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Article

Bioindicators for Assessing Soil Quality in Ecuador's Jun Jun Micro-Water Shed

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Abstract: The evaluation of soil quality in different altitudinal gradients, with vegetative and non-vegetative subareas, is crucial for proper soil functioning and optimal crop growth, thus contributing to the sustainability of agroecosystems. Although the altitudinal gradient significantly influences soil quality, the ability to predict this quality, expressed through an index, in soils with and without vegetative cover, is still insufficiently explored. This study employs the Simple Additive Index (SQI) method to evaluate soil quality in Ecuador's Jun Jun micro-watershed, Tungurahua region. Three altitude categories (<2800, 2800-2900, >2900 masl) were investigated, with 24 soil samples collected across both vegetated and non-vegetated areas. The indicators used included chemical and biological parameters such as soil organic carbon (OC), earthworm density (WD), earthworm biomass (WB), organic matter (OM), pH, and total nitrogen (TN). The results revealed that in areas with altitudinal gradients below 2800 masl, soil quality index values were higher compared to other altitudes. In vegetated areas, a decrease in index values was observed as the altitudinal gradient increased, indicating a deterioration in soil quality with high altitude. These findings are significant in providing a quantitative assessment of the effects of altitudinal gradient and vegetative cover influence on soil quality.

Keywords: bioindicators; earthworms; micro-watershed; soil quality; vegetation cover

1. Introduction

As the pressure on agriculture increases to meet growing food demands, understanding the interactions between different crops and the physical, chemical, and biological parameters of the soil becomes imperative. Inappropriate soil use can lead to nutrient depletion and a loss of quality, exacerbating soil contamination (Zhang et al., 2022). Soil quality plays a crucial role in the functioning of agroecosystems and in maintaining the health of organism's dependent on it, whether plants, animals, or microorganisms. Soil health and vitality are fundamental elements to ensure food security and the sustained success of agriculture (Bagnall et al., 2021).

In this context, the soil quality index emerges as an essential tool to assess soil capacity and ensure more efficient and sustainable crop management in agroecosystems. Soil fertility, its ability to retain water and nutrients, and its physical structure are determining factors in the viability of life and food production (Y. Li & You, 2022). However, soil quality is a complex concept to estimate, which makes it necessary to use many indicators to determine soil characteristics (Stocking, 2003). These physical, chemical and biological indicators of the soil are measurable parameters that, although they do not directly represent the quality of the soil, offer information about its properties and functions. (Ozsahin et al., 2017). The variety of parameters and indicators highlight the need to consider multiple dimensions to assess soil quality (Fazekašová, 2012), as each indicator offers unique information about specific aspects of the soil, and their combination provides a more comprehensive assessment (AbdelRahman et al., 2019).

The estimation of the soil quality index is a complex and challenging process, given the numerous factors influencing soil quality, such as texture, organic matter, and biological activity (Bonilla-Bedoya et al., 2023). Although considerable progress has been made in estimating the soil

quality index, especially in specific soil contexts and management practices (Armenise et al., 2013), most studies tend to employ a single method, indicating a trend toward simplification and standardization in soil quality assessment (Maaz et al., 2023). Simplification may not fully capture the complexity of agroecosystem soils, resulting in less accurate assessments (Bahena-Orsorio et al., 2023). Furthermore, standardization may not be applicable to all contexts, as soil characteristics can vary considerably from one place to another (Zuber et al., 2020). The difficulty in calculating the soil quality index underscores the importance of developing a credible and user-friendly index (Prior & Hagmann, 2014).

The ongoing degradation of micro-watersheds poses continuous environmental and social problems (Hubanks et al., 2018). Soil degradation contributes to the reduction of biological, chemical, and physical properties in extensive areas with agricultural potential, whose vulnerability increases with the intensification of land use in agricultural activities and other purposes (Magalhães et al., 2023). The evaluation of the soil quality index through indicators such as earthworm density and biomass, organic matter content, hydrogen potential, total nitrogen and organic carbon, becomes crucial to promote sustainable agricultural production by addressing the complex interaction between soil characteristics and crop yields (Isong et al., 2022). This study adopts a comprehensive approach, analyzing multiple altitudinal gradients of the soil and the influence of vegetation cover on crop yield in the Jun Jun micro-watershed in the Ecuadorian region. By considering variability across different soil layers, the aim is to capture the inherent complexity of soil processes in this specific region.

