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Fingerprinting using compound-specific $\delta^{13}\text{C}$ of n-alkanes indicates the relative importance of paddy fields as a temporary secondary source of watershed sediment loss

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Abstract:	<p>Sediment fingerprinting generates reliable sediment provenance information which supports policy or practical strategies for catchment sediment management. But the approach remains challenging in areas with complex landscape configuration. This investigation evaluated the seasonality of biomarker signatures and their variability among particle size fractions, and accordingly apportioned target time-integrated suspended sediment to land-use based sources in an intensive farming watershed with mosaic land use patch configurations and crop-specific farming practices. Source materials (i.e., topsoil) from dry croplands, paddy fields and citrus orchards were sampled, and time-integrated suspended sediment was collected at the watershed outlet. The absolute concentrations and compound-specific $\delta^{13}\text{C}$ of long-chain saturated n-alkanes (C23-C33) were determined for two absolute particle size fractions (i.e., $<25\ \mu\text{m}$ and $25-63\ \mu\text{m}$). The $\delta^{13}\text{C}$ of monomeric n-alkanes displayed no significant variabilities between the particle aize fractions nor across the whole sampling period. The MixSIAR Bayesian model was employed to quantify sediment source contributions. Due to human activities, paddy fields have become an important sediment source, but dry farmland remains the largest contributor. Based on sediment source information for the study watershed, a range of measures such as soil virginization, returning straw to fields, and pasture cultures in orchards are recommended.</p>
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Editor-In-Chief

International Soil and Water Conservation Research

Dear Editor,

Enclosed, please find a manuscript entitled “Fingerprinting using compound-specific $\delta^{13}\text{C}$ of n-alkanes indicates the relative importance of paddy fields as a temporary secondary source of watershed sediment loss”, which has been nominated for the WASWAC Youth Outstanding Paper Award 2024 (YOPA ID: YOPA-2024-037) and is now submitted to *International Soil and Water Conservation Research* for publication.

The Three Gorges Reservoir Area represents a hot-spot region for soil erosion due to mountainous and hilly landform, frequent erosive rainfall storms, and intensive surface land disturbance by human activities, where soil conservation is of utmost importance in terms of its potential adverse consequences to receiving freshwater impounded by the dam. Sediment fingerprinting generates reliable sediment provenance information which supports policy or practical strategies for catchment sediment management. But the approach remains challenging in areas with complex landscape configuration. This investigation evaluated the seasonality of biomarker signatures and their variability among particle size fractions, and accordingly apportioned target time-integrated suspended sediment to land-use based sources in an intensive farming watershed with mosaic land use patch configurations and crop-specific farming practices. Source materials (i.e., topsoil) from dry croplands, paddy fields and citrus orchards were sampled, and time-integrated suspended sediment was collected at the watershed outlet. The absolute concentrations and compound-specific $\delta^{13}\text{C}$ of long-chain saturated n-alkanes (C23-C33) were determined for two absolute particle size fractions (i.e., <25 μm and 25-63 μm). The $\delta^{13}\text{C}$ of monomeric n-alkanes displayed no significant variabilities between the particle size fractions nor across the whole sampling period. The MixSIAR Bayesian model was employed to quantify sediment source contributions. Due to human activities, paddy fields have become an important sediment source, but dry farmland remains the largest contributor. Based on sediment source information for the study watershed, a range of measures such as soil virginization, returning straw to fields, and pasture cultures in orchards are recommended.

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We thank you very much for considering this work and look forward to your response.

Yours sincerely,

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Declaration of conflict of interest

The authors declare that they have no conflict of interest.

Fingerprinting using compound-specific $\delta^{13}\text{C}$ of n-alkanes indicates the relative importance of paddy fields as a temporary secondary source of watershed sediment loss

Abstract

Sediment fingerprinting generates reliable sediment provenance information which supports policy or practical strategies for catchment sediment management. But the approach remains challenging in areas with complex landscape configuration. This investigation evaluated the seasonality of biomarker signatures and their variability among particle size fractions, and accordingly apportioned target time-integrated suspended sediment to land-use based sources in an intensive farming watershed with mosaic land use patch configurations and crop-specific farming practices. Source materials (i.e., topsoil) from dry croplands, paddy fields and citrus orchards were sampled, and time-integrated suspended sediment was collected at the watershed outlet. The absolute concentrations and compound-specific $\delta^{13}\text{C}$ of long-chain saturated n-alkanes ($\text{C}_{23}\text{-C}_{33}$) were determined for two absolute particle size fractions (i.e., $<25\text{ }\mu\text{m}$ and $25\text{-}63\text{ }\mu\text{m}$). The $\delta^{13}\text{C}$ of monomeric n-alkanes displayed no significant variabilities between the particle size fractions nor across the whole sampling period. The MixSIAR Bayesian model was employed to quantify sediment source contributions. Due to human activities, paddy fields have become an important sediment source, but dry farmland remains the largest contributor. Based on sediment source information for the study watershed, a range of measures such as soil virginization, returning straw to fields, and pasture cultures in orchards are recommended.

Keywords: Source apportionment; Sediment fingerprinting; Biomarkers; Compound-specific stable isotopes; Heterogenous landscape

1. Introduction

The global soil denudation system has been severely disrupted by expanding and intensifying human activities during the Anthropocene, such as large-scale deforestation, land clearance and reclamation, farming practices, infrastructure construction, dam operations, and urbanization (Jenny et al., 2019). Consequently, global modern soil erosion patterns have been unprecedentedly complicated by constant human modification of the earth, climate change impacts, and conservation actions over recent decades (Borrelli et al., 2020; Wuepper et al., 2020). Suspended sediment arising from the mobilization of eroded materials from distal upland sources or proximal within-channel sources acts as an essential structural and functional component of river systems (Apitz, 2012). Its transport dynamics influence the extent and intensity of geomorphological evolution (e.g., aggregation or degradation) of fluvial channels, and determines the continent-ocean delivery of terrestrial materials along river continuums (Ludwig et al., 1996; Walling, 2006). More recently, growing awareness has been directed to the ecological significance of suspended sediment in rivers since fine-grained particle dispersal acts as a critical vector for redistributing contaminants (e.g., nutrients, carbon, inorganic metals, and organic compounds, etc.), which deliver complicated detrimental consequences for water quality, river ecology and aquatic ecosystem health more generally (Walling and Fang, 2003).

Acquiring reliable sediment provenance information is critical for informing appropriate soil conservation policies or measures but is difficult to obtain in view of the diffuse nature of sediment mobilization and delivery and their concomitant variability in space and time. Sediment fingerprinting provides a useful approach for quantifying the relative contributions of sampled sources to target sediments at catchment scale via establishing a direct link between the tracer signatures of sources and sediment materials (Collins and Walling, 2004). The applicability of the approach to catchments with contrasting environmental contexts has been improved through the testing and adoption of an increasing number of tracers represented by inherent soil properties, including physical properties (Krein et al., 2003; Martinez-Carreras et al., 2010; Pulley et al., 2018), fallout radionuclides (Pulley et al., 2019), mineral magnetism (Kayvantash et al., 2017; Rowntree et al., 2017), stable isotopes (Fox and Papanicolaou, 2008; Mabit et al., 2018), geochemical elements (Chen et al., 2016; Chen et al., 2019), optical tracers (Verheyen et al., 2014; Tiecher et al., 2016) and biomarkers (Blake et al., 2012; Hancock and

