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Reconstructing the Changes in Sedimentation and Source Provenance in East African Hydropower Reservoirs: A Case Study of Nyumba ya Mungu in Tanzania

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Abstract: This study aimed to reconstruct the sedimentation rates over time and identify the changing sources of sediment in a major hydropower reservoir in Tanzania, the Nyumba ya Mungu (NYM). Fallout 210Pb measurements were used to estimate age of sediment deposits and broad changes in sedimentation rates were reconstructed. Sedimentation peaks were cross referenced to geochemical profiles of allogenic and autogenic elemental constituents of the sediment column to confirm a causal link. Finally, geochemical fingerprinting of the sediment cores and potential sources were compared using a Bayesian mixing model (MixSIAR) to attribute the dominant riverine and land use sources to the reservoir together with changes through recent decades. Reservoir sedimentation generally increased from 0.1 g cm−2 yr−1 in the lower sediment column to 1.7 g cm−2 yr−1 in the most recent deposits. These results correlated to changes in allogenic and autogenic tracers. The model output pointed to one of two major tributaries, the Kikuletwa River with 60.3%, as the dominant source of sediment to the entire reservoir, while the other tributary, Ruvu River, contributed approximately 39.7%. However, downcore unmixing results indicated that the latest increases in sedimentation seem to be mainly driven by an increased contribution from the Ruvu River. Cultivated land (CU) was shown to be the main land use source of riverine sediment, accounting for 38.4% and 44.6% in Kikuletwa and Ruvu rivers respectively. This study explicitly demonstrated that the integration of sediment tracing and dating tools can be used for quantifying the dominant source of sediment infilling in East African hydropower reservoirs. The results underscore the necessity for catchment-wide management plans that target the reduction of both hillslope erosion reduction and the sediment connectivity from hillslope source areas to rivers and reservoirs, which will help to maintain and enhance food, water and energy security in Eastern Africa.

Keywords: sediment; geochemical fingerprinting; sedimentation; tracers; mixing model; dating

1. Introduction

Hydropower reservoirs support both water and energy security, providing essential services required for development in East Africa [1–3]. They are essential for the production of climate neutral energy [3] that guarantees the long-term energy security for economic development in developing countries. Moreover, reservoirs in East Africa provide a source of irrigation and drinking water sources for agriculture purposes and aquatic habitats that are beneficial for fisheries and recreation [4,5]. Hydropower reservoirs, however, are increasingly threatened by changing hydrological and sediment dynamics in their wider
catchments [6,7]. The driving processes behind the observed increases in sediment delivery in the reservoirs are land use changes, climate change and natural climate variability [8], resulting in accelerating reservoir sedimentation. Subsequently, the water storage capacity is decreasing and energy production capacity also deteriorating [9]. Although dams reduce floods, siltation can cause flooding that may lead to loss of life and property, disruption of the local infrastructure and involuntary resettlement [10,11]. Additionally, siltation of reservoirs can also trigger eutrophication of surface water bodies through release of associated nutrients but also limits sediment transfer to the downstream reaches, thus leading to loss of biodiversity.

The changing demographics of East Africa create increased demand for land, food and water, leading to changes in land and water use [10–12]. These land use changes increase the vulnerability of soils to the impacts of hydroclimatic drivers of erosion by water [13]. An increased rate of soil erosion is one of the main drivers of land degradation in East Africa’s agropastoral landscapes [14], which often severe impacts on community resilience and socioeconomic development [15,16]. Furthermore, changes in precipitation under global climate change also alter the dynamics of river flow and discharge [17,18]. While some exploratory work has been performed on the potential impacts of climate change and land use in East African hydropower reservoirs [3,19–22] and the estimations of the sedimentation rate [19,23,24], there is little quantitative evidence available on the changing sedimentation rates and sediment sources contributing to the infilling of reservoirs. In this context, some pressing questions remain regarding hydropower management now and in the future. What are the major potential sources of sediments contributing to the infilling of the reservoir? What are the best techniques to assess reservoir sedimentation rates? What approaches can reduce the quantity of sediment incoming to the reservoirs from upstream? How might future climate change affect sediment supply? Radiometric dating using fallout radionuclides (FRNs) and geochemical fingerprinting (sediment tracing technique) can fill in this caveat in empirical data. These empirical models can subsequently be coupled with information on changing land use and climate to make meaningful deductions on the driving processes of increased sediment delivery. The present study therefore will reconstruct the sedimentation rate over time in the Nyumba ya Mungu reservoir (NYM) using fallout radionuclide (FRN)-based nuclear techniques (\(^{137}\)Cs and \(^{210}\)Pb dating techniques) and quantify the main riverine and land use sources to the sediment using geochemical fingerprinting.

Reconstructing changes in rates of reservoir sedimentation since dam closure is crucial for evaluating the extent of siltation problems and therefore the durability of hydropower reservoirs. Sediment tracing techniques evaluate the similarities and dissimilarities between the physical or chemical traits of downstream sediments and the catchment potential sediment sources [25–27]. Consequently the geochemical composition of downstream reservoir sediments depends on the relative contributions and geochemical properties of different tributaries [28–30]. The proportional attribution of the tributary sources to downstream sediment can therefore be obtained through integration of the multivariate source and mixture geochemical fingerprints within mixing models [30–33]. This is, however, only possible when the eroded soils of the fine particle fraction (e.g., particle size < 63 µm) transported from different watershed areas behaves conservatively from detachment, transport, deposition and after deposition.

The FRNs, \(^{210}\)Pb and \(^{137}\)Cs, can be employed to study erosional records of a catchment via dating of downstream sediment accumulation (e.g., in a reservoir) and, therefore, the effects of land use and climate by presenting data from different time windows during the last 100–150 years [34]. The fundamental ability of \(^{210}\)Pb to provide evidence on the chronology of a sediment deposit and thus estimate the sedimentation rate depends on its source in relation to local geology, its moderately long half-life, and its retrospective assessment that provides a longer-term (ca. 100 year) chronology or age–depth relationship [35]. The anthropogenic radionuclide, \(^{137}\)Cs, is from weapon testing fallout that peaked in early 1960s. However, its fallout in tropical Africa was low, challenging its application [36].
is a natural geogenic radionuclide—its deposition is continuous and basically constant from year-to-year [36]. Generally, the rate of decrease of $^{210}\text{Pb}_{\text{ex}}$ activity with depth in a sediment core offers the foundation for developing an age–depth correlation and for estimating sediment accumulation rates (SAR) [37]. From its activity profile, it is feasible to determine the sedimentation rate and in some conditions to reconstruct environmental changes [36] through time using numerous models comprising a variety of assumptions [36,38–40].

2. Materials and Methods

2.1. Description of the Study Area

NYM is a constructed reservoir that largely replaced a natural wetland when the river was impounded. The NYM catchment includes the highlands of Africa’s highest peak, Mt. Kilimanjaro (5985 m), and fifth highest peak, Mt. Meru (4566 m). The reservoir is part of the upper Pangani River Basin (PRB) and receives water from two main tributaries, the Kikuletwa and the Ruvu rivers (Figure 1). The Kikuletwa subcatchment covers about 6650 km$^2$ out of 12,000 km$^2$ and the Ruvu approximately 5350 km$^2$ of the total catchment. The dam was erected in 1965 for hydropower generation but in later years, irrigation potential was realized and integrated into strategies [23,41,42]. The NYM reservoir is about 150 km$^2$, has a live-storage capacity of about 875 million m$^3$ [43] and has a maximum depth of 40 m [44,45]. However, due to highly variable climate and changing sedimentation, these factors fluctuate seasonally and between years.

