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Article

Refinement and Validation of the SPEAR_{metal} Index for Assessing Ecological Impacts of Metal Contamination in the Nakdong River, South Korea

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Abstract: The SPEcies-At-Risk index for metals (SPEAR_{metal}) was refined using updated physiological sensitivity data and validated to assess the ecological impact of metal contamination on benthic macroinvertebrate communities in the upper Nakdong River, near a Zn smelter in Korea. Biosurvey and chemical monitoring were collected at 18 sites surrounding the smelter and nearby mines. Acute ecotoxicity tests on 20 indigenous species from the Korean peninsula were conducted and used to update taxon-specific metal sensitivity data. The refined SPEAR_{metal} index, based on this updated sensitivity, was significantly lower than previous versions, with most values below the severe impact threshold (0.5) in the mainstream. The correlation between hazard quotients in water and the SPEAR index improved, with the correlation coefficient increasing from 0.63 to 0.70. Despite consistently high benthic macroinvertebrate indices (BMI) across the study area, generic ecological indices such as total richness, EPT (Ephemeroptera, Plecoptera, Trichoptera), and Shannon diversity showed correlations with metal contamination levels. Principal component analysis identified SPEAR_{metal} as the primary indicator associated with metal contamination in both water and sediment. These findings highlight the improved performance of the refined SPEAR_{metal} index as a more sensitive and specific tool for assessing the ecological status of metal-impacted aquatic ecosystems compared to traditional indices.

Keywords: metal; trait-based bioindicator; macroinvertebrates; ecological impact; SPEAR

1. Introduction

Metal contamination in aquatic ecosystems poses significant challenges to ecological health, necessitating the development and application of comprehensive assessment methods. Ecological risk assessments at a metal contaminated site typically involve three complementary approaches: prospective assessments, ongoing ecotoxicological monitoring, and bioassessment [1–3]. Prospective assessments, utilizing chemical monitoring data, predict potential risks, while ecotoxicological monitoring evaluates the current status of aquatic ecosystems. Bioassessments, conducted in the field using ecological bioindicators, are essential for confirming the actual ecological impact of metals based on causal associations [4].

The integration of these approaches provides a more comprehensive understanding of metal contamination effects by incorporating multiple lines of evidence. This integrated framework aids in diagnosing chemical, ecotoxicological, and ecological conditions, offering a holistic view of environmental health. Moreover, it is particularly valuable in identifying key toxicants and stressors that may not be apparent through predictive models or current monitoring alone [5].

Traditionally, ecological indices such as species diversity and richness have been widely employed as general indicators of ecological status. However, these indices assume that all species contribute equally to the ecological system based solely on their presence-absence and abundance [6]. This assumption may lead to an oversimplification of complex ecological dynamics and potentially obscure the specific impacts of environmental stressors [7].

To address these limitations, more sophisticated trait-based bioindicators have been developed. The Species At Risk (SPEAR) index represents a significant advancement in this field [8–10]. SPEAR integrates trait information, such as physiological sensitivity, into community structure data, allowing

for a more nuanced understanding of ecosystem health, particularly in relation to pesticide and metal contamination.

Another example of a trait-based index is the Benthic Macroinvertebrate Index (BMI), which includes measures such as the Biological Monitoring Working Party (BMWP) [11] and the Korean Benthic macroinvertebrate Index of Biological Integrity (KB-IBI) [12,13]. These indices incorporate trait information on tolerance against biological oxygen demand, providing insights into organic matter pollution and low dissolved oxygen levels in aquatic ecosystems.

The incorporation of trait information in these indices represents a significant improvement over traditional metrics. By considering species-specific characteristics and sensitivities, these advanced bioindicators enable the identification of key stressors responsible for diagnosed ecological impacts [8, 14]. This trait-based approach provides a more comprehensive and accurate assessment of ecosystem health, allowing for more effective environmental management and conservation strategies.

Despite these advancements, the practical application of metal-specific bioindicators, such as $SPEAR_{metal}$, to field datasets has been limited [15]. [16] made a pioneering contribution by applying existing metal sensitivity data to field observations. Their study revealed a discrepancy between laboratory-derived physiological sensitivity and observed field effects, underscoring the complexity of natural ecosystems and the limitations of extrapolating laboratory results to real-world scenarios.

Interestingly, the study by [16] found that predator ratios served as a more meaningful community descriptor in elucidating the effects of metal contamination in field conditions. These findings emphasize the critical need for further validation of the $SPEAR_{metal}$ index across diverse metal-contaminated regions. Such validation would help to establish the robustness and applicability of the index across different ecological contexts and contamination scenarios.

To enhance the reliability of the $SPEAR_{metal}$ index, further validation across diverse metal-contaminated regions is essential, along with the incorporation of additional sensitivity data, particularly for a wider range of taxa, with an emphasis on insect species [16]. Expanding the sensitivity database would allow for the derivation of more refined sensitivity values at specific taxonomic levels, potentially increasing the precision and broadening the applicability of the $SPEAR_{metal}$ index in different ecological contexts.

This study also calculates and compares the $SPEAR_{metal}$ index across different sections of the watershed, including the main stream and tributaries. Such site-specific analyses provide a deeper understanding of the spatial variability in metal contamination impacts, potentially revealing distinct contamination patterns, identifying areas of concern, and highlighting zones of ecological resilience within the watershed. This refined trait-based approach would better bridge the gap between laboratory-derived data and field observations, ultimately leading to more effective environmental management strategies.

The present study aims to validate the $SPEAR_{metal}$ index as an effective tool for assessing the ecological impacts of metal contamination in aquatic ecosystems. Specifically, the refined $SPEAR_{metal}$ index is applied to the upper Nakdong River, a region significantly affected by industrial activities, including zinc smelting and mining operations. This refined index incorporates newly obtained acute toxicity data from 20 indigenous species of the Korean peninsula, thereby enhancing its ecological accuracy and regional applicability. By evaluating the $SPEAR_{metal}$ index across various sections of the watershed, including the main stream and tributaries, this study offers a comprehensive analysis of metal-induced ecological impacts and identifies key stressors and pollution sources within this critical river system in South Korea.