Revill, 2013; Chen et al., 2016; Chen et al., 2017). In some cases, the diversity of properties available has provided a basis for generating estimates of the contributions from a wide range of sources (e.g., spatial locations, erosion processes, land use categories or landscape units). Equally, the temporal coverage and resolution of the information provided by sediment fingerprinting is flexible on the basis of using a variety of targeted sediment sampling strategies including, for example, snapshot suspended sediment sampling throughout the hydrographs of individual storm events, time-integrated suspended sediment sampling, channel bed sediment remobilization or the coring of long-term floodplain deposits (Owens et al., 1999), or historical sedimentary archives preserved in lakes or reservoirs (Ben Slimane et al., 2013; Kim et al., 2013; Fang, 2015; Chen et al., 2017; Huang et al., 2019). Alongside the increasing number of case study applications, the most recent work has revisited the major assumptions underpinning sediment fingerprinting to improve the reliability of the source estimates obtained, for example, reexamining the conservative behaviour or diagnostic ability of both traditional and novel tracers by evaluating the spatial and temporal within-source variability of tracer signatures (Collins et al., 2019; Du and Walling, 2017; Reiffarth et al., 2016; Reiffarth et al., 2019), or testing the appropriateness of conventional source classification schemes and providing methodologies for refining source classification strategies (Zhou et al., 2016; Pulley et al., 2017; Vercruysse and Grabowski, 2018).

One of the most recent advances concerns increasing adoption of compound-specific stable isotope (CSSI) tracers since, in some catchments, these have the potential to generate reliable estimates of the relative contributions of land use-based sources to sampled target sediment. The basic technological premise here lies in the fact that land use or vegetation cover can imprint soil with a specific $\delta^{13}\text{C}$ signature through organic matter inputs associated with litter decomposition (Upadhayay et al., 2017). Biomarkers, including for example, long-chain saturated n-alkanes with a strong odd/even predominance derived from terrestrial higher plants, are found to be highly resistant to microbial decomposition and persist in the environment alongside sediment mobilization and storage, thereby permitting their use as tracers for apportioning sediment contributions from land use-based or plant/crop-specific sources (Blake et al., 2012; Cooper et al., 2015). Here, however, as landscape fragmentation can potentially result in greater within-source variability of tracer signatures, the diagnostic power of

biomarkers needs to be evaluated carefully (Reiffarth et al., 2019; Upadhayay et al., 2020).

A variety of fingerprint tracers has been employed according to the physical environmental context of the research catchment in question (e.g., different geologies, land uses, etc.). For example, in areas where lithology controls erosion, geochemical fingerprints are usually used in sediment fingerprinting framework (Krein et al., 2003; Fox and Papanicolaou, 2008; Martinez-Carreras et al., 2010; Verheyen et al., 2014; Tiecher et al., 2016; Rowntree et al., 2017; Mabit et al., 2018; Pulley et al., 2019); whereas in situations where geological variations are small or where different land uses span geological boundaries, biomarkers are increasingly shortlisted as potentially useful fingerprint properties (Blake et al., 2012; Hancock and Revill, 2013; Chen et al., 2016; Tiecher et al., 2016; Chen et al., 2017; Mabit et al., 2018). Alongside the increasing number of tracer properties applied in source fingerprinting studies, the number and structures of un-mixing models for estimating source proportions has also expanded. Here, models are either frequentist or Bayesian but can also differ within either category with regards the inclusion or exclusion or various correction factors and weightings (Collins et al., 1997a; Collins et al., 2017). Whereas as early studies typically used the <63 μm fraction regardless of catchment setting or tracer selection (Collins et al., 1997a; Collins et al., 2017), it is now recommended that careful attention be directed towards adoption of the most relevant size fraction for both (Collins et al., 2017). In some cases, this has resulted in the use of the <10 μm fraction (Theuring et al., 2015).

Sediment source fingerprinting in many previous cases was undertaken in homogeneous landscapes with relatively larger field sizes and simple farming practices. But, in China, most landscapes in the hilly and mountainous regions have been heavily fragmented by mosaic land use patches under the household-based farming system, where crop selection has been diversified and corresponding farming practices differ substantially. Paddy fields represent the dominant land use in south China which was traditionally viewed as a temporary sediment sink, but, from a longer-term perspective, could be a secondary sediment source resulting from the remobilization of deposited sediment. Fragmented landscapes with complex land use patterns and histories, such as those in many parts of China, potentially present a particular challenge to the capacity to identify and apply diagnostic source fingerprints (Tang et al., 2019). In the specific case of testing biomarkers in such settings, two issues requiring further investigation

concern the comparison of CSSI source apportionment using different particle size fractions and the assumption that biomarkers remain stable over the long-term (Hancock and Revill, 2013; Chen et al., 2017). To this end, this investigation examined the temporal variability of biomarker tracers closely related to land use and vegetation cover and their variability among different particle size fractions. In so doing, we also quantified the relative contributions of land-use based source units to target time-integrated suspended sediment samples in a typical hilly farming landscape. Paddy fields are a popular land use practice for rice planting in the humid region of southern China, and although these have conventionally been viewed as sediment sinks along hydrological pathways, they were assumed to be a secondary sediment source in our study.

2. Materials and methods

2.1 Site description

The Three Gorges Reservoir Area is spread across the transitional area from the Sichuan depression to middle-lower Yangtze plain (Fig. 1a). The landforms are characterized by hilly landscapes in the upper and middle reaches, while mountains and valleys prevail in the lower reaches. The climate is dominated by humid temperate monsoons, with mean annual rainfall being 1200mm. Land use is closely related to topography and human occupation. Forests and grasslands prevail in the mountainous areas where population density is low, while cropping land is predominantly in the hilly areas with high population density. Agriculture is managed by small-sized household crop farms, mainly consisting of dry croplands and paddy fields. Significant land use change has taken place due to dam construction, urbanization, infrastructure expansion, as well as a reforestation campaign, wherein arable lands have been converted to woodland or urban areas. Human relocation and water impoundment arising from the dam construction have elevated land resource scarcity, leading to the expansion of farming on marginal range lands and more intensive cultivation practices. The Three Gorges Reservoir Area has exhibited high rates of soil erosion, which seriously threaten sustainable agricultural land use, as well as aquatic ecosystem health. Soil conservation is thus a high management priority aiming to prevent the world's largest reservoir from experiencing rapid sedimentation and water quality deterioration from diffuse contaminant export associated with sediment

mobilization and delivery (Fu et al., 2010). This region is earmarked as a high priority for hillslope soil conservation, given the essential role of sediment transport in exporting diffuse contaminants, which would threaten the environmental quality and ecosystem health of the large reservoir, as well as rapid sedimentation eliminating the operational capacity of the reservoir.

This study was conducted in the Xinzheng catchment (0.25 km²) in Zhong County in the central Three Gorges Reservoir Area (Fig. 1b). The catchment landform is hilly with elevations ranging from 209 to 322 m a.s.l. The purple soil, which is developed from the Trias-Cretaceous system of sedimentary rocks and classified as Regosols in the FAO Taxonomy or Entisols in the USDA Taxonomy (He et al., 2009), is highly susceptible to rainfall detachment and runoff dispersal. Annual average rainfall is 1172 mm and a major proportion occurs in the rainy season from May to September. Catchment land use consists of dry croplands (36%), citrus orchards (35%), paddy fields (19%), rural residences (5.7%), paved roads (1.9%), water (1.7%) and exposed bare rock (0.7%) (Fig. 1c). The catchment uplands are predominantly occupied by sloping or terraced dry croplands (Fig. 2a) which are intensively cultivated throughout the year. Canola (*Brassica napus* L.) and cabbage (*Brassica oleracea* L. var. capitata L.) are key winter crops while maize (*Zea mays* L.) and sweet potato (*Ipomoea batatas* (L.) Lam.) are important summer crops. Citrus orchards situated along the middle slopes were converted from dry croplands more than 15 years ago (Fig. 2c). Citrus orchards have experienced the least intensity of land disturbance as no tillage is practiced but receive large inputs of agricultural chemicals such as pesticides, herbicides, and fertilizers, and are therefore likely to be a high-risk source for diffuse pollution generation. Paddy fields distributed on the valley floors where rice is planted during April to August (Fig. 2b) remain uncultivated and impounded with shallow water in the winter season, but undergo mechanical tillage before rice reseeding in April. There is a small pond in the upper section of the study catchment, receiving runoff from upstream (Fig. 2a).