The catchment of the NYM reservoir is located between latitudes 3°00′00″ and 4°3′50″ south, and longitudes 36°20′00″ and 38°00′00″ east, and its altitude ranges between 700 and 5825 m.a.s.l. The ice cap at the peak of Mount Kilimanjaro forms the highest ground in the catchment. The catchment occupies a total land and water area of about 13,000 km$^2$ [46] and experiences a tropical climate that provides high levels of precipitation with average annual rainfall (AAR) of 900–2200 mm/year at 800 m.a.s.l and 2200 m.a.s.l, respectively [44,45,47,48], (Appendix A). The catchment experiences bimodal rainfall, occurring mainly in March to May with short rains in November and December [49,50]. The temperature gradient of the catchment is closely related to altitude, ranging from 15 to 33 °C, with the maximum and minimum temperatures occurring during February and July, respectively [48] (Appendix B). The catchment land cover types change in response to the changing elevation, ranging from montane forests on the higher altitudes to semiarid in the lower slopes. The major land cover types include natural forests, woodlands, grassland thickets with emergent trees, bushland and plantation forests [51]. The geology upstream of the dam is characterized by volcanic rocks composed of the olivine and alkaline basalts, phonolites, trachytes, nephelinites and pyroclastics [43,48,52]. The majority of the population settlements are located on the lower slopes between 900 and 1800 m.a.s.l, where most agricultural activities are concentrated. The ever increasing demand for food with an increasing population in the NYM catchment within the Pangani River Basin, Northern Tanzania, has led to rapid expansion of agricultural land, thus accelerating soil loss rates and downstream siltation of the reservoir [30,53–57].
2.2. Sampling Strategy

Three 28 to 32 cm cores were taken from the reservoir, two close to the mouth of the adjacent river input and one more distant (Figure 2). The core locations were limited by logistics and chosen so that they included a temporal record of sediment deposited in the reservoir. Frequent flooding to the villages nearest the headwaters of the reservoir was an indication of the high siltation in the inlets. Shallow depth in the middle of the lake and presence of mostly coarse (sand to gravel) sediments in some locations in the middle was the reason for the abandonment in sampling in the middle of the reservoir. In addition, due to the presence of hippopotamus, certain areas in the lake were also avoided. To this end, efforts were mostly concentrated close to the inlets to account for spatially specific and localized sedimentation effects. Water depths of the core locations spanned between 1.6 and 3.0 m (site AC1 = 1.8 m; AC2 = 1.6 m; AC3 = 3.0 m). A corer, fixed inside with PVC tubes (60 cm length and 8 cm in diameter) was manually pushed into the sediment by divers. The cores were kept in upright positions in the boat for some hours where their overlying waters were gently decanted. The PVC tubes were consequently opened from one end, after which each core was subsectioned at 2 cm intervals to obtain a sufficient temporal resolution, while keeping a high analytical accuracy [58]. The core sections were stored in polyethylene bags and transported to the laboratory where they were weighed and then oven-dried at 50 °C to constant weight, in order to determine dry bulk density. The dried sliced cores were subsequently disaggregated using a mortar and pestle and then sieved.

The tributary riverbed sediment samples (DS) were collected from lower reaches of the two major tributaries, Kikulewa (KL) and Ruvu (RV), assuming that the transported and deposited sediments offer a representative sample of the composite mixture from their respective sources in the entire catchment. Fourteen and eighteen samples of the respective tributaries were collected over a range length of about 200 m to include potential spatial differences in riverine sediment deposition [30,59,60]. DS sample collection depended on the environmental and logistical constraints in the system. Due to the ephemeral nature of the rivers, DS samples were generally collected in the dry season from the exposed beds. Land use soil samples were recovered from agricultural topsoil (CU), bushland topsoil (BS), channel banks (CB) and mainstem river banks (RB). The agricultural and bushland topsoils...
involved surface soil (0–5 cm) sampled from areas presumed vulnerable to water erosion and their connectivity to river network. Sampling of subsurface/channel bank material was done in upstream areas characterized by exposed banks devoid of vegetation with actively eroding bank sections due to flow incision by high water energy released during heavy rainfall. Eroding mainstem river banks were also sampled. Sampling locations for the land use samples depended on the accessibility, necessary permits and safety. At each site, samples comprised a composite of 10 to 15 random scoops pooled into a single composite sample to ensure the representativeness of the corresponding fingerprint property datasets. A total of 57 samples were collected to characterize four main potential sediment sources: (i) bushland (BS, \( n = 15 \)), (ii) channel banks (CB, \( n = 15 \)), (iii) cultivated agricultural land (CU, \( n = 14 \)) and (iv) mainstem river banks (RB, \( n = 15 \)), all collected in one-year season. Soil and tributary sediments were also oven-dried at 55–60 °C to constant weight and consequently disintegrated using a mortar and pestle and then sieved.

![Figure 2. Sediment cores sampling locations (red marks).](image)

2.3. Radiometric and Geochemical Laboratory Analysis

Radiometric analysis and the geochemical analysis of the cores in loose powder form were done at the Consolidated Radioisotope Facility (CoRiF) Laboratory of the University of Plymouth certified under ISO9001:2018, while the geochemical analysis of the potential sediment sources in pressed pellets was done at the Tanzania Atomic Energy Commission (TAEC).
2.3.1. Radiometric Analysis

The sliced sections of the sediment cores were filled into aluminum canisters and sealed for a minimum of 21 days to account for the secular equilibrium between $^{226}\text{Ra}$ and its progenies. The activity concentrations of the radionuclides in the subsequent section cores were counted for 24 h, analyzing the gamma spectra of the natural radionuclides $^{210}\text{Pb}$ (46.5 keV), $^{226}\text{Ra}$ (via $^{214}\text{Pb}$ and $^{214}\text{Bi}$ peaks at 352 and 609 keV, respectively) and for the anthropogenic radionuclide $^{137}\text{Cs}$ (662 keV). The samples with low masses were counted for 48 h, and their results were cited with a $2\sigma$ counting error. Excess $^{210}\text{Pb}$ ($^{210}\text{Pb}_{\text{ex}}$) activity was determined as the difference between total $^{210}\text{Pb}$ and supported $^{210}\text{Pb}$ (in equilibrium with the parent radionuclide $^{226}\text{Ra}$). A natural homogenized soil, with low background activity and spiked with a radioactive traceable standard solution (80717-669 supplied by Eckert & Ziegler Analytics, Atlanta, GA, USA), was used to perform the calibration of the gamma spectrometer. GammaVision software was used to establish the geometry-specific calibration relationships. Analytical performance was assessed by participation in IAEA worldwide proficiency using example soils (IAEA-CU-2009-03 and soil IAEA-TEL-2012-03).