This study aims to assess soil quality and its implications for sustainable crop management in agroecosystems, driven by the increasing pressure on agriculture to meet growing food demands. We seek to understand the interactions between different crops and soil parameters to address issues such as nutrient depletion, quality loss, and soil contamination, ultimately contributing to food security and the sustained success of agriculture. This research was conducted in the Tungurahua region in Ecuador, and the principles and methodologies employed can be adapted and applied to other regions facing similar challenges in soil management and agricultural sustainability. The novelty of this research lies in its comprehensive approach, analyzing multiple altitudinal gradients and vegetation cover influences on crop yield, providing insights into the complex dynamics of soil processes in agroecosystems. It emphasizes the importance of considering various soil parameters, including physical, chemical, and biological indicators, to accurately assess soil quality and inform sustainable agricultural practices. Additionally, it addresses the challenges and limitations in estimating the soil quality index, highlighting the need for a credible and user-friendly index that captures the complexities of agroecosystem soils. By focusing on the Jun Jun micro-watershed in Ecuador, we contribute to understanding soil degradation issues and promote sustainable agricultural production practices in this specific region.

2. Materials and Methods

2.1. Study Area

This study was conducted in the Tungurahua region, Ecuador, where the predominant soils have been shaped by volcanic eruptions and ash deposition, resulting in diverse altitudinal gradients and areas with and without vegetative cover. Vegetative cover influences soil structure, organic matter, and other soil aspects. Comparing soil quality in areas with and without vegetative cover allows evaluating how this influence affects soil's capacity to sustain plant life and provide ecosystem services. Studying soil quality indices with and without vegetative cover in this area is essential for understanding terrestrial ecosystem functioning and managing them sustainably for the benefit of both agriculture and the environment. The topography of the area exhibits distinctive features due to the presence of mountains, which exert a significant influence on temperature, precipitation, and wind speed. This influence creates microclimates that foster a wide diversity of ecosystems (Zehetner & Miller, 2006). The region is characterized by irregular topography, with altitudes ranging from 2700 to 3200 meters above sea level, and has annual precipitation of 549.5 mm, with an average annual

temperature ranging between 7.6 and 18.7 °C (Villacís et al., 2008). The period of most significant precipitation extends from February to July, while the months with lower precipitation levels are between August and January. These soils are primarily used for the cultivation of short-cycle plants, herbs, and pastures, as well as the planting of eucalyptus trees and the formation of Andean grasslands (Buytaert et al., 2007) (Figure 1).

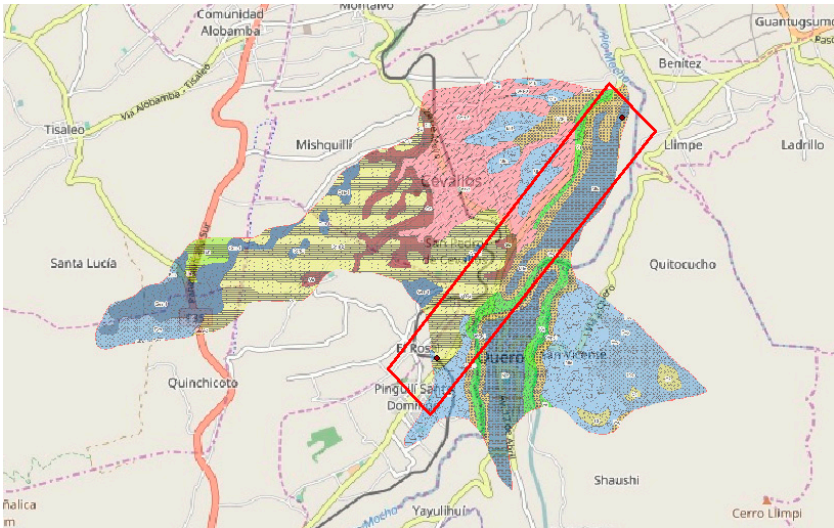


Figure 1. Location of the study area in the Jun Jun micro-watershed in the Ecuadorian region (adapted from Google. (n.d.)).

2.2. Soil Quality Index Calculations

To analyze the soil quality index using the Simple Additive Index (SQI) method, a set of 24 soil samples and a series of soil quality indicators were involved as parameters. The 6 parameters used to develop SQI were: earthworm density (WD), earthworm biomass (WB), organic matter (OM), hydrogen potential (pH), total nitrogen (TN), and organic carbon (OC) (Table 1).

