger forested and cropland streams in the Neckar sub-catchment, where higher annual flux contributions occurred primarily in the low discharge period (Fig. 6A). However, this pattern did not hold true for cropland streams with the wastewater inflows in the same catchment, with the sites showing an 82% increase in annual emissions during the high and medium discharge periods (Fig. 6 B, C).





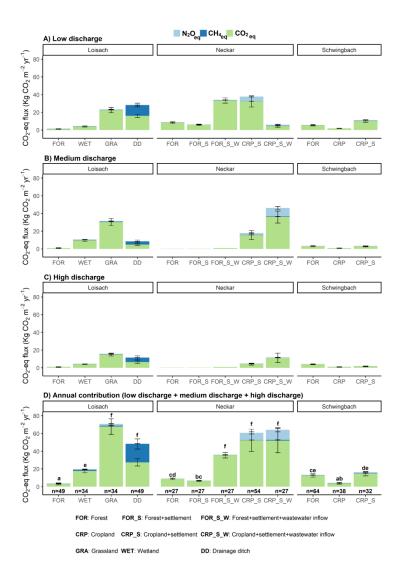


Fig. 6: Areal CO_2 -equivalent fluxes (mean $\pm SE$) grouped by GHG type for each land use class during A) low, B) medium, and C) high discharge periods. D) represents the total annual fluxes by summing up contributions from the three discharge periods. Letters on the bar graphs represent significant differences (p<0.05) in the annual areal emissions amongst the land use classes across the three catchments based on Tukey post-hoc analyses from the linear mixed-effects models' results (Table 2)





496 4 Discussion

4.1 Seasonal variability in GHG concentrations and fluxes

The GHG fluxes quantified from headwater streams and ditches in the three catchments in central and southern Germany add to the growing evidence that both aquatic ecosystems are significant net emitters of GHGs to the atmosphere. Seasonal trends in *in situ* GHG concentrations and fluxes were mainly linked to substrate availability (C and N), discharge and temperature, similar to previous studies on other streams in temperate climates (Dismore et al., 2013; Herreid et al., 2021). However, the observed seasonality of GHG fluxes from streams and ditches in our study was further impacted by land use across the three investigated catchments, with sub-catchments dominated by wetlands or forested land uses exhibiting lower seasonal variabilities than sub-catchments dominated by agricultural land use or affected by wastewater inflow (Fig. 3).

The low *in situ* CO₂ concentrations (< 100% saturation) during summer (Table B2) suggested elevated photosynthetic uptake within the streams and ditches, which is in line with the results of a recent meta-analysis on lotic ecosystems (Gómez-Gener et al., 2021). The decline in CO₂ concentrations in summer was most obvious at the non-forested stream sampling points, with higher canopy cover in the forested areas likely limiting *in situ* stream photosynthesis due to shading effects. We also found that stream ditch waters were oversaturated with CO₂ in autumn and winter. These seasons are characterized on the one hand by low discharge and low stream velocity, conditions which likely reduce degassing rates, and on the other hand by elevated *in situ* C metabolism, as supported by low DO concentration in autumn, which indicates respiratory O₂ consumption (e.g., Borges et al., 2018). We attribute the lack of seasonality in CO₂ fluxes (Table B2) to the compensatory effects of seasonally varying stream velocities and CO₂ source strengths. For example, high CO₂ concentrations and low gas transfer velocities in autumn and vice versa conditions in spring, resulted in comparable CO₂ fluxes amongst the two season (Table B2).

N₂O concentrations also varied significantly across seasons, but the pattern differed from that of CO₂. In autumn, forested lower-order streams in the Loisach and Schwingbach catchments mainly showed N2O concentrations below atmospheric background concentrations and were temporary sinks of N₂O (Fig. 3). This finding could be related to increased inputs of organic matter in these headwater catchments due to leaf fall, providing additional organic carbon for microbial metabolism in this period, which likely increased the demand for terminal electron acceptors such as O2, NO3, as well as N2O. This conclusion is also supported by lowest DO and NO₃-N concentrations during autumn, which could suggest the dominance of complete denitrification in the streams (Quick et al., 2019). With decreasing temperatures towards winter, lower productivity and N demand within the streams resulted in the accumulation of NO₃-N, which seemed to favor internal N₂O production as seen by the positive relationship between the two variables (Fig. 5C). The high sensitivity of the N₂O reductase to low temperatures might have further supported elevated N₂O concentration and fluxes during winter (e.g., Holtan-Hartwig et al., 2002). A similar finding of high winter N₂O concentrations and fluxes was also found in other temperate streams, alluding to similar controls of temperature and nutrient availability (Herreid et al., 2021; Galantini et al., 2021). Thus, based on our results, winter periods can significantly contribute to annual N2O emission budgets. Yet, to the best of our knowledge, temperate studies covering the winter period are still scarce.





In contrast to CO_2 and N_2O , neither CH_4 concentrations nor fluxes showed any seasonal trends. Such a finding is similar to what was found in a global meta-analysis (Stanley et al., 2016), where multiple controls related to substrate availability, geomorphology and hydrology were shown to result in high spatial-temporal variance of CH_4 , thus masking any seasonal emission patterns.

4.2 Effect of human impacts on GHG concentrations and fluxes

Anthropogenic-influenced streams and ditches draining predominantly agricultural and settlement areas showed higher CO₂-equivalent GHG emissions than forested streams (Fig. 6). Such a finding is similar to other studies in the temperate region (e.g., Borges et al., 2018; Galantini et al., 2021). The high GHG emissions of streams and ditches in agricultural and settlement areas is likely due to elevated hydrological inflow (e.g., via groundwater and interflow) of nitogen and labile carbon (Lambert et al., 2017, Mwanake et al., 2019) or terrestrialy originating dissolved GHGs linked to lower vegetation cover compared to forested catchments (e.g., Mwanake et al., 2022). This interpretation could be supported by the significant positive relationships that we found between percentage agriculture and stream CO₂, CH₄ and N₂O, as well as nitrate concentration, and a positive trend for DOC (Figure 5).