2.3.2. Geochemical Analysis

Before analysis, all dried samples were homogenized and sieved to <63 µm fraction to minimize particle size effects on tracer signals that can bias fingerprint property [13,32,61,62]. The elemental concentrations are generally enriched in the fine, < 63 µm, particle size fraction in comparison to <2 mm bulk fraction of the soil [13,63,64]. For comprehensive reviews on issues of particle size effects on sediment fingerprinting, readers are referred to [61]. Subsequently, about 4 g of dried and sieved sample material was mixed with about 0.9 g of cellulose binder (FLUXANA®), homogenized in a pulverizer and pressed into a pellet of approximately 32 mm diameter. The method was validated by using the IAEA Soil 7 certified reference materials (CRM) described in [65].

The core samples were analyzed for minor and major elemental geochemistry by wave length dispersive X-ray fluorescence (WD-XRF; PANalytical Axios Max; OMNIAIN application) as loose powder. The sediment certified reference material was used to validate the analyses (GBW07318, LGC, Middlesex, UK). The dried soil and tributary potential sediment sources were analyzed by an energy dispersive X-ray fluorescence (EDXRF) spectrometer coupled with Xlab ProTM software as pressed pellets. For assessment of the analytical variability and sample homogeneity, triplicates were made from arbitrarily selected samples about once every 3 samples. Only those elements returning measurements above the limit of detection (DL) were employed in the analysis (DL varies with the element and depends upon several factors including the sample matrix). The difference in the analytical methods for the major and minor elements may have influenced the accuracy of the model.

2.4. Data Analysis

2.4.1. Sediment Chronology and Mass Accumulation Rates

The rate of change in $^{210}\text{Pb}$ activity with mass depth in a sediment core provides the basis for an age–depth relationship and for estimating sediment mass accumulation rates (MARs) [30,41,66,67]. The comparatively long half-life of $^{210}\text{Pb}$ ($t_{1/2} = 22.23$ years) provides the basis for the age determination processes of up to 5–6 half-lives, i.e., ~100 years. For the purpose of reconstructing variations in sedimentation rates over time, the constant rate of supply (CRS) model was used because it assumes a constant $^{210}\text{Pb}$ flux but allows the sediment supply to vary [38] (Appendix C). This model was successfully applied in a nearby complex catchment that had experienced catchment-wide environmental changes [30]. However, a major limitation of this technique in complex East African catchments is that a significant fraction of $^{210}\text{Pb}_{\text{ex}}$ supply to the sediment deposits may originate from older catchment material [68]. In large catchments there might be natural
variability in the terrestrial geological prevalence of $^{238}$U that can influence the $^{210}$Pbex activity from the secondary $^{210}$Pbex activity. In addition, the natural variability might be caused by the dissimilarity in prevalent erosion processes [69] that can change the fraction of topsoil vs. subsoil in the transported sediment material, affecting the $^{210}$Pbex activity [30,36,70,71]. Consequently, whether the atmospheric $^{210}$Pbex flux to the reservoir environment acts steadily stable over time, the arriving secondary $^{210}$Pbex fingerprints from deposited sediment might differ significantly [68]. Owing to potential differences in $^{210}$Pbex fluxes, the CRS model outcomes were scrutinized through comparison with another independent marker, $^{137}$Cs ($t_{1/2} = 30.17$ years), from its 1965 peak fallout [72] in the southern hemisphere using the fitting approach (CRS-fitted) as described in Appleby [34] (Appendix D). However, due to the low levels of $^{137}$Cs fallout in tropical Africa and the known date of dam construction, the deepest sediment layers were also fitted to the reference date of 1969 using the fitting approach (CRS-fitted) as described in Appleby [34] and in Appendices C and D. Finally, the geochemical profiles of the cores were also scanned for distinct changes or peaks that could be linked to hydrological or sedimentological changes [30,73].

2.4.2. Bayesian Mixing Model (BMM) for Source Apportionment

After the broad spectrum geochemical analysis, each sample can be represented as a multielemental concentration data point. The fingerprints of the potential sediment sources and the lake mixture form multivariate concentration matrices on which the model is drawn to quantitatively compare different sources with the lake mixture. A Bayesian mixing model was built in the open-source MixSIAR framework [74–76], as first demonstrated by Blake et al. [31] for river basin sediment source apportionment. For comprehensive details of the mathematical formulation of MixSIAR, readers are referred to Stock et al. [74]. The deconvolutional MixSIAR (D-MixSIAR) methodology was used to hierarchically unmix the tributary sediment against the core sediments and subsequently unmix the land use source fingerprints against the tributary sediment. For the BMM to accurately represent the system, it depends on the following four assumptions: (1) the model includes all dominant sources contributing to the sediment, (2) the value of the tracers are known in both sources and mixture, (3) tracers behave conservatively throughout the mixing processes and (4) fingerprint variability between sources is larger than within sources.

Tracer Conservation Test

In order to make a direct comparison of the properties of the sediment samples with those of the potential source materials, tracers need to behave independently and conservatively (assumption 3) in the environment [31,77]. This implies that the chemical composition of the tracers does not alter during detachment, transport or after deposition [61,78,79].

The source apportionment results were achieved using a tracer selection procedure that only excluded tracers on the basis of nonconservative behavior from literature evidence of potential mobility in aquatic systems [80]. Initially, all elemental concentrations from the depositional samples that fell outside the minimum detection limits were removed. Thereafter, the basic tracer screening approach of Blake et al. [31] and Sherriff et al. [81] was adopted with additional evaluation of geochemical behavior. For each set of sources and associated mixtures for all tracers, boxplots were produced and the means of the mixture data assessed to see if they largely fell within or outside of the mean concentrations of the different sources [31] (Appendix E). Tracers wherein the mixture fell outside the source range were removed. In addition, the tracers that were found to be higher in intrasource variance than the intersource variance were also removed. Finally, the normality assessment using the Shapiro–Wilk test for the individual tracer mixtures was done because the model assumes normal distribution of the mixture tracer data [74]. Seven tracers passed the range test (P, Ti, Mn, Fe, Zn, Sr, Nb) and 7 borderline tracers (S, Co, Ni, Cu, Ga, Ba, Hf) were also retained and engaged. Model efficacy was evaluated using the Gelman–Rubin diagnostics [82]. Seventeen tracers (Na, Mg, Al, Si, Cl, K, Ca, Cr, Br, Rb, Zr, Ce, Pb, Th,
La, Y, Sn) were eliminated from the analysis based on the evidence of nonconservative behavior or high intrasource variability. These exclusions warrant some geochemical clarification. Sr and Rb may have been caused by their known variability in the soil depth as a function of weathering processes and mixing of soil horizons by cultivation [83]. Nonconservative behavior of Mg, Na, Ca, Cl, F, K and Br can be explained by their tendency to form highly soluble salts [31] driven by evaporation in the lake [84]. La, Ce and Th were removed because of the observed intersource variability, which are potentially an artifact of analytical challenges due to low abundance or high variability in the terrestrial source concentrations [30]. Various elements, such as Al, are known to exhibit nonconservative behavior during fluvial transport and short-term storage in river channels [85], while other tracers (Si Cr, Y, Pb and Zr) are known to undergo transformations in medium- to long-term storage elements such as floodplains, lakes and wetlands due to changes in redox, pH, salinity and other environmental conditions [86–88]. Concentrations of Al, Si and Zr may be due to wider fluvial sorting, e.g., textural controls on mineral composition, i.e., changing proportions of silt versus clay minerals in mixtures, which has been shown to exert a strong influence on sediment concentrations [89]. The older sediment deposits, however, may have undergone diagenetic processes within the sediment column [90].