Table 1. Indicators for estimating soil SQI collected under two types of vegetation cover across three altitude categories.

		With plant cover						Without plant cover					
Indicators	Units	< 2800 masl	SD	2800- 2900 masl		> 2900 masl		< 2800 masl	SD	2800- 2900 masl		> 2900 masl	
					SD		SD				SD		SD
pH		7,09	0,11	6,96	0,65	6,80	0,31	7,08	0,31	6,76	0,60	6,77	0,59
OM	%	3,28	0,19	3,04	0,24	2,60	0,13	2,57	0,34	2,95	0,25	2,43	0,22
OC	%	1,26	0,39	1,03	0,22	0,58	0,55	1,14	0,34	0,90	0,14	0,58	0,12
TN	%	0,19	0,08	0,47	0,20	0,40	0,06	0,16	0,13	0,47	0,07	0,50	0,08
WD	worm/m ²	15,25	1,98	13,00	1,00	6,25	0,6	6,50	0,20	6,00	0,23	5,75	0,06
WB	g/m ²	7,45	0,23	5,90	0,70	2,83	0,69	2,85	0,05	2,25	0,02	2,02	0,33

Note: WD: earthworm density, WB: earthworm biomass, OM: organic matter, pH: hydrogen potential, TN: total nitrogen, OC: organic carbon, and SD: standard deviation.

The SQI was estimated following the method described by Mukherjee and Lal, (2014).

$$\sum SQI = \sum Individual\ soil\ parameter\ index\ values$$

Where SQI is the sum of the values of the individual indices included in the estimation of soil parameters. This SQI is useful for evaluating soil quality, despite its sensitivity to the unit values of the parameters it uses.

2.3. Normalizing Soil Indicators

The normalization (0-1) of the SQI value for each individual soil was calculated using the following equation:

$$SQI - 1 = \left(\sum SQI - SQI_{MIN} \right) / (SQI_{MAX} - SQI_{MIN})$$

where SQI-1 is the normalized value of the indicator, SQ_{MIN} is the minimum SQI value, and SQ_{MAX} is the maximum SQI value of the total data set. The ranges of soil parameter values to estimate the SQ_{MAX} and SQ_{MIN} were determined by assigning threshold values primarily based on literature review and expert opinion of the authors (Table 2).

The values of the indicators were normalized using a scale from 0 to 1. SQ_{MIN} and SQ_{MAX} represent the worst and best conditions from the point of view of quality, respectively. In agricultural and degraded soils, the maximum value of the indicator represents the ideal value to be reached or the best soil quality situation. The minimum value represents the minimum desirable or the minimum acceptable quality.

The normalized indicator value is the indicator score, whose scores represent the contribution of each indicator to soil quality. Quality scores were classified on 5 soil class scales, ranging from 0 to 1 (class 1: very high quality, class 2: high quality, class 3: moderate quality, class 4: low quality and classes 5: very low quality) according to sensitivity to soil quality.

Table 2. Soil indicators and SQI threshold values.

Indicator	Units	Range	References
OM	%	2.0-6.0	(Estrada et al., 2017)
		3.5-5.0	(Desta, 2010)
		1.29-4.5	(Akram et al., 2014)
pH	-	5.5-7.2	(Amacher et al., 2007)
		7.2-8.0	
		5.5-7.2	(Mukherjee y Lal, 2014)
		7.2-8.0	
		5.5-7.0	(Cantú et al., 2007)
		5.0-8.5	(Prieto et al., 2013)
		5.5-7.0	(Cantú et al., 2009)
TN	%	0.2-0.3	(Feiza et al., 2011)
		0.1-0.5	(Amacher et al., 2007)
		0.09-0.12	(K C, 2013)
WD	Number/square foot ²	5-15	(Desta, 2010)
		15-70	(Demetrio et al., 2019)
WB	g/m ²	1.82-9.66	(Morel y Acosta, 2022)
		2	
		2.0-3.0	(Feiza et al., 2011)
OC	%	0.6-2.5	(M. P. Cantú et al., 2007)
		0.6-2.5	(M. Cantú et al., 2009)
		1.0-1.5	(Prieto et al., 2013)
		1-1.5	(Amacher et al., 2007)

Note: Range: Threshold values based primarily on literature review and expert opinion, WD: worm density, WB: worm biomass, OM: organic matter, pH: hydrogen potential, TN: total nitrogen, and OC: organic carbon.