Low DOC:DON ratios have been previously linked to more labile and less aromatic forms of dissolved organic matter (DOM) (Sebestyen et al., 2008, O'Donnell et al., 2010). We found significantly lower DOC:DON ratios in streams and ditches in agricultural and settlement areas than in forested streams, suggesting that the more bioavailable DOM in the human-influenced ecosystems, favored elevated GHGs production through heterotrophic processes (e.g., Bodmer et al., 2016). Such differences in DOC:DON ratios were also found amongst forested streams, with a decreasing trend from Loisach, Neckar to Schwingbach catchments, which may also explain the differences in their GHG emissions (Fig. 6). The differences in the DOM bioavailability of forested streams in the three catchments may suggest differences in DOM flowpaths during terrestrial-groundwater-stream interactions. We contend that the moderately sloping streams of the Neckar and Schwingbach catchments, likely had lower DOC:DON ratios due to longer water residence time and higher contributions of groundwater inflow (e.g., Sebestyen et al., 2008) than those in the steeper forested catchments of the Loisach (Table B3). Distinct difference in water stable isotope signatures, i.e. the shift of precipitation vs. stream water seasonality across the three catchments (data not shown), further supported the difference in water residence times and their relationships with stream slope (e.g., Zhou et al., 2021).

In addition to land use influences, wastewater inflows into streams in agricultural and settlement areas further increased GHG concentrations and fluxes. The two sampled wastewater effluents, which drained into the Steinlach and Ammer streams of the Neckar sub-catchments, showed higher GHG concentrations than the stream water upstream of the inflows (Fig. A5, Table B1), which mainly lead to increased GHG concentration and fluxes also downstream of the wastewater inflows. This finding is similar to what was found in other temperate studies comparing stream GHG concentration upstream and downstream of wastewater inflows (e.g., Marescaux et al., 2018; Aho et al., 2022). However, due to higher background GHG fluxes in the cropland than the forested sub-catchments (Fig. 4), differences in the total GHG emissions before and after wastewater inflow were more pronounced in the forested sub-catchments (Fig. 6). In addition to the pronounced differences in the quality of the wastewater effluent (Table B1), this finding also shows the importance of background GHG fluxes as influenced by catchment land use in assessing how wastewater inflows affect riverine GHG emissions.





Apart from land use influences, GHG fluxes from streams have been previously shown to decrease with stream order, as dissolved GHG inputs from groundwater and terrestrial sources also decrease (e.g., Hotchkiss et al., 2015, Turner et al., 2015, Mwanake et al., 2022). While our study design was not meant to explicitly asses stream order influences due to limited replication across a wide range of stream orders, we did find an opposite trend with stream order. For example, higher-order streams (stream orders> 5) in the Neckar sub-catchments dominated by croplands and with wastewater influences had mostly higher GHG fluxes than lower-order streams (stream orders < 3) in the Loisach and Schwingbach catchments. We therefore show a potential breakdown of stream order-GHG relationships in highly human-impacted lotic ecosystems, with disproportionately higher GHG emissions than in more natural ecosystems. We also show that significant nutrient and labile carbon supplies to higher-order streams, which create ideal conditions for GHG production and emission, may outweigh the physical disadvantages (e.g. lower surface area to volume ratio) of higher-order streams relative to lower-order streams.

Drainage ditches, characterized by low flow velocities and high DOC:DIN ratios, functioned as strong sources of CO₂ and CH₄ fluxes compared to streams. We assume that the low DO, high DOC, and low NO₃-N concentrations, along with high water retention times, supported high *in situ* CH₄ production rates in sediments, resulting in the overall highest contribution of CH₄ fluxes to total annual GHG emission budgets than streams(Figure 6). This interpretation is further supported by a significant negative relationship between CH₄ and DO, as well as NO₃-N concentrations, and a positive relationship with DOC concentrations, associations which have also been previously linked to *in situ* methane production in fluvial ecosystems (e.g., Baulch et al., 2011b; Schade et al., 2016). High CH₄ fluxes from drainage ditches were also found in other studies from both forested and wetland areas (e.g., Schrier-Uijl et al., 2011; Peacock et al., 2021b). Contrastingly, ditches were only weak sources or even sinks for atmospheric N₂O. This finding suggests N₂O reduction to N₂ via complete denitrification, an interpretation already made in previous studies on lotic ecosystems (e.g., Baulch et al., 2011; Mwanake et al., 2019).

4.3 Comparison of GHG flux magnitudes with other regional studies

This study reported among the highest fluvial CO₂ emissions compared to other studies, with significant mean fluxes of up to 51 g-C m⁻² d⁻¹ (Table 4). We attribute this finding to moderate-steep slopes such as those quantified in the mountainous streams of the Loisach catchment or diffuse and point terrestrial dissolved GHG inputs from the more human-influenced Schwingbach and Neckar catchments, translating to higher fluvial GHG fluxes (Fig. 6). However, our high CO₂ fluxes are comparable with those quantified from other temperate streams in Canada and Switzerland with similar moderate-steep slopes and considerable dissolved CO₂ inputs from terrestrial landscapes (e.g., Mcdowell & Johnson. 2018; Horgby et al., 2019). The CH₄ fluxes from streams in this study are comparable with those previously found in temperate sub-catchments with similar land uses and altitudes, but are lower than those reported from permafrost streams in China (Zhang et al., 2020). Our N₂O fluxes from cropland, settlement, and wastewater-influenced streams, are higher than those previously reported in a mixed landuse catchment (Schade et al., 2016), but our forest N₂O fluxes are in the same range as those of other temperate forested streams (Aho et al., 2022). That said, these comparisons may be hampered, particularly for fluvial N₂O fluxes, by the limited number of studies currently available (Table 4).