Principal Component Analysis (PCA)

PCA was performed to accommodate assumption 4, which requires the intersource differences of fingerprints to be larger than the intrasource differences in fingerprints. It is in this context that the PCA was performed to analyze variance in multitracer datasets and reduce dimensionality [90] from the potential sediment sources data into land use specific and tributary categories according to their geochemical composition.

Model Build and Running Protocols

(1) Error formulation: Eroded soils with distinct geochemical properties form the wider catchment mix from the source to a mixture through mixing process. However, in these complex systems it is impossible to capture the total variability in sediment source sample by sampling. Therefore, a “residual error” formulation was integrated in the model. Since the transport of sediment from channel networks to the reservoir is random and constant, a “process error” was not included [74,75].

(2) Since there are no other sources of empirical information concerning the sediment source contributions to the reservoir, an uninformative prior was used: (1, 1, 1, 1) and (1, 1) for land use and tributary sources, respectively.

(3) A mixture of sediment samples was analyzed without fixed or random effects to infer the proportions of the tributaries and land uses to the “total” reservoir sediment. Afterward, a fixed categorical effect of the sediment mixture from reservoir sampling locations was established to infer the proportions of different land uses and tributaries to the specific sampling location in the reservoir. The ascription of sediment delivery over time was inspected by using individual sediment cores whereby the “age” was introduced as a fixed continuous effect. Subsequently, the sediment core slice samples were grouped into distinct classes whereby the depth was established as fixed categorical effect. The model outputs were assessed under different modes of covariate structure.

For all model runs, the following stipulations were used: a residual error term only and an uninformative Dirichlet prior ($\alpha = 1$). Model convergence was assessed by the Gelman–Rubin diagnostic (variables < 1.05), rejecting model output if >5% of total variables were above 1.05. Model convergence indicates that the model found a singular solution to the problem. The MIXSIAR model using the selected 14 tracer fingerprints on the sediment sources passed the Gelman–Rubin convergence diagnostic with the parameters of the Markov chain Monte Carlo (MCMC) chain run length set as follows: chain length = 1,000,000, burn = 700,000, thin = 300, chains = 3.
3. Results and Discussion

3.1. $^{210}$Pb$_{ex}$ and $^{137}$Cs Vertical Profiles

The $^{210}$Pb$_{ex}$ profiles (Figure 3a–c) of all three cores did not follow an exponential decline with depth and were instead characterized by multiple peaks and troughs indicating episodic increase in sedimentation rate, which would have diluted the $^{210}$Pb$_{ex}$ activities of the more recently deposited sediments [30,34]. However, a number of processes could have also resulted in the vertical mixing in the sediment cores. First, high energy water flows and wave action at the river inlets could have led to vertical mixing of the sediment deposits and a flattening of the $^{210}$Pb$_{ex}$. Second, human and biological activities might also have led to mixing of the upper sediment layers in the shallow parts of the reservoir. Third is the variation in the remineralization of sediment particles downcore. Furthermore, previously discussed variations in the direct atmospheric depositional flux of $^{210}$Pb due to fluctuations in the annual amount of rainfall and variation in the secondary input of the $^{210}$Pb from the catchment could have influenced the $^{210}$Pb$_{ex}$ profile [68].

The $^{137}$Cs activity concentration, where detected, was low in all of the cores, with most of the sections below the limit of detection. These results are similar to the global $^{137}$Cs fallout estimates in tropical regions [35]. The low $^{137}$Cs activities obstructed the assessment of $^{137}$Cs peak integrity and the comparison of $^{210}$Pb radiometric dates with independent $^{137}$Cs peaks [3]. Contrary to core AC2 and AC3, core AC1 had a significant single measure of $^{137}$Cs activity above the detection limit (Appendix F). The detectable layer was assumed to be the 1965 southern hemisphere peak deposition [35,91]; however, the “peak” might also be related to secondary transport of catchment surface material to the reservoir, meaning that it could be younger than 1965 [68,92]. For these reasons, the $^{137}$Cs dating and fitting the CRS models of the NYM reservoir cores using the $^{137}$Cs dates would result in a high level of uncertainty. Therefore, two versions of the CRS model were run. The first “standard” approach estimated dates based solely on fallout $^{210}$Pb$_{ex}$. In a second “fitted” model, it was opted to fit the deepest sediment layers to the construction date of the dam. In the former case, the assumption is that the core penetrates the post impoundment soft sediment into the original substrate. In the latter, there is an assumption that the coring captured the full soft sediment profile.

Under CRS standard approach model version, the observed changes in sedimentation dynamics could be associated with the construction of the dam, but also by higher levels of soil erosion following deforestation and agricultural degradation. The changes in 1970s correspond with the adoption of the “villagization policy” in Tanzania, which led to practicing of a cooperative economy (communal production) through collective farming [93–95]. The sharp MAR peaks 0.86 and 1.01 g cm$^{-2}$ yr$^{-1}$ in the mid 1990s in cores AC1 and AC3, respectively, may be linked to ENSO, specifically, the 1997/1998 El Niño events [96] (Figure 3d,f). These general trends of sedimentation dynamics in cores AC1, AC2 and AC3 in the 2000s to 2010s is related to growing urbanization, extension of agricultural activities and the loss of permanent vegetation through the rapid expansion of agricultural land [12,22,96–98].

The alternative CRS-fitted geochronology model approach of Appleby [34], using 1969 as a fixed date, showed a similar trend of changing sedimentation dynamics as the CRS-standard approach. While the MAR of cores AC1, AC2 and AC3 and age–depth relationship using the standard CRS approach were found to be relatively similar (Figure 3d–f), the MAR-fitted trends were also found to be similar between the different cores. There were some interesting differences observed between the CRS-standard and CRS-fitted approaches in each individual core. For instance, the MAR-fitted trends in core AC3 were higher than those in core AC1 and core AC2. Due to the smaller time range, the MAR-fitted rates and peaks were much higher, giving rise to a more punctuated profile. However, the general observed trend of increasing sedimentation over time was observed in both approaches. While extreme rainfall during the ENSO 1997/1998 El Niño events [95] seems to have contributed to higher rates of sedimentation in the NYM reservoir, this was not observed for the higher rainfall in 1978/79 (Appendix A). Under this model scenario, this
seems to confirm that the increase in sedimentation in the NYM reservoir was driven by a complex interaction between natural rainfall variability and increasing vulnerability to soil erosion through land use change. The differences observed in the timing and heights of the MAR peaks between the cores might be due to spatial differences in sedimentation due to the dynamics of sediment transport from the dominant riverine sources, as will be further explored in Section 3.3. Difference in particle settling velocity due to different locations of the cores might be another reason for the variations in sedimentation rates. However, the main limitation of the CRS-fitted approach fitting the curve to 1969 is that it might have included the activities from the floodplain of the natural wetland before the date the reservoir was commissioned in 1969.

