A model-based assessment of the potential use of compound specific stable isotope analysis in river monitoring of diffuse pesticide pollution

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Abstract

Compound-specific stable isotope analysis (CSIA) has, in combination with model-assisted interpretation, proven a valuable approach to quantify the extent of organic contaminant degradation in groundwater systems. CSIA data may also provide insights into the origin and transformation of diffuse river pollutants such as pesticides and nitrate at the catchment scale. While CSIA methods for pesticides have increasingly become available, they have not yet been deployed to interpret isotope data of pesticides in surface water. We applied a coupled subsurface-surface reactive transport model (HydroGeoSphere) at the hillslope scale to investigate the usefulness of CSIA in the assessment of pesticide degradation. We simulated the transport and transformation of a pesticide in a hypothetical but realistic two-dimensional hillslope transect. The steady-state model results illustrate a strong increase of isotope ratios at the hillslope outlet, which resulted from degradation and long travel times through the hillslope during average hydrological conditions. In contrast, following an extreme rainfall event that induced overland flow, the simulated isotope ratios dropped to the values of soil water in the pesticide application area. These results suggest that CSIA can help to determine whether pesticides enter the stream via groundwater exfiltration or via surface runoff. Simulations with daily rainfall and evapotranspiration data and one pesticide application per year resulted in small seasonal variations of concentrations and isotope ratios at the hillslope outlet, which fell within the uncertainty range of current CSIA methods. This implies a good reliability of in-stream isotope data in the absence of transport via surface runoff or other fast transport routes, since the time of measurement appears to be of minor importance. The analysis of simulated isotope ratios also allowed quantifying the contribution of two different reaction pathways to the overall degradation, which gave further insight into transport routes in the modelled system. The simulations supported the use of the commonly applied Rayleigh equation for the interpretation of CSIA data, since this led to an underestimation of the real extent of degradation of less...
than 12% at the hillslope outlet. Overall, the model results emphasize the applicability and usefulness of CSIA in the assessment of diffuse river pollution.

1 Introduction

Modern agriculture makes use of a variety of pesticides to increase crop yield and reduce pests and the growth of weeds. As a result, pesticides have become ubiquitous organic contaminants in agricultural catchments. Diffuse pollution by pesticides can pose a risk for the terrestrial and aquatic environment and human health. Pesticide residuals and their metabolites have been found in groundwater and surface water and affect drinking water quality (Donald et al., 2007; Kolpin et al., 1998; Kjær et al., 2005). It is therefore vital to assess the fate of diffuse pollutants, and to identify major pesticide sources in a catchment.

After application, pesticides are subject to various transfer, transformation, and transport processes (Gavrilescu, 2005; Flury, 1996). Important transfer processes are volatilization (the transfer of compounds from the solid or liquid phase to the gas phase), and sorption (the transfer from the liquid to the solid phase of the soil matrix). Pesticide molecules that have not been volatilized can undergo transformation processes on the ground surface and in the soil, which leads to destruction of the compound. The transformation of pesticides is either ascribed to microbial activity or abiotic processes. Microbial degradation in the soil matrix can occur under aerobic or anaerobic conditions. The transport of pesticides in the aqueous phase towards surface water bodies can occur via surface runoff or subsurface flow. If the pesticide is not directly removed by surface runoff, it leaches into the subsurface. Drainage systems, preferential flow (e.g. in earthworm burrows and cracks) and subsurface storm flow can thereupon cause fast transport to surface water (Gavrilescu, 2005; Müller et al., 2003; Leu et al., 2004b). In contrast, pesticide leaching to groundwater represents a slow subsurface transport mechanism (Holvoet et al., 2007; Flury, 1996).
Previous research has shown that rainfall events increase the risk of surface water contamination by pesticides (Taghavi et al., 2011; Müller et al., 2003). The highest pesticide loads in response to rainfall events happen shortly after pesticide application, and are attributed to surface runoff and preferential flow (Leu et al., 2004a). A secondary contribution to the overall pesticide export is ascribed to the baseflow component (Louchart et al., 2001; Squillace and Thurmanz, 1992).

Monitoring of pesticide concentrations in rivers enables us to assess the extent of diffuse pollution at the catchment scale. However, concentration data do not provide clear evidence of degradation processes, since a reduction in concentration might, for example, indicate the degradation of the contaminant, changes in the application pattern and amount, or dilution. In contrast to degradation, non-destructive processes such as dilution do not reduce the contaminant loads to subsurface and surface water.

In this context, compound-specific isotope analysis (CSIA) has emerged as a valuable tool for the analysis of the transformation of organic contaminants.

CSIA is the measurement of the isotopic composition, i.e. the ratio between the abundance of a heavy and a light stable isotope of an element in a compound. This isotope ratio tends to increase during degradation since chemical bonds that contain a heavy isotope are less amenable to degradation processes than those that solely consist of light isotopes (Meckenstock et al., 2004; Elsner, 2010). This phenomenon, called isotope fractionation, does not occur during dilution (van Breukelen, 2007a), and it is only relevant under specific conditions for other physical non-destructive processes such as diffusion (van Breukelen and Rolle, 2012) and sorption (van Breukelen and Prommer, 2008). In contrast to concentration monitoring, CSIA thus serves as a direct indicator of the transformation of a contaminant and, by measuring multiple elements, even allows for the determination of the degradation mechanism that leads to the transformation (Elsner, 2010).

CSIA has been applied to analyze the fate of various groundwater pollutants, such as monoaromatic and polyaromatic hydrocarbons (Blum et al., 2009; Vieth et al., 2005; Griebler et al., 2004), MTBE (Kolhatkar et al., 2002; Zwank et al., 2005), chlorinated
ethenes (Hunkeler et al., 2005; Sherwood Lollar et al., 2000), perchlorate (Sturchio et al., 2012), nitrate (Deutsch et al., 2006; Ging et al., 1996; Seiler, 2005; Zhang et al., 2012), and chromium(VI) (Wanner et al., 2012a). However, except for the identification of nitrate sources, it has so far not been used to examine diffuse pollution at the catchment scale. Similarly, although analytical methods for CSIA of different elements have been developed for several pesticides (Hartenbach et al., 2008; Meyer et al., 2008; Penning and Elsner, 2007; Badea et al., 2009), isotope data of pesticides have not yet been continuously monitored and analyzed in rivers.

Recently, the use of reactive transport models that incorporate isotope fractionation effects (isotope fractionation reactive transport models, IF-RTMs) has become a popular method to model and interpret CSIA data from point-source pollutants in groundwater systems (Atteia et al., 2008; D’Affonseca et al., 2011; Pooley et al., 2009; Prommer et al., 2009; van Breukelen et al., 2005; Wanner et al., 2012b). In particular, assumptions concerning reaction kinetics can be validated by comparing model results to measured concentration and CSIA data (D’Affonseca et al., 2011; Pooley et al., 2009; Prommer et al., 2009; Atteia et al., 2008). IF-RTM has also proven useful to study the transformation of the agrochemical pollutant nitrate (Chen and MacQuarrie, 2004; Green et al., 2010). As opposed to the simulation of concentrations of a degrading contaminant, which solely yields a certain concentration reduction between the emission source and a downgradient point, these studies have demonstrated that IF-RTMs allow for the quantification of in-situ degradation and the distinction between destructive and non-destructive processes. Moreover, IF-RTMs permit the assessment of the performance of the Rayleigh equation, which is the mathematical basis of CSIA. The Rayleigh equation tends to underestimate the extent of biodegradation in real flow systems because of the attenuation of isotopic enrichment due to dispersion or mixing processes (Abe and Hunkeler, 2006; van Breukelen and Prommer, 2008).

The aforementioned IF-RTM studies have solely focussed on groundwater systems. In contrast, the current study presents the first incorporation of isotope fractionation effects into a reactive transport model of a coupled surface-subsurface system, which
involves more complex transport routes. The aim of the modelling study was to analyze whether CSIA measured in surface water can help to identify transport routes and determine the extent of degradation of diffuse agricultural pollutants in subsurface-surface systems. We modelled a hypothetical but realistic situation of pesticide application, transport, and degradation, including isotope fractionation at the hillslope scale. To this end, we applied a distributed physically-based model, which allowed for the simulation of hydrological processes in different flow domains in a detailed and spatially explicit way (Kampf and Burges, 2007) and the description of pesticide fluxes and concentrations (Thorsen et al., 1996). This study thus ties in with the virtual experiment approach in hillslope hydrology of Weiler and McDonnell (2004) and Hopp et al. (2009), and additionally considers reactive transport and isotope fractionation. We opted for a hillslope transect because hillslopes are a fundamental landscape element and the basic hydrological unit of catchments. Hence, understanding processes at the hillslope scale is an important and relevant first step towards interpreting CSIA at the catchment scale.