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The average ditch CH4 fluxes in this study are higher than those reported for forest and wetland draining
ditches in boreal and temperate regions (Table 4: Schrier-Uijl et al., 2011, Peacock et al., 2021b) and the global
mean provided by Peacock et al., (2021), which includes estimates from large canals. In contrast, N_2O fluxes
from ditches in this study are lower than those quantified from NO ₃ -N-rich agricultural ditches in tropical and
temperate regions (Table 4: Harrison & Matson, 2002; Reay et al., 2003).





Table 4: Compilation of GHG emissions from temperate streams and ditches with comparable land use, climate, and altitude ranges.

					Number of study Nun	Number of	CO_2 -C flux (g m ² d ¹)	$CH_{\text{d}}\text{-}C\text{flux}\;(\text{mg m}^{\text{-}2}\text{d}^{\text{-}1})$	$N_2O\text{-}N\text{flux}(\text{ng}\text{m}^2\text{d}^1)$	
Land use/ land cover	Climate	Country	Geographical coordinates Altitude (m)		reaches obse	reaches observations Duration of study	Range Mean	Range Mean	Range Mean	Reference
Forest/Loisach streams	Temperate	Germany	Table 1	616 - 2963	3 51	Annual, 2022	-0.05 - 17.4 2.4	-0.4 - 164 10.5	-9.2 - 20.3 1.1	This study
Forest/Schwingbach streams	Temperate	Germany	Table 1	176 - 480	5 27	Annual, 2022	0.08 - 33.4 9.5	-0.02 – 54.6 9.9	-1.6 - 9.6 2.1	This study
Forest/Neckar rivers	Temperate	Germany	Table 1	319 - 610	1 80	Annual, 2022	0.6 - 14.7 6.6	0.6 - 28.9 9.1	-6.9 - 5.9 0.3	This study
Forest+settlement/Neckar rivers	Temperate	Germany	Table 1	319 - 610	1 27	Annual, 2022	0.6 - 14.9 4.9	0.4 - 17.3 3.9	-7.7 - 6.0 2.2	This study
Forest+settlement+wastewater/Neckar rivers	Temperate	Germany	Table 1	319 - 610	1 27	Annual, 2022	12 - 71.7 28.3	1.4 – 15.2 6.5	-2.8 - 17.1 3.9	This study
Wetland/Lois ach streams	Temperate	Germany	Table 1	616 - 2963	2 34	Annual, 2022	2.8 - 25.2 13.3	17.2 - 237.5 101.7	-1.6 - 2.9 0.8	This study
Grassland/Loisach streams	Temperate	Germany	Table 1	616 - 2963	2 34	Annual, 2022	6.1 - 115.9 50.7	1.3 - 324.5 73.2	-0.8 - 25.5 12.4	This study
Cropland/Schwingbach streams	Temperate	Germany	Table 1	176 - 480	3 48	Annual, 2022	0.3 - 9.0 2.1	0.07 - 5.6 0.9	-0.8 - 18 1.9	This study
Cropland+settlement/Schwingbach streams	Temperate	Germany	Table 1	176 - 480	2 32	Annual, 2022	0.6 - 32.0 8.6	0.6 - 52.6 14.9	-0.8 - 22.4 6.5	This study
Cropland+settlement/Neckar rivers	Temperate	Germany	Table 1	319-610	2 54	Annual, 2022	4.5 - 181.3 39.1	1.6 - 77.5 21	8.4 - 165.7 46.9	This study
Cropland+settlement+wastewater/Neckar rivers	Temperate	Germany	Table 1	319 - 610	1 27	Annual, 2022	1.1 - 129.9 38.8	0.8 - 301.9 58.2	6.3 - 198.2 67.6	This study
Forest streams	Temperate	USA	43.0760° N, 107.2903° W 1211 - 3311	V 1211-3311	1 253	June - October, 2014	1.5 - 6.79 1.3	14.4 – 576 28.8		Kuhn et al., 2017
Forest streams	Temperate	USA	40.2140° N, 105.4332° W 2780 - 3505	V 2780-3505	2 111	June - July, 2013	0.2 - 1.6 0.49	0.3 - 7.8 2.1		Crawford et al., 2015
Forest streams	Temperate	USA	41.6032° N, 73.0877° W	270 - 810	7 608	4 years, 2016 – 2019			-0.4 - 29	Aho et al., 2022
Forest streams	Temperate	USA	41.6032° N, 73.0877° W	270 - 810	7 608	4 years, 2016 – 2019	-1.2 - 152 3.4	0.3 - 2870 28.7		Aho et al., 2021
Forest streams	Temperate	Canada	49.270°N, 122.560°W	1200 - 3050	1	Annual, 2017	8.7 - 1980 55.9			Mcdowell and Johnson. 2018
Mixed streams	Temperate	USA	43.123°N, 71.1219°W	165 - 348	3 37	Annual, 2012	0.4 - 1.1	6 - 43.8	-0.6 – 6.0	Schade et al., 2016
Mixed streams	Temperate	Switzerland	46.1512° N, 7.0634°E	1190 - 3051	1 300	Annual, 2016	13.3 - 494.5 31			Horgby et al., 2019
Mixed streams	Temperate	Europe			34 107	Annual, 2017	-0.8 – 5.8			Attermeyer et al, 2021
Grassland drainage ditches	Temperate		Table 1	616 - 2963	3 50	Annual,2022	2-63.3 13.7	116.6 - 7933 1532	-0.8 - 7.1 1.2	This study
Wetland drainage ditches	Temperate	Netherlands	52.2200°N, 4.5300°E	1 – 10	7 14	June - July, 2009	0.8	9.909		Schrier-Uijl et al., 2011
			111 0000 0 11 00000 27	00		1 14 1				1 9000

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617 Conclusions

Compared to forests and wetlands, streams and ditches in agricultural and settlement areas were characterized by significantly higher GHG fluxes with greater intra-annual variabilities. A combination of wastewater inflows and agricultural land use resulted in the highest riverine CO_2 and N_2O fluxes, particularly during high discharge periods with substantial contributions of external dissolved GHGs. In general, anthropogenic activities resulted in a potential breakdown of the expected decrease of the GHG source strengths with increasing stream order, as higher-order streams in the Neckar sub-catchments with cropland and settlement influences had higher concentrations and areal fluxes than small streams in the Loisach and Schwingbach catchments. As most studies use stream order to upscale local and regional riverine fluxes, we show from our results that caution must be taken in applying the methodology, particularly across catchments differing in land use intensity.

