



Renewal of Archival Legacy Soil Data: A Case Study of the Busia Area, Kenya

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Much older soils information, collectively known as *legacy soils data* lies idle in libraries or in the personal collections of retired soil scientists. The probability is very high for this legacy data to be lost or destroyed. We demonstrate the stepwise process of bringing legacy soils data “back to life” using the *Reconnaissance Soil Survey of the Busia Area (quarter degree sheet No. 101)* in western Kenya as an example. The first step, *site identification*, involves meeting and deliberating with key institutions to identify a setting for the study. The second step, *data archeology*, involves locating and cataloging legacy soil data from key institutions, which often requires numerous site visits and the assistance of individuals familiar with the target data. The third step, *data rescue*, involves converting paper copies of data into a digital format by scanning the maps, narrative descriptions, and tables, and storing the information in a database. The fourth step, *data renewal*, consists of bringing the data to modern standards by taking advantage of technological and conceptual advances in geoinformation technology. In our example, the resulting digital (scanned) soil map of the Busia Area is a significant upgrade from the fragile paper map. The fifth step, *data interpretation*, entails careful interpretation of the soil information available within the legacy soil survey to provide additional agronomic information. This allowed us to produce 10 land quality maps showing the ability of the land to perform specific agronomic functions, and 18 different crop suitability maps that were not previously available. The rescued maps and their associated tabular and narrative data also provide crucial inputs for generating more detailed soil maps using digital soil mapping techniques that were unavailable when the original mapping was conducted.

Keywords: Busia, data archeology, data rescue, data renewal, land quality map, crop suitability map

INTRODUCTION

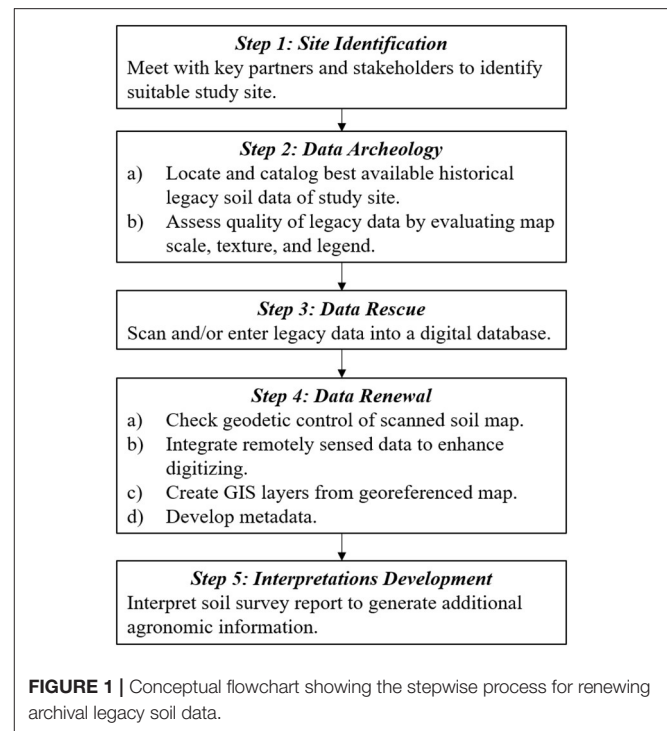
Most soil resources exist as traditional soil maps, soil survey reports, soil survey manuals, land evaluation frameworks, soil profile descriptions, and farm management handbooks, collectively known as legacy soil data (1). These soil resource inventories have been widely used as meaningful sources of soil information to support soil conservation or as major components of national environmental monitoring (2–4) and could still be useful as the demand for soil data is soaring (5). However, information on soils for much of Africa and most developing countries is sparse (6).

Kenya, where this study is focused, fortunately has considerable soils information (7, 8). Unfortunately, most legacy soil data often remains idle in libraries (6), and the probability is very high of such data being lost through natural, manmade, or political disasters, or simply neglect (9). In addition, the number of individuals involved in the description and interpretation of existing legacy soil data, especially in most African countries, is small and decreasing as many soil surveyors are retiring or moving to other positions. Our visits to the Kenya Soil Survey in the spring of 2016 confirmed this narrative. Most of the legacy soil data was left unused and stored in library shelves, some were in private collections of retired soil scientists, and those in digital format were largely underused or used only internally.

Rescuing legacy soil data is mainly driven by the fact that a lot of effort and resources went toward compiling, analyzing, and publishing them (6). In addition, information within legacy soil data often consists of spatial distribution of soils, land quality, crop suitability, and geolocated soil profile information with their respective laboratory data, geology, and land use types. This sort of data can be analyzed and used as a primary input for digital soil mapping (DSM), especially for countries with sparse soil data infrastructures (2, 10–14). In a highly competitive world for resources and funding, rescuing, and utilizing available soil data while identifying gaps in soil information not provided by the legacy data can be mutually beneficial (14). One potential solution to avoid the loss of the paper data and the information it contains is to take advantage of the advances made in geoinformation technology and bring this data from the library shelves to the light of the digital world.

Previous efforts, however, on the renewal of legacy soil data mainly have been focused on six major areas. (i) Protecting legacy soil data from getting lost due to unforeseen events, natural or manmade (6). (ii) Showcasing the stepwise process of renewing legacy soil data to support other research (9). (iii) Developing the criteria that can be used to assess renewed legacy soil data (15, 16). (iv) Using soil data mined from soil survey reports to either map soil properties and/or soil types using digital soil mapping techniques (6, 10, 13, 14, 17–21). (v) Utilizing legacy soil data to analyze and detect spatio-temporal changes in soil properties (22–29) and use of the point data to focus on resampling approaches to detect changes in soil properties (30–32). (vi) Extracting soil information from legacy soil data to build soil profile databases robust enough to be used for digital soil mapping (3, 33–36).

Even though a few studies have described the steps involved in the renewal of legacy soil data, there is still a gap in the literature on how to interpret information stored within legacy soil data for the provision of additional agronomic information. The aim of this study is to: (i) utilize the criteria described by Forbes et al. (15) and utilized by Rossiter (9), Odeh et al. (3), Cambule et al. (4), Arrouays et al. (6), and Rasaei et al. (12) to support the renewal of the best available soil survey report of a selected portion of Kenya into a digital format and (ii) utilize the information within the legacy soil data to provide additional agronomic information to meet current and future demands.



MATERIALS AND METHODS

The conceptual framework used for this study is summarized in **Figure 1** and consists of five steps, site identification, data archeology, data rescue, data renewal, and interpretation development.

Site Identification

Site identification, in our case, consisted of meetings with key institutions in Kenya, including the Academic Model Providing Access to Healthcare (AMPATH), Kenya Soil Survey, and the Kenya Agricultural and Livestock Research Organization (KALRO) to identify a setting for this study. Conversations with Dr. Joe Mamlin, AMPATH Field Director Emeritus, tasked us to identify a region in Kenya that has high population and poverty densities and high rates of people living with HIV/AIDS. These specifications are important for AMPATH because they want to have access to agronomic information and practices to test whether this will ultimately affect the quality of life of people living with HIV/AIDS from a nutritional standpoint. Discussions with the Kenya Soil Survey and KALRO were aimed at identifying a region in Kenya that meets the specifications described by AMPATH that also has detailed legacy soil data.

Data Archeology

The data archaeology step entails locating and cataloging legacy soil data available from key agricultural institutions. We retrieved legacy soil data from the Kenya Soil Survey in Nairobi. Kenya Soil Survey is the official government organization that stores all legacy soil data for the country.

TABLE 1 | Soil information contained within the *Reconnaissance Soil Survey of the Busia Area*.