2. Materials and Methods

2.1. Laboratory Acute Toxicity Test

2.1.1. Test Chemical and Test Organisms

Copper(II) chloride dihydrate ($\text{CuCl}_2 \cdot 2\text{H}_2\text{O}$, CAS No. 10125-13-0, > 99% purity) was purchased from Sigma (St. Louis, MO). Acute toxicity tests for copper were conducted with 20 indigenous species from Korean freshwater systems. These species belong to four taxonomic groups: Mollusca, Crustacea, Rotifera, and Insecta (further divided by order). The tested species included mollusks (*Limnoperna fortunei*, *Semisulcospira gottschei*, *Corbicula leana*, *Radix auricularia*, *Pomacea canaliculata*), crustaceans (*Gammarus sobaegensis*, *Moina macrocopa*, *Acanthocyclops vernalis*, *Branchinella kugenumaensis*), the rotifer (*Brachionus calyciflorus*), and insects: dipterans (*Glyptotendipes tokunagai*, *Chironomus flaviplumus*), trichoptera (*Lepidostoma albardanum*, *Hydropsyche orientalis*), ephemeroptera (*Baetis fuscatus*, *Ephemera orientalis*, *Epeorus pellucidus*), and odonates (*Anax parthenope*, *Sieboldius albardae*, *Ophiogomphus obscurus*). *B. kugenumaensis*, *G. tokunagai* and *M. macrocopa* were obtained by subculture from EH Research & Consulting Co. respectively. Other organisms were collected from Korean freshwaters and used for acute toxicity testing after acclimatization in dechlorinated tap water for 7 days. *G. sobaegensis*, *H. orientalis* and *L. albardanum* were acquired from the Gapyeong Stream in Gyeonggi-do. *A. vernalis*, *R. auricularia* and *P. canaliculata* were obtained from the Bernere Stream located in Bucheon City. *L. fortunei* was obtained from Namhan River in Chungcheongbukdo. Other taxa, including *L. fortunei*, *S. gottschei*, *C. leana*, *B. calyciflorus*, *C. flaviplumus*, *B. fuscatus*, *E. orientalis*, *E. pellucidus*, *A. parthenope*, *S. albardae*, and *O. obscurus*, were all collected from the Namhan River.

2.1.2. Toxicity Test Method

The water acute toxicity tests were conducted in the laboratory for twenty indigenous species from Korean freshwater systems. Test durations were 24 hours for *M. macrocopa*, *B. kugenumaensis*, and *B. calyciflorus*, and 96 hours for the remaining species. No food was provided during the exposure period. The stock solution was prepared by dissolving appropriate amounts of CuCl_2 in distilled water. Test concentrations were determined through preliminary experiments to achieve a mortality rate between 20% and 80%. Environmental factors, such as water hardness and temperature, were calculated as geometric means at the start and end of the tests. The tests were conducted following established guidelines, including OECD, US EPA, and others [17–24]. In acute toxicity testing, environmental factor, such as hardness and temperature, were determined as the geometric means of values measured at the beginning and end of the test. The results of the toxicity tests are summarized in Table S1.

2.2. Field Monitoring Data

The study area was centered on the smelter in Andong City, Korea. Since this area has a long history of mining and smelter, massive heavy metal contamination was found to extend 70 km downstream in the mainstream from the smelter. To assess the ecological impact of metal contamination, we utilized data on metal concentrations in sediments and surface water from Chung et al. (2024). Additionally, data on benthic macroinvertebrates were obtained from the National Aquatic Ecological Monitoring Program (NAEMP) database, managed by the National Institute of Environmental Research (NIER), South Korea. Benthic macroinvertebrates were sampled biannually (spring and fall) across the major river watersheds following the NAEMP protocol [25]. Data on sediments and surface water were integrated based on the sampling sites for benthic macroinvertebrates, resulting in a dataset comprising seven sites along the main stream and eleven sites in upstream tributaries.

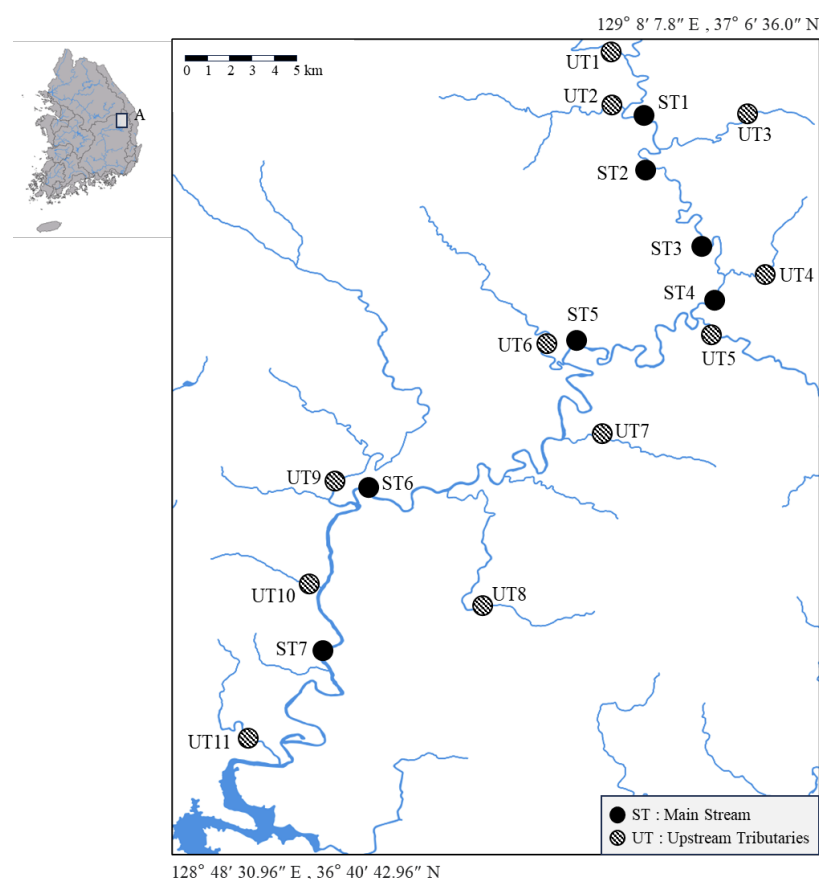


Figure 1. Location map of sampling stations.

2.3. Data Analysis

2.3.1. Calculation of Hazard Quotients (HQ) and Probable Effect Level Quotients (PELq)

To assess the toxic impacts of hazardous substances in the upstream Nakdong River, we calculated the Hazard Quotients (HQ) for surface water and the mean Probable Effect Level Quotients (mPELq) for sediment. HQ for surface water were calculated by dividing the measured concentrations of metals in surface water samples by the respective US EPA water quality criteria for each metal. This calculation provides a quantitative measure of the potential ecological risk posed by these metals. For metals such as Cu, Zn, Ni, Cd, and Pb, hardness-adjusted criteria values were utilized, based on the average hardness of 60 mg/L in the upstream Nakdong River. The adjustment was performed according to the formula provided by the US EPA for calculating hardness-adjusted criteria:

$$CMC_{(dissolved)} = e^{(m_A[\ln(hardness)] + b_A)} CF \quad (1)$$

where $CMC_{(dissolved)}$ is the criterion maximum concentration for dissolved metals, m_A is the slope of the hardness regression for the specific metal, b_A is the y-intercept of the hardness regression for the specific metal and CF is the conversion factor to convert from the total to the dissolved concentration. This adjustment is important because the toxicity of these metals can vary significantly with changes in water hardness, influencing their bioavailability and potential impact on aquatic life [26–28]. In contrast, for As, the criteria values were applied without any adjustment for hardness, reflecting the approach used in the assessment without considering hardness effects.

The mean Probable Effect Level quotient (mPELq) for sediments was calculated using the PEL quotient (PELq) values for each metal. PEL values for each metal in freshwater sediment were obtained from [29]. The PELq was determined by dividing the measured concentration of each metal by its

corresponding PEL value and then averaging these quotients across all metals present in the sediment sample. This method allowed for a quantitative assessment of metal concentrations in sediment samples, taking into account the specific PEL thresholds for each metal.