2.4. Sensitivity Index

The sensitivity index (SI) proposed by Sheidai Karkaj et al., (2019) was employed to assess the performance of the SQI method through the following equation:

$$\text{Sensitivity index} = \frac{SQI_{MAX}}{SQI_{MIN}}$$

where SQI_{MAX} and SQI_{MIN} are the maximum and minimum values of each SQI, respectively. The indicator with a higher sensitivity value is more susceptible to environmental conditions and management plans.

2.5. Sampling and Determination of Soil Quality Indicators

Samples were collected from the Quebrada Jun Jun Watershed along a transect, with coordinates: west longitude 78°59'16.50"; south latitude 01°34'32.31; west longitude 78°61'83.36" and south latitude 01°38'57.95" (Figure 2). In the transect, representative sampling points were selected considering the altitude and vegetation cover of the area. In terms of altitude, three zones were identified: zone 1 (< 2800 masl), zone 2 (2800–2900 masl), and zone 3 (> 2900 masl). Additionally, to consider the vegetation cover variable, samples were collected both in areas with and without vegetation cover in the three zones.

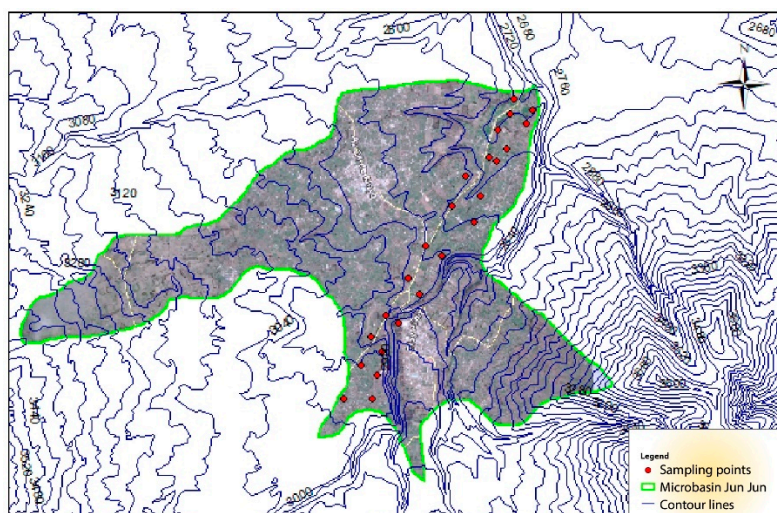


Figure 2. Sampling Zone of the Jun Jun Stream Watershed.

At each sampled point, four soil samples were collected randomly at a depth of 20 cm. Each soil sample was air-dried for two weeks without exposure to sunlight. Soil bioindicators were analyzed in each sample as follows: soil pH was estimated using the 1:2 soil/water ratio (25 g of dry soil and 50 ml of distilled water), employing a glass electrode and a potentiometer. Organic carbon (OC) and total nitrogen (TN) available in the soil were determined using the dry combustion method (Eyherabide et al., 2014 and Bellomonte et al., 1987). The quantification of organic matter (OM) was determined using the calcination or ignition method mentioned by Eyherabide et al., (2014), which directly quantifies the organic matter content and is based on determining the weight loss of the soil sample when subjected to high temperatures. The soil bioindicators in this research, worm density (WD), and worm biomass (WB), were estimated using the procedures described in USDA (1999).

3. Results and Discussion

Below are the results and data from three altitudes of the watershed, along with detailed discussions on the analyzed soil type. The average values and standard deviation of soil indicators based on altitude, with or without vegetation cover, are listed in Table 1 and Figure 1. The analysis

of variance of the data revealed no significant differences in the pH, OM, and WD variables for both altitude and trials with and without vegetation cover.

3.1. Organic Carbon (OC)

The OC content is the primary indicator of biological and chemical soil quality and strongly influences soil's physical quality indirectly. The OC in the soils with vegetation cover and without vegetation cover presented significant differences with the altitude variable. The results revealed that soils with altitudes below 2800 meters above sea level had the highest OC content at 1.26%; conversely, samples collected from altitudes greater than 2900 meters above sea level had the lowest amount of organic carbon at 0.58% (Figure 1). The organic carbon storage capacity in the soil of high Andean grasslands is influenced by the altitudinal gradient and soil temperature. As altitude increases, the soil exhibits a higher capacity for organic carbon storage (Beltrán-Dávalos et al., 2022). Additionally, soil cover (whether with vegetation or plant residues) mitigates water erosion caused by surface runoff, reduces soil compaction, increases organic carbon content, and promotes infiltration rates (Pedroza-Parga et al., 2022).

