Application of Geospatial Methods in Evaluating Environmental Interventions and Related Socioeconomic Benefits

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Abstract

Environmental interventions underpin the Sustainable Development Goals (SDGs) and the Rio Conventions. The SDGs are integrated and embody all three aspects of sustainable development—environmental, social, and economic—to capture the interlinkages among the three areas. The Rio Conventions—on biodiversity, climate change, and desertification, also intrinsically linked—operate in the same ecosystems and address interdependent issues, and represent a way of contributing to the SDGs. Assessing the results of environmental interventions and the related socioeconomic benefits is challenging due to their complexity, interlinkages, and often limited data. The COVID-19 crisis has also necessitated creativity to ensure that evaluation’s critical role continues during the crisis. Satellite and other geospatial information, combined with existing survey data, leverage open-source and readily available data to determine the impact of projects. Working with geospatial data helps maintain flexibility and can fill data gaps without designing new and often expensive data tools for every unique evaluation. Using data on interventions implemented by the Global Environment Facility in biodiversity, land degradation, and climate change, we present the application of geospatial approaches to evaluate the relevance, efficiency, and effectiveness of interventions in terms of their environmental outcomes and observable socioeconomic and health co-benefits.

Introduction

Environmental interventions are important mechanisms for delivering the objectives laid out in the Sustainable Development Goals (SDGs) and the United Nations Rio Conventions. The SDGs are integrated and embody all three aspects of sustainable development—environmental, social, and economic—with the intention of capturing the interlinkages among the three areas (United Nations Environment Programme [UNEP], 2013). The three Rio Conventions—on biodiversity, climate change, and desertification, also intrinsically linked—operate in the same ecosystems and address interdependent issues, and represent a way of contributing to the SDGs. The various activities related to sustainable development are linked through feedback mechanisms, resulting in both benefits and tradeoffs. This is where a systems approach that recognizes the dynamic, interdependent complexity of real-
world contexts across different scales, and recognizes dynamic shifts over time, is helpful in addressing environmental issues (Kass, 2019).

The Global Environment Facility (GEF) was set up in 1992 as a financial mechanism for the Rio Conventions. The GEF supports the implementation of projects in five focal areas—biodiversity, climate change, land degradation, international waters, and chemicals and waste—through 18 implementing agencies. Since 2010, the GEF has moved toward integrated programming that seeks to bring about changes in the multiple domains necessary to achieve the desired long-term transformation. These programs consider causes across the environment and different realms of human activity, generate benefits in two or more GEF focal areas, and generate social and economic benefits. This recent emphasis on multifocal and integrated programming presents its own sets of challenges, primarily those of evaluating the results and measuring other related benefits.

Drawing on GEF projects and programs in biodiversity, land degradation, and climate change, this chapter presents the application of geospatial approaches to evaluate the relevance, results, and sustainability of GEF interventions in terms of their environmental outcomes and their socioeconomic and health co-benefits. The first section of the chapter includes an introduction to geospatial data and analysis, the trends in its use, and the reasons behind the increase in the use of these approaches. The next section illustrates the usefulness of geospatial data in evaluation using examples of specific applications by the GEF Independent Evaluation Office (IEO). The final section discusses insights from geodata applications in environmental evaluations.

**Geospatial Approaches and Methods**

Geospatial data is unique because it contains spatially explicit information. The data can be collected from various sources such as remote sensing platforms, geotagged photographs, and ground sensors, or from survey data sets that include such information. Geospatial methods include the creation, collection, analysis, visualization, and interpretation of geospatial data.

Thus, geospatial data and methods can provide spatially explicit, synoptic, time-series data for various earth system processes, and have been used in the monitoring of environmental processes over the past 40 years (Awange & Kyalo Kiema, 2013; Melesse et al., 2007; Spitzer, 1986). Its application in environmental evaluations has gained traction in the last 2 decades. Evaluators initially used geographic information systems mainly to visualize and detect change in combination with other evaluation data (Renger et al., 2002). Others in evaluation have recognized the usefulness of spatial data for determining baselines, outputs, and monitoring of results over time (Azzam, 2013; Azzam & Robinson, 2013). Evaluators have employed quasi-experimental designs (Andam et al., 2008; Buchanan et al. 2016; Ferraro & Pattanayak, 2006) using geospatial data in impact evaluations of biodiversity and forestry interventions. Geospatial analysis has also been used recently in randomized control trials (Jayachandran et al., 2017).