2.3.2. Community Indices

To assess the biodiversity and species richness of the upstream Nakdong River, we calculated several ecological indices, including taxon richness (TR), EPT taxa richness (EPT), Shannon's diversity index (H'), and the Benthic Macroinvertebrate Index (BMI). Taxon richness (TR) was determined as the total number of distinct taxa collected at each monitoring site. EPT taxa richness (EPT) was calculated as the number of species belonging to the orders Ephemeroptera, Plecoptera, and Trichoptera within the same sites. Shannon's diversity index (H') was computed using the following formula:

$$H' = - \sum p_i \ln p_i \quad (2)$$

where p_i represents the proportion of individuals belonging to species i .

The Benthic Macroinvertebrate Index (BMI) was calculated using the following formula:

$$BMI = \left[4 - \frac{\sum_{i=1}^n s_i h_i g_i}{\sum_{i=1}^n h_i g_i} \right] \quad (3)$$

where n is the number of taxa, s_i is the unit indecision index of species i , h_i is the appearance value of species i , and g_i is the indicator weight of species i . The indicator weight of each taxon was determined according to the NAEMP protocol [13,25]. These indices provided insight into the macroinvertebrate community structure and overall ecological health of the streams investigated. The calculated indices are summarized in Table S2.

2.3.3. Calculation of Metal Sensitivity Value

In order to expand the applicability of the existing SPEAR for metal, it was necessary to improve the sensitivity value. A wide range of test data from the Ecotoxicology Database System (ECOTOX; <http://www.epa.gov/ecotox>) on March 14, 2024 was collected, and a total of 12,108 toxicity test records for 14 metal ions, including 12,108 toxicity test records for 13 metal ions, including Al, As, Cd, Co, Cr, Cu, Fe, Hg, Mn, Ni, Pb, Se, and Zn were obtained. Additionally, toxicity data from 20 species generated in our own experiments were included in the analysis.

The sensitivity index of freshwater macroinvertebrates to metals was determined according to the methodology outlined by [15]. The dataset underwent several preprocessing steps, including removing duplicates, filling missing values, and standardizing units. Taxonomic inconsistencies were harmonized to ensure consistency across studies, and the data were limited to freshwater macroinvertebrates. Furthermore, test data generated under different test conditions within the collected dataset were normalized to standard laboratory conditions (i.e., under 48 hours of exposure time, 20°C of temperature, and 50 mg/L CaCO₃ of hardness assumed in [15]). These standard laboratory conditions were normalized according to the following equation:

$$\log LC50_{norm} = \log LC50_{original} \frac{t}{0.5} \frac{0.58}{12.52 H^{-0.79}} \frac{0.33}{3 e^{-0.11 T}} \quad (4)$$

where $LC50_{norm}$ is the normalized LC50, $LC50_{original}$ is the original LC50 the literature, the t is exposure time, H is hardness values (mg CaCO₃/mL), and T is temperature during toxicity test.

The log-transformed normalized concentrations for multiple metals of the same aquatic organism and heavy metal were averaged. These values were then standardized according to:

$$S_{i(spe)} = \frac{\log LC50_{i(spe)} - \pi_i}{\sigma_i} \quad (5)$$

where LC50 is the species (spe) toxicity value, π_i the mean, and σ_i is the standard deviation for each heavy metal (i).

All data transformations, standardizations, and statistical analyses were conducted using the methodology described by [15] in the free open-source software R (version 4.3.3). To obtain an aggregated sensitivity value, we used a linear mixed-effect (LME) model to test for significant differences in $S_{i(spe)}$ values (species sensitivity to heavy metals), accounting for the unequal number of species representing each heavy metal, which created an unbalanced dataset. For this analysis, we limited the integration to eleven heavy metals, excluding Al and Mn, as they were deemed unsuitable for the aggregated S_{metal} calculations (Table S1, Figure S1).

2.3.4. Calculation of the SPEAR for Metals

The SPEAR (SPECies At Risk) index was developed through a dichotomous classification of taxa into “species at risk” and “species not at risk.” This classification is based on four biological traits: physiological sensitivity to metal contaminants, generation time, exposure potential during periods of maximum contamination, and migration ability [30].

Physiological sensitivity is assigned a score of 1 for taxa with relative sensitivity above a defined threshold, while those below receive a score of 0. Generation time sensitivity scores 1 for taxa with a generation time greater than or equal to a threshold, and 0 otherwise. Exposure sensitivity is scored 1 for epibenthic taxa and 0 for sediment-dwelling taxa. Lastly, migration sensitivity is 0 for organisms known to migrate rapidly, and 1 for all others. A taxon is defined as “species at risk” only if values for all four traits are equal to 1 [30].

The sensitivity value of freshwater macroinvertebrates to metals was determined using newly calculated values according to the methodology outlined by [15]. For the threshold value of physiological sensitivity, the median sensitivity score of 0.63 was determined by calculating the sensitivity values of 10 classes, 34 orders, and 166 families of resident taxa found in South Korea, obtained from the National Aquatic Ecological Monitoring Program (NAEMP) database (meaning that a taxon is considered sensitive only if its relative sensitivity score exceeds 0.63).

To compute the SPEAR index, the relative abundance of the “species at risk” is calculated for each sampling site. The formula for calculating the SPEAR for metals is given by:

$$SPEAR_{metal} = \frac{\sum_{i=1}^n \log(x_i + 1) y}{\sum_{i=1}^n \log(x_i + 1)} \quad (6)$$

where n is the number of taxa, x_i is the abundance of taxon i , and y is 1 if taxon i is classified as “at risk,” otherwise 0.

This calculation is performed at the lowest taxonomic level identified, extending down to species level where possible, to ensure an accurate representation of ecological risks associated with metal exposure. When data for a particular taxon was absent, sensitivity values were taken from the higher taxonomic levels. The derived SPEAR values allow for assessment of the ecological health of aquatic systems in relation to metal contamination, facilitating the identification of at-risk taxa and informing conservation and management efforts.

2.3.5. Statistical Analyses

Pearson’s correlation analysis was performed to assess the relationships between surface water metal HQ and biological indices, including TR, EPT, H' , and BMI. Additionally, correlations between $SPEAR_{metal}$ and both surface water metal HQ and sediment metal PELq were examined. These analyses aimed to evaluate the response of the $SPEAR_{metal}$ index to different levels of metal contamination in both surface water and sediment.

To investigate the relationships between biological indices and environmental variables, Principal Components Analysis (PCA) was conducted. PCA was chosen due to its effectiveness in describing continuous changes in complex systems characterized by multiple parameters, making it suitable for

this study [31]. PCA can provide insights into the complex relationships between metal contamination and various ecological indicators [32]. The analysis was performed using relevant libraries, following standard protocols for multivariate analysis.

Before conducting the PCA, all biological indices, including SPEAR_{metal}, TR, EPT, H', surface water metal HQ, and sediment metal PEL_q, were standardized to a zero mean and a standard deviation of one, as the indices represent different units of measurement. This standardization ensured comparability across indices, improving the reliability of the results. PCA was applied to the standardized biological indices, with environmental variables passively projected onto the ordination space. This unconstrained approach allowed for the exploration of variability in the biological indices, including variability that was not explained by the environmental variables. The interpretability of the PCA axes was assessed using the Broken Stick criterion [33]. All statistical analyses were conducted employing the scipy libraries for data manipulation and correlation analysis, using Python 3.11.5.