By means of scenario modelling, we examined the evolution of isotope ratios under average hydrological conditions, in response to an extreme rainfall event, and under transient, daily varying hydrological conditions. In order to make the transition from the virtual experiment to real applications, we give, based on the simulation results for this hillslope model, general guidelines for the monitoring and interpretation of isotope data in the context of diffuse pollution.

2 Methods

2.1 Model code description: HydroGeoSphere

We simulated isotope fractionation during the transport and transformation of a hypothetical pesticide with the software HydroGeoSphere (HGS) (Brunner and Simmons, 2012; Therrien et al., 2010). HGS is a fully-coupled, subsurface-surface flow and solute transport model. Surface flow is described by the diffusion-wave approximation of
the Saint Venant equation, while the variably saturated form of the Richards’ equation
is used for the subsurface. The Newton–Raphson technique is implemented to solve
the non-linear equations of variably-saturated flow. Solute transport is simulated by
solving the advection-dispersion equation; the degradation of the solute is modelled
as first-order kinetics. Furthermore, it is possible to include interception, transpiration
and evaporation processes. Applications of HGS include large watershed modelling
(Goderniaux et al., 2009; Li et al., 2008), simulations of aquifer-river interactions in hy-
pothetical model domains (Doble et al., 2012; McCallum et al., 2010), and the analysis
of contaminant transport in groundwater (Rivett et al., 2006; Sudicky et al., 2010).

2.2 Hillslope geometry and model grid

The model domain consists of a two-dimensional hillslope that is 200 m long, and
stretches 15 m in the vertical direction (Fig. 1). Conceptually, the hillslope represents
part of a headwater catchment. The hillslope outlet corresponds to a river monitoring
point. Inflow of water and pesticides from upstream parts was not considered in this
model. The hillslope is convex with an average gradient of 5 %. A river bank was incor-
porated as a vertical drop of 2 m over the last 5 m in the x-direction, which increased
the average slope to 5.8 %. The model domain does not represent, nor was calibrated
for a specific field site, but is supposed to resemble a realistic agricultural hillslope.

Figure 1a shows the model mesh and the three subsurface zones (see Sect. 2.3) in
the hillslope. The subsurface comprises a total of 43 layers. The vertical grid spacing is
a few centimetres in the topsoil and increases with depth, which results in a maximum
cell height of more than one meter at the bottom of the model domain. The mesh nodes
are 0.5 m apart in the horizontal direction, except for the zone near the river, which has
a horizontal discretization of 0.25 m (Fig. 1b).
2.3 Hydraulic properties and flow simulation

The hillslope consists of three zones with different properties: the topsoil extends 0.3 m below the ground surface, the subsoil is located between 0.3 and 2 m below the ground surface, and the remaining part of the subsurface represents bedrock. The saturated hydraulic conductivity for the three subsurface zones was set as follows: 1.0 m d\(^{-1}\) in the topsoil, 0.5 m d\(^{-1}\) in the subsoil, and 0.1 m d\(^{-1}\) in the bedrock (Table 1). These values are comparable to those chosen by Christiansen et al. (2004) to simulate pesticide transport in agricultural soil. The values for the porosity, the residual saturation, and the parameters \(\alpha\) and \(\beta\) (which determine the saturation-pressure relation in the Van Genuchten model) for the top- and subsoil were taken from a study of a silty sand site in Canada (Abdul, 1985). The values for the bedrock were set to those given by Mirus et al. (2011) for the Coos Bay (CB) catchment (Table 1). The specific storage coefficient was kept at its default value (1 \times 10^{-4} m\(^{-1}\)) for all three zones.

HGS explicitly represents the overland flow domain as a layer on top of the subsurface domain. The nodes of the overland flow domain coincide with the top elements of the subsurface domain. The overland flow parameters rill storage height and coupling length, which determine the depth of depressions on the ground surface and the flux between the surface and the subsurface domain, respectively, were both set to 0.01 m based on previous studies using HGS (Goderniaux et al., 2009; Therrien et al., 2010).

The lateral and bottom boundaries of the hillslope were set to zero-flux boundaries such that the water could leave the hillslope only via the overland flow domain. This was realized by applying a critical depth boundary at the hillslope outlet, which yields a time varying flux that is given by Manning’s equation (Therrien et al., 2010). The hillslope outlet was represented by two nodes in the surface layer, since the hillslope extends one cell into the \(y\) direction. Rainfall was applied at a spatially uniform rate onto the entire model domain.

Rainfall and potential evapotranspiration data were supplied as model input. The actual evapotranspiration in HGS depends on the moisture content and several
vegetation-related parameters. The difference between precipitation and actual evapotranspiration is then added to or abstracted from the model domain, according to which component is dominant. The parameters for evapotranspiration were kept at their default values, except for the leaf area index, which was set to 4.2 according to the value for temperate and tropical crops given in the HGS manual, and the interception by canopy, which was assumed to be negligible.

2.4 Reactive solute transport

We considered a soluble, non-volatile, and non-sorbing hypothetical pesticide, which are characteristics of the widely used compounds MCPA, bentazone, metam-sodium and clopyralid (PPDB, 2013), as well as nitrate. We therefore neglected volatilization and sorption processes. The pesticide was applied to the cells at the ground surface between $x$ coordinates 50 and 130 m (Fig. 1a). Depending on the simulated scenario, the solute transport boundary condition was either set to a specified concentration or to a specified mass flux (see Sect. 2.6). The transport parameters were chosen as follows: an aqueous diffusion coefficient of $7.8 \times 10^{-5}$ m$^2$ d$^{-1}$, which is the value that was assumed for the diffusion of the herbicides metolachlor and alachlor in groundwater by Lee and Benson (2004); and the default values of dispersivities in HGS of 1.0 m in the longitudinal direction and 0.1 m in the vertical transverse direction.

The hypothetical pesticide was assumed to degrade via two different pathways: an aerobic reaction in the topsoil and an anaerobic reaction in the subsoil and bedrock. A pesticide half life of 51 days, corresponding to an overall degradation rate constant of 5 per year, was chosen for the aerobic reaction (Table 2), which is within the range of reported half lives of widely used pesticides (e.g. linuron and clopyralid; PPDB, 2013). The half life for the anaerobic reaction was assumed to be much longer (1265 days; rate constant of 0.2 yr$^{-1}$).

The model yielded concentrations for the entire model domain. Additionally, for each time step and solute, the concentration at the hillslope outlet was obtained by tracking the rates of mass flux and water flow through the boundary notes. These concentra-
The simulations were run in finite difference mode with upstream weighting. The adaptive time stepping scheme was used; the maximum time step was set to one day for the steady state simulations and to 0.01 days for the transient simulation (see Sect. 2.6). A maximum concentration change of 0.1 in each time step relative to the source concentration was allowed for the steady state simulation; this value had to be increased to 500 for the transient simulation because of the temporary high mass load.

2.5 Simulation of isotope fractionation effects

Isotope fractionation effects were included into the model by simulating the concentration of light and heavy isotopes of the pesticide. Since the assumption of two degradation pathways requires the analysis of two isotopic elements (van Breukelen, 2007b), this resulted in the simulation of four solutes. The two-dimensional isotope analysis was implemented by simulating carbon and hydrogen isotopes because carbon represents the element with the most isotope data available for pesticides (e.g. Annable et al., 2007; Meyer et al., 2009; Penning et al., 2010), and the cleavage of chemical bonds in pesticide molecules with hydrogen atoms induces a significantly stronger isotope fractionation effect than, for example, with nitrogen atoms (Hartenbach et al., 2008; Penning et al., 2010).