Type of soil information	Scale
Reconnaissance soil map of the Busia Area (colored)	1:100,000
Reconnaissance soil map of the Busia Area (black and white)*	1:100,000
Land evaluation key**	
Soil profile characteristics significant for soil classification	
Soil profiles and analytical data descriptions	
Land quality ratings for soil map units**	
Soil engineering map of the Busia Area (black and white)**	1:100,000

*Both the digital and paper copies were available.

**Retrieved during second visit to the Kenya Soil Survey.

Quality Assessment of Legacy Data

The guidelines described by Forbes et al. (15) were used to assess the quality of the legacy data by evaluating the map scale, texture, and legend. More specifically, this process is aimed at addressing: (i) whether the soil map is legible enough to represent the smallest land area of interest on paper maps to the user, (ii) whether the soil map conveys sufficient soil property information, (iii) whether control points and areas can be accurately located on the ground or map, and (iv) whether the map captures the soil development process for the target scale. In other words, for criterion iv, does the soil map use the catena concept to capture the soil variability within the landscape?

Map Scale and Texture

Map scale refers to “the relationship between the distances on the map and the corresponding distances on the ground,” whereas *map texture* refers to “the sizes and pattern of delineations on the map and determines the map’s overall legibility” (15). Both the map scale and texture of the soil map were evaluated to assess the legibility and capability to represent the smallest area of interest. To do so, two map parameters were used. First was the minimum legible area (MLA) (Equation 1) that represents the smallest land area that can be represented on the map at its published scale using the criterion of a minimum legible delineation (MLD) of 0.4 cm². The MLD is independent of the map scale and is conventionally defined to be a roughly circular area of 0.4 cm². Smaller delineations are considered illegible for two reasons: (i) there is not enough room inside the delineation to legibly write the map unit symbol and (ii) the proportion of the delineation covered by the bounding line becomes significant. Second was the index of maximum reduction (IMR) (Equation 2), which refers to the factor by which the map scale can be reduced before the average size delineation (ASD) (Equation 3) would become equal to the MLD, i.e., before more than half of the map would become illegible. A large IMR implies that the survey area is represented on a map that is physically larger than necessary (15). The ASD is estimated using portions of a map with a given map texture by randomly sampling map areas with circles or squares of known areas and then converting the delineation counts in several of these areas (37). For this study, a transparent overlay with a 2.5 cm radius circle was used to count the number of delineations

within the circle.

$$MLA = \frac{\left(\frac{1}{RF}\right)^2}{2.5 \times 10^8} \quad (1)$$

where the abbreviation RF stands for the representative fraction, which is the amount of any unit of distance measured on the ground that is represented by that unit on the map and is written as a ratio of two distances. For the Busia area, the RF is 1:100,000 meaning that 1 cm on the Busia soil map represents 100,000 cm (1000 m) on the ground.

$$IMR = \sqrt{2.5 \times ASD} \quad (2)$$

$$ASD_{2.5 \text{ cm radius circle}} = \left\{ \left(\frac{\text{sum of 5 counts}}{142} \right) - 0.1 \right\} \quad (3)$$

The “sum of counts” refers to the number of delineations randomly selected within the transparent overlay.

Map Legend

The *map legend* identifies the map units based on the soil classification used (38). It refers to a full description in the associated survey report and may also provide a brief description and various interpretations. The legend can be identified by the symbols printed inside the map unit polygons. A descriptive legend gives information about each map unit, whereas the map unit names and definitions in descriptive and interpretative legends dictate the level of usefulness of the information. The map legend may be evaluated either in terms of a specific use of the soil inventory or by a more general criterion, such as a soil classification system (15). Map unit information was evaluated based on the availability of information on soil classification. Information is considered adequate if map unit descriptions included the diagnostic information such as horizons and properties, or the soil classification.

Data Rescue

Data rescue involves the transformation of the primary legacy data existing in paper format to an up-to-date archival format either by scanning or direct entry into a database (9, 39). Luckily, both the colored and the black-and-white soil maps of the study area were available in the European Soil Data Center (ESDAC) database <http://esdac.jrc.ec.europa.eu/resource-type/national-soil-maps-eudasm> (40). These maps had been scanned using a wide-format color scanner and were stored at 200 dpi (dots per inch) in TIFF (Tagged Image File Format), a lossless format that holds more detail than most other formats. Further processing of the digital maps such as cropping, distortion elimination, rotation, and color adjustment was carried out using an image-processing scanning software for wide-format images. Photo processing software was used to convert TIFF files into both JPEG and PDF formats (41). This process was replicated for the remaining Busia Area legacy soils data existing in paper format and not found in the ESDAC database (Table 1). The specifications of scanners, especially the color quality and resolution in terms of pixels per square are important

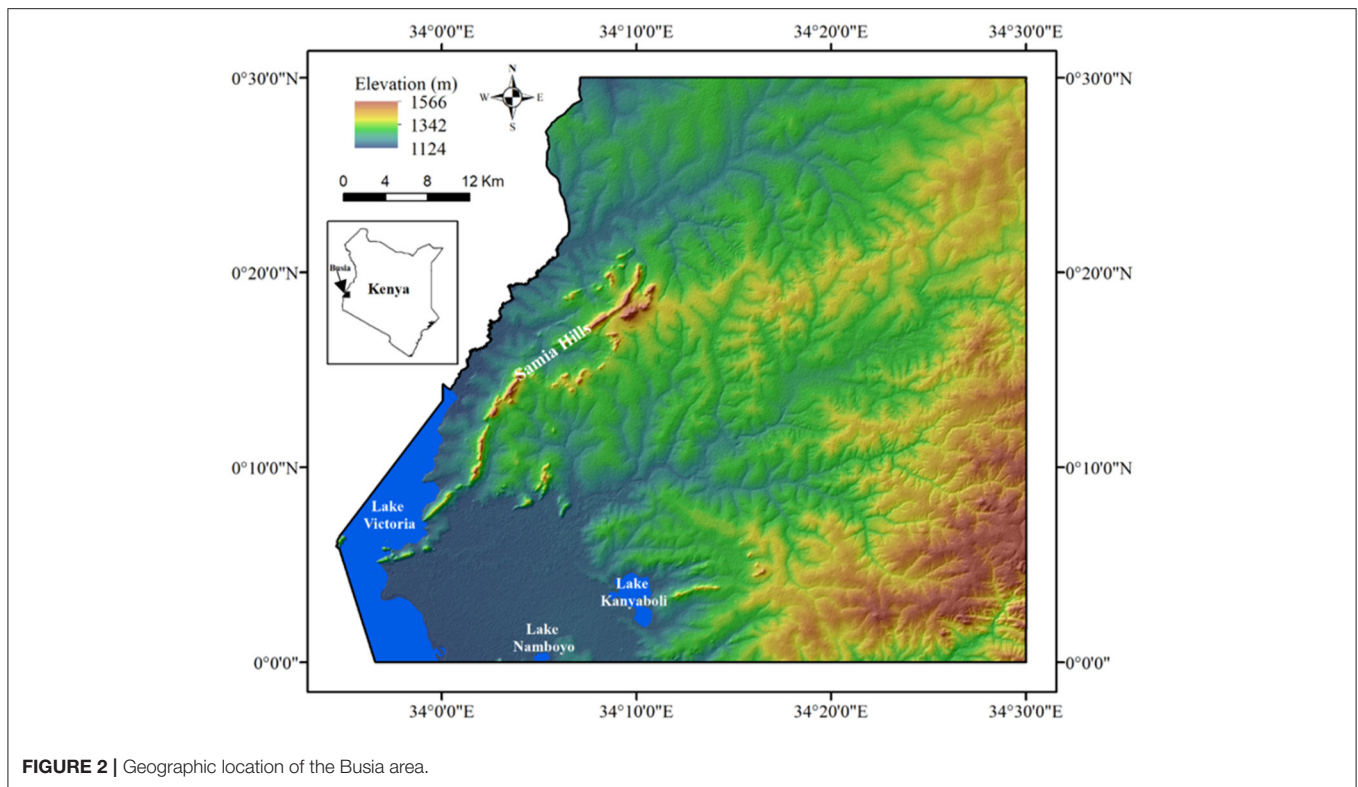


FIGURE 2 | Geographic location of the Busia area.

considerations for preserving the authenticity of the original maps. In addition, all the available soil information contained within the scanned soil survey report [Appendices 3–6 in (42)] were manually entered into separate spreadsheet documents, saved as CSV (Comma Separated Values) text files, and stored for later use.