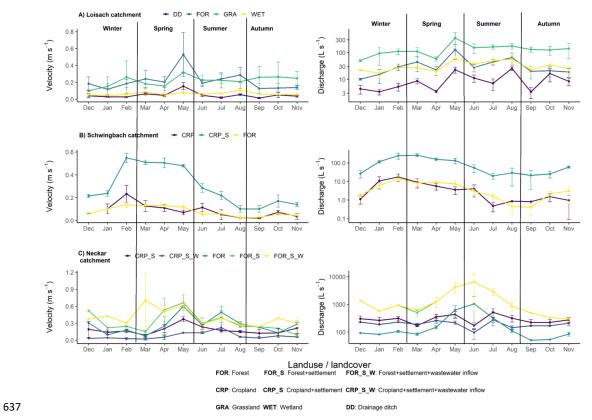
In general, our findings indicate that future work should focus more on human-influenced headwater stream ecosystems, as they contribute disproportionately large annual fluxes and are more temporally variable than natural ones. Our study also found higher winter N_2O fluxes, emphasizing the need for continuous sampling regimes covering full years in order to reduce uncertainty in annual GHG emission estimates. Combining continuous sampling regimes of all three biogenic GHGs (CO_2 , N_2O , and CH4) across catchments with contrasting land uses will further constrict riverine emissions and aid in developing targeted emission reduction mitigation strategies.





635 Appendices

636 Appendix A: Figures



B) Schwingbach and C) Neckar catchments.





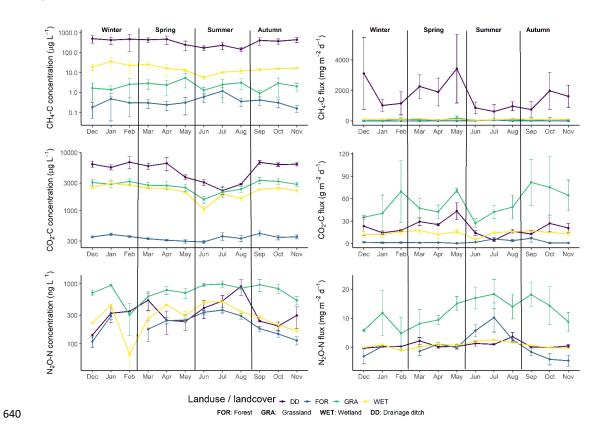
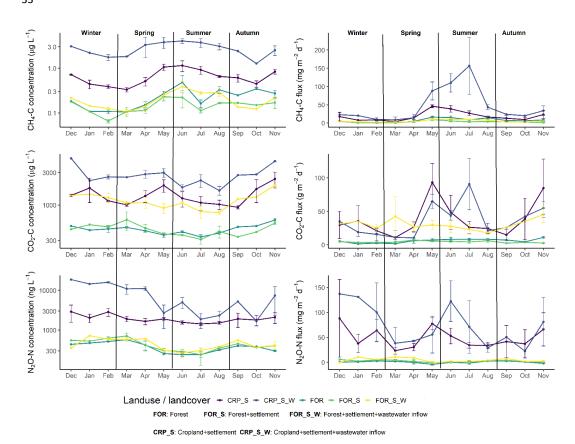


Fig. A2: Monthly mean ± SE CO₂, CH₄ and N₂O concentrations and fluxes at forested (FOR), wetland (WET),
 grassland (GRA) and ditch (DD) sites in the **Loisach** catchment (see Table 1 methods).







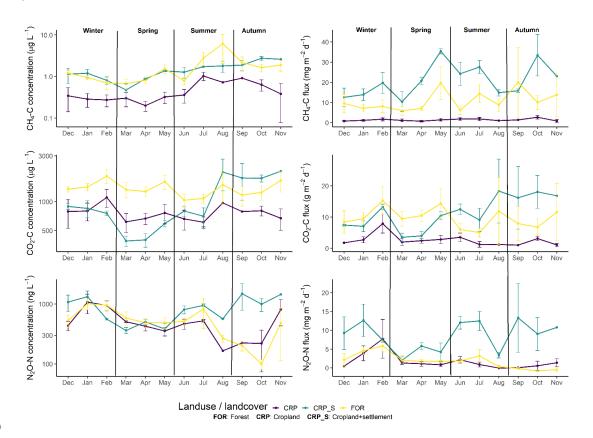
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Fig. A3: Monthly mean \pm SE CO₂, CH₄ and N₂O concentrations and fluxes at forested (FOR), forested + urban (FOR_S), forested + urban + wastewater (FOR_S_W), cropland + urban (CRP_S) and cropland + urban + wastewater (CRP_S_W) sites in the **Neckar** catchment (see Table 1 methods).







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Fig. A4: Monthly mean ± SE CO₂, CH₄ and N₂O concentrations and fluxes at forested (FOR), cropland (CRP)
 and cropland + urban (CRP_S) sites in the Schwingbach catchment (see Table 1 methods).