Figure 3. Cont.
Figure 3. Cont.
The geochemical profiles of the sediment cores offer a useful indication of changes in tributary sediment delivery to NYM reservoir over time to help constrain and interpret the different geochronological models. Comparing the broad trends of the CRS output of all model versions with their geochemical profiles allowed for catchment-based contextualization of the modeled MARs [30]. The recent MAR peaks in all cores closely matched with peaks in many elements connected to allogenic sediment origins and minima in elements linked to evaporative autogenic tracers, as shown in Appendix G [84]. The allogenic tracers Fe, Ti, Zr and Nb, Rb, Ba and Ga and the evaporative autogenic tracers Mg, K, Na, Ca and Sr have corresponding maxima and minima at the same depth (4 cm) in 2016 and 2017 for cores AC1 and AC3, respectively, as the MAR peak. However, in core AC2, the allogenic tracers Al, Ti, Ni, Cr and Fe and the autogenic tracers Na, S, Cl, Sr and Ca have distinct maxima and minima at 2 cm in 2019 and 5 cm in 2016, respectively, while the evaporative autogenic tracers Mg and Na have the minima corresponding to maxima of the allogenic tracers. This high correlation between geochemical tracers of allogenic sediment delivery and reconstructed sedimentation peaks in both cores reciprocally validates both evidence bases, making it highly likely that NYM reservoir recently experienced extreme sedimentation rates driven by increased erosion and sediment transport from the catchment. The difference in concentrations was observed between the cores; the variation in geochemical concentration is not only caused by changes in absolute sediment delivery, it may be attributed by the difference in source contribution (e.g., from different tributaries) and erosion process (e.g., subsurface vs. surface) which also cause changes in profile. The difference in concentration can also be due to location-specific sedimentation effects (e.g., more or less sedimentation and/or higher contribution from one of the tributaries). Interestingly, the difference in concentrations was also observed between the deepest older sections and upper recent sections of the autogenic (Na, S, Cl, Sr, Ca) and the allogenic (Al, Ti, Ni, Cr, Fe) tracers. The concentrations of the autogenic tracers were much higher in the deepest sections and lower in the upper recent sections and vice versa for the allogenic tracers. The possible reason is that the deepest core sections are from before reservoir construction, i.e., “standard” geochronology model, or the older sediment deposits may have undergone diagenetic processes within the sediment column [90].

An additional consideration to the limitations of ascribing exact dates to the sediment column is that the $^{210}\text{Pb}_{\text{ex}}$ profiles are vulnerable to sediment reworking through biological and physical activities such as bioturbation and fishing [99]. Following these occasional events, the $^{210}\text{Pb}_{\text{ex}}$ profiles are interspersed and might be difficult to determine [100]; however, in many cases, an indication of broad rates of SAR change can still be determined with notable evidence for changes in sediment provenance being a key observation [37].
3.2. PCA for Statistical Analysis of Data

The temporal and spatial distinctiveness of the source fingerprints identified show distinct fingerprint clusters between the two tributaries (Figure 4) and a low level of overlap between the land use sources, the RB and CB, while CU and BS show distinct fingerprints (Figure 5). The reduced discrimination between the RB and CB may be due to source signatures to resemble a mix of RB and CB, given both are subsurface materials and presumably less weathered. Overall, however, the signatures provide a clear basis for sediment attribution [101,102].

![Figure 4](image_url)

**Figure 4.** Ordinance biplots visualization of the geochemical drivers of variance in the fingerprint of the tributary sources detailing the intra- and intervariance from a mixture pool.

![Figure 5](image_url)

**Figure 5.** Ordinance biplots visualization of the geochemical drivers of variance in the fingerprint of land use sources detailing the intra- and intervariance from a mixture pool.

3.3. Proportional Tributary and Land Use Contribution

The BMM outputs revealed that the Kikuletwa River contributed 60.3% of “total” reservoir sediment and the Ruvu River 39.7% (Figure 6). While the Kikuletwa River is the dominant contributing tributary to the total reservoir sediment, cores AC2 and AC3 have a higher proportional contribution of Ruvu sediment with 55.4% and 51.8%, respectively.
The dominance of Ruvu in AC3 is counterintuitive since the location of AC3 is farther from the Ruvu inlet. However, sedimentation dynamics are also regulated by other factors besides distance to the inlet, such as the dominant flow direction and velocity at the river inlets. Overall, the contribution seems to be well balanced between both rivers.

Figure 6. Sediment source apportionment using MixSIAR, where KL accounted with 60.3% and RV with 39.7%.

The sign of changes in tributary sediment delivery to the NYM reservoir over time is evidenced by the geochemical stratigraphic record of the sediment cores that experienced distinct periods of high and low sedimentation associated with changes in tributary sediment delivery (Figure 7). Unmixing the core sections over time yielded changing proportional contributions of the river systems. The proportional contribution of KL in core AC1 seemed to have increased in the lower core before reverting back to almost 50%. RV decreased from the lower to mid core and then remained stable over time before a slight increase in the upper core. In core AC2, the proportion of KL showed a continuous decrease from 83.1% in the lower core to approximately 50% in the upper, and vice versa for RV. Core AC3 also shows an increase of KL over time, after experiencing a distinct drop in contribution in the upper core (Appendix H). Integration of the changing proportional contribution and reconstructed MAR seems to indicate that the most recent increase in sedimentation is mostly driven by increased sediment delivery from the RV system. This finding seems to indicate that increased erosion and sediment transport from the RV system are driven by increased land use change in the catchment. This result corresponds with previous research by Mzuza et al. [43] using magnetic properties showing that RV is the most recent contributing tributary. This association of land use change was also observed using the older and the recent core sections of AC1 and AC3, respectively (Appendix H). The comparison was made following the trend of increasing sedimentation rate in core AC1 and AC3 (Figure 4e,f) with the relative source contribution (Appendix H) and (Figure 8).

Unmixing of the core sections directly against the land use pattern demonstrated a significant increase of sediment contribution from the agricultural land (CU) and decrease of the BS, CB and RB in older and younger sections of cores AC1 and AC3. CU increased from 47.6% (lower core) to 59.6% (upper core) while BS, CB and RB decreased from 11.6%, 20.5%, and 20.3% to 10.1%, 15.4% and 14.9%, respectively, in core AC1 (Appendix H) and (Figure 8). A similar trend was observed in core AC3 where CU increased from 53.5% to 71.0% while BS, CB and RB decreased from 16.1%, 15.1% and 15.3% to 8.7%, 10.3% and 10.0%, respectively (Appendix H) and (Figure 8). The increase in contribution from CU corresponds to the previous studies that observed high levels of deforestation and the loss of permanent vegetation through the rapid expansion of agricultural land and growing urbanization [12,22,96–98]. The decrease in the contribution of the BS, CB and
RB is proportional to the increase in CU and the increase in sedimentation in younger core sections, increasing the evidence for changing dynamics of soil erosion and sediment deposition in the reservoir during recent years.

**Figure 7.** Changing proportional sediment contributions of tributary rivers over time to cores AC1, AC2 and AC3, based on age estimates from the standard CRS model used here as relative age markers.