Following Hunkeler et al. (2009), van Breukelen et al. (2005) and van Breukelen and Prommer (2008), the differential equations for the degradation of the light carbon and hydrogen isotopes were simulated as a first-order reaction with the degradation rate constant $k$:

$$\frac{dL\cdot C}{L\cdot C} = k \cdot C$$

(1)
with $^L C$ being the concentration of the light carbon or hydrogen isotopes, respectively. The reaction kinetics of the heavy isotopes, involving a different degradation rate constant $^H k$, were specified as:

$$\frac{d^H C}{H C} = ^H k \cdot ^H C$$

(2)

where $^H C$ is the concentration of the heavy carbon and hydrogen isotopes, respectively.

The kinetic isotopic fractionation factor, $\alpha$, represents the ratio between the degradation rate constants of the heavy and light isotopes (i.e. $^H k / ^L k$). It thus determines the strength of the isotope fractionation effect for a specific reaction. Since it typically has a value close to one, it is reported in per mil ($\%_o$) as the kinetic isotopic enrichment factor $\varepsilon (\%_o) = (\alpha - 1) \cdot 1000$. The degradation rate constant of the light isotopes ($^L k$) was set to the overall degradation rate constant ($k_{eff}$, Table 2). Given $\alpha = ^H k / ^L k$ and $^L k = k_{eff}$, the degradation of the heavy carbon and hydrogen isotopes was thus simulated as

$$\frac{d^H C}{H C} = k_{eff} \cdot \alpha \cdot ^H C = k_{eff} \cdot \left( \frac{\varepsilon}{1000} + 1 \right) \cdot ^H C.$$

(3)

The enrichment factors ($\varepsilon$) for aerobic and anaerobic degradation for the carbon isotopes were chosen to be representative for fractionation effects during biotic pesticide degradation (Meyer et al., 2009; Penning et al., 2010). Data about fractionation effects for hydrogen isotopes are much scarcer; the enrichment factors for hydrogen were therefore chosen such that they followed the general trend of stronger enrichment in hydrogen compared to carbon isotopes (Hunkeler and Elsner, 2009).

In a system with two transformation pathways, the ratio between the enrichment factors of two isotopic elements, for example of hydrogen and carbon ($\varepsilon_H / \varepsilon_C$), can be indicative of a specific degradation mechanism (Meyer et al., 2009; Fischer et al., 2008). The enrichment factors for the two reaction pathways were therefore chosen
such that they yield distinct $\varepsilon_H/\varepsilon_C$-ratios (Table 2). This is in agreement with considerably diverging $\varepsilon_H/\varepsilon_C$-ratios that were, for example, observed for aerobic and anaerobic biodegradation of MTBE (Zwank et al., 2005).

Combining and integrating Eqs. (1) and (2) leads to the Rayleigh equation, which allows for the quantification of in-situ degradation on the basis of isotope ratios. Its simplified form can be expressed as:

$$\frac{R_{\text{sample}}}{R_{\text{source}}} = f(\alpha - 1)$$

where $R$ represents the ratio between the abundance of a heavy and a light isotope of an element in a compound for a sample ($R_{\text{sample}}$) and at the emission source ($R_{\text{source}}$), respectively, and $f$ denotes the non-degraded fraction of the compound in the sample with respect to the emission source.

To facilitate inter-sample comparison, the isotope ratio of a sample ($R_{\text{sample}}$) is expressed in the $\delta$-notation, which is the relative difference of $R_{\text{sample}}$ from a standard ratio $R_{\text{standard}}$ (Schmidt and Jochmann, 2012):

$$\delta_{\text{sample}} = \frac{R_{\text{sample}} - R_{\text{standard}}}{R_{\text{standard}}} = \frac{R_{\text{sample}}}{R_{\text{standard}}} - 1$$

where the $\delta$ value or isotopic signature is commonly reported in per mil ($\%\delta$). The standard ratios that were used in this study are the international standards for carbon and hydrogen isotope ratios, i.e. Vienna Pee Dee Belemnite (VPDB; $R_{\text{standard}} = 0.011237$) and Vienna Standard Mean Ocean Water (VSMOW; $R_{\text{standard}} = 1.5575 \times 10^{-4}$), respectively.

The $\delta^{13}C$-value at the source was fixed at $-30 \%\delta$, which corresponds to typical values for carbon isotopes in pesticides (Annable et al., 2007; Kawashima and Katayama, 2010; Badea et al., 2009). Since no typical source values were available for hydrogen isotopes in pesticides, a value of $-100 \%\delta$ was chosen, which is consistent with $\delta^{13}C$. 8800
δ²H values reported for other organic contaminants (Mancini et al., 2008; Wang et al., 2004). These initial δ values were used to determine the ratio of the concentrations of light and heavy isotopes at the pesticide source (Eq. 5). The simulated concentrations of the light and heavy isotopes were used to calculate the δ values for the entire hillslope domain and the hillslope outlet.

As sorption and volatilization processes were disregarded, the model did not account for the influence of these processes on the isotopic composition of the pesticide. We assumed that volatilization-induced isotope fractionation is negligible, as rainfall in the simulated scenarios (see Sect. 2.6) would lead to a rapid mobilization of the recently applied compound. Similarly, while sorption during transport might induce small fractionation effects (Kopinke et al., 2005), this is only relevant at the front of a spreading pollution plume and in the absence of degradation (van Breukelen and Prommer, 2008). We also disregarded diffusion-induced isotope fractionation effects, as their importance for the simulated system should be minor given the widespread nature of the emission and the relatively large spatial scope.

### 2.6 Simulated scenarios

In a preliminary model run, a recharge rate of 250 mm yr⁻¹ was applied to the whole surface domain in order to achieve a steady-state flow field. This value was considered representative for average hydrological conditions for the Netherlands and northern Germany (Otto, 2001; Querner, 2000) and did not cause any surface saturation, except for a few nodes at the hillslope outlet that represent the river bank and bottom. The distribution of the hydraulic head values at the end of this simulation was used as the initial condition for subsequent simulations with solute transport. We simulated three different scenarios with solute transport: scenario 1 represents steady state conditions in order to analyse the pattern of isotopic enrichment of the pesticide during transport and degradation under average hydrological conditions; scenario 2 focuses on the response of concentrations and isotope ratios to an extreme rainfall event to de-
termine the effect of surface runoff; and scenario 3 incorporates both periods of base-
flow conditions and extreme rainfall events to study the transient pattern of pesticide
concentrations and isotopes in the course of the year.

2.6.1 Scenario 1: steady-state flow conditions

The first scenario was designed to mimic diffuse pollution under average hydrological
conditions. The emission source was implemented as a specified concentration bound-
ary with a constant relative concentration of \( C_0 = 1.0 \) for the sum of the light and heavy
isotopes of each isotopic element. Two conservative tracers with \( C_0 = 1.0 \) were applied
between \( t = 0 \) and \( t = \infty \) across the entire surface of the model domain and at the
application area only to allow for the calculation of the mean travel time of ground-
water. This scenario was run until the concentrations at the hillslope outlet reached
steady-state.

2.6.2 Scenario 2: extreme rainfall event

As a hillslope system is in reality exposed to varying hydrological conditions, it was sub-
sequently studied how concentrations and isotope ratios responded to a single extreme
rainfall that leads to surface runoff. To facilitate the occurrence of surface runoff, the
coupling length for this scenario was increased from 0.1 to 0.8 m. Rainfall was applied
with a uniform intensity of 60 mm h\(^{-1}\) for 30 min. Based on a rainfall depth-duration-
frequency curve derived for the Netherlands (Overeem et al., 2009), the return period
for such an event is more than 58 yr. For the rest of the simulation time, the recharge
rate was held constant at the same value as for the steady-state scenario (scenario 1).
Solute transport was initialized with the concentration results from the steady-state sim-
ulation; the concentration in the source area was kept at a constant value of \( C_0 = 1.0 \)
for the total concentration of each isotopic element.
2.6.3 Scenario 3: transient simulation for water flow and solute input

Finally, a time-varying hydrological system was realized by applying the daily rainfall and evapotranspiration data of 2010 from the meteorological station “Twente” (Database of the Royal Netherlands Meteorological Institute, Koninklijk Nederlands Meteorologisch Instituut KNMI) for 20 yr. This time series (Fig. 2a) includes an extraordinary rainfall event at the end of August with a maximum intensity of 27.6 mm h$^{-1}$ and a total rainfall amount of 106.4 mm; the rainfall event lasted for 20.8 h. Based on Overeem et al. (2009), the return period of such an event exceeds 260 yr. The use of this data thus allowed for the simulation of a broad range of possible rainfall intensities and runoff responses for Mid-European climates; it resulted in a net groundwater recharge of approximately 360 mm yr$^{-1}$ (Fig. 2b). The annual rainfall and evapotranspiration dataset was repeated for a total of 20 simulation years. This facilitated the interpretation of the results as opposed to an actual dataset for 20 consecutive years, which would have produced a different hydrological and isotopic response for every simulation year.