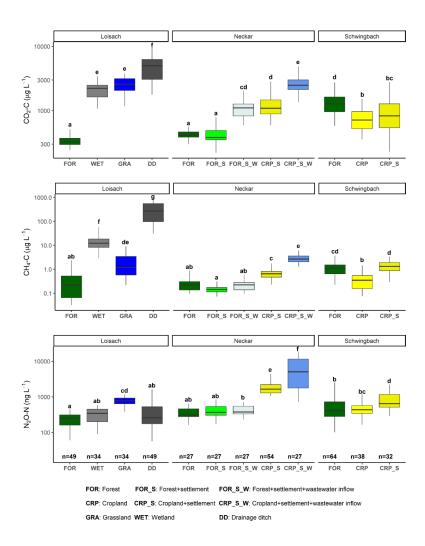


Fig. A5: Boxplots of CO_2 , CH_4 , and N_2O concentrations in stream and ditch waters in the three catchments grouped by dominating land uses (see Table 1 methods). Letters on top of the boxplots represent significant differences (p<0.05) amongst the land use classes across the three catchments based on Tukey post-hoc analyses from the linear mixed-effects models' results (Table 3).





Appendix B: Tables

Table B1: Annual means (+SE) of water chemistry variables and gas concentration measured in the effluents of the Ammer (WWA) and Steinlach (WWS) wastewater treatment plants.

Wastewater effluent quality from inflow zones (Annual Mean ± SE)

	wastewater enfuent quanty from n	mow zones (Annuai Mean ± 5c)
Water quality variables and discharge	Ammer WWA	Steinlach WWS
Temperature (° C)	13.85 ± 0.61	13.72 ± 0.65
pH	7.58 ± 0.07	7.37 ± 0.09
DO (mg L ⁻¹)	6.01 ± 0.32	5.99 ± 0.34
Specific Conductivity	1017.96 ± 63.08	776.68 ± 63.48
NO_3 -N (mg L ⁻¹)	7.57 ± 0.6	6.33 ± 0.47
NH_4 - $N (mg L^{-1})$	0.14 ± 0.02	0.09 ± 0.03
DOC (mg L ⁻¹)	6.8 ± 0.33	5.66 ± 0.58
$TDN (mg L^{-1})$	8.43 ± 0.88	7.58 ± 0.88
CO ₂ -C concentration (µg L ⁻¹)	4020.08 ± 192.75	4529.3 ± 224.37
CH_4 -C concentration ($\mu g L^{-1}$)	2.13 ± 0.3	0.73 ± 0.09
N ₂ O-N concentration (ng L ⁻¹)	9255.11 ± 1563.23	483.23 ± 61.35





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Table B2: Seasonal means (+SE) of water physico-chemical variables, gas concentration and flux measured in the Loisach, Neckar and Schwingbach catchments. Letters beside the means represent significant differences (p<0.05) amongst the seasons across the three catchments based on Tukey post-hoc analyses from the linear mixed-effects models' results (Table 2).

	Summer	Autumn	Winter	Spring
Temperature (° C)	$14.04 \pm 0.2 \mathbf{d}$	$9.83 \pm 0.32 \mathbf{c}$	$5.55 \pm 0.21 \text{ a}$	$8.38 \pm 0.22 \mathbf{b}$
pH	$7.85 \pm 0.03~\textbf{a}$	$7.88 \pm 0.04~\textbf{ab}$	$7.98 \pm 0.04~\textbf{b}$	$7.96 \pm 0.04~\textbf{ab}$
DO (mg L ⁻¹)	$8.71 \pm 0.18~\textbf{a}$	$8.55 \pm 0.29~\textbf{a}$	$9.63 \pm 0.27~\textbf{b}$	$9.85 \pm 0.22~\textbf{b}$
Specific Conductivity	$612.03 \pm 21.8 \mathbf{a}$	$606.91 \pm 28.44~\textbf{b}$	$600.86 \pm 32.62 \text{ ab}$	$555.63 \pm 24.03 \mathbf{a}$
NO ₃ -N (mg L ⁻¹)	$2.54 \pm 0.22~\textbf{a}$	$2.14 \pm 0.29~\textbf{a}$	$2.86 \pm 0.28~\textbf{b}$	$2.6 \pm 0.22~\text{ab}$
NH ₄ -N (mg L ⁻¹)	$0.11 \pm 0.01~\textbf{a}$	$0.14 \pm 0.02~\textbf{a}$	$0.13 \pm 0.02~\textbf{a}$	$0.1 \pm 0.01~\textbf{a}$
TN (mg L ⁻¹)	$2.9 \pm 0.22~\textbf{a}$	$2.49 \pm 0.3~\mathbf{a}$	$3.01\pm0.36\textbf{b}$	3 ± 0.29 ab
DON (mg L ⁻¹)	$0.5\pm0.07 \ \mathbf{a}$	0.75±0.15 a	1.56±0.26 a	1.3±0.24 a
DOC (mg L ⁻¹)	$4.37 \pm 0.24~\textbf{a}$	$4.26\pm0.36~\textbf{a}$	$4.1 \pm 0.31~\textbf{a}$	$4.66\pm0.26~\textbf{a}$
DOC:DIN	$11.45 \pm 2.9~\textbf{b}$	$7.21 \pm 1.37~\textbf{ab}$	$4.14 \pm 0.75~\textbf{a}$	$7.21\pm1.81~\textbf{b}$
DOC:DON	$103.91 \pm 56.91 \ \mathbf{a}$	$183.33 \pm 140.18 \ \mathbf{a}$	$13.19 \pm 2.37 \; \mathbf{a}$	$28.33 \pm 7.31 \; \mathbf{a}$
Stream velocity (m s ⁻¹)	$0.18 \pm 0.01~\textbf{ab}$	$0.12 \pm 0.01~\textbf{a}$	$0.16 \pm 0.01~\textbf{ab}$	$0.24 \pm 0.02~\boldsymbol{b}$
Discharge L s ⁻¹	$526.41 \pm 171.4 \ ab$	$86.25\pm13.07~\textbf{a}$	$157.3 \pm 31.58~\textbf{ab}$	$384.08 \pm 96.29~\textbf{b}$
CO ₂ concentration (µg-C L ⁻¹)	$1198.93 \pm 71.66 \mathbf{a}$	$2222.22 \pm 208.63 \ \mathbf{c}$	$1869.06 \pm 185.95 \; \mathbf{c}$	$1666.03 \pm 148.04 \; \boldsymbol{b}$
CH ₄ concentration (µg-C L ⁻¹)	$20.94 \pm 5.36~\textbf{a}$	$58.08 \pm 17.8~\textbf{a}$	$46.98\pm18~\textbf{a}$	$40.94 \pm 13.03 \; \mathbf{a}$
N ₂ O concentration (ng-N L ⁻¹)	$816.06 \pm 75.58~\textbf{ab}$	$796.45 \pm 169.08 \ \mathbf{a}$	$1691.19 \pm 400.62 \; \boldsymbol{b}$	1021.38 ± 185.45 ab
$k_{600} \text{md}^{-1}$	$32.31 \pm 3.09 \text{ ab}$	$22.71 \pm 2.8~\textbf{a}$	$24.54 \pm 3.36~\textbf{ab}$	$33.92 \pm 3.42 \mathbf{b}$
CO_2 flux (mg-C m ⁻² d ⁻¹)	$17008.98 \pm 1876.63 \; \mathbf{a}$	$22710.21 \pm 3422.95 \ \mathbf{a}$	$14836.51 \pm 1835.54~\mathbf{a}$	$20592.21 \pm 2563.97 \mathbf{a}$
CH_4 flux (mg-C m ⁻² d ⁻¹)	121.65 ± 30.93 a	$233.99 \pm 84.4 \mathbf{a}$	$157.33 \pm 73.04 \mathbf{a}$	$262.87 \pm 89.31 \mathbf{a}$
N_2 Oflux (mg-N m ⁻² d ⁻¹)	$13.69 \pm 2.22 \mathbf{b}$	$9.63 \pm 2.86 \mathbf{a}$	$16.12\pm4.05~\textbf{b}$	$10.64 \pm 2.11 \text{ ab}$