Data Renewal

Data renewal involves the process of bringing legacy soil data to modern standards by taking advantage of technological and conceptual advances in geoinformation technology. These steps include georeferencing, digitalization, formatting for compatibility with GIS software, and metadata.

Geodetic Control

One drawback with most legacy soil maps is the lack of geodetic control points (4) for accurate georeferencing. This step involves looking for clearly labeled latitude and longitude points on the soil map to be used as the geodetic control points. Additional control points can also be identified using clues such as road intersections and rivers that are clearly visible on the available legacy soil map. These control points are used together with a suitable transformation available in ArcGIS (43) to shift and warp the scanned raster image from its existing location to the spatially correct location.

For georectification, satellite imagery was used as a basemap because it is freely available and provides the best currently available, up-to-date, georeferenced imagery of the study area.

For this study, we used the satellite Imagery with Labels basemap from ArcGIS Online service. Key features on the soil map such as road intersections and natural features such as rivers were clearly visible on the scanned map and were used as additional control points for georectification. Forty control points evenly spread out across the study area were used for georectification and a third-order polynomial transformation was used to shift and warp the scanned map to its spatially correct location ($RMSE = 6.63 \times 10^{-4}$ m).

Integration of Remotely Sensed Data

Digitizing legacy soil maps often requires the integration of ancillary data, mostly remotely sensed data (9). Such products include Landsat imagery, vegetation cover, and terrain attributes that are generated from a digital elevation model (44). Soil is in part related to topography and vegetation (45, 46) and therefore the borders of some soil map units may align with remotely sensed data (13).

Creation of GIS Layers

This step involves the conversion of the georeferenced scanned soil paper map to a GIS layer by digitization. In cases where a digital copy of the soil map exists, this process can be omitted. One should, however, take caution with such digital copies and manually check if all the soil map units have been digitized and the soil map unit boundaries of the digital copy correctly overlay on the scanned soil map unit boundaries.

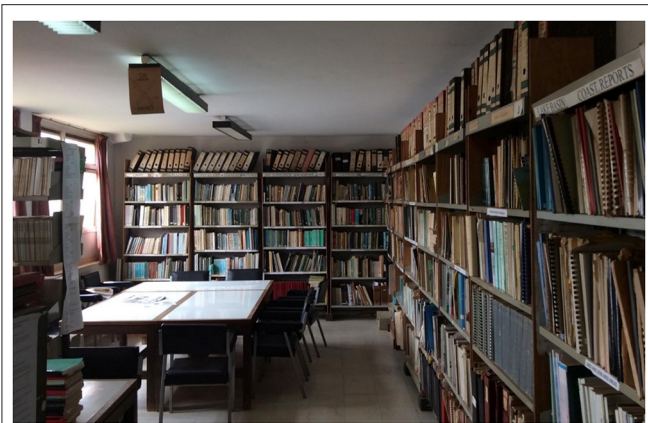


FIGURE 3 | The Kenya Soil Survey library, June 6, 2017. Photo by Joshua O. Minai.

Development of Metadata

This step involves the development of appropriate metadata to include key identification information such as spatial data source, spatial reference, attributes, information on data quality, and description(s) of methods used to renew the data. The metadata also should include the explanation of key semantics used.

Interpretation of Soil Survey Information

This final step involves interpreting the soil classification information within the soil survey report to show and rate the best uses of the soil resources. This is because legacy data such as soil survey reports can provide basic information on soil and land characteristics that can be useful for various purposes such as determining the suitability for various types of practices for agricultural, range, and forestry land use.

For the Busia Area, the “Framework for Land Evaluation” prepared by FAO (47), was followed to develop additional agronomic information. Fundamental in this approach is that land can be classified meaningfully for clearly defined uses termed *land use alternative types* or *land utilization types* considered relevant for the survey area. Each land use type was defined by specific, quantifiable factors that have a marked influence on performance that is integral for defining crop suitability maps.

The Kenya Soil Survey prepared proposals for rating land qualities for the Busia Area [internal communication Nos. 7 and 29 (48, 49)]. In this rating system, land qualities were classified into three to five grades ranging from *very low* to *very high* based on the most limiting factor for the land qualities. The next step involved the establishment of specifications for the land qualities that will define the suitability class levels for each land use alternative. These steps were followed to generate land quality and crop suitability maps of the Busia Area. The suitability evaluation of each soil map unit for each land use alternative was carried out by comparing the land quality ratings of each soil map unit to the specifications for each land use alternative.

RESULTS AND DISCUSSIONS

Site Identification

Meetings with AMPATH, KALRO, and the Kenya Soil Survey resulted in choosing the Busia Area as the setting for this study (Figure 2). This is because: (1) it has accessible detailed legacy soil data at a scale of 1:100,000 (42); (2) agriculture is the main economic activity in the area (50); (3) it has high population and poverty densities, and therefore provision of agronomic information is needed to revitalize agriculture in the area (51); (4) it has high rates of HIV/AIDS infections (52); and (5) the main author is from the study area and is familiar with it (13, 14). The site identification phase included numerous site visits with local partners who are familiar with the study area.

Data Archeology

Available legacy soil data for the Busia Area, whether paper or digital formats, were manually retrieved by going through all field soil survey reports within the Kenya Soil Survey library (Figure 3). This effort resulted in the retrieval of the *Reconnaissance Soil Survey of the Busia Area* (quarter degree sheet No. 101) as the primary legacy soil data for this study (42). Table 1 shows the information contained within the soil survey report. This step requires care because information contained within the legacy soil data packet can easily be missed. After studying the soil survey report, maps, and tables obtained during our first visit, we identified missing materials, which required a second visit to the Kenya Soil Survey to obtain the missing information (Table 1). The best way to ensure that all the information is retrieved during the initial visit is by going through the table of contents of the soil survey report and paying attention to the appendix to ensure that all information is included the soil survey report packet. A reconnaissance visit to the study area was also conducted to familiarize ourselves with the area.

Quality Assessment of Legacy Data

Map Scale and Texture

The minimum legible area for the soil map of the Busia Area was 40 hectares (ha), which represents the smallest land area that can be represented on the map. The maximum location accuracy was 25 m, meaning that the inherent uncertainty on the ground of well-defined map points was 25 m. This directly affects the accuracy with which points on the ground may be represented. For a map scale to be adequate, the maximum location accuracy must be numerically smaller than the accuracy to which the user wishes to locate points on the ground and therefore depends upon the intended uses of the survey. A well-defined ground point can be plotted on the map sheet with an accuracy of 0.25 mm (53). The index of maximum reduction was 3.2, indicating that the map is very legible, and the map scale could be substantially reduced without impairing legibility.