2.2 Sampling strategies

Catchment-wide generic sediment sources were classified *a priori* as dry croplands, citrus orchards, and paddy fields. Mosaic bare rock, woodland, rural residences, paved roads, and

water were not prescribed as potential sediment sources in this study given their relatively minor importance to soil erosion and sediment supply. Each topsoil (0-2 cm) source sample was composed of 10 random subsamples collected in the immediate vicinity using a stainless-steel corer. Composite samples were collected to improve the spatial representativeness of the source material samples. Six composite topsoil samples were collected for each source category (18 source samples in total) in late March 2017 before the ploughing of the paddy fields and rice reseeded, while 14 composite topsoil samples were taken (42 source samples in total) in late April 2018 after paddy field ploughing and rice reseeded.

Time-integrated suspended sediment was collected at the catchment outlet over the period from March 2017 to September 2018 using four independent Phillips samplers (Phillips et al., 2000). All samplers were deployed in May 2017 during the first field campaign. The first sample retrieval was conducted in July 2017 when one composite sample was obtained by mixing four independent samples. For each suspended sediment sampling period, large organic detritus was removed manually, and sediment was settled in a dark cool room for 7–10 days. Excess water was siphoned off without disturbing the settled sediment. There were nine samples collected in 2017 and a further six samples in 2018 (Table 1).

2.3 Laboratory analysis

Both source and target time-integrated suspended sediment samples were freeze-dried, manually disaggregated, and sieved through a 63 μm mesh. To select the most appropriate particle size fraction for the source fingerprinting work, time-integrated suspended sediment samples were analyzed for absolute grain size with a Malvern Mastersizer laser diffraction granulometer, following pre-treatment with hydrogen peroxide and ultrasonic dispersion post the application of sodium hexametaphosphate as a dispersal agent. This analysis indicated that the d_{90} of the absolute particle size of the target suspended sediment samples ranged from 41–60 μm , with a corresponding average of 48 μm and a median of 49 μm . Because it is also informative to consider any potential effects of tracer distribution within particle size fractions of source and target sediment samples, especially where the d_{90} is reasonably coarse, all samples were separated into <25 μm and 25–63 μm fractions. This provided a basis for comparing source fingerprinting results for two absolute particle size fractions.

Sample extraction and n-alkanes fractionation were performed using the following procedure. Approximately 10 g of sample was weighed into a thimble and soxhlet extracted with 250 ml of dichloromethane (DCM): acetone (9:1, v/v) for 24 h. Before the extraction, a 100 μ l of 1 mg/ml C₁₉ and C₃₄ n-alkane internal standard was added. The total lipid extract (TLE) was evaporated to dryness using a rotatory evaporator (*Buchi*), then re-dissolved in 3 x 2 ml DCM: acetone (1:1), transferred to a vial and re-evaporated to dryness under a gentle stream of nitrogen (N₂) at 37 °C. To obtain a neutral lipid extract (NL), the TLE was dissolved in a ml of chloroform, the 0.5 ml of the extract was eluted through an activated silica column 60A, particle size 35-70, (Fisher Scientific), with 5 ml of chloroform and then blown down to dryness with N₂. To obtain the hydrocarbon fraction containing the n-alkanes, the NL was dissolved in 2 ml hexane, 0.5 ml was eluted through another silica column with 4 ml of hexane and then blown down to dryness with N₂.

The absolute abundance of C₂₃-C₃₃ n-alkanes was quantified using an Agilent 7890A GC with a flame ionization detector (FID), 7963 autosampler and splitless injector at 300 °C (Agilent Technologies). Prior to analysis, the hydrocarbon fraction was re-dissolved in 2 ml of hexane and 100 μ l transferred to a 2 ml vial. The oven programme was hold at 40 °C for 1 min, then ramp to 130 °C at 20 °C per minute, then 4 °C per minute to 300 °C and hold for 10 min. The GC column was an Agilent HP-5 (30 m x 320 μ m x 0.250 μ m) with helium carrier gas at a flow rate of 1 ml per minute. The odd n-alkanes (C₂₃-C₃₃) were initially identified by retention times and distinctive odd/even patterns. The concentrations were determined relative to the C₃₄ internal standard. The identity of the n-alkanes was confirmed using GC-MS. Here, the n-alkanes spectra were compared to those in the National Institute of Standards spectral library (NIST, US Gov.) and characteristic base peaks identified.

The compound-specific $\delta^{13}\text{C}$ signatures ($\delta^{13}\text{C}_{23}$, $\delta^{13}\text{C}_{25}$, $\delta^{13}\text{C}_{27}$, $\delta^{13}\text{C}_{29}$, $\delta^{13}\text{C}_{31}$, $\delta^{13}\text{C}_{33}$) were determined using a Finnigan Mat 6890 GC coupled to a Finnigan Mat Delta Plus IRMS via a Combustion III interface, with oxidation reactor containing platinum/copper oxide and nickel oxide at 940 °C (Thermo Fisher Scientific, Bremen, Germany). Samples were introduced using a PAL AS200 autosampler (CTC analytics, Switzerland) via splitless injector at 300 °C with helium carrier gas at 1.4 ml per minute. The oven programme was the same as above, but the column was a Varian CP-SIL 5CB (50 m x 320 μ m x 0.12 μ m) (Varian Inc. California, US). The

$\delta^{13}\text{C}$ ratio was determined relative to CO_2 reference gas of known $\delta^{13}\text{C}$ and N5.5 grade purity (BOC, Guildford, UK) previously calibrated by Iso-Analytical. The reference gas was injected directly to the source just prior to the n-alkane peaks of interest, and four times at the beginning and end of each run. The $\delta^{13}\text{C}$ was expressed relative to Vienna Pee Dee Belemnite (VPDB). During the runs, the known $\delta^{13}\text{C}$ values of the C_{19} and C_{34} internal standard (determined in house) in the samples was monitored as well as running the n-alkane standards at the beginning and end of each batch. The operation of the system was regularly checked using a certified n-alkane compound mixture $\text{C}_{16}\text{-C}_{30}$ (Indiana State University, US) with variable $\delta^{13}\text{C}$ values over a range of concentrations. The $\delta^{13}\text{C}$ should be $>1\text{‰}$ from the reference value. The stability and linearity of the system were better than 0.06‰ . The $\delta^{13}\text{C}$ standard deviation from the standards was $\pm 0.35\text{‰}$.