**Drivers of Increased Use**

Recognition of the role of geospatial science by intergovernmental agencies and major environmental and development policy frameworks is growing as countries move toward more evidence-based policy decisions (Lech et al., 2018). The United Nations Convention to Combat Desertification (UNCCD) has recommended using indicators obtained from remote sensing to monitor progress toward reversing and stopping land degradation and desertification (Minelli et al., 2017). The United Nations Framework Convention on Climate Change (UNFCCC) and the Convention on Biological Diversity (CBD) have also endorsed the use of objective indicators, many of which are derived through geospatial methods.

Other factors have influenced the increased use of geospatial data and analysis. First, we have seen an unprecedented flow of spatial data from multiple sources, including satellite data.
Moderate and coarse resolution data is free, and high-resolution data is becoming less expensive and more widely available. The recent developments in data science have influenced the availability and cost of data. The infrastructure and tools to work with large quantities of geospatial data or big geodata have increased substantially. The availability of application programming interfaces (APIs), cloud-based services, and browser-based development environments have allowed access to geospatial data and analysis without the need for significant computational infrastructure (Lech et al., 2018). Traditional statistical tools are often incapable of dealing with the volume and variety of geospatial datasets, thus paving the way for machine learning and artificial intelligence algorithms in the last 5–7 years.

Use of Geospatial Data and Analysis in Monitoring and Evaluation

Evaluators often encounter methodological challenges and data issues during the course of evaluations, including lack of baseline data, sampling bias, difficulties in selecting appropriate counterfactuals, and accounting for the impact of multiple scales and contexts on processes and interventions. Geospatial approaches and tools can effectively address these gaps and can be applied to evaluate environmental and socioeconomic outcomes, to measure environmental change. We can also combine them with existing qualitative and quantitative methods and the results of interventions over time, while recognizing and accounting for the complex interrelationships across the various factors.

Assessing the Relevance of GEF-Supported Interventions to Combat Land Degradation and Desertification

As the financial mechanism of the UNCCD, the GEF uses land degradation focal area strategies consistent with the UNCCD global priorities, including its focus on combating desertification in Africa, emphasis on drylands and non-drylands, and achieving land degradation neutrality. The GEF is gradually moving toward integrated approaches in this area to deliver global environmental benefits in multiple focal areas while generating local environmental and development benefits. A 2017 land degradation focal area study conducted by the IEO analyzed 618 land degradation projects or multifocal area projects with a land degradation component. The study looked at the relevance and performance of GEF’s investments in addressing land degradation.

The IEO used geospatial analysis to assess the relevance of GEF interventions at global, country, and site levels. The analysis involved a feature overlay of the GEF-supported land degradation projects with the areas of land degradation severity. The analysis showed that the GEF implemented interventions to address land degradation in all developing regions of the world (Fig. 1) with Africa appropriately receiving the highest share of land degradation focal area project financing (37%), followed by Latin America and the Caribbean with 24% (GEF IEO, 2017). Africa has the largest share of land with extreme degradation in semi-arid areas (United Nations Environmental Programme [UNEP], 2002). Degraded soils are also found in regions undergoing deforestation such as Indonesia and Brazil and areas with high population pressure such as China, Mexico, and India (UNEP, 2002).

The study also noted that India, Mexico, Brazil, Indonesia, and China received the majority of land degradation financing from the GEF and the majority of national projects focused on forest and agricultural lands and rangelands. Overall, the results from this study showed that the GEF was supporting land degradation projects where most needed and relevant (GEF IEO, 2017).
Assessing Impacts in GEF-Supported Protected Areas

Protected areas are among the critical strategies for biodiversity conservation (DeFries et al., 2005). Global commitments and targets recognize the importance and role of protected areas in biodiversity conservation, ecosystem services management, poverty reduction, and generation of economic benefits. As the financial mechanism for the United Nations Convention on Biological Diversity (CBD), the GEF applies a strategy consistent with the CBD’s strategic plan, reflected in its support to protected areas over the last 26 years. Between 1991 and 2015, the GEF provided $3.4 billion in grants to 618 projects involving protected areas, matched by $12.0 billion in cofinancing, to help protect almost 2.8 million km² of the world’s non-marine ecosystems (GEF IEO, 2016). These figures exclude the support provided by GEF outside of the protected area systems.

Assessing the effectiveness and impact of GEF-supported protected areas is challenging mainly due to the scale, different timelines, and difficulty in collecting primary data due to the remoteness of protected areas. The IEO addressed these challenges using remote sensing data in the Evaluation of GEF Support to Protected Areas and Protected Area Systems (GEF IEO, 2016).¹ At the global level, the evaluation used observations from satellite data, conducting geospatial analysis for the period 2001–2012 in GEF-supported protected areas and their buffers at 10 and 25 km to compare the extent of forest loss in these areas. The study examined forest change for 37,000 protected areas in 147 countries using a global dataset derived from satellite data analysis (GEF IEO, 2016).