Since pesticides are not applied at a constant rate throughout the year, a more realistic setup comprises short emission pulses. To this end, the pesticide was applied once a year on a dry day in spring (11 April; day 100), using a specified relative mass flux of 1 for the sum of light and heavy isotopes of each isotopic element for every grid cell at the pesticide source. The specified mass flux was chosen as the boundary condition for this scenario because the pesticide application was supposed to occur on a day without precipitation. If the boundary condition had been set to a specified concentration, the pesticide would not have been transferred into the subsurface because of the lack of infiltration on dry days. The application of the pesticide occurred on the same day in every simulation year; the model domain did not contain any pesticide at the beginning of the simulation. In order to assure the presence of pesticide in the system, which is required for the simulation of degradation, the model run was started on the first application day.
2.7 Post-model calculations

For the steady-state conditions, the cumulative transit time distribution at the hillslope outlet was derived from the concentration of the conservative tracer that was applied to the entire surface of the model domain using the relation between the transit time distribution and conservative solute breakthrough described by Duffy and Lee (1992) and further explored by Eberts et al. (2012). The mean travel time (MTT) of the system was calculated by integrating the transit time distribution. The MTT for the pesticide at steady-state was obtained from the concentration at the outlet of the conservative tracer that was only applied at the pesticide application area.

Two-dimensional CSIA permits the assessment of the proportional contribution of each transformation pathway to the overall degradation in a system of two competing reactions. For the simulated system, which involves an aerobic and an anaerobic reaction affecting carbon and hydrogen isotopes, this contribution was calculated according to van Breukelen (2007b):

\[
F = \frac{\Phi \cdot \varepsilon_{C_{\text{anaerobic}}} - \varepsilon_{H_{\text{anaerobic}}}}{\left(\varepsilon_{H_{\text{aerobic}}} - \varepsilon_{H_{\text{anaerobic}}}\right) - \Phi \cdot \left(\varepsilon_{C_{\text{aerobic}}} - \varepsilon_{C_{\text{anaerobic}}}\right)}
\]

(6)

where the subscripts C and H indicate the enrichment factors for carbon and hydrogen isotopes, the subscripts aerobic and anaerobic denote the respective pathway, and \( \Phi \) is the ratio of the isotopic shifts \( \Delta^2H \) and \( \delta^{13}C \) for carbon and hydrogen: \( \Phi = \Delta^2H/\delta^{13}C \).

The isotopic shift, i.e. the change in the \( \delta \) value, is defined as

\[
\Delta = 1000 \cdot \ln \left(\frac{10^{-3} \cdot \delta_{\text{sample}} + 1}{10^{-3} \cdot \delta_{\text{source}} + 1}\right)
\]

(7)

where \( \delta_{\text{sample}} \) and \( \delta_{\text{source}} \) are the \( \delta \) values for the carbon or hydrogen isotopes at a point in the model domain and at the pollution source, respectively.
With the relative share $F$ of one pathway given by Eq. (6), the non-degraded fraction of the compound at a point in the model domain, $f_{\text{deg}}$, was obtained from the Rayleigh equation for two-dimensional CSIA (van Breukelen, 2007b):

$$f_{\text{deg}} = \left( \frac{\delta_{\text{sample}} + 1000}{\delta_{\text{source}} + 1000} \right)^{\frac{1000}{C_{\text{aerobic}} + (1-F)C_{\text{anaerobic}}}}. \quad (8)$$

The fraction $f_{\text{deg}}$ yielded the extent of degradation based on the Rayleigh equation approach:

$$B(\%) = (1 - f_{\text{deg}}) \cdot 100. \quad (9)$$

The exact value of the fraction of non-degraded pesticide, $f_{\text{deg}, m}$, was calculated from the simulated concentrations of the pesticide and the conservative tracer. It was used to quantify the underestimation of the extent of degradation that would result from the application of the Rayleigh equation to the simulated isotope data (Abe and Hunkeler, 2006). Given that the rate constant, $k$, of a first-order reaction satisfies $-kt = \ln(f)$, this underestimation, $\theta$, was calculated as (van Breukelen and Prommer, 2008):

$$\theta(\%) = \left( 1 - \frac{k_{\text{Rayleigh}}}{k_{\text{model}}} \right) \cdot 100 = \left( 1 - \frac{\ln(f_{\text{deg}})}{\ln(f_{\text{deg}, m})} \right) \cdot 100. \quad (10)$$

The simulation results were also used to derive the residual fractions of the pesticide that remain at the hillslope outlet after dilution ($f_{\text{dil}}$) and degradation ($f_{\text{deg}}$) following the assumption of an open system (van Breukelen, 2007a). The parameter $f_{\text{deg}}$ was determined by applying Eq. (8) to the simulated isotope ratios at the hillslope outlet. The dilution factor $f_{\text{dil}}$ was then obtained as

$$f_{\text{dil}} = \frac{f_{\text{tot}}}{f_{\text{deg}}}. \quad (11)$$

where $f_{\text{tot}}$ is the ratio between the concentration at the hillslope outlet and the source concentration.
3 Results

3.1 Scenario 1: steady-state flow conditions

The constant recharge rate in scenario 1 produced a steady-state flow and transport regime with a mean travel time (MTT) of 6.7 yr for the groundwater and of 5.0 yr for the pesticide. At the hillslope outlet, this resulted in a steady-state concentration of 0.41 for the conservative tracer that was applied at the pesticide application area and of 0.09 for the degrading pesticide relative to the initial concentration of 1.0 at the pollution source. Accordingly, concentrations of the pesticide were lower than for the conservative tracer in the entire model domain. Pesticide concentrations were the highest below the application area and decreased with depth and distance from the source (Fig. 3a).

In contrast to the shallow subsurface beneath the application area, the shallow soil layers at the hillslope bottom were characterized by low solute concentrations. Figure 3b shows the carbon isotope ratios in the hillslope domain, which became progressively enriched with distance from the source area. Hence, with a shift of up to 4 % from the initial value of −30 %, the shallow subsurface at the lower hillslope was characterized by the largest enrichment in δ¹³C. Hydrogen isotopes show an enrichment of about 20 % in the shallow soil layers close to the hillslope outlet (Fig. 3c). In contrast to carbon isotopes, the strongest fractionation effects occurred in the deeper bedrock, which displays an enrichment of up to 25 %.

Degradation during transport induced isotope fractionation, which resulted in a steady state isotope ratio of −26.7 % for δ¹³C and −80.8 % for δ²H at the hillslope outlet. According to the two-dimensional Rayleigh equation, this corresponds to an extent of degradation of 73 % at the hillslope outlet. It follows from Eq. (11) that the concentration reduction due to degradation was approximately 27 % (fdeg = 0.27 in Eq. 11), while the residual fraction after the effect of dilution was 34 % (fdil = 0.34 in Eq. 11). Dilution therefore contributed slightly less to the concentration decrease than degradation. The extent of degradation, derived from the simulated isotope values (Eq. 9), increased with depth and distance from the pollution source (Fig. 3d). With...
a value of 80%, it was the highest in the deep bedrock and in the shallow soil layers at the footslope. The $\theta$ value, which compares the extent of degradation given by the two-dimensional Rayleigh equation to the model-based extent of degradation (Eq. 10), was largest below the application area (up to 25%). Apart from this area, the underestimation remained below 20% and decreased to a value of about 5% in the deep bedrock (Fig. 3f). It reached a value of 11.5% at the hillslope outlet.