2.4 Sediment source fingerprinting procedure

Prior to un-mixing source contributions to the target time-integrated suspended sediment samples, tracer conservation was assessed using a bracket test. Selected conservative tracers were then tested individually for their source discrimination efficiency using the Kruskal-Wallis rank sum test (Collins et al., 1997b). Here, any properties returning a statistical significance of $P > 0.05$ were excluded from further analysis. Properties passing the Kruskal-Wallis test were then used in a stepwise discriminant function analysis (DFA) to identify a composite set of signatures for un-mixing modelling (Collins et al., 1997c). The estimation of the relative contribution of each potential sediment source to the target time-integrated suspended sediment samples was assessed with the concentration-dependent MixSIAR model (Upadhyay et al., 2018a). This model has been widely used for un-mixing compound-specific stable isotopes (CSSI) of odd long-chain n-alkanes ($\text{C}_{23}\text{-C}_{33}$). MixSIAR was implemented in the R package linked to the JAGS 3.3.0 library for Bayesian data analysis using a Gibbs sampling Markov chain Monte Carlo (MCMC) algorithm (Plummer, 2003) and a fractionation factor of zero was used. The Markov Chain Monte Carlo (MCMC) parameters in MixSIAR were set as follows: number of chains = 3, chain length = 5,000,000 (extreme), burn = 1,500,000, thin = 500. The convergence of un-mixing models was evaluated using the Gelman-Rubin diagnostic, rejecting the model output if any variable was above 1, in which case the chain length was

increased. Furthermore, a diagnostic matrix plot of posterior source contributions was used to evaluate the quality of source discrimination. Density plots of the proportional source contributions are reported along with a mean, median and standard deviation. MixSIAR simulations were run by considering different levels of prior informativeness defined in this research as the sum of the prior input into the model. The input priors ranged from low informative priors in which the sum of the priors input was 1, medium informative priors in which the weight of the prior was set as the number of sources, to very informative priors in which the sum of the priors correspond to 100 (Upadhayay et al., 2018b).

3. Results

3.1 The behaviour of biomarker signatures among source categories and particle size fractions

Source materials from paddy fields exhibited the highest n-alkane concentrations of 4863.8 ± 1116.2 ng/g in the <25 μm fraction and 5855.9 ± 2294.3 ng/g in the 25-63 μm fraction, followed by those from citrus orchards, with 5513.5 ± 3137.8 ng/g in the <25 μm fraction and 3999.8 ± 1241.4 ng/g in the 25-63 μm fraction. Dry croplands exhibited the lowest n-alkane concentrations of 2122.5 ± 793.6 ng/g in the 0-25 μm and 2447.5 ± 1506.3 ng/g in the 25-63 μm fraction. N-alkane concentrations were higher in the finer particle size fractions in the soil samples collected to characterize citrus orchards. However, in the case of samples collected to characterize the signatures of dryland farming and paddy fields, this trend was not exhibited (Fig. 3). All source soil samples showed obvious odd-number n-alkane predominance, wherein the dominant peaks were C_{31} , C_{33} and C_{29} in orange orchard samples, C_{27} , C_{31} and C_{29} in paddy field samples, and C_{29} , C_{31} and C_{33} in dry farmland samples. These three n-alkanes accounted for more than 60% of the total n-alkanes. Differences in the $\delta^{13}\text{C}$ of n-alkanes among the different sediment source categories were not obvious (Fig. 3). T tests was employed to examine the presence of any differences between the $\delta^{13}\text{C}$ of n-alkanes or n-alkane content of the two particle size fractions for each source category. These tests revealed some statistically significant differences for the paddy field and orange orchard sediment sources. On this basis, the <25 μm fraction was used to fingerprint sediment sources to minimize any particle size effects. Molecular biomarkers are potentially influenced by the conversion of organic matter and farming practices between years. We therefore tested for statistically significant differences

between the $\delta^{13}\text{C}$ of n-alkanes for the two sampling years (Table 2). The results indicated that there were no significant differences and, accordingly, we considered it appropriate to use the soil samples collected from each source in both 2017 and 2018 when estimating the proportional contributions to the time-integrated suspended sediment samples.

3.2 Tracer selection and sediment source discrimination

All six fingerprint properties taken forward in the data processing procedure returned a P value of <0.01 from the Kruskal-Wallis H test, with the H-values ranging from 9.59 to 33.52 (Table 3). Individually, the $\delta^{13}\text{C}$ of all odd-numbered n-alkanes returned statistically significant differences among the three potential sediment sources (P-value < 0.01 ; Table 4). DFA indicated that the discriminatory efficiency of the individual tracers ranged between 51.7% and 65.0%. The $\delta^{13}\text{C}_{25}$ was the only individual fingerprint property excluded from the final composite signature which yielded a discriminatory efficiency of 90% (Table 4). On this basis, a final composite signature combining $\delta^{13}\text{C}_{33}$, $\delta^{13}\text{C}_{23}$, $\delta^{13}\text{C}_{29}$, $\delta^{13}\text{C}_{31}$ and $\delta^{13}\text{C}_{27}$ was used in MixSIAR to estimate the sources of the time-integrated suspended sediment samples collected in the study catchment.

3.3 Source contributions to watershed sediment loss

The above results showed that the particle size and organic matter contents have little effect on the carbon isotope contents. The results indicated that for the 12 sediment samples, dry farmland contributed the most ($47.4 \pm 16.5\%$), followed by paddy fields and orange orchards, which contributed $32.4 \pm 14.7\%$ and orchards $20.2 \pm 3.7\%$, respectively. There are slight differences in the source contributions between the sampling years, but sloping dry farmland was consistently the most important source of suspended sediment. We analyzed the errors associated with the estimated contributions from each source of sediment and found that the standard deviation varied from 12.7%- 24.9%. Here the standard deviation for orchard contributions varied from 12.7%-20%, for paddy fields from 14.3%-24.9%, and for sloping dry farmland from 16.8%-23.6% (Fig. 4).

In 2017, dry farmland contributed 43.4% of the sampled suspended sediment, whereas paddy fields and orange orchards contributed 36.3% and 20.3%, respectively. In 2018, dry

farmland contributed 53.0% of sampled sediment, whilst paddy fields and orange orchards contributed 27.0% and 20.0%, respectively. The rice in this area is harvested in late August, and the corresponding contributions of sediment differed before and after the rice harvest. Before the rice harvest, the contribution of the paddy fields was 11.7% higher (37.3%) than that after the harvest (25.6%). After the rice harvest, the sediment contribution from the dry farmlands increased by 41.6% and 55.7%, respectively. During the study period, the sediment contributions from the orange orchards did not change significantly, remaining at ~20%. However, in August when the rice was harvested, paddy fields were the main source of sediment in the study catchment (51.7%), whilst sloping fields and orchards contributed 26.0 and 22.3%, respectively. Overall, the results showed a specific pattern: August and October were the peak periods of sediment contributions from paddy fields; September, April and July were the peak periods of sediment delivery from dry farmland, while the contributions from orchards remained comparatively stable throughout the year.

4. Discussion

4.1 Factors affecting the distribution and contents of biomarkers

OA waxy layer often develops on the surface of terrestrial higher plant stems, fruits, petals and leaves to protect plant cells from ultraviolet rays, fungi and insects and maintain water balance. Among them, the epidermal waxy layer of plant leaves has the highest content, and the composition of plant leaf wax is very complex, including long-chain alkanes, alcohols, ketones, acids, lipids and acids ([Eglinton and Hamilton, 1967](#)). Plants synthesize organic matter through photosynthesis by absorbing water and CO₂ in the surrounding environment, and complex isotope fractionation occurs in the process of biosynthesis. Plants perform photosynthesis by absorbing CO₂ in the atmosphere and soil, and there are obvious differences in the carbon isotope of organic matter produced by plants with different photosynthetic pathways ([Tippie and Pagani, 2007](#)).