**Figure 8.** Changing proportional contribution of land use from older to the recent core sections of sediment cores in AC1 and AC3.
Unmixing of the land use types against the riverine sediment revealed that the Kikuletwa tributary had a greatest contribution of sediments from CU (38.4%), and the other types also had significant contribution, RB with 25.6%, CB with 25.5% and BS with 10.5% (Figure 9, left, and Appendix H). The CU on Ruvu River accounted for 44.6% of the sediment contribution, CB with 31.0%, RB with 18.0% and BS with 6.4% (Figure 9, right, and Appendix H). These findings corroborate with previous research showing that both subcatchment have high soil erosion risk due to unsustainable farming practices [12,22,47,96–98]. The farming practices in both subcatchments may have increased the sediment connectivity to the channel networks due to less buffering of the soil [103]. Although most of the irrigated large-scale plantations are located on the lowlands, sediment sources might have originated from the sloped small-scale agricultural plots and not from the lowland irrigation agriculture. In addition, the use of floodplains for agriculture might have increased the sediment routing to the lake. The evidence of eroding river and channel banks as important sediment sources in fluvial systems [104–107] is underscored by the concentration of the geochemical tracers in the Ruvu River (Appendix H). The steady increase in the concentration of the Mn, Sr and S [85,108,109] with depth supports the hypothesis of the channel banks and gully erosion. The anthropogenic land use changes or natural alteration of geomorphology dynamics of channel and river bank dimension in the watershed may have influenced the subsurface erosion of the allogenic sediment delivery to the reservoir. In addition, the changes in the catchment land cover types from montane forests on mountain slopes to semiarid grasslands [12] in response to climate change impacts [15] and land use pressures [17,18] may have increased the hillslope erosion, thus increasing the structural sediment connectivity to the channel networks.

![Figure 9](image-url)

**Figure 9.** D-MixSIAR of the landuse from KL and RV tributaries where CU accounted for 38.4%, CB with 25.5%, RB with 25.6% and bushland with 10.5% for KL (left) and where CU accounted for 44.6%, CB with 31.0%, RB with 18.0% and BS with 6.4% of the total RV sediment (right).

Overall, the fingerprinting analysis shows that cultivated land was the dominant source of the riverine and total reservoir sediment. However, these results should be inferred as estimates and not complete observations because of the temporally and spatially constrained representations of geochemical fingerprints, the geochronological model assumptions and structure and tracer selection protocols. Nevertheless, the compatibility of
model output proves the general strength of the methods and the significance of sediment input from this system.

A major challenge of the approach used here is that the riverine sediment fingerprint originates from samples spanning one year, while the cores integrate >60 years. The riverine sediment thus does not include potential variations in sediment fingerprint over time. However, given the strong difference in signature between the Kikuletwa and Ruvu river sediment, it is assumed that intrasource variations will remain smaller than intersource difference. In addition, the model may have difficulty in distinguishing between clusters with similar geochemical fingerprints. For example, RB samples are often a natural integration due to deposition from upstream sources, often leading to large overlaps with other sources. Differences in tracer concentrations between land uses replicated the processes distinguishing physical and chemical properties of subsurface soils. The prevalence of deeply weathered soils at the study site could have led to the overlap between CB and RB [110,111]. CB comprised mostly subsurface soil, however, due to sampling of the entire bank profile it might include a mixture of some surface and subsurface soil.

4. Conclusions

This study demonstrated the potential use of geochemical profiles as a valuable alternative tool for independent confirmation of changing sedimentation rates in the context of disturbance in East African hydropower catchments. The allogenic maxima and autogenic minima of NYM reservoir freshwater have reciprocally aligned with $^{210}$Pbex reconstruction of sedimentation dynamics over time. This study successfully deployed a quantitative sediment fingerprinting technique to apportion recent potential sediment sources in relation to land use change. The integration of geochemical fingerprinting within a BMM framework pointed toward one of two tributaries, the Kikuletwa River, as the dominant contributing tributary to the total reservoir sediment. In addition, the fingerprinting analysis shows that cultivated land was the dominant source of the riverine and total reservoir sediment. Moreover, the integration of the changing proportional contribution and reconstructed accumulation rates seems to indicate that the most recent increase in sedimentation is mostly driven by increased sediment delivery from the other tributary, the RV system. This result corresponds with previous research by Mzuza et al. [43] using magnetic properties that RV is the most recent contributing tributary, which indicates the pace of land use change in this system is having a profound effect on sediment supply. The assertion is also backed by sedimentary evidence from cores AC2 and AC3, which had a higher proportional contribution of Ruvu-derived sediment, 55.4% and 51.8%, respectively. This finding seems to indicate that increased erosion and sediment transport from the RV system are driven by increased land use change in the catchment. Overall, the study revealed major changes in the sedimentation dynamics over time, which is probably driven by a complex interaction between land use changes, climate changes and natural rainfall variability. Since sedimentation is always highly localized in large reservoirs, evaluation of localized sedimentation effects would provide a full representation of spatially specific sedimentation issues and a deeper understanding of the driving processes within the catchment. To this end, future studies using BMM in large reservoirs should aim to include spatial factors in the model setup concerning receptor sediment mixtures; thus, a better representation of complex sedimentation dynamics in space could be obtained. Furthermore, sediment control strategies require confirmation on the relative and absolute contributions of sediment from different sources. The information on the sediment sources and sedimentation rate in the reservoir alone are not enough since the sediment delivery process is complex; it involves the interaction of multiple factors and processes on different spatial and temporal scales. Therefore, understanding the processes that result in erosion and its connectivity to the river channel, storage in hillslopes, floodplains and sediment accumulation in the reservoirs is vital for the sustainable management of the reservoir. For instance, not all eroded materials reach the reservoir; some is stored in the hillslopes and floodplains. The sediment budget approach provides such a holistic perspective by accounting for the
various sediment sources, transport, sinks and redistribution when the sediment is routed through that catchment. Thus, this study recommends the integration of several different techniques such as remote sensing GIS models, sediment source fingerprinting, FRNs for dating and soil redistribution etc. The integration of the techniques will mutually offer the required information on sediment mobilization, redistribution, transport and storage within a catchment. The results underscore the necessity for targeted erosion mitigation strategies on the potential sources to limit soil erosion and reduce further impact on the reservoir water quality.

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**Conflicts of Interest:** The authors declare that there are no conflict of interest.

**Appendix A**

A: Rainfall fluctuations

Figure A1. Changes in annual precipitation (mm) measured at Moshi Airport Moshi Municipal, Kilimanjaro station No: 9337004 coordinate 3°21’0” S and 37°19’48” E.
Appendix B

Figure A2. Temperature fluctuations—Changes in annual temperature (°C) measured at Moshi Airport Moshi Municipal, Kilimanjaro station No: 9337004 coordinates 3°21′0″ S and 37°19′48″ E.