3.2. Total Nitrogen (TN)

Similarly, the TN in soils with vegetation cover and without vegetation cover presents significant differences with the altitude variable. The results revealed that soils with altitudes above 2900 meters above sea level had the highest TN content at 0.50%; conversely, treatments with altitudes below 2800 meters above sea level had the lowest amount of organic carbon at 0.16% (Figure 1). This demonstrated that altitude is a determining factor in the properties and processes of ecosystems in mountains (He et al., 2016). Mountains serve as useful "indicators" of how climate changes can affect non-mountainous terrestrial ecosystems (Beniston, 2003). The nutrient status can differ among ecosystem components and nutrients along altitudinal gradients (Sundqvist et al., 2013). Additionally, it was observed that soils without vegetation cover had elevated levels of nitrogen due to fertilization received from legume crops previously in the soil.

3.3. pH Analysis

The pH presented significant differences in terms of altitude and vegetation cover of the soils; It was shown that the pH decreases with increasing altitude (Figure 1). In general, the results revealed neutral soils with values ranging from 6.7 to 7.1. According to the literature, pH values decrease with increasing altitude (Oliveras et al., 2020) and (Llambí et al., 2020). However, other results show that soil pH values increase with increasing altitude (L. Li et al., 2016). The increase in pH in these cases may be related to the soil's parental material. An illustrative example of this phenomenon occurs during soil formation from limestone or calcareous rocks containing high levels of calcium carbonate. Over time, these rocks can gradually dissolve due to the influence of water and chemical reactions in the soil. As a result, the dissolved calcium carbonate contributes to an increase in soil alkalinity, leading to a rise in pH in the surrounding area. In the altitudinal gradient from 871 m to 4550 m, the soil pH decreases at an altitude of 3000 m, while, at higher altitudes than 3000 m, the pH increases (Peters et al., 2019).

3.4. Organic Matter (OM)

OM showed no significant differences in terms of altitude and vegetation cover. However, it was evident that organic matter content increased with altitude (Figure 1). The increase in organic matter content with altitude is associated with the decrease in temperature at altitudes above 3000 m.a.s.l. For altitudes between 1500 and 2000 m.a.s.l., it corresponds to increases in precipitation (Figure 1). Similarly, under conditions of extreme acidity and fungal proliferation, it can be associated with a decrease in bacterial activity and a slowdown in the mineralization process, determining high organic matter contents in the soils (Hou et al., 2021). Organic matter, organic carbon, total carbon, total

nitrogen, and the soil C/N ratio experience an increase until reaching 3000 meters in altitude; however, beyond this altitude, these values show a decrease (Zhang et al., 2021).

3.5. Bioindicators

Regarding bioindicators, the results of WD and WB showed significant differences in terms of altitude and vegetation cover. In this way, the more the elevation of the basin decreases, the greater the number of worms and grams of biomass was obtained. Additionally, the best results for WB and WD were obtained in trials with vegetation cover (Figure 3). Earthworm abundance and species richness decrease with increasing altitude, showing a significant negative correlation (Zool et al., 2022). Similarly, the richness and diversity of plant species along altitude gradients generally show a gradual and continuous pattern of species decline from lower to higher altitude regions. It was also evident that the only bioindicator present was earthworms (*Lumbricus terrestris*); other bioindicators were not present, indicating that their community is affected by anthropogenic activities and abiotic factors in the environment (Zhu et al., 2019). Finally, biomass can be related to the nutrient availability for earthworms in covered soils, as higher vegetation and relative humidity lead to greater material production, which will be available (rapid decomposition) for earthworm communities to grow in size (Al-Maliki et al., 2021).