3. Results

3.1. Updated Metal Sensitivity and SPEAR Index Calculation

Metal sensitivity was updated using new toxicity test results for twenty indigenous species in Korea, primarily focusing on insect taxa (Table S1). The total number of families with available toxicity data increased from 122 in 2012 to 411 in the current study. The median value of metal sensitivity shifted from 1.01 in 2012 to 0.63 in the present study (Figure 2). The LME model analysis revealed no statistically significant differences among the rankings of the 11 heavy metals examined (Figure S1 and Table S2). Consequently, an overall heavy metal sensitivity value (S_{metal}) was calculated for each taxon (Table S2 and S3). The most sensitive species were identified as *Villorita cyprinoides* (-2.89), *Physella gyri* (-2.58), and *Leptodea leptodon* (-2.00), with families Mycetopodidae (-1.88) and Hyalellidae (-0.97) also showing high sensitivity. Among insect orders, Coleoptera (-0.31) and Ephemeroptera (-0.009) were most sensitive, while Odonata (1.82) were relatively tolerant. Other taxa such as Plecoptera (1.18) and Trichoptera (1.19) exhibited intermediate sensitivity.

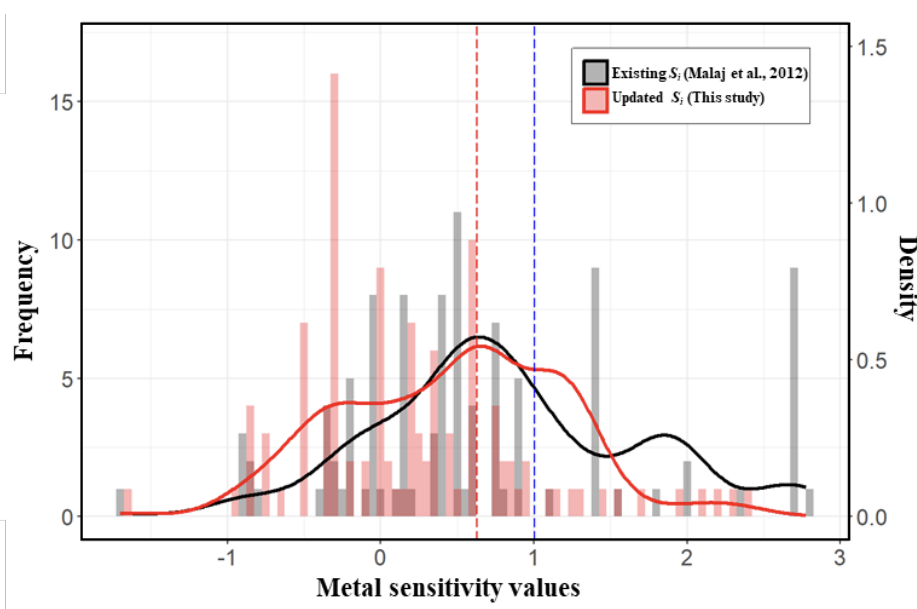


Figure 2. Histogram of the metal sensitivity values for SPEAR. The black bar and solid line were the frequency of sensitivity values using the updated ecotoxicity dataset. The red bar and solid line were the frequency of sensitivity values in the existing SPEAR for metal.

In the upper Nakdong River, 81 families across 6 classes of species were recorded, with Insecta representing the highest abundance at 94%. The most abundant orders within Insecta were Ephemeroptera

(50%), Diptera (17%), and Trichoptera (8.1%). The taxon-specific coverage of the metal sensitivity index in this study accounted for 46% of the recorded taxa, compared to 31% in the previous sensitivity index reported by [15]. Insecta had the lowest coverage, with 37% of the 63 families covered in this study, while the previous research reported only 24% coverage. Other taxa in this study showed coverage rates exceeding 60% (Figure 2).

Figure 3 illustrates the impact of the updated ecotoxicity database on taxon coverage and metal sensitivity. The percentage of families with available toxicity data increased from 44% to 63% of the total number of families recorded in the study area (Figure 3-a). Figure 3-b demonstrates the correlation between metal sensitivities derived from the existing and updated ecotoxicity databases.

The analysis of sensitivity indices revealed distinct patterns across different datasets and stream types. For the whole dataset, the median value of S_i -2024t exceeded that of S_i -2012, indicating an overall increase in sensitivity with the updated index. In the main stream, S_i -2024t demonstrated a higher median value compared to S_i -2024s. However, this difference was not observed in the tributaries, where the median values of S_i -2024t and S_i -2024s were comparable.

For instance, in the Plecoptera order, the previous study assigned a uniform sensitivity value (1.39) to all lower taxa, whereas our study calculated different sensitivity values for each family: Chloroperlidae (0.78), Nemouridae (0.42), Perlodidae (1.45), and Perlidae (2.27). Conversely, Capniidae, Leuctridae, and Taeniopterygidae were assigned the sensitivity value of the higher taxonomic group (1.18). The median sensitivity value for species in the upper Nakdong River was 0.63 in our study, compared to 1.01 derived from the previous sensitivity index.

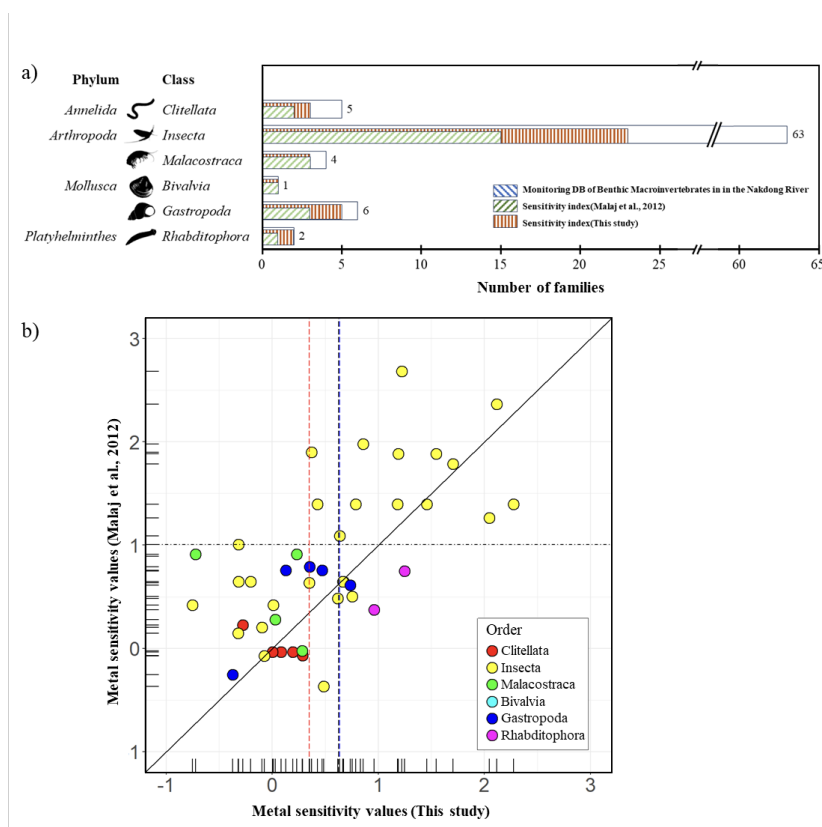


Figure 3. Comparison of metal sensitivity values using the existing [15] and updated (this study) SPEAR indices in the upper Nakdong River. (a) Number of macroinvertebrate families in the upper Nakdong River and those with available ecotoxicity data. (b) Scatter plots comparing metal sensitivity values between the existing and updated indices. Black dashed lines represent the median metal sensitivity of the existing index. Red dotted lines indicate the mainstream, and blue dotted lines indicate the tributaries.