Results of satellite data analysis of GEF-supported protected areas demonstrated that these protected areas experienced less forest loss than their surrounding 10 km buffer zones (see Fig. 2). In the 2001–2012 period, GEF-supported protected areas had up to four times less forest cover loss than the overall respective country averages, and at least two times less than protected areas not supported by the GEF in the same biomes and countries.

Using analysis through the biome lens, the evaluation found the greatest loss in protected areas in tropical and subtropical moist broadleaf forests, followed by tropical and subtropical conifers and tropical and subtropical dry broadleaf forest biomes (see Fig. 3). The findings confirmed the global trend of the most extensive forest loss in the tropics, followed by boreal and subtropical forests. The percentage loss of forest cover was highest in temperate conifers and temperate grassland, followed by tropical and subtropical grasslands, savannas, and scrublands; and then tropical and subtropical dry broadleaf forests. These results are consistent with global

¹The evaluation was conducted in collaboration with the Independent Evaluation Office of the UNDP.
trends of tropical and subtropical forests exhibiting the most significant loss, followed by temperate and boreal forests (Hansen et al., 2013). These findings indicate the GEF’s relevance to protected areas and affirm that the funding is going to areas experiencing a significant loss in protected forests—an important indicator of ecosystem integrity.

The results also showed that the median percent forest loss in GEF-supported protected areas was 1.2% percent while averaging 4.1% in the countries. The highest net percent forest losses were seen in Côte d’Ivoire (14.72%), South Africa (6.75%), and Guatemala (5.37%), and the highest net area forest losses were observed in Nicaragua (2528.76 km²), Honduras (1592.78 km²), and Bolivia (1072.29 km²). The results also showed that the median percent forest loss in GEF-supported protected areas was 1.2% percent while averaging 4.1% in the countries. The highest net percent forest losses were seen in Côte d’Ivoire (14.72%), South Africa (6.75%), and Guatemala (5.37%), and the highest net area forest losses were observed in Nicaragua (2528.76 km²), Honduras (1592.78 km²), and Bolivia (1072.29 km²).

Assessing Socioeconomic Co-Benefits

Despite widespread interest and extensive research into the socioeconomic impacts of environmental interventions in the past few decades, evidence remains inadequate and inconsistent (Awange & Kyalo Kiema, 2013; Melesse et al., 2007; Spitzer, 1986). Studies that have attempted to generate analytical insights and evidence have faced challenges, including the varying nature of co-benefits attributable to environmental initiatives, the typology and breadth of implementation approaches, and the data and methods used to assess co-benefits. The difference in methodology, data, and temporal and spatial scales also makes it tricky to draw overarching insights from these studies (Alpízar & Ferraro, 2020; Naidoo et al., 2019).

Studies have begun applying satellite and other spatial data sources to assess the co-benefits of development initiatives. These studies have
demonstrated how to leverage satellite-based data sources for evaluating environmental outcomes. Building on the recent developments in research and impact evaluation, the GEF IEO conducted a study to estimate the global and local-level contributions of GEF environmental initiatives and their related benefits.
IEO evaluators used a geospatial approach to determine the socioeconomic benefits associated with GEF-supported sustainable forest management (SFM) interventions. The GEF has a long history of providing support to improve the sustainability of forestry resources to increase environmental benefits and deliver socioeconomic co-benefits. This evaluation assessed the impacts of GEF-supported SFM interventions on biophysical and ecological variables and co-benefits measured in terms of socioeconomic indicators, and estimated monetary values of ecosystem services using the principle of natural capital accounting (Runfola et al., 2020). To examine the socioeconomic effects, the study used both a portfolio-wide approach (based on night light activity3) and a recent case study from Uganda, which was the first attempt to combine geospatial data with other survey data. To detect the impact of GEF projects on proximate (within 50 km) households, evaluators used the World Bank’s Living Standards Measurement Survey of in-country household information (see Fig. 4).

The evaluation used the geographic locations of GEF SFM projects and data on the measurements of environmental outcomes based on suggested indicators from the CBD (2016) and UNCCD (2015). Night lights are a frequent proxy for socioeconomic outcomes, and the study used satellite-based measurements of nighttime light intensity over time. It also utilized a quasi-experimental approach to analyze GEF interventions’ effectiveness along both environmental and socioeconomic dimensions. Details of this evaluation’s methods and approaches are available in the original evaluation report (GEF IEO, 2019).