Figure 3e shows the relative contribution $F$ of the aerobic reaction to the overall degradation of the pesticide according to Eq. (6). $F$ decreased rapidly with depth from a fraction of more than 0.9 in the topsoil below the pesticide application area to less than 0.2 in the deeper bedrock. At the hillslope outlet, it reached a value of 0.39.

### 3.2 Scenario 2: extreme rainfall event

Figure 4a illustrates the pesticide concentrations and mass flux at the hillslope outlet in response to the extreme rainfall event. It can be seen that with the onset of rainfall, concentrations first dropped and subsequently reached a distinct maximum. In contrast, the pesticide mass flux shows two peaks, with the second being much more pronounced than the first. In response to the rainfall event, the isotope signatures at the hillslope outlet dropped from their steady-state values of $\delta^{13}C = -26.7$‰ and $\delta^2H = -80.8$‰ to $\delta^{13}C = 30.0$‰ and $\delta^2H = -100.0$‰ (Fig. 4b). The latter values are the characteristic isotopic signatures of the pollution source. The minima in the isotope ratios coincided with the maximum in pesticide concentrations and the second peak in pesticide mass flux (Fig. 4a).

### 3.3 Scenario 3: transient simulation for water flow and solute input

The changing hydrological conditions and the annual pesticide application in scenario 3 induced a transient concentration response in the subsurface. We consider the first fifteen years of the transient simulation as spin-up period and focus on the results of the last five simulation years. Figure 5 shows a comparison of the pesticide concentration
at two different times: after 19 yr (panel a) and after 19 yr and 50 days (panel b), thus 50 days after the last pesticide application. It illustrates that the pesticide plume from the new application mixed with the residual plume from the previous applications.

The inclusion of evapotranspiration into the simulation resulted in several periods of negative net infiltration, which followed the time of pesticide application (shaded areas in Fig. 6). Hereafter, this phase will be denoted as the dry period. The dry period was characterized by an increase in pesticide concentrations till the extreme rainfall event in Aug (Fig. 6b). Two other concentration peaks occurred later in each simulation year (beginning of November and end of February), while concentrations sharply dropped following increased rainfall in November. After the third concentration peak in February, the pesticide concentrations decreased again until the next pesticide application. In contrast to the steady state scenario, the pesticide concentrations at the hillslope outlet thus varied significantly and followed the same seasonal pattern every year. In particular, some of the days with a high rainfall amount induced distinct minima in the pesticide concentrations.

Figure 6c shows the $\delta^{13}$C- and $\delta^2$H-values, which alternated between periods of enrichment and depletion. Moreover, despite the spin-up period of fifteen years, the isotope ratios still show a slight increase. The hydrogen isotope ratios declined to their minimum in the middle of the wet season and reached the most enriched value at the end of the dry season (shaded area). In contrast, the peak in carbon isotope ratios occurred at the onset of the dry season. While the sharp drops in pesticide concentrations correlate with the minima in $\delta^2$H, they correspond to maxima in $\delta^{13}$C.

The time course of the contribution of the aerobic degradation to the overall mass destruction ($F$) mirrored the seasonal pattern of the $\delta^{13}$C-values: The maximum occurred at the beginning of the dry season (Fig. 6d). In contrast, the extent of degradation reached its maximum at the end of the dry season before it declined again following the extreme rainfall event in August and subsequent precipitation. It thus shows a similar response as the $\delta^2$H-values (Fig. 6c). In addition, both parameters reflect the
response of the isotope ratios to rainfall events: While the distinct peaks in $\delta^{13}\text{C}$-values correspond to maxima in $F$, the drops in $\delta^2\text{H}$-values coincided with minima in $B$.

The underestimation of the extent of degradation ($\theta$) that results from the use of the two-dimensional Rayleigh equation approach increased during the wet period and with subsequent simulation years. It reached a value of 9.2% at the hillslope outlet at the end of the transient simulation (Fig. 6e). In addition, $\theta$ shows distinct peaks that correlate with the concentration minima during rainfall events. With respect to the entire model domain, the maximum in $\theta$ occurred in the shallow subsurface below the application area in response to every pesticide input. Subsequently, this zone of the highest $\theta$ moved downgradient with the plume centre (not shown).

4 Discussion

4.1 Scenario simulations

4.1.1 Pesticide movement through the hillslope

Scenario 1 yielded an average travel time of five years for the pesticide, which implies a relatively long response time to the emission of diffuse pollutants. This is consistent with elevated atrazine concentrations in a spring in a small catchment even several years after the use of the herbicide had been abandoned (Gutierrez and Baran, 2009). Accordingly, the pesticide reached the hillslope outlet approximately 1.4 yr after the first application in the spin-up period of the transient scenario (scenario 3). This, in turn, highlights the potential time lag between the first use and the first detection of a newly introduced pesticide in stream flow. Analogous to the continuous detection of an abandoned compound, the extent of contamination caused by a new compound might only manifest itself after several consecutive application years. Moreover, the isotope ratios and derived parameters in the transient scenario did not reach an oscillatory steady state at the end of the 20 yr simulation. Correspondingly, the extent of degradation ($B$)
still increased over the course of the last five simulation years (Fig. 6d). This is another indicator of the long response time of the modelled hydrological system to diffuse pollution.

At steady state, the pesticide concentrations in the shallow soil layers at the footslope were low (Fig. 3a), which resulted from their position above the water table during steady-state conditions: The transfer of pesticides into the unsaturated shallow soil close to the outlet could only occur via dispersion and diffusion from deeper layers, whereas the transport through the saturated zone was driven by faster advective transport. The part of the pesticide that was transported to the shallow subsurface at the footslope had, therefore, undergone more degradation than the pesticide that was present directly below the water table (Fig. 3d). Correspondingly, the shallow subsurface soil at the footslope also shows more enriched isotope ratios relative to other parts of the model domain (Fig. 3b and c).

The first two periods of increasing concentrations in each simulation year in transient scenario 3 ended abruptly in response to rainfall events (Fig. 6b). These rainfall events thus led to a pronounced dilution effect in the pesticide concentrations. The third concentration peak in every simulation year was, in contrast, not followed by the occurrence of a large rainfall event. It is, instead, related to a phase of generally wetter conditions (Fig. 6a), which induced a stronger dilution of pesticide concentrations. In summary, concentrations seem to be rather driven by the hydrological conditions than by the application pattern of the pesticide. However, the application pattern might have influenced the second phase of rising concentrations following the extreme rainfall event because the pesticide reached the hillslope outlet 1.4 yr after the first application. The second increase thus coincided with the arrival of the pesticide plume from the previous year. Consequently, the hydrological conditions immediately affected the concentration pattern, whereas the response to the emission of the contaminant into the system was much more delayed.

The results of the steady state simulation illustrate how the relative contribution of the aerobic degradation pathway ($F$) rapidly decreased with depth (Fig. 3e). This is
due to the fact that aerobic conditions solely occurred in the topsoil. Accordingly, flow lines that display long travel times through the bedrock exhibit a significant extent of degradation under anaerobic conditions. However, even directly below the application area, $F$ did not attain its maximum value of one because of the effect of vertical dispersion that caused pesticide transport from the anaerobic subsoil into topsoil layers. $F$ reached a steady-state value of 39.0% at the hillslope outlet, although the topsoil accounts for less than 3% of the total subsurface volume. This mainly resulted from the higher degradation rate in the topsoil ($5 \text{ yr}^{-1}$) than in the subsoil and bedrock ($0.2 \text{ yr}^{-1}$), which disproportionately increased the contribution of the aerobic reaction to the overall degradation. Furthermore, the pesticide traversed the aerobic zone during infiltration and exfiltration, which enhanced the imprint of the aerobic degradation pathway. In contrast, the influence of the anaerobic transformation was significantly lower than expected from its large volume. This indicates that, in addition to the slower degradation rate of the anaerobic transformation, a considerable part of the pesticide was transported via shallow flow paths, which did not traverse the deeper bedrock and thus had a relatively short residence time in the anaerobic zone. The relatively large contribution of shallow flow paths to the hillslope discharge resulted, in turn, from the decrease in saturated hydraulic conductivity with depth. Hence, the chosen hydraulic properties also influenced the relative contribution of the aerobic reaction pathway to the overall degradation.