Map Legend

Soil map units were explicitly labeled and categorized in the map legend (Table 2). The construction of the map unit legend indicated physiographic land types (such as hills, footslopes, uplands, etc.) based on physiographic photointerpretation. These

TABLE 2 | Samples of soil map legend tables of the Busia Area, at 1:100,000 scale.

Cartographic unit	Physiography	Geology	Soil depth (cm)	Soil characteristics	Dominant soils***
HIP*	Hills	Igneous	0–50	Overlying hard rock	Lithosols (l)
AA1**	River terraces and floodplains	Alluvium			Ferralic arenosols (Qt), Chromic vertisols (Vc)
PSb1	Plains	Sandstone		Brown	Orthic acrisols (Ao), Orthic ferralsols (Fo)
VXC2	Minor valleys	Various parent materials (i.e. plural)		Complex	Ferralic cambisols (Bf), Dystric gleysols (Bd), Vertic fluvisols (Jv)
UGr4M****	Uplands	Granite	0–50	Shallow and red, over petroplinthite	Rhodic ferralsols (Fr)

* "P", soils over hard rock.
 **AA1, integer numbers 1, 2, and 4 indicate sequence of soil map units with almost identical features.
 *** Major soil grouping according to the FAO/UNESCO legend of the Soil Map of the World (38).
 **** "r", red soils at depth specified by the letter "M" (M, shallow).

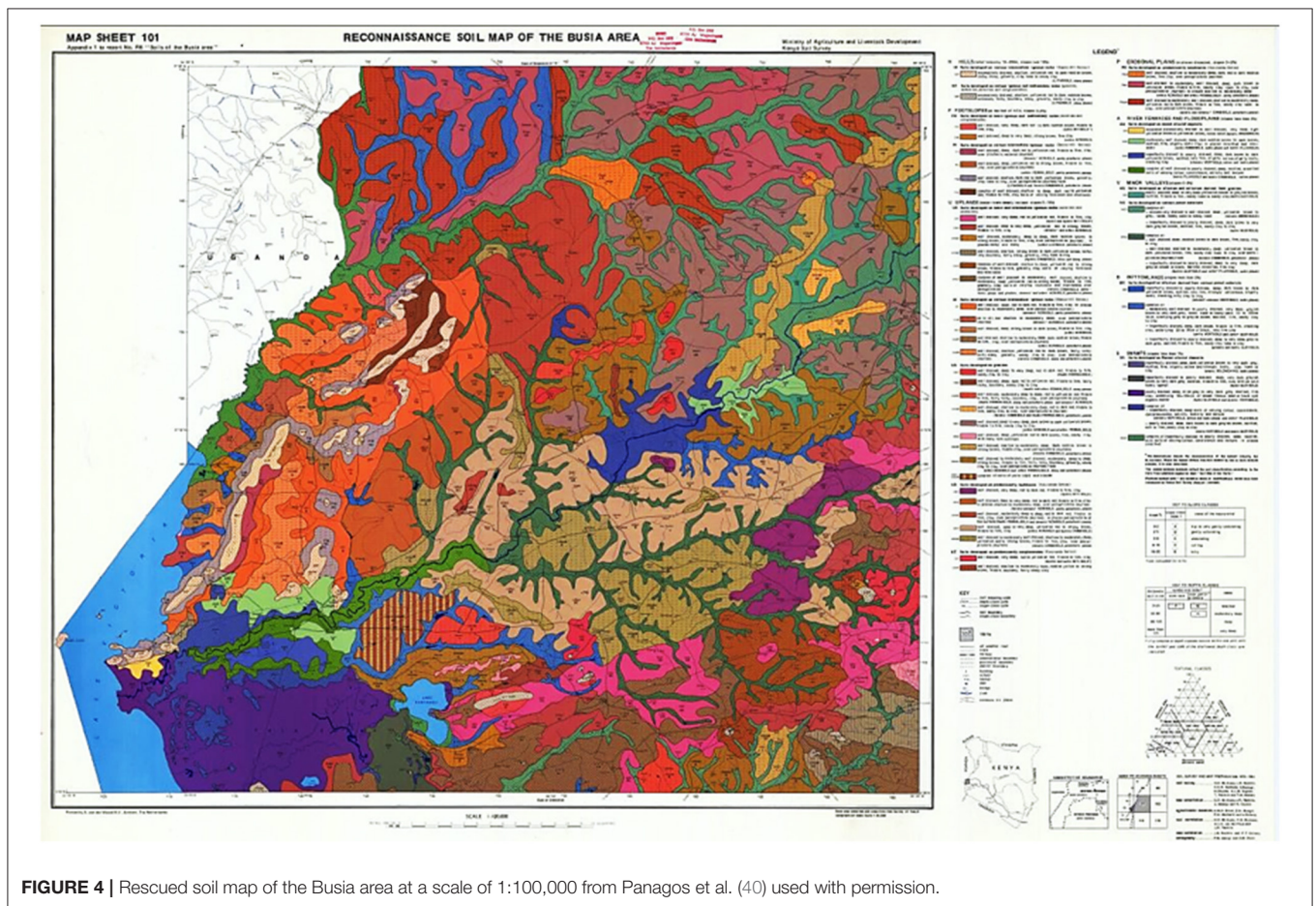


FIGURE 4 | Rescued soil map of the Busia area at a scale of 1:100,000 from Panagos et al. (40) used with permission.

land types were further subdivided according to the underlying parent material on which the soils were developed, described as either the stratigraphy or underlying rocks such as dolerites, granites, etc. At the third level the map units were broken down and described based on important soil profile characteristics including drainage conditions, depth, color, consistency, texture, etc. (42). This was then followed between brackets (dominant

soils column in **Table 2**), by the classification of the main soils described according to the FAO/UNESCO nomenclature in the legend of the Soil Map of the World (38).

All the map units were explicitly described and therefore map units were adequately defined because the information within the map units provides sufficient specific information relative to the land use so that the map unit's suitability for a specific use may be