number of observations in each land use class is represented by "n" in brackets. Letters beside the means represent significant differences (p<0.05) amongst the land forest, WET: wetland, GRA: grassland, and DD: drainage ditches), the Neckar (FOR, FOR_S: forest+settlement, FOR_S_W: forest+settlement+wastewater inflow, CRP_S: cropland+settlement, and CRP_S_W: cropland+settlement+wastewater inflow, and the Schwingbach catchment (FOR, CRP: cropland and CRP_S). The Table B3: Annual mean ± standard errors of measured water physico-chemical variables, GHG concentration, and flux for land use classes in the Loisach (FOR: use classes across the three catchments based on Tukey post-hoc analyses from the linear mixed-effects models' results (Table 2).

			Loisach				Neckar	.			Schwingbach	ų,
	FOR (n=49)	FOR (n=49) WET (n=34)	GRA (n=34)	DD (n=49)	FOR (n=27)	FOR_S (n=27)	$FOR_S \ \ (n{=}27) \ FOR_S_W \ (n{=}27)CRP_S \ (n{=}54)$	CRP_S (n=54)	CRP_S_W (n=27)	FOR (n=64)	CRP (n=38)	FOR (n=64) CRP (n=38) CRP_S (n=32)
Temperature (° C)	8 ± 0.5 a	8.6 ± 0.4 ab	9.5 ± 0.2 bd	9 ± 0.5 bc	10.44 ± 1.01 bd 11.6 ± 1.01 de	11.6 ± 1.01 de	12.14 ± 0.85 ef	11.7 ± 0.41 e	13.06 ± 0.63 f	9.7 ± 0.5 be	9.9 ± 0.7 cdef 9.8 ± 0.8 be	9.8 ± 0.8 be
Hd	8.3 \pm 0.01 de 7.7 \pm 0.01 b	$7.7\pm0.01~\mathbf{b}$	$7.6\pm0.01~\mathbf{b}$	$7.3 \pm 0.01 \text{ a}$	8.45 ± 0.05 e	$8.44 \pm 0.05 \mathbf{e}$	$8.07 \pm 0.05 \text{cd}$	$8.13 \pm 0.05 \mathbf{cd}$	$7.72 \pm 0.08\mathbf{b}$	$7.7 \pm 0.01 \mathbf{b}$	$8 \pm 0.01 c$	$8 \pm 0.1 c$
$DO(mgL^{-1})$	$11\pm0.1~\text{de}$	8.3 ± 0.2 c	$7.4\pm0.2\mathbf{b}$	$4.2 \pm 0.3 \mathbf{a}$	11.49 ± 0.39 de	11.57 ± 0.33 de	10.62 ± 0.31 d	$11.65\pm0.17\mathbf{e}$	$8.3 \pm 0.29 \text{ bc}$	8.8 ± 0.1 c	8.9 ± 0.1 c 9 ± 0.1 c	$9 \pm 0.1 \mathbf{c}$
Specific Conductivity	$365.1\pm8.1\mathrm{a}$	365.1 ± 8.1 a 436.9 ± 9.4 ab	447.7 ± 2.3 bc	484.9 ± 16.2 bcd	738.51 ± 51.37 g		$582.07 \pm 13.96 \text{de} \ \ 700.87 \pm 31.16 \text{fg} \ \ \ 1116.86 \pm 31.11 \text{i}$	$1116.86 \pm 31.11 i$	$971.46 \pm 41.76 \mathbf{h}$	$389.7 \pm 18.8 \text{ at}$	$389.7 \pm 18.8 \text{ ab} 597.2 \pm 13 \text{ ef} 566.4 \pm 20.2 \text{ ce}$	566.4 ± 20.2 ce
NO_{3} -N (mgL^{-1})	0.8 ± 0.01 cd	0.8 ± 0.01 cd 0.5 ± 0.01 b	0.8 ± 0.01 cd	0.1 ± 0.01 a	0.57 ± 0.04 bc	2.39 ± 0.13 e	$3.73 \pm 0.29 \text{ ef}$	$6.74\pm0.17\mathbf{h}$	$7.18 \pm 0.38 gh$	1.5 ± 0.1 d	4.9 ± 0.4 fg 2.3 ± 0.2 e	2.3 ± 0.2 e
NH_4 - $N (mgL^{-1})$	$0.01 \pm 0.001 \text{ ab}$	$0.01 \pm 0.001 ab$ $0.01 \pm 0.001 a$	$0.01 \pm 0.001 \mathbf{a}$	$0.3 \pm 0.001 d$	0.07 ± 0.02 bc	0.1 ± 0.01 cd	$0.11\pm0.02\text{cd}$	$0.12 \pm 0.01~\text{d}$	$0.14\pm0.02\mathbf{d}$	0.1 ± 0.01 d	$0.1 \pm 0.01 \mathbf{d}$ $0.1 \pm 0.01 \mathbf{d}$	0.1 ± 0.01 d
TN (mg L ⁻¹)	$0.7 \pm 0.01 \mathbf{b}$ $0.4 \pm 0.01 \mathbf{a}$	$0.4 \pm 0.01 \mathbf{a}$	$0.7\pm0.01~\mathbf{b}$	$0.9\pm0.1\mathbf{b}$	$0.73 \pm 0.06\mathbf{b}$	$2.3\pm0.11~\text{cd}$	$3.92 \pm 0.3 \text{ ef}$	$6.57 \pm 0.19~\mathrm{gh}$	$7.24\pm0.53\mathbf{h}$	$2.2\pm0.2~\mathbf{c}$	6.1 ± 0.5 fg	3 ± 0.3 de
DON (mg L ⁻¹)	0.08±0.02 ab 0.03±0.02 a	0.03±0.02 a	0.06 ± 0.03 acd	0.45±0.04 cd	0.3±0.05 d	0.26±0.08 bd	1.02±0.33 de	2.76±0.48 e	3.6±1.03 e	0.65±0.11 d	1.45±0.24 ce 0.75±0.1 de	0.75±0.1 de