3.2. Association between Metal Concentrations and SPEAR Index

The updated SPEAR index values for whole dataset (SPEAR-2024t) based on the new metal sensitivity data were lower than those derived from the previous version (SPEAR-2012) (Figure 4-a). The median values were 0.70 for SPEAR-2012 and 0.56 for SPEAR-2024t, with a mean difference of 0.14 ± 0.06 (mean \pm SD). While most SPEAR-2012 values exceeded 0.5, the majority of SPEAR-2024 values in the main stream fell below the severe level of SPEAR index (0.5) [10,34]. SPEAR-2024s (median value: 0.64) calculated separately for main stream and tributaries, yielded lower values than SPEAR-2024t (0.56), with a mean difference of 0.13 ± 0.15 . A significant difference was observed between SPEAR-2024t (0.56) and SPEAR-2024s (0.45) in the main stream (mean difference: 0.11 ± 0.0085), but not in the tributaries (0.04 ± 0.0042). These results mean that the separating calculation of SPEAR index is more sensitive to metal contamination.

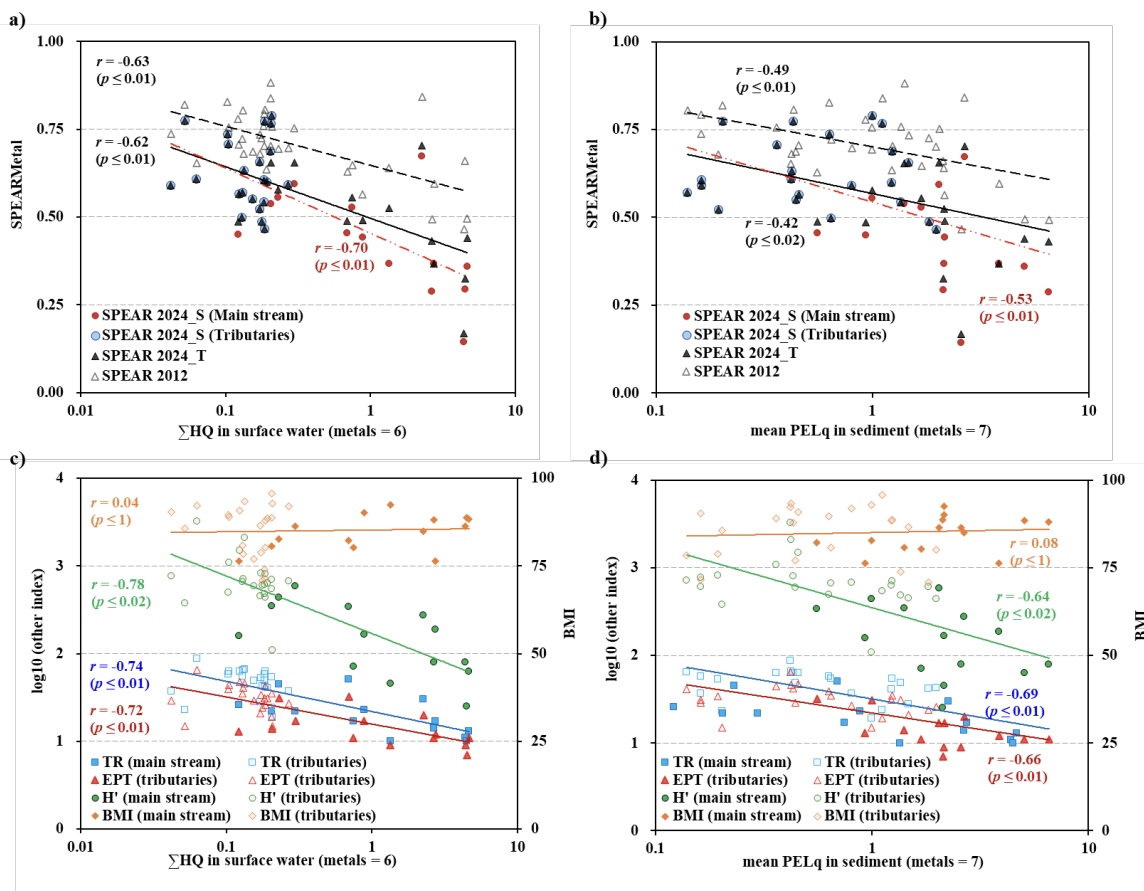


Figure 4. Site-specific concentration and ecological impact relationship between the sum of HQ or mean PELq and the ecological indices for macroinvertebrate community structure (TR, EPT, H', BMI and SPEAR for metals) in the upper Nakdong River. TR: taxa richness, H':Shannon diversity, EPT: Ephemeroptera, Plecoptera, and Trichoptera taxa, BMI: Benthic Macroinvertebrate. Index, Linear relationships between the logarithm of the sum of toxic unit (TU) and SPEAR index calculated using the existing (SPEAR-2012) and newly updated (SPEAR-2024) taxon-specific metal sensitivity. The SPEAR-2012 were calculated for the whole data, but SPEAR-2024 were calculated for the whole data (SPEAR-2024t) and the main stream (SPEAR-2024s), separately.

The correlation coefficients (r) between SPEAR indexes and the sum of HQ in surface water were -0.63 (SPEAR-2012), -0.62 (SPEAR-2024t), and -0.70 (SPEAR-2024s). The refined SPEARs showed a significant correlation with both the sum of HQ for six metals in water and mPELq for seven metals in sediment (Figure 4-b). The correlation coefficients (r) between SPEAR indexes and mPELq in sediment were -0.49 (SPEAR-2012), -0.42 (SPEAR-2024t), and -0.53 (SPEAR-2024s). The correlation coefficients (r)

between SPEAR-2024s were greater than those of SPEAR-2012 and SPEAR-2024t. These results mean that the separating calculation of SPEAR index is more specific to metal contamination.

3.3. Association between Metal Contamination and Generic Ecological Indices

The relationship between metal contamination and various ecological indices was examined to assess the specificity of observed impacts. Generic ecological indices, including TR, EPT, H', showed significant negative correlations with both ΣTU and mPELq (Figure 4c and 4d). Specifically, sites with higher metal concentrations exhibited reduced species richness ($r = -0.74$, $p < 0.01$), a lower proportion of EPT taxa ($r = -0.72$, $p < 0.01$), and decreased H' values ($r = -0.78$, $p = 0.02$) compared to uncontaminated sites.