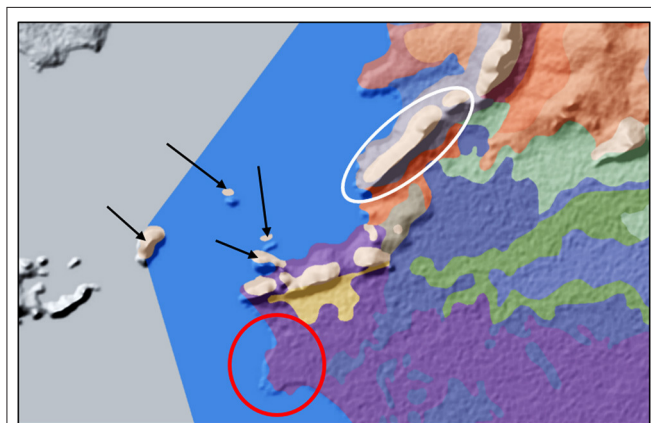
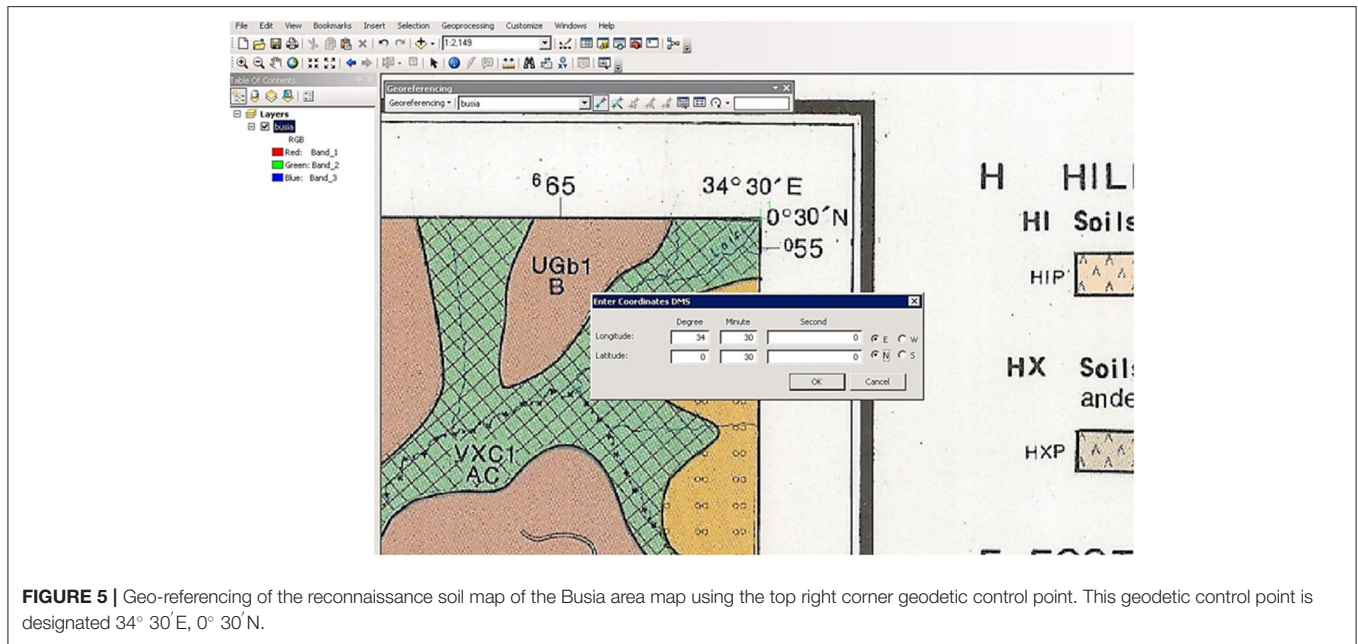


FIGURE 6 | Using a hillshade to identify and correct inaccurately drawn soil map units. Black arrows show inaccurately drawn soil map units occurring on islands. Soil map unit within the red circle shows an incorrect boundary between that soil map unit and the water body. The soil map unit within the white boundary shows a map unit that is meant to represent soils of the hills.

determined and are uniform in their suitability for the land use i.e., 85% of their total area will perform similarly for the use. Map units contained descriptions of acreage, agro-ecological zone, parent material, meso- and macrorelief, erosion, vegetation, land use, general soil description, color, texture, structure, consistence, chemical properties, clay mineralogy, diagnostic properties, and soil classification according to the Soil Map of the World (38, 42).

Data Rescue

Most of the data in the soil survey report existed in paper formats. The survey report consists of the published report of the soil resources of the Busia Area itself, three map sheets, and two large

folio sheets. The three map sheets included two soil maps of the Busia Area, one in color (**Figure 4**) and another in black-and-white [Appendices 1 and 2 in (42)], and a black-and-white soil engineering map [Appendix 7 in (42)]. The two large folio sheets included a land evaluation key [Appendix 3 in (42)] and soil profile characteristics significant for soil classification [Appendix 4 in (42)] (**Table 1**).

Data Renewal Geodetic Control

The geodetic control points were available and legible, and both the geographic and grid coordinates were printed on the margins of the maps (**Figure 5**). The map projection was not given explicitly, but conversations with the Kenya Soil Survey GIS expert confirmed that the Busia Area soil map was developed using the East Africa War System of Coordinates, Traverse Mercator projection, belt I on the Arc 1960 datum with Clarke (54) as the reference ellipsoid (54). The map was first projected to Arc 1960 and then georeferenced using the four geodetic control points printed at the four corners of the map. It was then projected to the WGS84 Web Mercator (Auxiliary Sphere) projection.

GIS Coverages

The digitized soil map, provided by the Kenya Soil Survey, showed that the soil map units often were inaccurately delineated and did not capture key features such as islands and hills (**Figure 6**). This is a common challenge with paper maps because the transfer of the lines from field sheets to basemaps was not performed by surveyors. Without the soil surveyor's expert eye, knowledge, and experience, soil boundaries that followed obvious landscape features may have not been reproduced correctly, as described for other cases by Rossiter (9). This

TABLE 3 | Differences in soil map unit acreage between the original and the edited digitized soil map of the Busia Area.

Soil map unit	Original digitized map (km ²)	Edited digitized map (km ²)
AA1	3.3	3.2
AA2	37.6	38.1
AA3	14.2	13.6
AAC	40.1	40.0
BX1	12.3	11.8
BXC	99.3	100.0
Flb	2.2	2.2
FIC	16.5	16.4
FIM	41.2	43.7
Flr	7.7	7.2
FXb	0.6	0.6
FXr	5.5	4.9
HIP	45.1	44.2
HXP	7.6	7.7
PSb1	118.6	127.1
PSb2M	148.8	145.2
PSrM	16.0	16.2
SA1	1.3	1.1
SA2	20.4	20.9
SA3	132.9	128.8
SAC1	76.0	75.1
SAC2	44.3	47.7
UDb1	187.8	195.3
UDb2M	112.4	114.0
UDr1	40.4	40.3
UDr2	62.8	63.9
UDr3m	91.4	93.7
UGb1	199.9	207.2
UGb2	7.6	7.4
UGb3M	55.5	55.9
UGb4m	145.5	151.2
UGr1	7.9	7.4
UGr2	2.0	7.9
UGr2-UGb	24.2	23.8
UGr3m	27.1	27.4
UGr4M	12.8	13.0
Ulb1*	-	4.7
Ulb2M	17.8	12.9
Ulb3M	57.5	59.7
Ulr	100.7	103.6
UlrM	15.5	16.3
UVb1	84.8	75.9
UVb2m	10.2	10.4
UVb3P	26.9	27.8
UVC1	46.3	48.0
UVC2	122.0	123.9
UVr	91.0	95.2
UZbP	2.9	3.0
UZr	17.1	17.2
VG1	39.9	42.8

(Continued)

TABLE 3 | Continued

Soil map unit	Original digitized map (km ²)	Edited digitized map (km ²)
VXC1	195.2	185.8
VXC2	171.0	152.1
Unmapped**	0.0	-
UGAND**	0.3	-
Total	2867.6	2883.1

*Soil map unit that exists in the survey report but was not digitized in the original digital copy provided to us by the Kenya Soil Survey.

**Polygons that were digitized in the original digitized soil map but that do not represent actual soil map.

is expected because basemaps used by soil mappers available in the early 1980s were not as accurate compared to what is available today.