The portfolio-level, global scope analysis of economic and social co-benefits of GEF SFM projects indicated a small, positive impact on socioeconomic benefits as indicated by nighttime light intensity. The study found that projects implemented since 2010 showed a positive effect on nighttime lights (+0.24), a proxy for economic development, that had not been observed in prior years. The study noted that, in the absence of precise geographic location information, these findings could have been an underestimate of the actual impacts across the GEF SFM portfolio. The study recognized that results from the nighttime lights at the portfolio level were not evident and expanded the analysis to include local-level data. The local-scale case study in Uganda using survey data helped fill the portfolio-level analysis gap and further explore the impact of GEF SFM projects on socioeconomic outcomes. The results showed that GEF SFM projects were associated with an increase in household assets. By matching the longitudinal survey data locations from the World Bank household survey that were close to GEF interventions to those farther away from GEF intervention sites, the evaluation found that GEF SFM projects were associated with increased household assets between $163 and $353 (within 40–60 km, respectively). The Uganda case study showed that households proximate to a GEF implementation site tended to experience average improvements in assets of approximately $310 (within 50 km) as compared to those that were not close to a GEF implementation site. Although results from a single case study cannot be considered representative of the entire portfolio, the study provided useful insights that help in understanding the main dynamics taking place in these areas.

Assessing Health Co-Benefits

GEF-supported projects and programs seek to influence positive environmental outcomes across critical areas such as biodiversity, land degradation, and climate change by generating global and local environmental benefits. It is well understood that improved environmental outcomes such as cleaner air, water, and soil undeniably contribute to better living conditions and health. The COVID-19 pandemic has compelled a reexamination of the consequences of environmental destruction and its direct implications for human health. Anthropogenic activities leading to land use mismanagement,
fragmentation and destruction of natural habitats, and overexploitation of wildlife have fundamentally created more opportunities for the spread of infectious and zoonotic diseases (Liu et al., 2013; Olivero et al., 2020). Besides zoonotic diseases, poor water and air quality are still the leading causes of mortality worldwide. Therefore, human health issues cannot be separated from the environmental agenda and actions.

Building on the earlier work on SFM in Uganda, the GEF IEO undertook a further analysis to examine and quantify the association between GEF interventions and health co-benefits. The study looked at the health conditions of children under the age of 5 in Kenya,
focusing on health measures including the prevalence of diarrhea and coughs (Fig. 5). The study explored whether improving environmental and socioeconomic co-benefits through GEF-supported projects led to improved health outcomes. It utilized the health survey dataset from the Kenya Department of Health Services (DHS, 2014), which contained 1594 survey clusters, each of which represented 19–25 households. The analyzed projects were drawn from the GEF’s biodiversity, land degradation, climate change, and sustainable forest management focal areas and programs. Only projects implemented before 2014 were considered for the analysis. Evaluators used a quasi-experimental geospatial interpolation (QGI) method on Kenya’s health data to quantify the association between GEF interventions and children’s health conditions. The QGI method has three parameters: sample density, upper distance bound, and maximum matching difference. It uses a propensity-matching approach to pair treated and controlled survey clusters based on covariates. Runfola et al. (2020) provide more details on the QGI approach.

The study observed localized associations in both variables tested, with a 17% reduction in the occurrence of coughs within 10 km of the GEF intervention areas, and a 9% reduction in the occurrence of diarrhea within a distance of less than 3 km. Besides these direct measures of health outcomes, GEF-supported projects also had positive impacts on water access, including the access to source water in dwellings and the presence of water at hand-washing facilities. The results were found to be stronger in clusters closer to GEF interventions (see Fig. 6).

Assessing Outcome Sustainability in Fragile and Conflict Situations

Assessing the sustainability of outcomes is challenging because, in most cases, projects do not have the resources or the mandate to look at the project results after closure. Examining outcome sustainability can be more of an issue in fragile and conflict-affected situations due to logistical and safety concerns. Geospatial analysis using remote sensing data can help in such situations where field visits and primary data collection are not possible.

The IEO used satellite-based data to assess the sustainability of environment-related project outcomes, part of the evaluation of GEF support in fragile and conflict-affected situations (GEF IEO, 2020). The IEO analysis looked at the trends in the change of forest cover in Sapo National Park, Liberia (Fig. 7). Evaluators compared the loss in forest cover for different periods (before, during, and after the project) to those periods in areas outside the protected areas and to trends in the overall national forest cover loss.