Isotope fractionation has nothing to do with organic matter content but only with species and environments. The individual carbon isotope ratios of n-alkanes have been widely used to identify vegetation with different photosynthetic pathways ([Sun et al., 2016](#)), including terrestrial higher plants due to photosynthetic pathways (C₃/C₄). The distribution range of the monomer

carbon isotope values of different n-alkanes produced is obviously different. The single carbon isotope ratio of n-alkanes is used the most widely to indicate the change in the relative abundance of C₃/C₄ vegetation in a study area.

The single carbon isotope of n-alkanes can also be used to explore the environmental conditions of plant growth, such as temperature and humidity changes. Previous studies have shown that the wax carbon isotope values of C₃ plants are related to the water supply during photosynthesis (Kohn, 2010; Suh and Diefendorf, 2018; Wang et al., 2018); and, the correlation between carbon isotope changes of C₄ plants and water is not obvious. In our small study catchment, orange and rice are C₃ vegetation, while corn is C₄ vegetation. Orange and rice differ in their n-alkane monomer C isotopes due to the differences in species. In previous research, it has been reported that the monomer carbon isotope content of C₄ vegetation is higher than that of C₃ vegetation (Chikaraishi and Naraoka, 2003; Meyers, 2003; Bi et al., 2005). Our results also proved this, since dry farmland has the largest carbon isotope ratio, followed by the paddy fields, and orange orchards.

The compound-specific $\delta^{13}\text{C}$ of n-alkanes exhibited no significant difference between the two particle size fractions and across the whole sampling period, which showed that the monomer carbon isotope of n-alkanes is a good sediment fingerprint. There are several possible explanations for the difference in n-alkane contents between different particle size fractions. Most of the organic carbon is stored in microaggregates (Wang et al., 2017), but the forms of organic matter stored between different particle size fractions are different. The organic components of > 20 μm mainly consist of plant detritus, and the decomposition of these components is slower. Additionally, the morphology of these components is easier to identify in the coarse component than in the fine component; however, the organic constituents in the 20 μm component are mainly sapropelized substances with long transformation times (Feller and Beare, 1997; Baldock and Skjemstad, 2000). In addition, fine sediment has a larger specific surface area and can absorb more organic matter. Because of high intensity tillage, land cover and other reasons, dry farmland is more prone to rill erosion, and the particle size selectivity of rill erosion is poor, meaning that both coarse and fine particles will be transported along with runoff (Shi et al., 2012; Wang et al., 2017). Furthermore, different soil erosion processes can cause significant differences in the concentration of organic carbon in sediments by

redistributing sediment particles of different sizes (Nadeu et al., 2011). In our study catchment, soil erosion selectively delivers finer particles of sediment (0-25 μm) to the channel and this selectivity could reflect the intensity of soil erosion. In addition, fine particles may be easier to transport in the paddy fields, while coarser particles mobilized from upstream may be more prone to deposition. However, these results do not prove that individual biomarkers cannot be used as fingerprints, because this study did not analyze their contents in sediment of each particle size fraction, so there may be no significant difference between them overall. In a region with a relatively stable farming system and land use, biomarkers can still be considered as a reliable fingerprinting tracer (Hancock and Revill, 2013; Chen et al., 2016; Chen et al., 2017; Galoski et al., 2019).

4.2 Land use effects on sediment source contributions

Our findings do not conflict with conventional knowledge that dry farmlands can often be a principal sediment source. But our results do highlight the dual roles that paddy fields act either as a temporary sediment sink along the delivery continuum or as a secondary proximal suspended sediment source. Paddy fields therefore both trap freshly eroded sediment carried by storm runoff flowing through these part of the catchment, whilst also releasing sediment which is transported to the catchment outlet. By way of comparison, in the Wangjiagou catchment, Fuling, paddy field high ridges are sufficient to intercept excess runoff, reduce runoff speed, and reduce the direct impact of raindrops on the field surface soil (Chen et al., 2019). However, in our investigation, we found that from the end of March to the beginning of May, the study catchment is characterized by rice planting. In the initial stage, the paddy fields would be subjected to high-intensity mechanical cultivation, and coarse particles are broken into finer particles. Since the water storage height is minimal, paddy soils are prone to splash erosion after ploughing. When there is runoff, there is a propensity for fine particles of sediment to be mobilized from the paddy fields (Fig.2). Therefore, in April, paddy fields contributed a large proportion of sediment, as shown in Fig. 4. Any interannual variability in sediment contributions rates may be due to the spatial heterogeneity of rainfall and resultant erosion (Renschler et al., 1999; Renschler and Harbor, 2002).

Primary sediment sources are site-specific and vary occasionally with catchment size and

sediment connectivity determined by catchment properties. In the Three Gorges Reservoir area, rice is generally harvested at the end of August. Due to technological innovation, rice is harvested by machines in mountainous areas. Even small machines cause greater damage to the surface than traditional farming operations. Coupled with the recycling of rice stalks, the ground is exposed due to the lack of mulch and erosion therefore increases. In the same way, dry farmlands, where maize is harvested earlier than rice, contributed the most sediment to the study catchment outlet due to the lack of plant cover and the steeper slopes (Chen et al., 2019). Citrus orchards also contributed to suspended sediment loss. As labour costs increase, farmers generally spray herbicides on the ground to kill ground weeds in the orchards to reduce nutrient competition. Despite this, the orchard trees still provide some surface cover, which reduces the splash erosion by raindrops and sediment transport by surface runoff to a certain extent. Here, the retention of the citrus canopy reduces the kinetic energy of raindrops, which is also an important factor in reducing sediment mobilization from the citrus orchards.

Monthly sediment contributions from the three sources were also different, which is related to seasonal rainfall, plant growth cycles and human activities. For example, the rice grows well in July, and the dry farmland has already been harvested, meaning that dry farmland soils contributed the most in July, followed by September and October, due to obvious human activities. In April, the plum rain season has just started, and the early rainfall intensity is not high, meaning that only the finest grain size fraction of sediment is transported. When September comes, the rape planting season begins, and dry farmlands and paddy fields are ploughed to sow rape. Bare topsoil, seedlings and mechanical tilling provide ideal conditions for soil erosion in the study catchment at this time. Fortunately, after September, the Three Gorges Reservoir area enters the dry season, and the rain intensity and rainfall amounts are not as severe as the rainy season.

4.3 Informative options for soil conservation and sediment management

Since the launch of the Three Gorges Dam, increasing attention has been directed to environmental challenges associated with intensive crop farming on cultivated hillslopes of the surrounding landscapes in the Three Gorges Reservoir Area. Here, the short-distance connectivity for inflowing sediment and associated diffuse pollutants poses a large threat to

water quality and ecosystem health of the receiving waters impounded by the reservoir (Fu et al., 2010; Zhang and Lou, 2011; Li et al., 2013b; Xu et al., 2013). Farming has been extended to marginal hillslopes and the cultivation intensity has been intensified over the post-dam period. The sustainability of the local community therefore requires precision land use planning and refined conservation schemes to temper the conflicts between ecosystem conservation and rural livelihood improvement when limited land resource is available. According to the 2019 Yangtze River Sediment Bulletin and Chinese national dynamic monitoring of soil erosion, since the reservoir was put into operation in June 2003, the reservoir has accumulated 1.83 billion tons of silt. In 2018, the area of soil erosion in the Three Gorges Reservoir area was 19,200 km², accounting for 33.34% of land. Eutrophication has been a serious problem, and the tributaries are frequently subjected to algal blooms. The statistical evaluation of 649 groups of data from 28 tributaries from 2004 to 2007 shows that the proportion of tributaries suffering light eutrophication is 75% (Wang and Qian, 2009). One of the most important reasons for this is that the sediment in the agricultural catchment surrounding the reservoir area is directly delivered into the Yangtze River.