Integration of Remotely Sensed Data

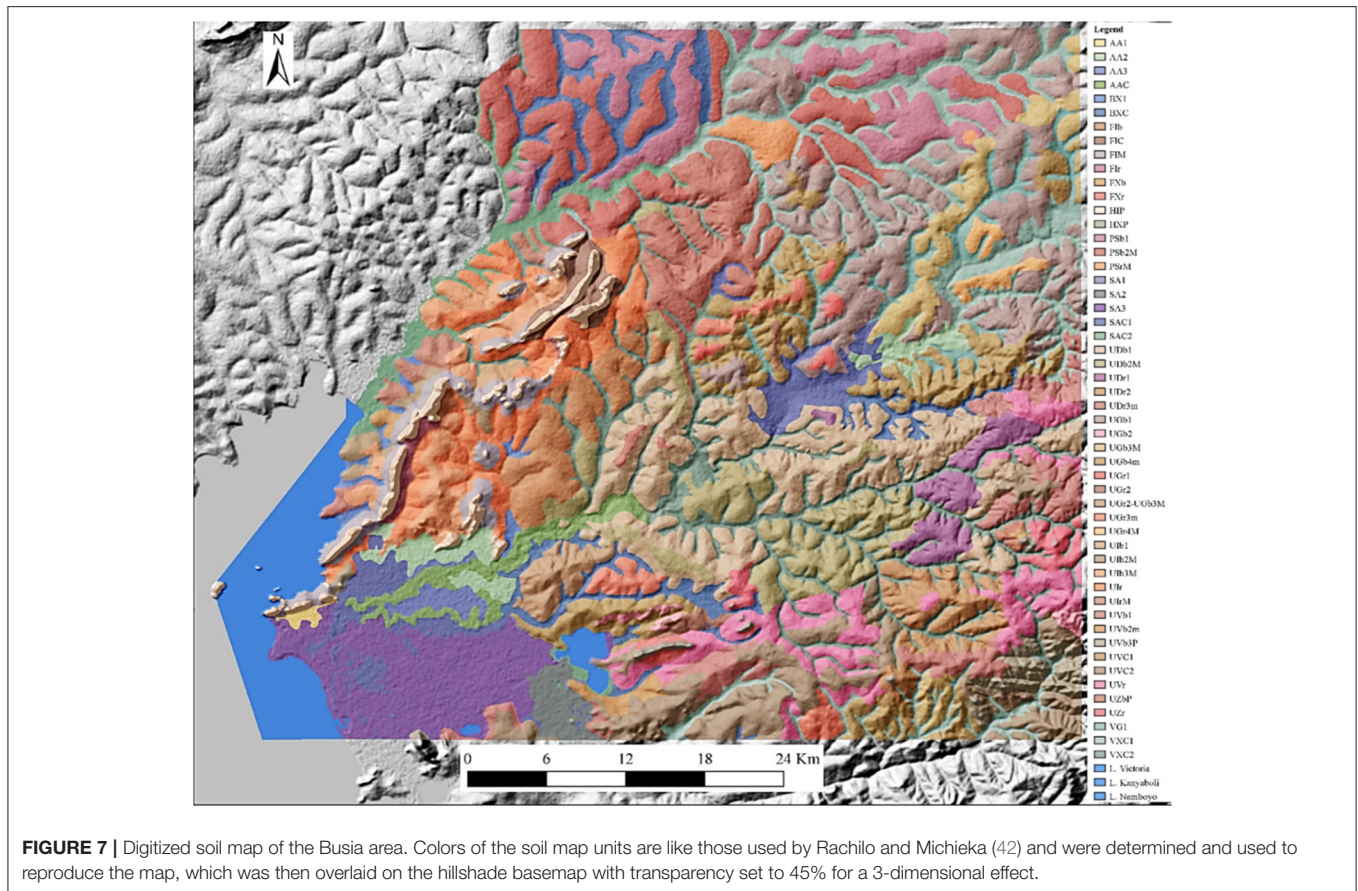
Rossiter (9) proposed the use of both satellite imagery and terrain attributes to adjust soil map unit boundaries. To ensure that each soil map unit captured its respective landscape features, satellite imagery and the hillshade generated from NASA's 30 m Shuttle Radar Topographic Mission (SRTM) (55) were used to manually adjust the polygon boundaries for some soil map units. Satellite imagery proved useful in correcting the soil map units that occurred on river terraces and swamps and for correcting boundaries between soil and water bodies, whereas the hillshade was used to adjust soil map unit boundaries on islands and hills (Figure 6). The integration of remotely sensed data resulted in a slight increase in the acreage of the soil map from 2,868 to 2,883 km² (Table 3). This change is not considered significant and is likely to be due to either the projection of the digitized map into the Web Mercator (Auxilliary Sphere) projection, and/or to the fact that the surveyors probably determined soil map unit areas manually using a planimeter.

Creation of GIS Layers

The result of the process described above was an accurately georeferenced, digitized soil map of the Busia Area with 348 polygons belonging to 52 different soil map units (Figure 7) broadly grouped into twelve soil orders (Table 4). Soil orders were determined from the taxonomic names of the soil classes in the soil map units. Most of the soils are moderately deep to deep, yellowish red to reddish brown, non-calcareous and predominantly kaolinitic in clay composition with few weatherable minerals remaining, and with evidence of a weak argillic horizon (42).

Development of Metadata

Metadata included spatial data source, spatial reference, and processing steps (Table 5). Such documentation is useful because it allows other users to access the data, evaluate its usefulness for their intended purposes, and assist others in similar efforts to renew legacy soil data.



Soil Property Data

Two soil property datasets were also mined from the survey report. The first dataset consisted of 76 georeferenced locations for which A and B horizons were sampled [Appendix 4 in (42, 57)]. The A horizons were collected between depths of 0–30 cm for fertility analysis, whereas the B horizons were sampled to an unspecified depth. The soil properties included: texture, Munsell color, structure, consistence, the presence or absence of clay cutans, clay type, bulk density, porosity, soil classification according to both the FAO-UNESCO Soil Map of the World (38) and Soil Taxonomy (58), soil organic carbon (SOC), base saturation (at pH 7 and 8.2), exchangeable sodium percentage at pH 8.2 (ESP), and electrical conductivity (42). Of all these soil properties, data for SOC for the A horizon, texture for the A and B horizons, B/A clay ratio, and soil classification were available for all 76 profile pit locations (14).

The second set of soil property data consisted of detailed descriptions and analytical data for 48 georeferenced profile pits [pages 158 to 256 in (42, 57)]. Up to 15 soil properties were provided at different soil horizon depths. Latitudes and longitudes (in West Africa War System of Coordinates, Transverse Mercator projection, belt I on the Arc 1960 datum) for this dataset are contained within the soil profile descriptions [see page 159 in (42) as an example]. For more details see Hinga et al. (59).

Development of Interpretations

Land Quality Maps

Interpretation of the Busia Area soil survey report showed that land quality maps could be generated depending on the agro-ecological zone within which a specific soil type occurs. To show how these land qualities were generated using the information within the soil survey report, we demonstrate how the available soil moisture for plant growth was generated through the interpretation of the data. Even though in many publications the availability of moisture is defined as a *soil property*, the Busia Area soil survey report defines it as a *land quality*. However, the interpretation of the available soil moisture for plant growth needs to be related to the agro-ecological zones and the distribution of which may have been affected by climate change since the original soil survey was conducted. While available soil moisture capacity (AWC) is relatively stable the available soil moisture is dynamic and changes during the season depending on the interactions between crop type and climate.

Availability of Moisture for Crop Growth

This land quality expresses the period that a plant has adequate available soil water to support normal productive growth. The adequate available soil water is measured in terms of the presence of humid months without limitations for plant growth. The length of the accumulated growing months determines the suitability for a specific plant or crop (42).

TABLE 4 | Soil orders and soil classes of the Busia area according to the Legend for the Soil Map of the World (38) and correlates to the Soil Taxonomy, 12th edition (13).