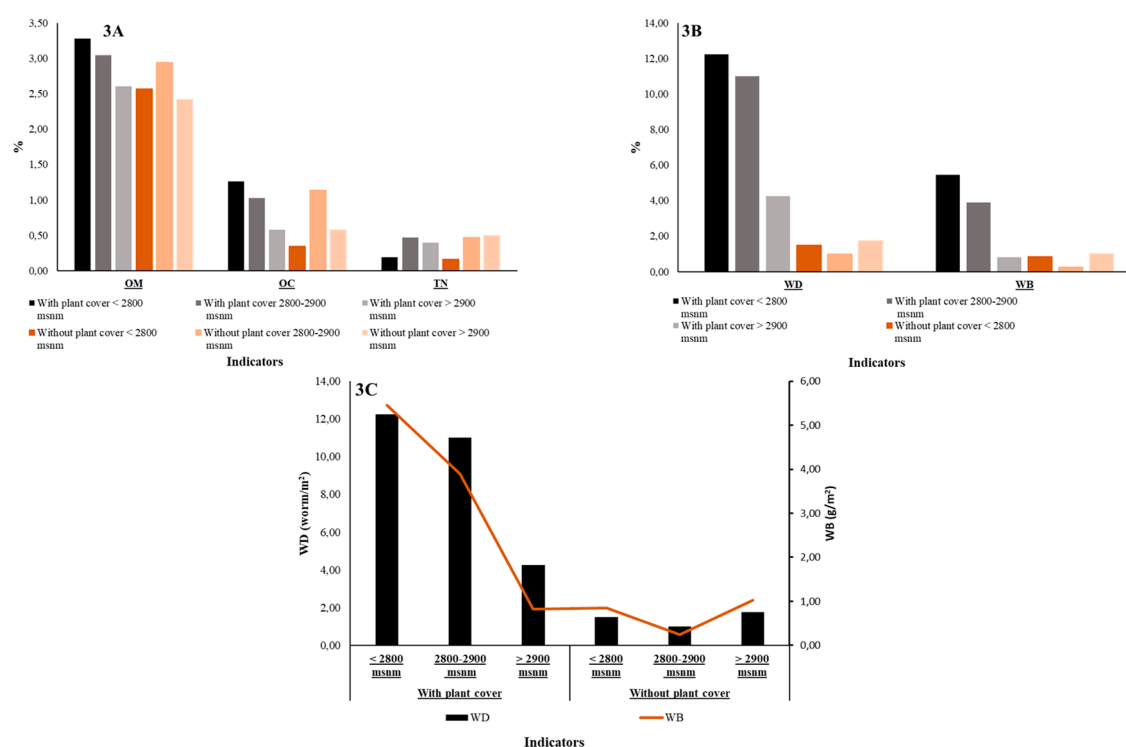


Figure 3. Indicators for estimating Soil Quality Index (SQI) collected under two types of vegetation cover across three altitude categories.

3.6. Soil Quality

Within the framework of SQI, six soil quality indicators were numerically integrated after scoring them primarily using information from literature reviews (Table 2). However, data scoring for some biological indicators is limited in the literature, making it challenging to establish a sufficiently robust methodology (Puig-Girones & Real, 2022). In this regard, maximum and minimum values were established in different ways for each indicator. For some attributes, especially under optimal conditions, thresholds were considered based on values from reference soils, while for others, theoretical criteria were applied.

A Soil Quality Index was developed by averaging the values of all selected indicators. To facilitate its interpretation, a transformation scale that defines five classes of soil quality was applied.

The creation of this scale allowed the standardization and integration of the indicators, which in turn made possible the quantitative evaluation of soil quality through numerical values, assigning each soil a value within the different quality categories.

Subsequently, the indicators were normalized on a scale from 0 to 1, respectively representing the most unfavorable and optimal conditions from the perspective of soil quality, regardless of the absolute values measured for each indicator.

Table 3. Soil quality classes for the Jun Jun watershed.

Soil Quality Classes	Scale	Classes
Very High Quality	0.80-1.00	1
High Quality	0.60-0.79	2
Moderate Quality	0.40-0.59	3
Low Quality	0.20-0.39	4
Very Low Quality	0.00-0.19	5

Figures 4 and 5 present the normalized values of the calculated indicators, as well as the resulting soil quality index. It is observed that soils with vegetation cover, on average, have an index of 0.58, classifying them as moderate quality (class 3). Similarly, soils in the area without vegetation cover had an average of 0.45, also indicating moderate soil quality (class 3). Soils exhibit moderate variability in the study area, showing a heterogeneous spatial distribution. This variability can be attributed to various factors, such as diversity in land use, variations in soil depth, specific terrain characteristics, and topography, among others (Fang et al., 2012).