Small catchments are the basic unit of sediment yield, sediment transport and soil erosion control. Carrying out comprehensive management of water and soil erosion in small watersheds is the key to reducing the inflow of sediment, improving the ecological environment, and maximising reservoir life span. In order to reduce soil erosion in agricultural catchments, the farming system should be optimized based on increasing the implementation of water and soil conservation measures, such as the promotion of soil virginization technology and changing deep tillage to shallow tillage. This can reduce the artificial fragmentation of soil aggregates. Soils with many water-stable aggregates are more resistant to the formation of soil crusts, have good infiltration, and have less runoff and erosion (Fox et al., 2004). Especially in the case of paddy fields, the implementation of paddy field soil virginization, which can change the micro-topography, thicken the active soil layer for roots and make the capillary pores of the soil connect normally to improve water and air flow represents an important management intervention.

In addition, the strategy of returning straw to the field should be promoted. Returning straw to fields can effectively promote microbial activity which degrades the organic components in

the straw, increase soil organic matter, optimize the physical structure of the soil, and improve soil fertility (Liu et al., 2014; Hamm et al., 2016; Li et al., 2019). At the same time, straw mulch can intercept rainfall, prevent water volatilization, and form a natural farmland reservoir. Covering the ground with straw can also effectively reduce rainfall kinetic energy, intercept runoff, and greatly reduce soil erosion (Shi et al., 2013). The technology of returning straw to the fields can also greatly increase soil carbon sequestration and reduce greenhouse gases emitted into the atmosphere

For the orange orchards, the method of pasture culture is recommended (Li et al., 2013a). The specific measures of this intervention are to interplant *Leguminosae* or *Poaceae* between the rows of orchards or beneath orchards in their entirety, cut them regularly, cover the underside of the trees with the cut stalks, and allow them to decay and decompose naturally (Richardson, 1986). These actions are designed to improve soil structure and soil fertility and reduce soil erosion. Studies on the Three Gorges Reservoir area show that vetch (*vicia villosa* Roth var.) is an ideal understory vegetation (Wen et al., 2011; Han et al., 2014). In conjunction with this intervention, herbicides are applied in orchards, which protects the environment. At the same time, as a legume, it can not only fix nitrogen, but also increase soil organic matter after decay (Yang et al., 2020).

4.4 Limitations of this study and future research directions

At present, the mainstream biomarkers used in source fingerprinting are mainly fatty acids and n-alkanes (Cooper et al., 2015; Gibbs, 2008; Reiffarth et al., 2016; Mabit et al., 2018; Upadhayay et al., 2018b; Reiffarth et al., 2019). In contrast to fatty acids, n-alkanes are not easily decomposed. Therefore, the use of n-alkane concentrations to track sediment sources has been reported in the literature (Chen et al., 2016; Chen et al., 2017). Monomer carbon isotopes of fatty acids and n-alkanes have been widely used. Decomposition of fatty acids and n-alkanes affects the concentration of total organic matter, but the $\delta^{13}\text{C}$ isotopic signature of that compound does not change. Isotopes of short-chain fatty acids, which dissolve in water and migrate with water, were used first (Gibbs, 2008); these biomarkers (C16:0 and C18:0) have applicability to areas with deep soil erosion (C16:0 and C18:0). Tunnel erosion exists in the Loess Plateau, and, therefore, short-chain fatty acids may be a new way to monitor

cavitation erosion (Zhu, 2012). Longer-chain fatty acids and n-alkanes are insoluble in water and do not migrate downward with water, so they are more suitable for most areas of severe surface erosion (Reiffarth et al., 2016; Mabit et al., 2018; Upadhyay et al., 2018b; Reiffarth et al., 2019).

This study only analyzed the C isotope of n-alkanes. Future work could also include $\delta^2\text{D}$ isotopes. If fatty acid is used as the fingerprint, the three CHO isotopes can also be tested. Most fingerprint models have an optimal solution, but the solution set interval of these methods is frequently too small. An optimal solution or a very approximate solution may therefore only represent one possibility for the sediment transport processes of a catchment (Collins et al., 1997a; Collins et al., 2010; Chen et al., 2016; Chen et al., 2017;). In practice, due to the complex interplay of surface soils, land use, rainfall, and hydrology, the same magnitude of sediment loss can be produced by a variety of source contributions, which is the 'black box' model of soil erosion (Borrelli et al., 2021). The MixSIAR model makes the solution set space a normal distribution and considers the likelihood of extreme sediment source contributions. In this case, the sediment source contributions are not restricted which is more consistent with the real situation (Parnell et al., 2010; Xue et al., 2009; Garzon-Garcia et al., 2017). The absolute contributions from the study catchment sources, including paddy fields, were not calculated due to the lack of suspended sediment load monitoring data.

5. Conclusion

Exploring the applicability of sediment source fingerprinting in complex landscapes represents a recently identified research priority. Our investigation therefore examined the behaviour of biomarkers (absolute abundance and compound-specific stable isotope ratio of n-alkanes) and their variability among different particle size fractions, and apportioned target time-integrated suspended sediment collected at the study catchment outlet to upstream land-use based sources. The seasonality of the absolute abundance of n-alkanes (C_{23-33}) in surface soils, especially in paddy fields, was revealed, which is likely to be related to farming practices, plant growth dynamics and the redox environment. The timing of sampling should be carefully considered when using organic tracers. Repeated sampling could be undertaken as an option to overcome the temporal representativeness of the samples collected. However, the

applicability of compound-specific $\delta^{13}\text{C}$ as a conservative tracer to apportion sediment contributions to catchment land-use based sources has been confirmed, as no variability among seasons and particle size fractions was observed. Our fingerprinting with the compound-specific $\delta^{13}\text{C}$ of n-alkanes highlights the relative importance of paddy fields as a temporary secondary sediment source, especially during the early Spring when ploughing and rice planting are usually performed. Paddy fields were traditionally assumed to be sediment sinks given the inflow of turbid runoff for irrigation and a major proportion of suspended sediment is likely to be trapped by plant stems.

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Table 1. Summary of the sampling scheme for potential sources and time-integrated target suspended sediment samples collected using Phillips traps over a period spanning 2017-2018.

Catchment-wide potential sediment sources			Target suspended sediment samples	Sampling time
Orange orchards	Dry croplands	Paddy fields		
6	6	6		Mar 2017 ^a
			1	Jul 2017 ^b
			4	Aug 2017
			4	Oct 2017
14	14	14	4	Apr 2018 ^c
			2	Sep 2018

^a Source samples from paddy fields were taken before ploughing and rice reseeded and the sediment samplers were installed during this field trip.

^b One composite sediment sample was obtained by mixing sediments collected from four independent samplers.

^c Source samples from paddy fields were collected after ploughing and rice reseeded.

Table 2. Fingerprint property T tests for the $\delta^{13}\text{C}$ n-alkane and n-alkane values measured in the two different absolute particle size fractions (<25 μm and 25-63 μm) of each sediment source category for each sampling year.