Soil order	Soil classification system		Frequency (%)
	legend for the soil map of the World	Soil taxonomy, 12th ed.*	
Acrisols	Chromic Acrisols	Ultic Hapludalfs/ Typic Hapludults/ Typic Kanhapludults	16.1
	Orthic Acrisols	Ultic Hapludalfs/ Typic Hapludults/ Typic Kanhapludults	
	Plinthic Acrisols	Typic Plinthudults/Plinthic Kanhapludults	
Cambisols	Chromic Cambisols	Dystric Eutrudepts	21.4
	Dystric Cambisols	Typic Dystrudepts	
	Eutric Cambisols	Dystric Eutrudepts	
	Ferralic Cambisols	Oxic Dystrudepts	
	Vertic Cambisols	Vertic Dystrudepts/ Vertic Eutrudepts	
Vertisols	Chromic Vertisols	Chromic Dystruderts/Typic Dystruderts/Chromic Hapluderts/Typic Hapluderts	7.1
	Pellic Vertisols	Typic Dystruderts/ Typic Hapluderts	
Gleysols	Dystric Gleysols	Humic Endoaquepts	14.3
	Eutric Gleysols	Mollic Endoaquepts/Typic Endoaquepts	
	Humic Gleysols	Typic Humaquepts/Humic Endoaquepts	
	Plinthic Gleysols Vertic Gleysols	Plinthic Petraquepts Vertic Endoaquepts	
Lithosols	Dystric Lithosols	Extremely shallow (< 10 cm to hard rock) Lithic Udorthents	7.1
Nitrisols	Dystric Nitrisols	Typic Kandiodults/ Typic Paleodults	7.1
	Eutric Nitrisols	Typic Kandiodalfs/ Typic Paleodalfs	
Fluvisols	Eutric Fluvisols	Typic Udifluvents/ Typic Fluvaquents/ Mollic Fluvaquents	3.6
	Vertic Fluvisols	Vertic Udifluvents/ Vertic Fluvaquents	
Histosols	Eutric Histosols	euic reaction classes of Typic Haplofibrists/ Typic Haplohemists/ Typic Haplosaprists	1.8
Arenosols	Ferralic Arenosols	Typic Udipsamments	3.6
Solonchaks	Gleyic Solonchaks	Typic Halaquepts	1.8

(Continued)

TABLE 4 | Continued

Soil order	Soil classification system		Frequency (%)
	legend for the soil map of the World	Soil taxonomy, 12th ed.*	
Ferralsols	Orthic Ferralsols	Typic Eutrudox/ Typic Kandiodox/ Typic Hapludox	14.3
	Plinthic Ferralsols	Plinthic Eutrudox/ Plinthic Kandiodox/ Plinthic Hapludox	
	Rhodic Ferralsols	Rhodic Eutrudox/ Rhodic Kandiodox/ Rhodic Hapludox	
Luvisols	Orthic Luvisols	Typic Hapludalfs	1.8

*Classification correlates according to the Keys to Soil Taxonomy, 12th edition for soils in a udic soil moisture regime (56). Classes in the FAO-UNESCO (38) system often fall into two or more subgroups in Soil Taxonomy (13).

TABLE 5 | Metadata information in the GIS layer of the Reconnaissance Soil Map of the Busia Area.

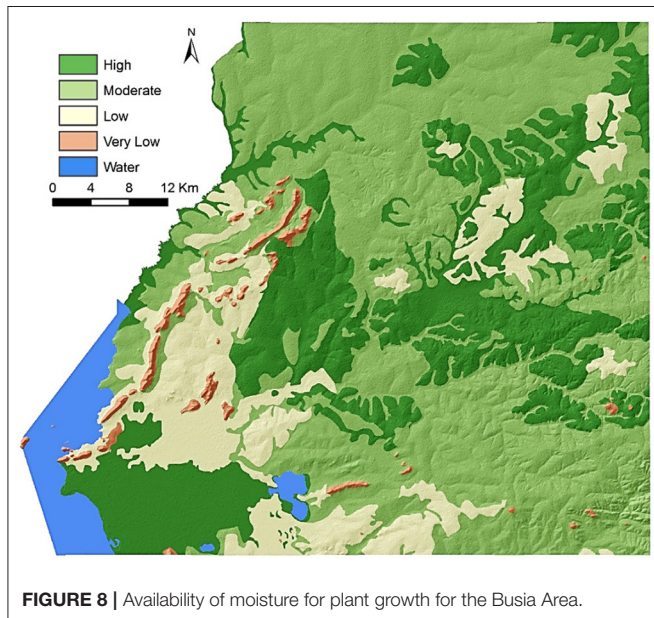
Item	Detail	Description
Data source	Description	Data was obtained from the Kenya Soil Survey library.
ID information	Description	The soil map was created to enhance a systematic inventory of soil and land resources for multipurpose land use planning for the Busia area.
Spatial reference	Description	East Africa War System of the Coordinates Traverse Mercator projection belt I on the Arc 1960 datum. The reference Clark 1880 (54). This projection is named as Arc 1960 UTM Zone 36N under the current ESRI's list of projections.
Data quality	Processing steps	This GIS layer was created by (1) downloading the scanned map from the European Digital Archive of Soil Maps (EUDASM) (40), (2) georeferencing using satellite imagery and the hillshade as basemaps, (3) editing the digital soil map to ensure polylines are placed on top of the soil line units' borders, and (4) attributing, editing, and repopulated tables with information from the legend and soil survey report.

Three different rooting depths, 0–50, 0–80, and 0–120 cm, were used to determine the total available soil moisture capacity for the crop growing months. This approach was explicitly described in the legacy soil survey report [pages 127 and 128 in (42)]. Since soil moisture also is climate-dependent, the agro-climatic map of Kenya was used to map out the different agro-climatic zones in the Busia Area. Four different agro-climatic zones (I, II, III, and IV) were mapped (60). For example, agro-climatic zone I has 11 growing months available for 0–50 and 0–80 cm soil depth and 11.5 months for 0–120 cm soil depth. For agro-ecological zones II to IV, see Tables 11–13 in Rachilo and Michieka (42).

TABLE 6 | Ratings for the availability of moisture for plant growth [Table 14 in (42)].

Month(s) per growing season	Rating*
>11	1
9–5	2
6–9	3
4–5.5	4

*1, High; 2, Moderate; 3, Low; 4, Very low.

**FIGURE 8** | Availability of moisture for plant growth for the Busia Area.

The length of growing season(s) in months was further grouped according to the land quality rating (Table 6). The length of the growing season in this example represents available soil moisture under the assumption that a longer growing season translates to more available soil moisture during the growing season. We computed the land quality ratings for each soil map unit [Appendix 6 in (42)] and used this to map out the availability of soil moisture for the study area (Figure 8). The map of available soil moisture is consistent with what we would expect on this landscape. Hills have low available moisture because the soils are very shallow, consisting of Lithosols with stony phases. Conversely, river terraces and swamps have soils with high available moisture since these are depositional areas where water accumulates.

Similar stepwise approaches were used to map nine additional land quality categories for the study area including: (i) temperature, (ii) availability of nutrients for plant growth, (iii) salinity hazard, (iv) sodicity hazard, (v) erosion susceptibility, (vi) availability of oxygen in the root zone, (vii) flooding hazards, (viii) seedbed preparation and cultivation potential, and (ix) availability of foothold for roots. See Appendix A in Minai (61) and Minai and Schulze (62) to view and download these land quality maps. Although the renewed maps are more consistent from the soil-landscape relationship perspective, their utility is still limited due to lack of field validation.

Crop Suitability Maps

The ratings for the 10 land qualities developed by Rachilo and Michieka (42) were further used to determine suitability classes for specific crops [Tables 32–49 in (42)]. All the land qualities were applied to individual soil mapping units (plural) to determine their suitability for specific crops. We utilized this information in the form of decision matrices to delineate suitability classes for agronomic crops suitable for the study area.