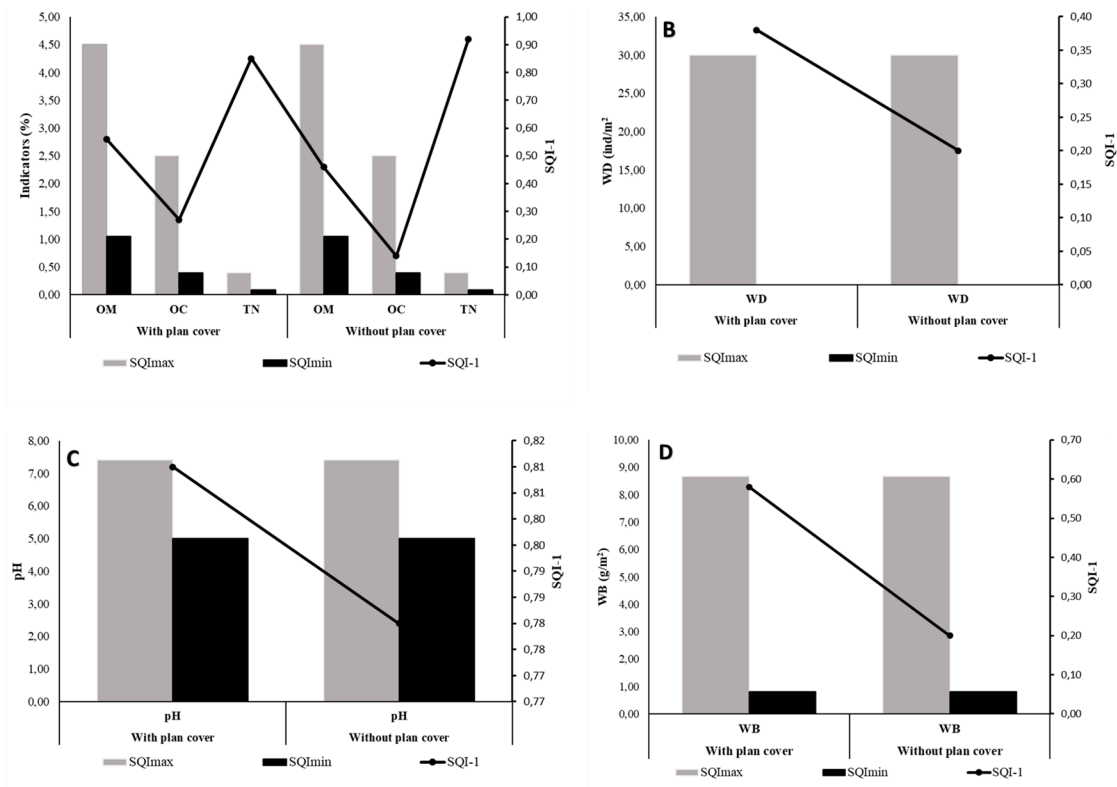


Figure 4. Indicators for estimating Soil Quality Index (SQI) collected under two types of vegetation cover across three altitude categories.

The determination of the soil quality index is influenced by the values obtained in each of the indicators. In this sense, some indicators had higher values than the thresh-old values in the literature, which caused the value of the total index to rise. Thus, the indicator with the lowest value

was Organic Carbon (OC), while the highest corresponded to Total Nitrogen (TN). The significant reduction in Organic Carbon in the area could be attributed to practices such as excessive grazing, deforestation, intensive tillage, and the use of burns in agricultural fields (Gangopadhyay et al., 2021).

The pH indicator has a quality value close to 7; however, in soils without vegetation cover, a decrease in pH is observed compared to reference soils. This trend has been previously recorded by other researchers (Barrow & Hartemink, 2023) and can be attributed to the loss of organic matter, the removal of minerals during crop harvesting, erosion of the surface layer, and the effects of nitrogen and sulfur fertilizers. Despite this, most measured values are still considerably far from the toxicity point established for most crops in the region. The pH indicator value approaches the maximum quality, thus significantly reflecting the local soil conditions.