4.1.2 Carbon and hydrogen isotope ratios

In general, isotope ratios increased with travel distance, as more degradation leads to a more pronounced isotope fractionation. This becomes apparent when considering the progressive enrichment in carbon and hydrogen isotope ratios with distance from the source area during steady state (Fig. 3b and c). However, carbon and hydrogen isotopes show a different strength of enrichment in the deep bedrock, which can be ascribed to the choice of enrichment factors for the aerobic and the anaerobic zones. In contrast to the hydrogen isotopes, the isotopic enrichment of carbon was stronger...
for the aerobic reaction pathway in the topsoil than for the anaerobic reaction pathway in the subsoil and bedrock. Hence, the pesticide that crossed the topsoil for the second time during exfiltration shows the most pronounced enrichment in $\delta^{13}$C. In addition, this re-entrance into the topsoil was driven by a dispersive flux, which is, as explained before, associated with a higher degree of degradation than the advective flux. In contrast, the most enriched $\delta^2$H-values occurred in the deep bedrock, since long travel times through the subsurface together with the high hydrogen enrichment factor under anaerobic conditions induced the strongest fractionation effect for the hydrogen isotopes. Corresponding to these different areas of largest isotope fractionation for carbon and hydrogen, the two-dimensional Rayleigh equation (Eq. 8) yielded a high degree of degradation for the deep bedrock and the shallow soil layers at the footslope (Fig. 3d).

Similar to scenario 1, the different choice of enrichment factors also explain the differences in the seasonal patterns of the hydrogen and carbon ratios in scenario 3. Since the enrichment for hydrogen was assumed to be stronger in the anaerobic than in the aerobic zone, the increase in $\delta^2$H during the dry season reflects the rising contribution of deeper groundwater. This agrees with the shape of the curve of the extent of degradation (Fig. 6d), which resembles more the pattern of $\delta^2$H than of $\delta^{13}$C. In general, an increasing contribution of deeper flow paths, which implies longer travel times, also caused a higher degree of fractionation for the carbon isotopes. Nevertheless, a smaller relative contribution of the aerobic reaction pathway can account for a less pronounced overall fractionation effect for carbon because of the higher enrichment factor in the topsoil. Accordingly, rising $\delta^{13}$C-values correspond to a relative increase in aerobic degradation during the wet season (Fig. 6d), which demonstrates that shallow flow through the topsoil was more pronounced under wet conditions. This finding agrees with Brown et al. (1999) and Rozemeijer and Broers (2007), who emphasized the increasing importance of shallow subsurface flow during storm events in headwater catchments and lowlands, respectively. Pesticide transport via shallow layers favoured carbon isotope fractionation and thus led to a stronger enrichment in $\delta^{13}$C than the
transition to pesticide transport via deeper flow paths during the dry period. Nonetheless, the overall extent of degradation increased with longer residence times, which were the highest at the end of the dry season.

4.1.3 Responses to rainfall events

The response of concentrations and isotope ratios to the extreme rainfall event in scenario 2 highlight the advantages of combined measurements of concentration and CSIA data. Since the mass flux peaked, the concentration minimum at the onset of the rainfall event has to be ascribed to initial dilution by rainwater (Fig. 4a). The subsequent concentration peak occurred shortly after the rain had ceased. Given that the mass flux shows a second maximum, this concentration peak stemmed from increased pesticide transport. However, it cannot be concluded from the concentration and mass flux data which mechanism led to this pulse. In contrast, the analysis of isotope ratios serves as an indicator of the underlying mechanism. The CSIA data reveal the occurrence of surface runoff, which is reflected in the drop of $\delta^{13}C$ and $\delta^2H$ to the source values of $-30$ and $-100\%$, respectively (Fig. 4b). As the time lag between the release from the pollution source and the arrival at the hillslope outlet was too short (less than 30 min) to allow for significant degradation, the pesticide did not undergo any detectable fractionation. Hence, the dominance of pesticide transport via overland flow caused the isotope ratios at the hillslope outlet to decrease to the source signature.

The peak in pesticide concentrations shows a slight delay relative to the onset of the rainfall, as the pollution source is located uphill from the pesticide-free area at the lower hillslope. Consequently, the extreme rainfall event initially induced discharge of clean water from the part of the hillslope downstream of the application area, which was then followed by contaminated surface runoff that originated from the pollution source. However, the increase in mass flux at the beginning of the rainfall event suggests that enhanced pesticide transport via exfiltration of shallow groundwater proceeded pesticide transport via surface runoff, which illustrates the emission of “old” contaminant residues with pre-event water at the onset of rain (Burt and Pinay, 2005).
be inferred from the analysis of the parameter $F$, which indicates a slightly enhanced contribution of topsoil degradation after the onset of rainfall and prior to the discharge of the contaminated surface runoff (not shown).

Surface runoff in scenario 2 occurred as infiltration excess overland flow, which can be an important mechanism of pesticide emission from agricultural land (Doppler et al., 2012). The infiltration excess overland flow was generated by increasing the coupling length from 0.1 to 0.8 m. A large coupling length results in a decoupling of the overland flow domain from the subsurface domain, which can be considered analogous to the decreased infiltration capacity as a result of surface sealing during high-intensity rainfalls. This approach was also applied in the HGS simulation by Verbist et al. (2012) to generate infiltration excess overland flow. A layer of low hydraulic conductivity at the surface would have had the same effect, but this would have required a modification of the model domain. Alternatively, the rainfall intensity could have been increased, but without changing the coupling length, a sufficiently long period of overland flow could have only been achieved with an unrealistically high rainfall amount (data not shown).

With a return period of nearly 60 yr, the simulated rainfall event in scenario 2 corresponds to an exceptional event for mid-European climate. This extreme rainfall event was required to generate continuous overland flow from the area of the pesticide application to the hillslope outlet and thereby produce a discernible response in the isotope ratios in the stream. Nonetheless, scenario 2 can be considered representative of situations where the surface is sealed or has been disturbed, or where a pesticide product is spilled on impermeable areas such as paved farmyard. In this case, subsequent surface runoff from these contaminated areas may lead to concentration peaks even during relatively dry periods (Holvoet et al., 2007; Kreuger, 1998). Similarly, preferential flow to drainage systems represents a fast transport route of pesticides that leads to elevated pesticide concentrations in the absence of surface runoff at the application area (David et al., 2003). Pesticide transport via drain flow and surface runoff from impermeable areas can thus also occur in response to rainfall of lower intensity. The response in isotope ratios would be comparable to the one simulated in scenario 2.
This mechanism has already been observed for oxygen isotope ratios of nitrate: Ging et al. (1996) found elevated $\delta^{18}$O-values in a storm sewer, which were indicative of an atmospheric nitrate source, and ascribed them to surface runoff from impervious areas. In addition, CSIA data that is measured during storm events could be used for forensic source determination, i.e. for the distinction between different pesticide products and application areas that show varying isotopic compositions.

The sharp drops in pesticide concentrations in response to large rain events in scenario 3 (Fig. 6b) result from dilution. Moreover, these drops also occurred on days with smaller events, provided that the rainfall (in combination with previous wet conditions) was sufficient to saturate the footslope. This had a similar dilution effect as high intensity rainfalls. The concentration minima correlate with minima in hydrogen isotope ratios, as high intensity rainfall and footslope saturation were accompanied by the discharge of more shallow groundwater, which had a shorter residence time and thus caused less hydrogen isotope fractionation. In contrast, the concentration minima correspond to peaks in the carbon isotope ratios, since they are associated with a relatively large contribution of topsoil degradation and thus more pronounced carbon isotope fractionation. Consequently, analogous to the seasonal pattern of isotope ratios, the different strength of isotope fractionation for the aerobic and the anaerobic reaction pathway resulted in the opposite behaviour of carbon and hydrogen isotope ratios during single rainfall events. This demonstrates how CSIA can give an insight into transport routes of a contaminant in a hydrological system, provided that reaction pathways can be attributed to different subsurface zones. The rainfall events in scenario 3 that led to footslope saturation were accompanied by minima in the extent of degradation (Fig. 6e), since they induced a higher emission of more recently applied and thus less degraded pesticide. This agrees with previous findings about nitrate contamination where rainfall events led to a fast mobilization of soil nitrate with a relatively depleted $\delta^{15}$N-value, while increasing denitrification in soil between rainfall events resulted in progressively enriched $\delta^{15}$N-values (Kellman and Hillaire-Marcel, 2003). In contrast to the simulation of the extreme rainfall event in scenario 2, the isotope ratios did not drop...
to the initial values of the pesticide application area in scenario 3, which implies the absence of pesticide transport via surface runoff. This can be explained by the time lag of 137 days between the pesticide application and the extreme rainfall event in August, which allowed for infiltration and degradation of the pesticide prior to intense rainfall as opposed to the simultaneous application of rain and pesticide in scenario 2.