Suitability Class Map for Rainfed Maize

Table 7 [also Table 40 in (42)] was used as the decision matrix to determine the suitability for each soil map unit to support rainfed maize (*Zea mays* L.) growing under intermediate technology. Using this decision matrix, we generated *if-then* (conditional) statements to determine the suitability of each soil map unit in the Busia Area for rainfed maize. The *Join* and *Relate* tools in ArcMap were used to combine the newly created suitability class table with the Busia Area soil map attribute table to delineate the suitability for rainfed maize under intermediate technology for the study area (Figure 9).

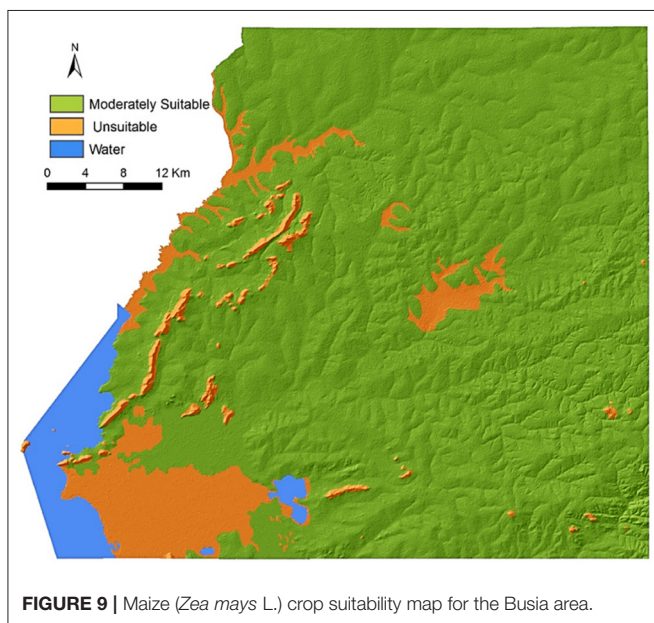
Rachilo and Michieka (42) developed a detailed land evaluation key and decision matrix, showing land suitability classifications for various activities and land use types for all the soil map units within the study area across all four agro-ecological zones [Appendix 3 in (42)]. This decision matrix was used, following similar approaches for developing the suitability map for maize, to produce suitability class maps for additional 18 crops, including (1) sugarcane (*Saccharum officinarum* L.), (2) cabbages (*Brassica oleracea* var. *capitata* L.), (3) kale (*Brassica oleracea* var. *viridis* L.), (4) onions (*Allium cepa* L.), (5) tomatoes (*Lycopersicon esculentum* L.), (6) wetland and upland rice (*Oryza glaberrima* Steud.), (7) citrus guava (*Psidium guajava* L.), (8) cotton (*Gossypium hirsutum* L.), (9) groundnuts (*Arachis hypogaea* L.), (10) maize (*Zea mays* L.), (11) finger millet (*Eleusine coracana* L.), (12) cassava (*Manihot esculenta* Crantz), (13) common beans (*Phaseolus vulgaris* L.), (14) sunflower (*Helianthus annuus* L.), (15) Robusta coffee [*Coffea canephora* var. *robusta* (L. Linden) A. Chev)], (16) forestry, (17) fodder crops, and (18) areas suitable for grazing. See Appendix B in (61), and (63) to view and download the crop suitability maps, respectively.

These maps offer the first attempt to generate additional land quality and crop suitability maps generated from interpretation of an existing soil survey report. These products, however, may need substantial improvement because they are based on agronomic data collected over 40 years ago. Since then, several of the underlying factors used to determine the suitability classes for these crops have likely changed significantly (64). Additional field studies are needed to generate new land quality maps for the study area. For example, current climatic conditions such as temperature and rainfall amount and distribution may not reflect past climatic conditions (65). Similarly, chemical soil property data used to determine land quality ratings for the study area are not static (66). Many soil properties such as organic matter, available phosphorus, exchangeable K, Ca, and Mg, pH-H₂O, can vary even within a growing season (67).

TABLE 7 | Decision matrix for the suitability classification of soils for rainfed maize growing under the intermediate technology option [Table 40 in (42)].

Suitability class	Land qualities								
	Temp	AoM	AoN	SH	Sod	SE	Ox	FH	SPC
Highly suitable (S1)	1	1–2	2–3	1	1	1–2	1–2	1–2	1–2
Moderately suitable (S2)	2, 3, 4	3	4	2–3	2–3	3–4	3–4	3–4	3–4
Marginally suitable (S3)	5, 6	4	4	4	4	5	5	5	5
Unsuitable (NS)	7	4	4	4	5	5	5	5	5

Temp, Temperature; AoM, Availability of moisture for plant growth; AoN, Availability of nutrients; SH, Salinity hazard; Sod, Hazard of sodicity; SE, Susceptibility to erosion; Ox, Availability of oxygen for root growth; FH, Flooding hazard during the growing season; SPC, Possibility of seedbed preparation and cultivation. Intermediate technology refers to the “level of technology where certain inputs such as fertilizers, insecticides, and mechanized land preparation are used on a modest scale” (42).

**FIGURE 9** | Maize (*Zea mays* L.) crop suitability map for the Busia area.

Digital Soil Mapping

The availability of georeferenced soil property from the survey report allowed us to map soil organic carbon and texture using digital soil mapping techniques (14, 61). Similarly, a soil landscape rule-based approach was used to disaggregate the traditional soil polygon map of the survey area into individual soil classes at a spatial resolution of 30 m (13).

CONCLUSION

This study aimed at renewing the best available soil survey report of a selected portion of Kenya into a digital format and using the information in the legacy data to support the provision of additional agronomic information. The *Reconnaissance Soil Survey of the Busia Area* (quarter degree sheet No. 101) was used as an example. Interpretation of the agronomic information in the legacy report resulted in the development of 10 land quality maps and 18 different crop suitability maps that were previously not available.

We demonstrated the feasibility of delivering some of these maps via the cellphone network in rural Kenya. The Exploratory

Soil Map of Kenya (1982), the Reconnaissance Soil Map of the Busia Area, and three additional maps of the Busia area were published to a server and then made available in the Soil Explorer mobile app for iOS and Android devices and in the SoilExplorer.net website. We were able to access these maps in the field on an Apple iPad Mini connected to the Safaricom cellphone network (61, 68).

The process of *legacy data rescue* first resulted in georeferenced soil property data. These data allowed us to map selected soil properties (soil organic carbon and texture for the surface horizon) using digital soil mapping techniques (14). The same information was also used to disaggregate the traditional soil map into individual soil classes within soil map units at a spatial resolution of 30 m using a soil landscape rule-based approach (13). These map products currently are the best available high-resolution soil maps for soil organic carbon and texture for the Busia Area and exemplify the ability to map soil properties using digital soil mapping techniques solely based on data from a previously published soil survey report.

We hope that the procedures described above will provide a blueprint for the rescue of legacy soil data for other areas. Additionally, digital soil mapping can benefit from renewed legacy data as the legacy soil data can provide inputs into digital soil mapping projects.

DATA AVAILABILITY STATEMENT

The raw data supporting the conclusions of this article are available without reservation as per citations in the article.

AUTHOR CONTRIBUTIONS

JM, DS, and ZL contributed to the conception of the study. JM conducted reconnaissance visits, legacy data archeology, and write-up of the manuscript with contributions of DS and ZL. All authors read and approved the submitted manuscript.

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