These correlations suggest a selective disappearance of metal-sensitive species from contaminated areas. In contrast, the Benthic Macroinvertebrate Index (BMI), which is primarily sensitive to organic pollution, showed no significant correlation with metal contamination levels ($r = 0.11$, $p > 0.05$). This lack of association between BMI and metal contamination indicators (ΣTU and mPELq) suggests that organic matter pollution was not a significant stressor in the studied system, further emphasizing the specificity of the observed ecological impacts to metal contamination.

3.4. Relationship between Individual Metal Concentration and SPEAR Index

In the main stream samples, Cd and Zn were identified as potential key toxicants in surface water through HQ analysis (Chung et al., 2024). SPEAR-2024s values for metals in the main stream and whole dataset also showed significant correlations with HQ values of Cd and Zn, with HQ ranges from 0.02 to 3.1 (Figure 5-a). Other metals did not show significant relationships with SPEAR-2024s. In sediment samples, while both Cd and As were previously identified as key toxicants contributing to sediment toxicity, Cd and Zn displayed a significant relationship with the $SPEAR_{metals}$ ($r = 0.39$, $p < 0.05$ for Cd; $r = 0.37$, $p < 0.05$ for Zn). As did not show a significant correlation. Notably, Cd PELq values exceeded 1 at several sites, while As PELq values remained below 1. Zn, despite PELq values below 1, demonstrated a significant relationship with the SPEAR index ($r = 0.08$, $p > 0.1$).

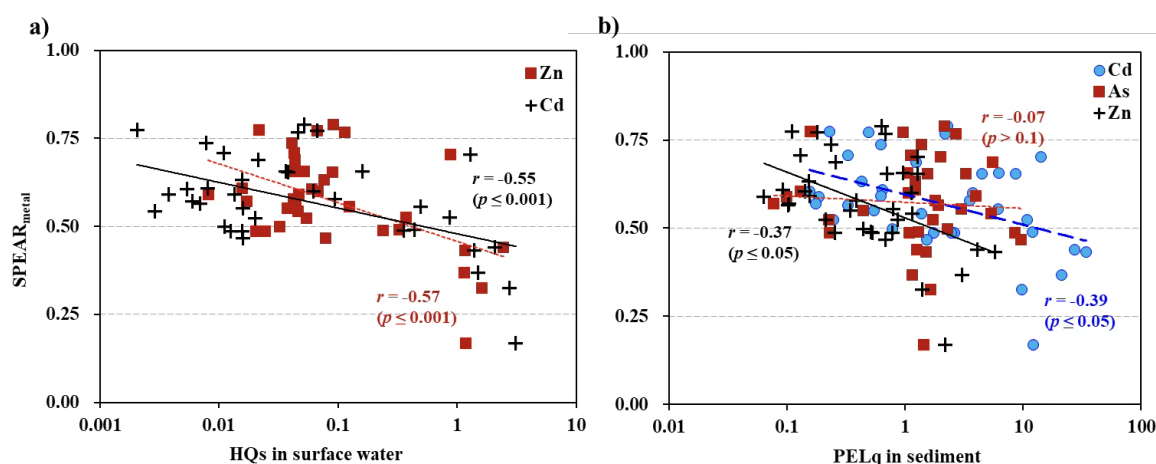


Figure 5. Site-specific concentration and ecological impact relationship between the HQ or mean PELq of individual metals and SPEAR for metals in the main stream of the upper Nakdong River.

A clear gradient of decreasing SPEAR index values was observed along the main stream, correlating with increasing distance from the Zn smelter. Near the smelter, HQ values exceeded 1, with both the sum and maximum HQ values following a similar pattern. Specifically, HQ values for Zn and Cd were greater than 1 in proximity to the smelter. In contrast, SPEAR values in the tributaries showed no discernible pattern, and the sum of HQ values remained below 1 (Figure A2).

In conclusion, Cd and Zn were identified as the primary stressors responsible for the ecological impacts of metal contamination in the upper Nakdong River near and downstream of the Zn smelter.

3.5. Multivariate Statistical Analysis to Analyze the Association Between Environmental Factors and Community Indices

Principal Component Analysis (PCA) was employed to elucidate the associations between environmental factors and community indices. Figure 6 presents the ordination plots from the PCA, illustrating the relationships between ecological indices and environmental factors for the total samples from the main stream and tributaries. This visual representation aids in the interpretation of the complex interactions between various environmental parameters and community indices in the studied ecosystem. These findings highlight the value of using multiple indices, including both generic and stressor-specific indicators like SPEAR, in comprehensively assessing the ecological impacts of metal contamination in aquatic ecosystems.

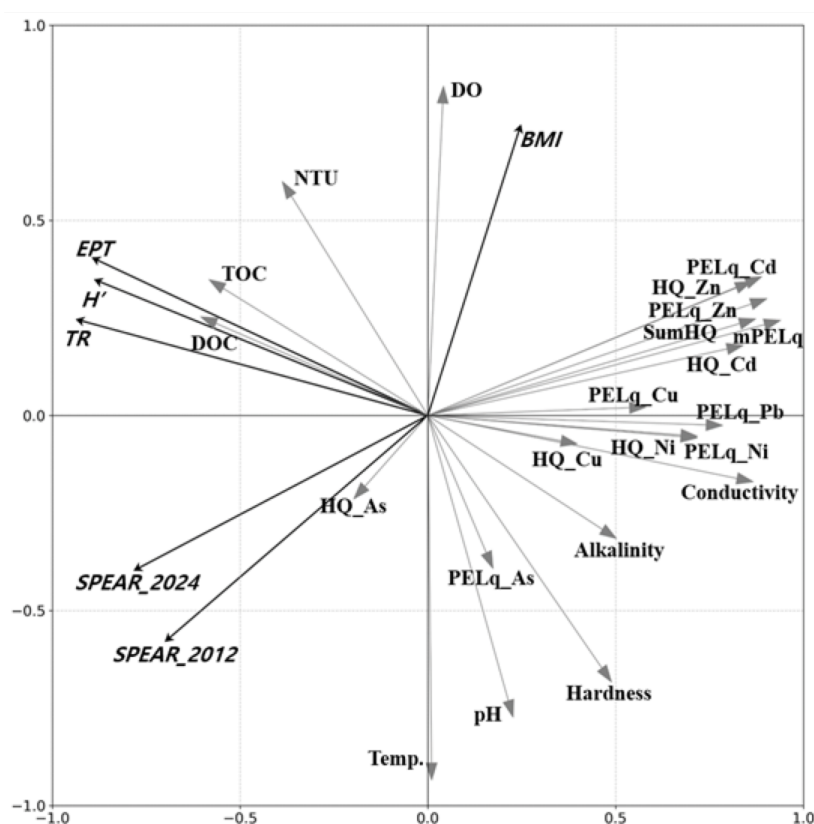


Figure 6. Ordination plot for the principle component analysis of the ecological indices and environmental factors projected into the ordination model for the total samples from the main stream and tributaries. TR: taxa richness, H':Shannon diversity, EPT: Ephemeroptera, Plecoptera, and Trichoptera taxa, BMI: Benthic Macroinvertebrate Index, SPEAR-2012: SPEAR for the existing metal sensitivity [15], SPEAR-2024s: SPEAR calculated in this study, sumHQ: the sum of Hazard Quotient, mPELq: mean Probable Effect Level quotient, and Temp: Temperature.