To determine the number of earthworms, density (worm/m²) was used, using the maximum number of earthworms found in the study soil as the maximum value, while the minimum value was set at zero. Other bioindicators, such as ants, arachnids, and centipedes, will not be considered in the quality assessment due to their total absence at the sample points. Earthworms, as bioindicators belonging to the macrofauna group, are closely linked to soil quality. Their presence promotes significant benefits to soil structure and positively impacts subsequent crops. However, the watershed soil has experienced various intensities of use, and these organisms show a noticeable sensitivity to changes in their environment, resulting in alterations in their communities. Therefore, the absence of these earthworms clearly indicates a decrease in soil quality (Sofa et al., 2020). The abundance of earthworms and species richness decreased monotonically with increasing altitude. The distribution pattern of earthworms is explained by the fact that, with increasing altitude, habitat conditions become less favorable, establishing a wide range of environmental barriers and leading to a limitation of taxon spread (Singh et al., 2019).

Soils with vegetation cover have an average quality index of 0.58, while soils without vegetation cover have a quality index of 0.45 (Figure 6). These results show that vegetation and soil have a direct and close relationship, as soil and vegetation mutually influence each other over time (Xiang et al., 2023). Vegetation contributes to nutrient retention, improves soil structure, and promotes microbial activity, key factors for soil quality (Koudahe et al., 2022). In this sense, vegetation improves soil structure, water retention, promotes microbial activity, and prevents erosion. Plant roots can contribute to the formation of soil aggregates, improving porosity and nutrient retention capacity (Sekaran et al., 2021). The drastic decrease in soil quality index in areas without vegetation cover could be due to the loss of these associated vegetation benefits. The absence of plants can lead to soil compaction, loss of organic matter, and a decrease in microbial biodiversity, negatively affecting the soil's ability to sustain plant life (Rahman et al., 2021). These findings support the importance of conservation and restoration of vegetation cover as key strategies to improve and maintain soil quality. Additionally, they may have significant implications for land use management, sustainable agriculture, and soil degradation mitigation. It is crucial to consider these results when designing policies and soil management practices to ensure the long-term health of terrestrial ecosystems.

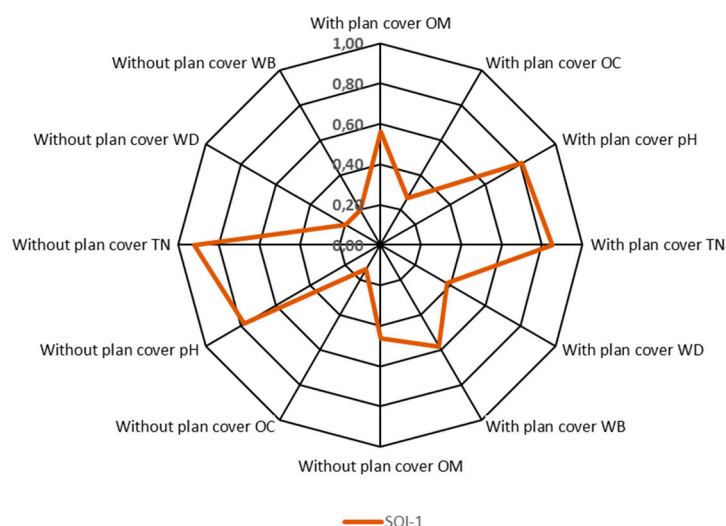


Figure 6. Soil quality classes in the Jun Jun watershed under the influence of various chemical and biological parameters.

4. Conclusions

Understanding and characterizing soil quality are fundamental aspects of sustainable soil management. In the study area, the soils exhibited quality indices that varied between 45 and 58 for soils with and without vegetation cover, respectively, indicating the soil studied has a moderate and high quality throughout the basin. The influence of the altitudinal gradient is evident, revealing a significant relationship between altitude and soil quality. Soil quality index values show a tendency to decrease with increasing altitudinal gradient, suggesting that soil quality tends to deteriorate at higher altitudes. The presence of vegetation cover emerges as a positive factor, as soils with this cover exhibit higher soil quality index values compared to soils without vegetation cover. The spatial distribution of soil quality is heterogeneous in the study area, influenced by factors such as land use, topography, and soil depth. The decrease in organic carbon, especially in soils without vegetation cover, indicates possible degrading practices such as excessive grazing, deforestation, and intensive tillage. In summary, the study highlights the importance of assessing soil quality considering multiple factors, such as altitudinal gradient and the presence of vegetation cover. These findings offer key insights into soil dynamics in the Jun Jun watershed. Furthermore, they underscore the urgent need to adopt sustainable management practices to preserve and improve soil quality in this region, thereby contributing to the long-term sustainability of local agroecosystems.

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