DOC (mg L ⁻¹)	$2.9 \pm 0.3 \mathbf{b}$ $1.8 \pm 0.1 \mathbf{a}$	1.8 ± 0.1 a	$1.5 \pm 0.1 \mathbf{a}$	$9.5 \pm 0.7 g$	5.9 ± 0.67 fg	$4.22 \pm 0.35 \text{bc}$	4.12 ± 0.39 cdf	3.47 ± 0.17 bd	$4.67 \pm 0.23 ef$	4.8 ± 0.2 ef	3.8 ± 0.1 cde 4.7 ± 0.2 cf	4.7 ± 0.2 of
DOC:DIN	4.23±0.46 ef	4.23±0.46 ef 4.48±0.73 ef	2.06±0.22 d	45.14±8.27 h	13.19±2.32 g	1.84±0.24 cd	1.64±0.23 cd	0.62±0.03 a	0.85±0.09 ab	5.89±1.1 f	1.25±0.17 bc 2.82±0.3 de	2.82±0.3 de
DOC:DON	694.26±615.24	694.26±615.24 gl 861.15±610.89 h	93.39±57.03 cfh	37.84±3.02 fg	60.73±30.87 efg	46.02±16.38 dfg	18.06±10.65 acd	5.68±1.9 a	6.65±2.54 ab	37.19±15.88 df	37.19±15.88 df 9.02±2.67 ac 13.13±2.9 bcde	13.13±2.9 bcde
Stream velocity (m s ⁻¹)	0.22 ± 0.03 cd 0.07 ± 0.01 b	0.07±0.01 b	0.22 ± 0.02 cd	0.05±0.01 a	0.3±0.04 de	0.34±0.04 ce	0.4±0.04 e	0.19±0.02 d	0.09±0.02 ab	0.09±0.01 ab	0.09±0.01 ab 0.1±0.01 ab 0.29±0.03 de	0.29±0.03 de
Discharge Ls ⁻¹	$37.7 \pm 7.3 c$	34.5 ± 3.2 cd	$142.1 \pm 20.6 ef$	$11.1 \pm 1.4 \mathbf{b}$	290.56 \pm 109.66 e	290.56 ± 109.66 ef 2053.15 ± 705.38 g 2117.15 ± 730.03 g 318.55 ± 32.65 f	; 2117.15 ± 730.03 g	; 318.55 ± 32.65 f	$196.67 \pm 14.43 f$	$5.4 \pm 0.7 a$	$5.4 \pm 1.3 \mathbf{a}$	94 ± 15.5 de
CO_{2} -C concentration (µg L ⁻¹) 337.9 ± 9.1 a	$337.9 \pm 9.1 a$	$2075.3\pm107.8\mathbf{e}$	2559.5 ± 123.8 e	$4913.5 \pm 285.4 f$	$423.85 \pm 14.6 \mathbf{a}$	$426.67\pm24.18\mathbf{a}$	1093.04 ± 71.11 cd	$423.85 \pm 14.6 \mathbf{a}$ $426.67 \pm 24.18 \mathbf{a}$ $1093.04 \pm 71.11 \mathbf{cd} 1372.92 \pm 104.52 \mathbf{d}$ $2586.47 \pm 191.08 \mathbf{e}$	2586.47 ± 191.08 e	$1350 \pm 65.3 d$	$748.9 \pm 45.1 \mathbf{b}$	1350 ± 65.3 d 748.9 ± 45.1 b 1018.1 ± 117.6 bc
CH_4 -C concentration ($\mu g L^{-1}$) 0.4 ± 0.1 ab	$0.4 \pm 0.1 \text{ ab}$	$16.2\pm2.2\mathbf{f}$	$2.4\pm0.4~\text{de}$	338 ± 37 g	0.25 ± 0.03 ab	$0.25 \pm 0.03 ab$ $0.15 \pm 0.01 a$	$0.23 \pm 0.02~\text{ab}$	$0.72 \pm 0.06\mathbf{c}$	$3.01\pm0.25\mathbf{e}$	1.5 ± 0.2 cd	$0.4 \pm 0.1 \mathbf{b}$ $1.5 \pm 0.1 \mathbf{d}$	$1.5\pm0.1~{\bf d}$
N_2O-N concentration (ng L ⁻¹) 240.9 ± 16.3 a 323 ± 25.1 ab	240.9 ± 16.3 a	$323 \pm 25.1 \text{ ab}$	771.1 ± 42.2 cd	431.3 ± 64.9 ab	355.91 ± 24.26 ab	355.91 ± 24.26 ab 405.94 ± 32.61 ab 421.75 ± 28.5 b	$421.75 \pm 28.5 \mathbf{b}$		1846.46 ± 106.37 e 6600.11 ± 1121.92 f	$569 \pm 59.6 \mathbf{b}$		540 ± 64.5 bc 864.5 ± 89.4 d
$k_{600} \text{md}^{-1}$	$80.9\pm10.6\mathbf{f}$	10.5 ± 0.7 bc	31.5 ± 3.1 df	$6.5\pm0.6\mathrm{a}$	$52.58 \pm 5.1 \mathrm{f}$	$37.66 \pm 3.56 ef$	$43.41 \pm 3.2 \text{ ef}$	$36.26 \pm 2.54 ef$	19.95 ± 2.62 cd	11.7 ± 1.1 ac	7.1 ± 0.9 ab 22.9 ± 1.8 de	$22.9\pm1.8~\text{de}$
CO ₂ -C flux (g m ⁻² d ⁻¹)	$2.39 \pm 0.4\mathbf{a}$	13.33 ± 0.9 df	$50.71 \pm 5.3 \mathrm{g}$	$20.52 \pm 1.9 \text{ ef}$	$6.66 \pm 0.8 \text{ cd}$	$4.89 \pm 0.55 \text{bc}$	$28.26 \pm 2.8 fg$	$39.16 \pm 6.3 fg$	$38.81 \pm 6.5 \text{fg}$	9.54 ± 0.9 cd	2.8± 0.4 ab	10.96± 1.3 cde
CH ₄ -C flux (mg m ⁻² d ⁻¹)	10.5 ± 4.3 ab 101.7 ± 8.3 f	$101.7 \pm 8.3 f$	73.2 ± 15.7 de	$1532.9 \pm 244.8 \mathbf{g}$	9.09 ± 1.5 bc	$3.88 \pm 0.7 \text{ ac}$	$6.54 \pm 0.81~\text{bc}$	$21.09 \pm 2.37 \mathbf{d}$	58.23 ± 13.33 e	9.9 ± 1.3 c	$1.5 \pm 0.2 \text{ a}$	21.5 ± 2 de
$N_2O-N \text{ flux } (\text{mg m}^{-2} d^{-1})$ 1.1 ± 0.9 a	$1.1 \pm 0.9 a$	$0.8 \pm 0.2 a$	$12.4 \pm 1.4 c$	$1.2 \pm 0.4 a$	$0.32 \pm 0.63 a$	$2.2 \pm 0.64 \mathbf{a}$	$3.96 \pm 0.85 ab$	$46.92 \pm 5.02 d$	$67.59 \pm 11.34 d$	$2.1 \pm 0.3 a$	$1.9 \pm 0.6 a$	8.8 ± 1.1 bc