The first principal component (PC1) demonstrated strong negative associations with most community indices, except for BMI. PC1 also showed negative correlations with metal contamination indicators, with the exception of As, and with conductivity. Notably, BMI did not show a significant association with PC1, suggesting its independence from metal contamination effects.

The second principal component (PC2) exhibited a strong positive association with BMI and weak positive associations with other community indices. Interestingly, generic community indices (TR, EPT, and H') showed weak positive associations with PC2, while SPEAR indices demonstrated

weak negative relationships. PC2 was positively correlated with dissolved oxygen (DO) and turbidity (NTU), and weakly associated with total organic carbon (TOC), dissolved organic carbon (DOC), Cd (both HQ-Cd and PEL_q-Cd), Zn (both HQ-Zn and PEL_q-Zn), ΣHQ, and mPEL_q. Conversely, PC2 showed negative relationships with temperature, pH, and water hardness.

The PCA results also revealed that metal contamination, particularly by Cd and Zn, had weak negative relationships with SPEAR indices but weak positive relationships with other community indices. This differential response highlights the specificity of the SPEAR indices in detecting metal contamination effects.

In summary, the multivariate analysis yielded the following key findings: (1) BMI did not show any correlation with metal contamination, but positive correlations with NTU and DO, weak positive associations with DOC and TOC, and negative correlations with temperature, pH, and water hardness. (2) Other community indices (TR, EPT, and H') were associated with metal contamination as well as conductivity and alkalinity. (3) SPEAR indices demonstrated strong associations with metal contamination, particularly with Cd and Zn. The main stressors, such as Cd and Zn, showed stronger associations with SPEAR indices compared to generic community indices, underscoring the specificity of SPEAR in detecting metal contamination effects.

3.6. Discussion

This study provides a comparative analysis of two fundamental approaches to assessing ecological risks from metal contamination: toxicological methods and eco-epidemiological techniques. Each method offers distinct advantages that, when integrated, allow for a more comprehensive understanding of how metal pollution affects aquatic ecosystems [35,36].

Toxicological approaches, such as the calculation of HQ based on controlled experiments, offer predictive insights into ecological impacts through dose-response relationships. While these methods are powerful for identifying potential risks, they may not fully account for the complex, real-world interactions among multiple stressors [35,37].

Eco-epidemiological methods, on the other hand, provide a retrospective perspective by focusing on the observed ecological state of ecosystems affected by contamination. These approaches, particularly through the use of bioindicators, can directly reflect the cumulative impact of stressors like metals on community structure and function [15,16,28,38].

Trait-based bioindicators, such as SPEAR index, add significant value to eco-epidemiological methods by targeting species traits sensitive to specific stressors. The SPEAR index has been shown to offer a more precise and focused diagnosis of metal-induced effects in aquatic ecosystems. However, limitations exist, particularly the potential for confounding factors, such as the presence of other pollutants or environmental variables, that could obscure the association between metal exposure and ecological impact [15,16].

The application of the SPEAR index for metals in the upper Nakdong River watershed provided significant insights into ecosystem responses to metal contamination. Both existing and updated SPEAR_{metal} indices showed robust associations with metal contamination levels in water (HQ) and sediment (mPEL_q) (Figure 4). Notably, this study is the first to demonstrate a correlation between SPEAR_{metal} and metal contamination in water and sediment, expanding the index's utility beyond water column assessments. This relationship suggests that SPEAR_{metal} is sensitive to integrated metal exposure effects from both water and sediment compartments, reflecting long-term exposure and bioaccumulation processes.

These findings underscore the robustness of SPEAR_{metal} as an indicator of metal contamination across different environmental compartments, enhancing its ecological relevance and potential for holistic ecosystem health assessment. The index's ability to bridge the gap between chemical monitoring data and ecological effects offers a more integrative approach to assessing metal impacts in aquatic ecosystems. Further research is needed to explore the mechanistic basis of the relationship

between SPEAR_{metal} and sediment contamination, and to validate these findings across diverse aquatic ecosystems with varying sediment characteristics and metal contamination profiles.

The updated metal sensitivity database, encompassing a substantial portion of families in the study area, enhanced the index's sensitivity (Figure 4). This improvement allowed for the detection of subtle shifts in community structure, which is crucial for early warning in environmental monitoring programs. A key finding was that calculating the SPEAR index separately for main streams and tributaries increased its sensitivity and diagnostic power. This result underscores the importance of considering habitat-specific variations when applying ecological indices, aligning with research advocating for the subdivision of study areas based on ecological or hydrological characteristics [34].

To elucidate the complex interactions between various environmental factors and metal contamination, Principal Component Analysis (PCA) was employed. This multivariate approach revealed intricate relationships between metal contamination and other environmental variables, providing a more comprehensive understanding of how multiple factors shape aquatic community structures [39,40].

The study highlights the importance of incorporating uncertainty analysis in the application of the SPEAR index. Reporting the percentage of family coverage in the watershed would provide transparency and improve decision-making regarding the index's reliability in different ecological contexts [15].

Interestingly, the predator ratio, previously proposed as a diagnostic tool for metal contamination, did not correlate with either the SPEAR index or the sum of hazard quotients. This discrepancy suggests that different bioindicators may respond differently depending on the ecosystem and stressor types, reinforcing the need for multiple lines of evidence in ecological assessments [2,41].

While the SPEAR index performed well in this study, the homogeneity of water chemistry across the watershed may have limited the role of metal bioavailability. This observation suggests the need for further validation in watersheds with more varied physicochemical conditions [42].

Future research directions should focus on validating the SPEAR index across watersheds with diverse habitat conditions, water chemistry profiles, and baseline community structures. Investigating the relationship between metal bioavailability and SPEAR index performance in different chemical environments, such as those with varying pH or dissolved organic carbon levels, would offer valuable insights. Furthermore, integrating the SPEAR index with other ecological indicators could yield a more holistic assessment of ecosystem health, potentially leading to the development of multi-metric indices that are more adaptable to specific environmental contexts.

3.7. Conclusions

This study demonstrates that the SPEAR index for metals is an effective and sensitive tool for evaluating the ecological impacts of metal contamination in aquatic ecosystems. The index's strong correlation with metal exposure concentrations and its ability to detect subtle changes in community structure make it a valuable bioindicator for environmental monitoring, particularly in watersheds impacted by multiple pollution sources such as zinc smelting and mining activities. While the SPEAR index has shown significant potential, further research is needed to validate its applicability across diverse environmental conditions. The findings underscore the importance of refining the index for use in varied ecological contexts and integrating it with other ecological indicators to create a more comprehensive tool for ecological risk assessment. Ultimately, the development and application of such refined bioindicators will enhance our ability to monitor, assess, and manage metal contamination in aquatic ecosystems more effectively.