Sampling year	Sediment sources	Independent-samples T test	$\delta^{13}\text{ n-alkanes}$				n-alkanes												
			$\delta^{13}\text{C}_{23}$	$\delta^{13}\text{C}_{25}$	$\delta^{13}\text{C}_{27}$	$\delta^{13}\text{C}_{29}$	$\delta^{13}\text{C}_{31}$	$\delta^{13}\text{C}_{33}$	C_{23}	C_{24}	C_{25}	C_{26}	C_{27}	C_{28}	C_{29}	C_{30}	C_{31}	C_{32}	C_{33}
2017	Orange orchards	F-value	1.23	0.12	0.04	0.16	0.12	1.56	3.12	0.31	0.03	3.87	0.69	0.09	0.10	0.10	0.05	1.46	0.20
		P-value	0.54	0.25	0.85	0.99	0.33	0.83	0.78	0.46	0.67	0.63	0.98	0.80	0.72	0.74	0.81	0.95	0.58
	Paddy fields	F-value	0.16	3.81	0.00	2.67	0.58	0.90	1.57	1.89	1.48	5.49	0.09	1.02	0.67	2.16	6.47	0.78	0.32
		P-value	0.19	0.51	0.41	0.67	0.57	0.49	0.35	0.40	0.99	0.05*	0.16	0.40	0.76	0.63	0.42	0.84	0.51
	Dry farmland	F-value	1.32	0.55	1.73	0.03	0.01	0.00	0.60	11.25	0.01	4.97	0.30	1.28	0.91	3.36	0.70	0.00	0.32
		P-value	0.41	0.58	0.79	0.69	0.27	0.54	0.94	0.71	0.53	0.54	0.53	0.71	0.76	0.62	0.75	0.78	0.64
2018	Orange orchards	F-value	2.22	2.47	1.41	1.30	0.25	2.01	14.05	20.96	12.15	15.18	3.88	12.95	3.18	11.82	1.20	1.08	0.34
		P-value	0.96	0.33	0.30	0.22	0.24	0.31	0.01**	0.00**	0.01**	0.00**	0.82	0.00**	0.14	0.01**	0.25	0.17	0.47
	Paddy fields	F-value	9.36	3.34	0.64	0.01	1.16	0.92	8.52	9.64	6.36	3.27	6.21	3.52	8.75	12.42	6.36	2.18	2.93
		P-value	0.07	0.06	0.06	0.17	0.39	0.08	0.22	0.31	0.22	0.03*	0.09	0.23	0.21	0.17	0.22	0.30	0.11
	Dry farmland	F-value	0.02	1.14	0.53	0.08	0.35	0.01	5.81	6.24	4.36	5.23	2.56	7.94	1.41	3.91	2.97	1.90	0.08
		P-value	0.97	0.56	0.86	0.75	0.99	0.92	0.25	0.18	0.36	0.23	0.33	0.13	0.46	0.98	0.38	0.50	0.80

* Significant difference at the 95% credible level.

** Significant difference at the 99% credible level.

Table 3. T tests for differences between the $\delta^{13}\text{C}$ n-alkane signatures (<25 μm fraction) of each sediment source category collected in 2017 and 2018.

Land use types	Independent-samples T test	Fingerprint property					
		$\delta^{13}\text{C}_{23}$	$\delta^{13}\text{C}_{25}$	$\text{C}^{13}\text{C}_{27}$	$\delta^{13}\text{C}_{29}$	$\delta^{13}\text{C}_{31}$	$\delta^{13}\text{C}_{33}$
Orange orchards	F-value	0.569	0.529	0.213	0.192	0.062	0.579
	P-value	0.108	0.201	0.256	0.364	0.055	0.066
Paddy fields	F-value	0.613	0.564	1.418	3.046	0.612	0.019
	P-value	0.241	0.063	0.064	0.133	0.068	0.197
Dry farmland	F-value	13.100	26.298	9.151	1.033	0.081	0.321
	P-value	0.322	0.987	0.348	0.061	0.405	0.277

Table 4. Kruskal–Wallis one-way analysis of variance and discrimination function analysis (DFA) results for sediment source discrimination.

Fingerprint property	Kruskal Wallis		Minimization of Wilks' lambda				
	H-value	P-value	Step	Wilks' lambda	F-value	% source samples classified correctly using individual fingerprint properties	Cumulative % of source samples classified correctly
$\delta^{13}\text{C}_{33}$	31.91	0.000**	1	0.419	39.57	63.3	63.3
$\delta^{13}\text{C}_{23}$	33.52	0.000**	2	0.158	42.47	53.3	83.3
$\delta^{13}\text{C}_{29}$	9.59	0.008**	3	0.102	39.05	51.7	88.3
$\delta^{13}\text{C}_{31}$	33.12	0.000**	4	0.089	31.70	65.0	90.0
$\delta^{13}\text{C}_{27}$	17.31	0.000**	5	0.075	28.08	51.7	90.0
$\delta^{13}\text{C}_{25}$	25.22	0.000**	-	-	-	-	-

** Significant difference at the 99% credible level.

“-“ indicates the fingerprinting property was excluded from the final composite signature.

Figure captions

Fig. 1. a. Map showing the spatial extent of the Three Gorges Reservoir extending along the upper Yangtze River and the topography of its surrounding landscapes; b. Remote sensing image generated by a drone flying over the study catchment in the middle reaches that directly drains to the reservoir, and; c. Land use in the study catchment.

Fig. 2. Images of the study landscape. a. dry croplands; b. paddy fields at the end of April, showing a low water level meaning that the soil disturbed by mechanical ploughing is prone to splash erosion; c. orange orchards converted from dry croplands for over 15 years, and; d. time-integrated suspended sediment sampler installed near the study catchment outlet.

Fig. 3. The biomarker results for the different absolute particle size fractions of the potential sediment sources.

Fig. 4. Sediment source contributions during 2017 and 2018. The error bars indicate the standard deviations (SD).