Appendix C

CRS-standard approach

With $s_n$ as the experimentally derived dry bulk density in section $n$, the cumulative dry mass $m_n$ above sediments at depth $x_n$ can be calculated as:

$$m_n = m_{n-1} + s_n(x_n - x_{n-1})$$  \hspace{1cm} (A1)

where $C_n$ is the experimentally derived $^{210}\text{Pb}_{\text{ex}}$ activity at layer $n$, the cumulative $^{210}\text{Pb}_{\text{ex}}$ inventory can be calculated using the trapezium rule:

$$\hat{A}_n = \hat{A}_{n-1} + \frac{C_{n-1} - C_n}{\ln \left( \frac{C_{n-1}}{C_n} \right)} (m_n - m_{n-1}).$$  \hspace{1cm} (A2)

The total $^{210}\text{Pb}_{\text{ex}}$ (in Bq m$^{-2}$) inventory $A(0)$ of the sediment core is then equal to the $\hat{A}_n$ value in the deepest layer. The residual $^{210}\text{Pb}_{\text{ex}}$ (in Bq m$^{-2}$) inventory in the sediment core below depth $n$ can subsequently be easily calculated by subtracting $\hat{A}$ from $A(0)$. Following the CRS model and with $\lambda_{\text{Pb}}$ as the $^{210}\text{Pb}$ radioactive decay constant of 0.03114 y$^{-1}$, the age $t$ of the sediment layer at depth $n$ can be estimated by:

$$t = \frac{1}{\lambda_{\text{Pb}}} \ln \left( \frac{A(0)}{A(n)} \right).$$  \hspace{1cm} (A3)

The sedimentation rate $r$ at depth $z$ can subsequently be calculated as follows:

$$r = \frac{\lambda_{\text{Pb}} A(n)}{C(n)}.$$  \hspace{1cm} (A4)

Appendix D

CRS-fitted approach

If $\hat{A}_{\text{ref}}$ denotes the entire $^{210}\text{Pb}_{\text{ex}}$ inventory above the reference level $t_{\text{ref}}$, the inventory below that level can be obtained by the following formula:

$$A_{\text{ref}} = \frac{\hat{A}_{\text{ref}}}{e^{\lambda_{\text{Pb}} t_{\text{ref}}} - 1}.$$  \hspace{1cm} (A5)
The total inventory is then:

\[ A(0) = \hat{A}_{\text{ref}} + A_{\text{ref}} \]  

Sediment dates and accumulation rates can subsequently be calculated using Equations (A3) and (A4) respectively.

Appendix E

Appendix E: Boxplots for tracer selection of the potential sediment sources and cores (mixture). In box plots, median is shown by central line, interquartile range by box, range by whiskers with circles indicating outliers.

Figure A3. Cont.
Figure A3. Boxplots for tracer selection of the potential sediment sources and cores (mixture).

Appendix F

Appendix F: The CRS dating results and 137Cs laboratory results for cores AC1, AC2 and AC3 respectively.
Table A1. The CRS dating results and $^{137}$Cs laboratory results for cores AC1, AC2 and AC3 respectively.

| Core      | Depth (cm) | Mass Depth (g/cm$^2$) | Date (y) | MAR (g/cm$^2$/y) | Date Fit (y) | MAR Fit (g/cm$^2$/y) | $^{210}$Pb$_{ex}$ (Bq/kg) | $^{137}$Cs (Bq/kg) |
|-----------|------------|-----------------------|----------|------------------|--------------|----------------------|---------------------------|------------------|
| AC1       | 2          | 2.26                  | 2019     | 0.879            | 2018         | 1.03                 | 60.81                     | 2.67             |
|           | 4          | 5.13                  | 2016     | 1.12             | 2015         | 1.32                 | 44.01                     | 2.30             |
|           | 6          | 8.26                  | 2014     | 1.52             | 2013         | 1.83                 | 29.75                     | 1.82             |
|           | 8          | 11.33                 | 2012     | 1.2              | 2012         | 1.45                 | 35.29                     | 2.14             |
|           | 10         | 14.62                 | 2009     | 0.91             | 2009         | 1.1                  | 42.98                     | 1.85             |
|           | 12         | 17.29                 | 2005     | 0.79             | 2006         | 0.99                 | 43.82                     | 1.83             |
|           | 14         | 20.17                 | 2001     | 0.65             | 2003         | 0.83                 | 47.5                      | 2.83             |
|           | 16         | 22.57                 | 1997     | 0.86             | 2001         | 1.18                 | 31.07                     | 2.38             |
|           | 18         | 24.83                 | 1994     | 0.42             | 1998         | 0.58                 | 57.42                     | 2.02             |
|           | 20         | 27.42                 | 1988     | 0.44             | 1994         | 0.63                 | 46.47                     | 2.75             |
|           | 22         | 30.18                 | 1981     | 0.404            | 1989         | 0.63                 | 41.05                     | 2.99             |
|           | 24         | 33.07                 | 1974     | 0.29             | 1984         | 0.48                 | 45.08                     | 2.80             |
|           | 26         | 36.01                 | 1962     | 0.17             | 1977         | 0.33                 | 53.41                     | 4.90             |
|           | 28         | 38.78                 | 1937     | 0.09             | 1968         | 0.27                 | 48.14                     | 2.40             |
| AC2       | 2          | 1.87                  | 2019     | 1.4              | 2018         | 1.25                 | 59.66                     | 2.32             |
|           | 4          | 3.99                  | 2017     | 0.865            | 2016         | 1.05                 | 67.26                     | 1.99             |
|           | 6          | 6.25                  | 2015     | 1.2              | 2014         | 1.48                 | 44.92                     | 2.41             |
|           | 8          | 8.84                  | 2013     | 0.863            | 2012         | 1.06                 | 58.66                     | 2.45             |
|           | 10         | 11.22                 | 2010     | 1.05             | 2010         | 1.34                 | 43.7                      | 2.83             |
|           | 12         | 13.82                 | 2007     | 0.771            | 2008         | 0.99                 | 55.31                     | 4.26             |
|           | 14         | 16.47                 | 2004     | 0.701            | 2005         | 0.92                 | 54.44                     | 3.06             |
|           | 16         | 18.96                 | 2000     | 0.649            | 2003         | 0.89                 | 51.9                      | 2.45             |
|           | 18         | 21.40                 | 1996     | 0.519            | 2000         | 0.73                 | 57.18                     | 2.53             |
|           | 20         | 23.85                 | 1990     | 0.404            | 1996         | 0.59                 | 62.74                     | 3.00             |
|           | 22         | 26.45                 | 1984     | 0.479            | 1992         | 0.77                 | 42.86                     | 3.12             |
|           | 24         | 29.04                 | 1978     | 0.61             | 1989         | 1.08                 | 28.12                     | 2.13             |
|           | 26         | 31.86                 | 1973     | 0.35             | 1986         | 0.65                 | 42.11                     | 2.37             |
|           | 28         | 34.81                 | 1964     | 0.27             | 1981         | 0.57                 | 40.71                     | 2.18             |
|           | 30         | 37.67                 | 1951     | 0.383            | 1977         | 1.08                 | 19.31                     | 2.47             |
|           | 32         | 41.0702               | 1942     | 0.27             | 1973         | 0.56                 | 32.53                     | 2.64             |
| AC3       | 2          | 3.37                  | 2019     | 1.69             | 2018         | 2.23                 | 42.77                     | 1.36             |
|           | 4          | 6.88                  | 2017     | 1.86             | 2016         | 2.26                 | 36.4                      | 2.00             |
|           | 6          | 11.37                 | 2015     | 2.12             | 2015         | 2.57                 | 30.15                     | 1.61             |
|           | 8          | 15.18                 | 2013     | 0.92             | 2012         | 1.12                 | 64.73                     | 2.32             |
|           | 10         | 19.69                 | 2008     | 1.12             | 2009         | 1.38                 | 46.53                     | 1.48             |
|           | 12         | 23.05                 | 2004     | 1.26             | 2006         | 1.67                 | 36.08                     | 2.00             |
|           | 14         | 26.5                  | 2001     | 0.85             | 2004         | 1.13                 | 49.38                     | 1.57             |
|           | 16         | 30.37                 | 1997     | 1.01             | 2001         | 1.39                 | 36.11                     | 1.72             |
|           | 18         | 35.04                 | 1993     | 1.21             | 1998         | 1.73                 | 26.54                     | 1.47             |
|           | 20         | 38.68                 | 1989     | 0.60             | 1995         | 0.86                 | 46.78                     | 1.64             |
|           | 22         | 41.86                 | 1982     | 0.47             | 1991         | 0.76                 | 49.13                     | 2.03             |
|           | 24         | 45.35                 | 1974     | 0.61             | 1987         | 1.13                 | 29.3                      | 1.49             |
|           | 26         | 49.63                 | 1968     | 0.46             | 1983         | 0.89                 | 32.5                      | 1.66             |
|           | 28         | 52.29                 | 1957     | 0.24             | 1979         | 0.58                 | 44.43                     | 1.82             |
|           | 30         | 55.49                 | 1943     | 0.19             | 1974         | 0.62                 | 35.01                     | 1.82             |
|           | 32         | 59.62                 | 1920     | 0.13             | 1968         | 0.70                 | 25.46                     | 1.55             |