Table B4: Indices highlighting the performance of the best-fit SEMs, which indicate significant interaction pathways of both direct and indirect drivers of in-situ GHG concentrations in temperate streams, rivers, and drainage ditches. The goodness of fit index (GFI), comparative fit index (CF1), Tucker lewis index, standardized root mean square residual (SRMR), and root means squared error of approximation (RMSEA) are measures of model goodness of fit, while the parsimony fit index (PNFI) compares the best-fit model to the theoretical-model.

	Perfor	mance in	dices for	the best-fit	SEMs		Model con	nparison
							PNFI	_
Greenhouse gas (GHG)	GFI	CFI	TLI	SRMR	RMSEA	r ²	Theoretic	al SEM Best-fit SEM
CO ₂ concentration (µg-C L ⁻¹	1.00	1.00	1.00	0.02	< 0.01	0.60	0.13	0.22
CH ₄ concentration (µg-C L ⁻¹	1.00	1.00	1.00	0.02	< 0.01	0.66	0.13	0.22
N ₂ O concentration (ng-N L	¹) 0.99	1.00	0.98	0.03	0.04	0.46	0.13	0.22

Best-fit SEM structure:-

- $1. \ Log \ GHG = DO + DOC + Log \ NO_3 \ + agricultural \ area + was tewater \ inflow + stream \ velocity$
- 2. Log NO₃ = DO + agricultural area + wastewater inflow + stream velocity
- 3. DOC = agricultural area
- 4. DO = DOC + stream velocity

Goodness of fit assesment:- GFI, CFI and TLI: 0.90 - 0.95; Good fit and >0.95 Excellent fit

 \boldsymbol{SRMR} and $\boldsymbol{RMSEA} \colon 0.05$ - 0.08; Good fit and $<\!\!0.05$ Excellent fit

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Data availability

Our institute is currently constructing an own data infrastructure to host all available data from its scientist and link it to the main KIT database. We are working currently with our institute's data department to provide a publicly accessible DOI link for all our research data as directed by our institute's policy. We hope the data will be soon available before the review process completes. Reviewers are however invited to ask for the data anytime during the review process and it will be provided by the corresponding author via email.

Author contribution

RM, RK, GG, CG, and KB designed the field experiments. RK, KB, TH, and LB provided the infrastructural funding and RM and EW did the field and laboratory work. RM did the statistical analysis, consulting with RK and GG. RM prepared the first draft manuscript, consulting with RK. All co-authors contributed to the final version.

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Declaration of competing interest

The authors declare that they have no conflict of interest.





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