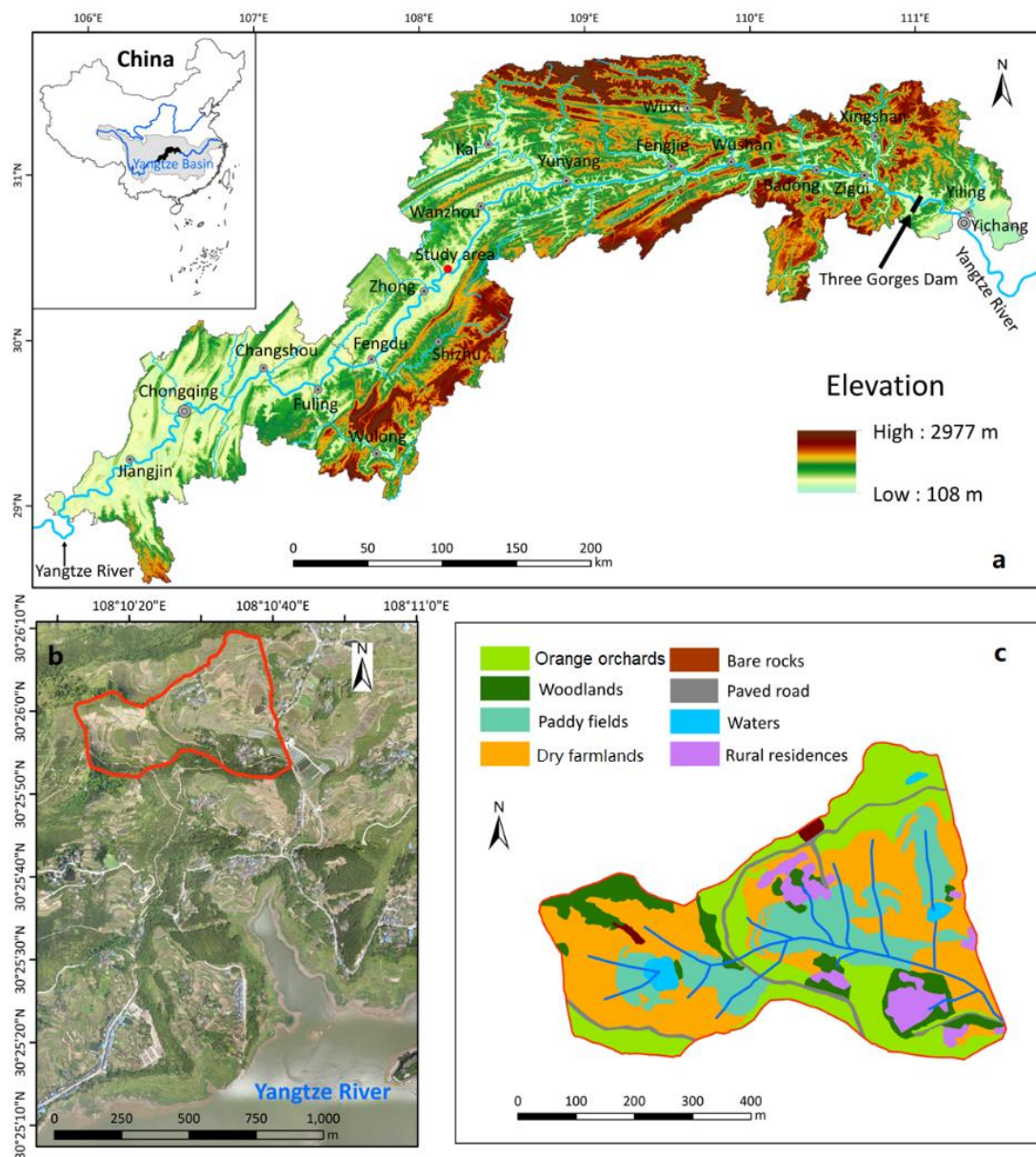


Figure 1



Figure 2

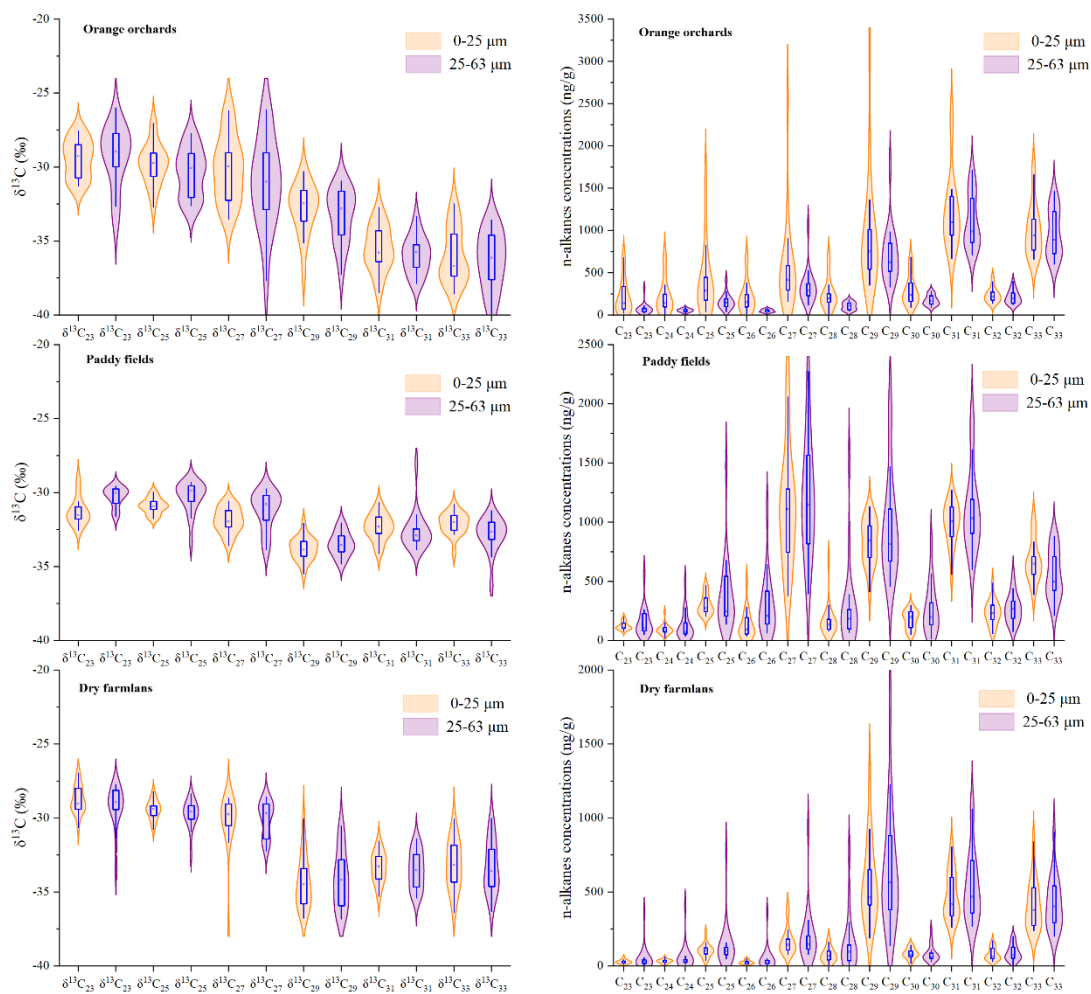


Figure 3

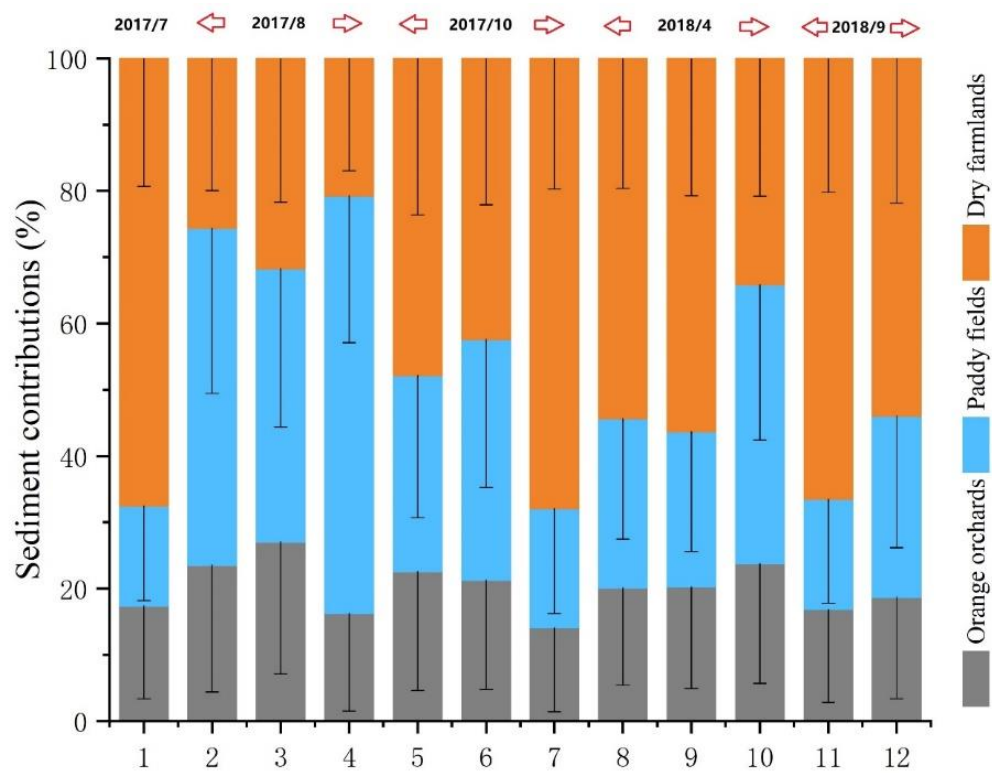


Figure 4

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