Appendix G

Appendix G: Geochemical depth profiles of cores AC1, AC2 and AC3 respectively. Concentration in ppm.
Appendix G: Geochemical depth profiles of cores AC1, AC2 and AC3 respectively. Concentration in ppm

Figure A4. Cont.
Figure A4. Geochemical depth profiles of core AC1, AC2 and AC3 respectively. Concentration in ppm.

Appendix H

Appendix H. Bayesian Mixing Model results.
Table A2. The mean values and Gelman diagnostics (Diag.) of the Bayesian Mixing model runs for both ‘total’ and ‘spatial’ model builds.

| Tributary | Total AC1 | AC2 | AC3 |
|-----------|-----------|-----|-----|
|           | Mean      | Diag| Mean| Diag|
| Kikuletwa | 0.603     | 1.002|0.572|1.002|
|           | 0.446     | 1.001|0.554|1.000|
| Ruvu      | 0.397     | 1.001|0.428|1.001|
|           | 0.518     | 1.002|0.518|1.002|

Table A3. The mean values and Gelman diagnostics (Diag.) of the Bayesian Mixing model runs for “landuse” in “tributaries’ model builds.

| Land Uses      | Kikuletwa     | Ruvu     |
|----------------|---------------|----------|
|                | Mean Diag     | Mean Diag|
| Bush (BS)      | 0.105 1.005   | 0.064 1.000|
| Channel Bank (CB)| 0.255 1.006  | 0.310 1.003|
| Agricultural land (CU)| 0.384 1.002 | 0.446 1.002|
| River bank (RB)| 0.256 1.002  | 0.180 1.001|

Table A4. The mean values and Gelman diagnostics (Diag.) of the Bayesian Mixing model output of core AC1, AC2 and AC3 specified for grouped core sections.

| Cores Core groups | Kikuletwa AC1 | AC2 | AC3 |
|-------------------|---------------|-----|-----|
|                   | Mean Diag     | Mean Diag | Mean Diag | Mean Diag | Mean Diag | Mean Diag | Mean Diag | Mean Diag |
| 1                 | 0.532 1.002   | 0.468 1.003 | 0.461 1.001 | 0.539 1.003 | 0.301 1.003 | 0.699 1.003 |
| 2                 | 0.693 1.002   | 0.307 1.001 | 0.469 1.001 | 0.531 1.002 | 0.392 1.003 | 0.668 1.003 |
| 3                 | 0.69 1.001    | 0.31 1.001 | 0.652 1.005 | 0.348 1.005 | 0.359 1.002 | 0.641 1.002 |
| 4                 | 0.677 1.002   | 0.323 1.001 | 0.679 1.004 | 0.321 1.004 | 0.367 1.001 | 0.633 1.002 |
| 5                 | 0.638 1.002   | 0.362 1.001 | 0.7 1.001 | 0.3 1.000 | 0.287 1.001 | 0.713 1.003 |
| 6                 | 0.713 1.000   | 0.287 1.002 | 0.806 1.005 | 0.194 1.005 | 0.317 1.000 | 0.683 1.001 |
| 7                 | 0.548 1.001   | 0.452 1.001 | 0.829 1.004 | 0.171 1.004 | 0.29 1.001 | 0.71 1.003 |
| 8                 | 0.831 1.000   | 0.169 1.001 | 0.247 1.002 | 0.753 1.002 |
| 9                 | 0.216 1.002   | 0.784 1.002 |

Table A5. The mean values and Gelman diagnostics (Diag.) of the Bayesian Mixing model output of cores AC1 and AC3, specified for recent and older core sections.

| AC1 | Bush (BS) | Channel Bank (CB) | Cultivated (CU) | River Bank (RB) |
|-----|-----------|-------------------|-----------------|-----------------|
|     | Mean Diag | Mean Diag         | Mean Diag       | Mean Diag       |
| Recent sections | 0.101 1.002 | 0.154 1.002 | 0.596 1.001 | 0.149 1.001 |
| Older sections  | 0.116 1.001 | 0.205 1.001 | 0.476 1.000 | 0.203 1.000 |
| AC3 | Bush (BS) | Channel Bank (CB) | Cultivated (CU) | River Bank (RB) |
|     | Mean Diag | Mean Diag         | Mean Diag       | Mean Diag       |
| Recent sections | 0.087 1.001 | 0.103 1.002 | 0.710 1.002 | 0.100 1.002 |
| Older sections  | 0.161 1.001 | 0.151 1.002 | 0.535 1.005 | 0.153 1.002 |
Table A6. Tributary source tracer values in means and standard deviations (SD) and the sample size n.

| Sources | MeanP | SDP | MeanS | SDS | MeanTi | SDTi | MeanMn | SDMn |
|---------|-------|-----|-------|-----|--------|------|--------|------|
| KL      | 950.26 | 48.20 | 326.33 | 84.79 | 1220.91 | 1051.49 | 1051.49 | 82.59 |
| RV      | 2053.06 | 258.82 | 187.17 | 16.75 | 1782.34 | 2159.02 | 2159.33 | 170.93 |

| Sources | MeanFe | SDFe | MeanCo | SDCo | MeanNi | SDNi | MeanCu | SDCu |
|---------|--------|------|--------|------|--------|------|--------|------|
| KL      | 49217.61 | 3679.79 | 94.91 | 29.39 | 92.20 | 13.11 | 70.27 | 83.57 |
| RV      | 83718.41 | 2917.35 | 168.96 | 35.11 | 116.79 | 18.39 | 87.95 | 170.18 |

| Sources | MeanZn | SDZn | MeanGa | SDGa | MeanSr | SDSr | MeanNb | SDNb |
|---------|--------|------|--------|------|--------|------|--------|------|
| KL      | 82.18 | 7.64 | 23.12 | 5.16 | 566.99 | 19.58 | 61.56 | 5.14 |
| RV      | 155.92 | 6.81 | 30.66 | 3.14 | 1131.87 | 52.35 | 244.73 | 15.60 |

| Sources | MeanBa | SDBa | MeanHf | SDHf | n   |
|---------|--------|------|--------|------|----|
| KL      | 900.10 | 218.46 | 14.36 | 8.81 | 18 |
| RV      | 1647.24 | 327.30 | 23.2   | 7.06 | 14 |

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