Supplementary Materials: The following supporting information can be downloaded at the website of this paper posted on [Preprints.org](https://www.preprints.org). Table S1: Summary of acute toxicity test results for selected freshwater species exposed to copper; Table S2: Summary of results for the fixed factors of the mixed effect model; Table S3: Physiological sensitivity values for As, Cd, Co, Cr, Cu, Fe, Hg, Ni, Pb, Se and Zn for each taxonomic level; Table S4: Summary of site-specific environmental parameters, macroinvertebrate community indices, and toxicity values in the Upper Nakdong River; Table S5: Information on benthic macroinvertebrate species found in the upper Nakdong River; Figure S1: Pairwise mean comparison of physiological sensitivity values for each taxon of the eleven heavy

metals expressed as S_{As} , S_{Cd} , S_{Co} , S_{Cr} , S_{Cu} , S_{Fe} , S_{Hg} , S_{Ni} , S_{Pb} , S_{Se} and S_{Zn} . Histograms display the distribution of standardized sensitivity values for each heavy metal. The upper panel shows the Pearson correlation coefficient (r) and the number of pairwise comparisons (n), while the lower panel presents the plotted values for each pair of heavy metals.

Author Contributions: DS.Hwang conceived the study, developed the methodology, performed formal analysis, wrote the original draft, and created the visualizations. JW.Kim conducted data analysis, performed statistical analysis, and contributed to the creation of visualizations. JW.Chung performed ecotoxicological tests, analyzed data, and contributed to the creation of visualizations. JH.Lee contributed to the methodology, conducted formal analysis, and reviewed and edited the manuscript.

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Data Availability Statement: The data and source code for statistical analysis that support the findings of this study are available from the corresponding authors upon reasonable request.

Conflicts of Interest: The authors declare no competing interests.

Appendix A. Detail Results of Data Analysis for Individual Metals

Appendix A.1. Gradient of Metal Concentration and SPEAR for Metals in the Main Stream and Tributaries

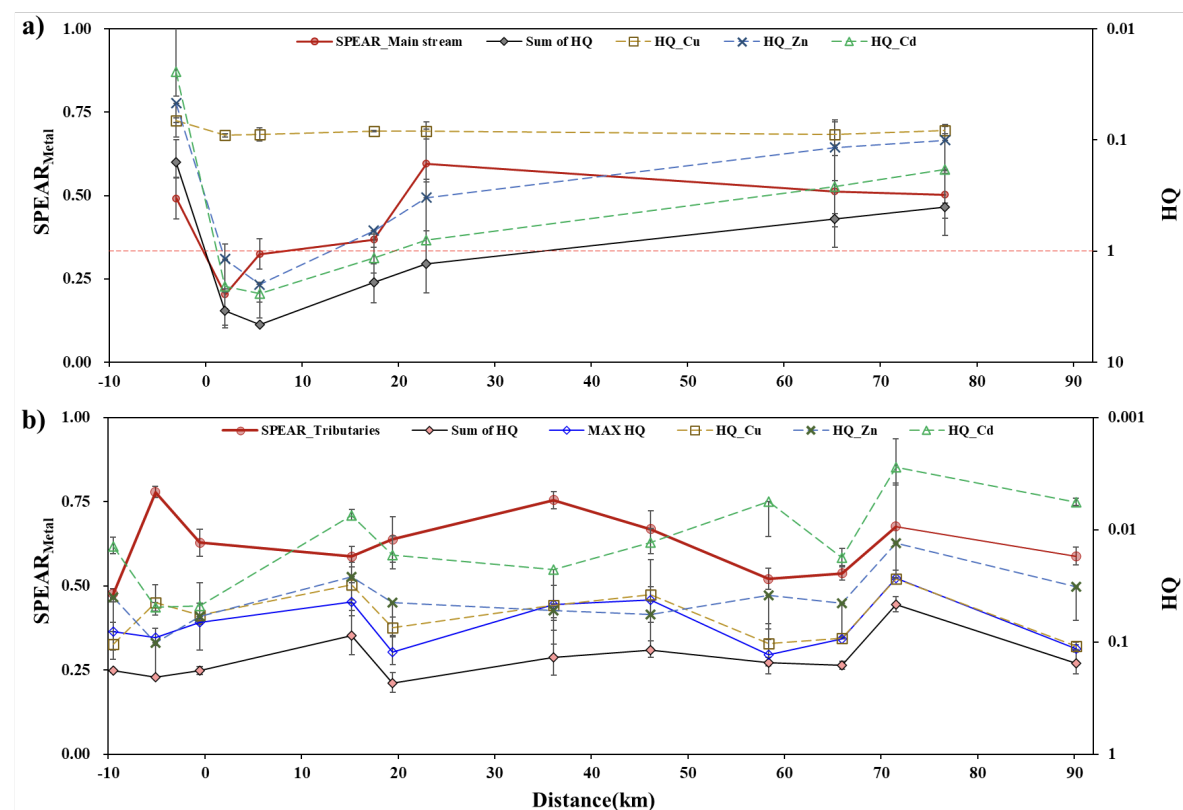


Figure A1. Gradient of metal concentration and SPEAR for metals in the main stream and tributaries.

Appendix A.2. Relationship of the Sum of Toxic Unit and SPEAR for Metals and Predator Ratio

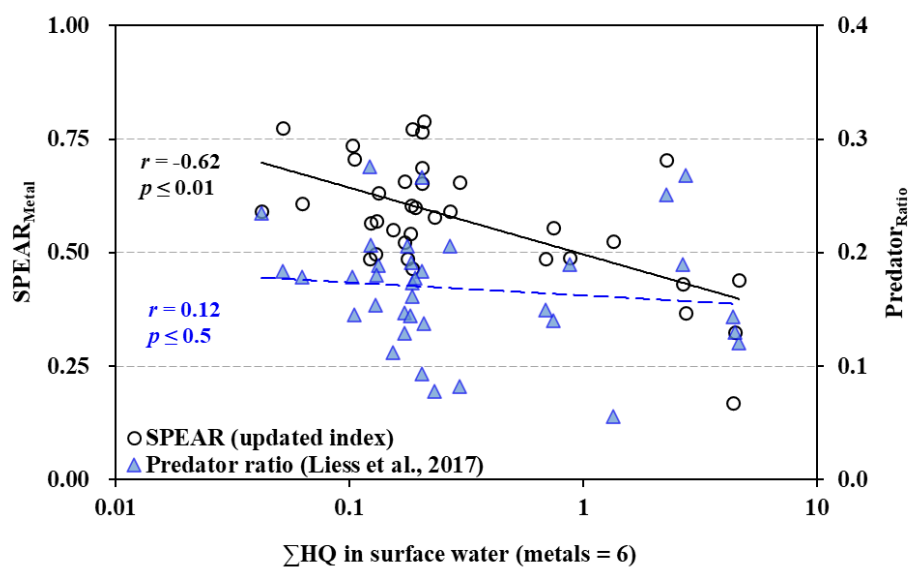


Figure A2. Relationship of the sum of toxic unit and SPEAR for metals and predator ratio.

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