### 4.2 Implications for the applicability of CSIA to assess pesticide transport and transformation

In order to examine the applicability of compound specific isotope analysis in the context of diffuse agricultural pollutants, the concentration decrease in scenario 1 needs to be compared to the difference between soil water concentrations and detection limits of CSIA. Initial soil water concentrations of up to a few milligrams per litre have been reported for pesticides at realistic application rates (Liu et al., 2012), while carbon isotopes of organic contaminants have been quantified at concentrations of 100 ng L\(^{-1}\) (Jochmann et al., 2006). Consequently, the difference between pore water concentrations and the detection limit for carbon isotope analysis can be four orders of magnitude. By way of comparison, the simulation yielded a concentration reduction by factor 10 between the application area and the hillslope outlet. Therefore, the model results indicate that, given appropriate sampling and preconcentration techniques, low environmental concentrations of pesticides would not impede CSIA of diffuse river pollutants.

The magnitude of the simulated enrichment between the source area and the hillslope outlet in scenario 1 exceeds the uncertainty range of CSIA. Isotope ratios increased by 4 \(\%\) for \(\delta^{13}C\) and by 20 \(\%\) for \(\delta^2H\), while the instrumental uncertainty is about 0.5 \(\%\) for carbon and 5 \(\%\) for hydrogen isotope analysis (Sherwood Lollar et al., 2007). This indicates that the enrichment in in-stream isotope composition relative to the source values is detectable, which supports the use of CSIA in the analysis of diffuse river pollution under average hydrological conditions. It should be noted, however,
that the magnitude of isotope enrichment depends, among others, on the site-specific travel times and isotope fractionation effects.

In contrast to the response to the extreme event in scenario 2, the isotope ratios in scenario 3 display only small variations, which did not exceed 0.3 ‰ for δ¹³C and 1.4 ‰ for δ²H during each simulation year. Hence, these small fluctuations would not be detectable based on the current accuracy of carbon and hydrogen CSIA. This suggests that in the absence of overland flow, the assessment of the extent of in-situ degradation on the basis of CSIA would yield the same results throughout the year. It follows that grab samples for CSIA measurements seem sufficient for a representative assessment of pesticide transformation between the pollution source and river monitoring point, except during events that result in direct transport via overland flow. This might, however, not apply to systems that show a larger temporal variation in the relative contribution of baseflow versus stormflow to streamflow, which could, for example, result from the activation of preferential flow paths during rainfall events or from a larger difference between the permeabilities of the soil and bedrock. A more pronounced variation in in-stream CSIA might also arise from pesticide use closer to the river than for the modelled hillslope, as rainfall events would then be more likely to lead to the emission of recently applied pesticide, while the dry season would be characterized by the emission of strongly degraded pesticide via baseflow. Moreover, catchments with well-drained soils, long mean travel times and a high groundwater contribution to streamflow tend to dampen solute input signals to a much larger extent than catchments that exhibit more responsive soils and short mean transit times (Hrachowitz et al., 2009). Consequently, the latter systems might show a significant seasonal variability in isotope data that reflects the short emission pulses of pesticide application and the fast subsurface transport mechanisms in response to rainfall events. The determination of factors that would cause larger seasonal variabilities in in-stream CSIA was not the objective of this study, but could be addressed in future research.

During extreme rainfall events, only a fine temporal sampling resolution can allow for the detection of transient surface runoff, as the latter might only occur for a short
time. According to the comparable response in concentrations and isotope ratios in scenario 2, its detection by CSIA requires the same temporal resolution as concentration measurements. We suggest that isotope ratios could be analyzed in addition to concentrations during storm events. This requires the use of automatic sampling devices, which would allow capturing surface runoff from potentially contaminated areas. If a concentration peak in the monitoring data occurs, CSIA could additionally be performed in order to detect isotope ratios that correspond to typical values for pesticide products.

The model results for the isotope ratios allowed for the detailed analysis of the seasonal pattern of the extent of degradation ($B$), the relative contribution of the aerobic reaction pathway ($F$), and the underestimation by the use of the Rayleigh equation approach ($\theta$). In the case of measured CSIA data, these parameters might, however, appear constant over time because of minor seasonal fluctuations in isotope ratios. Nonetheless, even a constant value of these parameters can facilitate the analysis of the underlying transport and transformation mechanisms in the studied flow system. Since CSIA represents a unique method for the determination of these parameters, this highlights an additional benefit of isotope analysis in the context of diffuse river pollutants.

The extent of degradation during pesticide transport through the hillslope was determined by applying the Rayleigh equation to the simulated isotope ratios. Nevertheless, the Rayleigh equation is in principle only applicable to closed and fully mixed systems (e.g. to degradation experiments in microcosms) (van Breukelen and Prommer, 2008). Hydrological systems such as the modelled hillslope are, however, open systems, which display a variety of transport and transfer processes. For example, dispersion leads to varying transport velocities and routes that characterize the same measurement point. Therefore, while some molecules might have been subject to strong isotope fractionation during the transport along a flow path, others might have travelled much faster via a different flow path. The latter molecules exhibit less fractionation because of less exposure to degradation processes (Aravena and Hunkeler, 2009).
other words, physical heterogeneities of the aquifer and hydrodynamic dispersion can result in the attenuation of isotope fractionation effects that are only inherent in a part of the sample mixture (van Breukelen and Prommer, 2008). This masking effect might also occur in areas where flow paths from different aquifer layers mix (Kopinke et al., 2005). Consequently, if the Rayleigh equation approach is applied in such open flow systems, it tends to result in an underestimation of the extent of in-situ degradation (Abe and Hunkeler, 2006; van Breukelen and Prommer, 2008). For the simulated hillslope system, the extent of degradation at steady state was significantly underestimated for the topsoil below the application area. This is due to the high degradation rate under aerobic conditions, which induced larger concentration gradients and thus larger dispersive fluxes than the slower reaction rate under anaerobic conditions. Large dispersive fluxes cause, in turn, an attenuation of the isotopic shifts and, therefore, result in a more significant underestimation of the actual extent of degradation (van Breukelen and Prommer, 2008). Similarly, the zone of the strongest underestimation in scenario 3 was located at the plume centre, as the pesticide plume is associated with the strongest concentration gradients in the system (not shown).

Although transient hydrological conditions are likely to cause enhanced mixing of different flow pathways, the degree of underestimation at the hillslope outlet in the steady state simulation was slightly larger than for the transient scenario. This might result from a greater importance of deeper flow paths for pesticide transport at steady state, which agrees with a smaller contribution of the aerobic pathway at steady state than in the transient scenario. Pathways with a long travel time thus controlled the isotope ratios at the hillslope outlet to a larger degree in scenario 1 than in scenario 3. Consequently, the contribution of pesticide that displays advanced degradation and thus strong isotope fractionation to the overall pesticide export was more significant at steady state. This, in turn, favoured the masking of strong isotope fractionation in scenario 1, which resulted in a higher $\theta$ for scenario 1 compared to scenario 3. Scenario 3 shows a higher $\theta$ during the wet period than during the dry period (Fig. 6e), which illustrates the effect of mixing of different flow paths on the accuracy of the Rayleigh
equation: the growing relative contribution of more shallow flow paths during wet conditions resulted in increased discharge of more recently applied pesticide, which led to an enhanced masking of the isotopic enrichment in deeper groundwater. The same mechanism caused the distinct peaks in $\theta$ following rainfall events.

The underestimation of degradation due to the use of the Rayleigh equation is typically below 5% for groundwater plumes (Abe and Hunkeler, 2006). However, it has been shown that the underestimation can exceed 50% at fringes of pollution plumes, especially for high degradation rates and at large distances from the contaminant source (van Breukelen and Prommer, 2008; van Breukelen and Rolle, 2012). In view of the large underestimation for groundwater systems, the maximum $\theta$ value at the hillslope outlet of 11.5 and 10.0% in scenario 1 and scenario 3, respectively, can be considered negligible. Therefore, the simulation results suggest that CSIA yields a good assessment of in-situ degradation, not only for aquifer systems, but also for more complex subsurface-surface systems that are subject to mixing of different flow pathways.

This study illustrated how coupled subsurface-surface models permit to investigate isotope fractionation effects related to diffuse pollution. Future studies might extend such an analysis to a three-dimensional catchment and include in-stream degradation and dilution. In addition, if the model setup comprises several pollution sources, CSIA may allow for source identification and apportionment. Finally, transport via suspended matter can be an important aspect in diffuse pollution by hydrophobic agrochemicals, which should be considered for highly sorptive pesticides and erosion-prone sites.

5 Conclusions

In this paper, we present a model study of compound-specific stable isotope analysis (CSIA) in the context of diffuse river pollution. The objective was to examine whether CSIA qualifies as a feasible and expedient technique for the analysis of transport pathways and the assessment of the extent of degradation of diffuse pollutants. We simulated reactive solute transport and isotope fractionation effects for a hypothetical hills-
lope. The model results support the usefulness of CSIA data in this context: the simulated isotope data allowed for the quantification of the extent of in-situ degradation and the relative contribution of two competing pathways to the overall degradation, which would not have been possible on the basis of simulated concentration data only.

The two-dimensional Rayleigh equation provided a reliable estimate of the overall extent of degradation under transient conditions. In particular, the inherent underestimation of the Rayleigh equation approach was small, considering the high degree of mixing of groundwater flow paths from different depths at the hillslope toe. The attenuation of isotope signals, which partly results from this mixing, did not exceed the degree of attenuation that has been reported in previous studies of groundwater pollution plumes.

The simulation of an extreme rainfall event illustrated how isotope data can reveal the occurrence of surface runoff and thereby indicate the immediate transport of a diffuse pollutant into a river. In this way, CSIA allowed for the distinction between pesticide emission via groundwater exfiltration and via overland flow, drains, or direct spillage. However, the simulation results also showed that surface flow might only be discernible in the CSIA data for a very short period, which requires the use of automated sampling procedures that respond to large rainfall events.

The simulation of transient hydrological conditions resulted in small seasonal variations in isotope ratios and derived parameters, which would not be detectable in actual CSIA data (with the exception of pesticide emission via surface runoff in response to storm events) because they would fall within the uncertainty range of current analytical methods. For systems with a larger seasonal variation in isotope ratios (resulting from, e.g. the activation of preferential flow paths in response to storm events), CSIA could yield a time-dependent estimate of the extent of degradation. In the case of a system with low seasonal variability such as the modelled hillslope, however, CSIA would give a stable result throughout the year, regardless of the temporal sampling resolution. This, in turn, supports the practicability of CSIA in the analysis of river contamination. Provided that the relevant underlying degradation mechanisms and associated isotope...
fractionation factors are known, CSIA thus offers a unique tool for the assessment of pesticide transformation, and even for the analysis of the interplay between different transport routes. In addition to modelling studies, it is crucial to evaluate the feasibility of CSIA of diffuse river pollutants in experimental studies, especially in view of the low environmental concentrations. These studies have not been conducted yet, but the model results advocate their applicability and advantages. In conclusion, this study emphasized the potential benefits of CSIA in the characterization of diffuse river pollution.

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References


PPDB (2013): The Pesticide Properties DataBase (PPDB) developed by the Agriculture & Environment Research Unit (AERU), University of Hertfordshire, funded by UK national sources and through EU-funded projects, 2006–2013.


### Table 1. Hydraulic properties of the subsurface domain.

<table>
<thead>
<tr>
<th>Property</th>
<th>Topsoil</th>
<th>Subsoil</th>
<th>Bedrock</th>
</tr>
</thead>
<tbody>
<tr>
<td>Saturated hydraulic conductivity [m d$^{-1}$]</td>
<td>1.0</td>
<td>0.5</td>
<td>0.1</td>
</tr>
<tr>
<td>Porosity [-]</td>
<td>0.37$^a$</td>
<td>0.37$^a$</td>
<td>0.12$^b$</td>
</tr>
<tr>
<td>Residual saturation [-]</td>
<td>0.18$^a$</td>
<td>0.18$^a$</td>
<td>0.01$^b$</td>
</tr>
<tr>
<td>Van Genuchten parameters</td>
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<td></td>
</tr>
<tr>
<td>$\alpha$ [m$^{-1}$]</td>
<td>1.9$^a$</td>
<td>1.9$^a$</td>
<td>4.3$^b$</td>
</tr>
<tr>
<td>$\beta$</td>
<td>6$^a$</td>
<td>6$^a$</td>
<td>1.3$^b$</td>
</tr>
</tbody>
</table>

$^a$ From Abdul (1985).
$^b$ From Mirus et al. (2011).
Table 2. Parameters for degradation and isotope fractionation.

<table>
<thead>
<tr>
<th></th>
<th>$k_{\text{eff}}^a$ [a$^{-1}$]</th>
<th>$\varepsilon_C^b$ [%]</th>
<th>$\varepsilon_H^c$ [%]</th>
<th>$\varepsilon_H/\varepsilon_C$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Aerobic (topsoil)</td>
<td>5</td>
<td>-5</td>
<td>-10</td>
<td>2</td>
</tr>
<tr>
<td>Anaerobic (subsoil and bedrock)</td>
<td>0.2</td>
<td>-1</td>
<td>-20</td>
<td>20</td>
</tr>
</tbody>
</table>

$^a$ Overall degradation rate constant (Eq. 3).
$^b$ Enrichment factor for carbon isotope fractionation.
$^c$ Enrichment factor for hydrogen isotope fractionation.
Fig. 1. Model domain with pesticide application area (red) and three subsurface zones: topsoil (yellow), subsoil (green) and bedrock (blue) (a). The discretization is finer in the upper soil and close to the hillslope outlet (b). The location of the hillslope outlet is represented by two boundary nodes that are marked with a red circle.
Fig. 2. Daily rainfall and evapotranspiration data for the meteorological station Twente (a); and net infiltration as the difference between precipitation and modelled actual evapotranspiration (b). The dataset shown in panel a was repeated for 20 yr in the transient simulation. The pesticide application (11 April; day 100) is marked with the dashed vertical line in red. The simulation was started on this day to ensure the presence of pesticide in the hillslope system.
Fig. 3. Steady-state results for the pesticide concentration (a); $\delta^{13}$C values (‰) of the pesticide (b); $\delta^{2}$H values (‰) of the pesticide (c); the extent of degradation (%), (d); the relative contribution of the aerobic reaction to the overall degradation of the pesticide (e); and the underestimation (%) of the true extent of degradation when the Rayleigh equation is used (f). Areas with a concentration reduction of more than three orders of magnitude relative to the source are blanked, because CSIA would not be possible due to detection limits. The black lines with arrow heads indicate streamlines of the steady state flow field. The purple line shows the position of the water table. Vertical exaggeration is five times.
**Fig. 4.** Response to the extreme rainfall event at the hillslope outlet immediately before and after the rainfall event: pesticide concentration and mass flux (a, note the logarithmic scale for the mass flux); and carbon and hydrogen isotope ratios of the pesticide (b). The timing of the rainfall event is shaded in grey.
Fig. 5. Pesticide plume before and after the last application: concentrations after 19 yr (a) and after 19 yr and 50 days (b). Areas with a concentration reduction of more than three orders of magnitude relative to a concentration of 1.0 are blanked because CSIA would not be possible due to detection limits. The black line indicates the position of the water table.
Fig. 6. Results of the last five years of the transient simulation at the hillslope outlet (after 15 yr of spin-up): application days and net infiltration (a); pesticide concentration (b); δ¹³C and δ²H isotope ratios (c); extent of degradation and relative contribution of the aerobic reaction pathway to the overall degradation based on the two-dimensional Rayleigh equation approach (d); and underestimation of the extent of degradation resulting from the use of the Rayleigh equation (e). Periods of a negative net infiltration (dry periods) are shaded in grey.