Sapo National Park is Liberia’s only national park and a biodiversity hotspot within the Upper Guinea Forest ecosystem. It has faced long-standing threats from illegal farming, hunting, logging, and mining. In postwar Liberia, GEF-supported programming illustrates its catalytic potential in situations affected by conflict and fragility. The project Establishing the Basis for Biodiversity Conservation on Sapo National Park and in South-East Liberia, approved in 2004, marked one of the earliest GEF-funded projects in postwar Liberia. The World Bank implemented the project, and Flora and Fauna International (FFI) executed the project in collaboration with the Forestry Development Authority (FDA) of Liberia. The World Bank’s re-engagement in Liberia started after the Second Liberian Civil War ended in 2003 (Independent Evaluation Group, World Bank [IEG], 2012, p. xiii.). Taking place from 2005–2010, the project was deemed successful, and project documents noted that “implementation occurred within a period of profound governance, environmental, institutional and societal changes in Liberia following a decade and half of the civil instability” (FFI, 2010, p i).

Since then, the GEF has supported various projects in Liberia in different focal areas. Two other relevant GEF-funded projects—Consolidation of Liberia’s Protected Area Network, from 2008 to 2012, and SPWA-BD: Biodiversity Conservation through Expanding the Protected Area Network in Liberia—
(EXPAN)—followed the first project and were also implemented by the World Bank. The Forest Development Authority of Liberia executed the projects. Both of these projects were “built on
successful GEF investments in Sapo NP” (World Bank, 2007, p. 4) and focused on biodiversity conservation, protected area management, community participation, and reducing rural dependence on forests and wildlife in Liberia. Drawing on these projects’ lessons, the World Bank continued its engagement with the forests and protected area interventions in Liberia, expanding the protected area systems and strengthening capacity to maintain them. Ultimately, the Government of Liberia received grant funding ($37.5 million) through the World Bank from the Government of Norway for the cost of the Liberia Forest Sector Project, 2016–2023, which expanded substantially on the initial GEF projects (World Bank, 2016). This project supports priority investments to strengthen the on-the-ground management of Sapo National Park, including physical demarcation, provision...
of vehicles and equipment, and updating the park’s management plans (World Bank, 2016).

The remote sensing analysis results in Fig. 7 indicate minimal forest loss, close to zero deforestation within the park boundary (flat dark line). This could be explained by the prohibition on all economic activities, including mining, within national parks, as per Liberia’s National Park legislation. Legal mining concessions are present in the buffer zone.

The results illustrate how efforts to protect Sapo National Park’s resources during the first project have been sustained beyond the project duration and supported through subsequent interventions. This trend inside the park contrasts with the phenomenal increase in forest loss outside the park borders (see Fig. 8) and in Liberia as whole, mainly driven by illegal activities such as mining and logging for sustenance in the post-war nation.

The Liberian economy is highly dependent on natural resource exports from the mining, forestry, and rubber sectors. According to an International Monetary Fund (2008) study, the small-scale mining sector for gold and diamonds in the country was estimated to involve as many as 100,000 artisanal miners in 2008, but only 48 artisanal and small-scale mining (ASM) miners (Small & Villegas 2012).

The two dips in the forest loss outside the Park (around 2005 and 2010) shown in Fig. 7 coincide with the eviction of illegal gold miners and settlers in Sapo National Park (FFI, 2010). The lack of financial, technical, and human resources, and lack of capacity and conducive legal environment in Liberia to effectively monitor ASM sites and other illegal activities also explain forest loss in the Sapo National Park’s buffer zone (World Bank, 2020).

Conclusions

This chapter demonstrates the utility of geospatial approaches and data to evaluate complex environmental interventions and assess their socioeconomic and health co-benefits. Geospatial analysis can help answer key evaluative questions on relevance, effectiveness, and sustainability of outcomes.

Geospatial methods can save financial and human resources and be very useful when working in hard-to-reach areas, especially in fragile and conflict situations or in a limiting context such as the COVID-19 pandemic. Geospatial analyses are also scalable and provide a cost-effective and efficient approach for meaningful studies at the project site, portfolio level, and global level. The results generated through these methods provide objective evidence and thereby aid transparency. The analyses can also reveal patterns that are not obvious and help in understanding complex processes. Geospatial methods and approaches work well in a mixed-methods framework and assist with common evaluation challenges such as lack of baseline, finding the right counterfactuals, and addressing accessibility issues. The GEF IEO has used geospatial tools for sharing evaluation results through 2-D maps and interactive maps and visualizations. These tools facilitate the communication of complex ideas and information.

As environmental programming becomes complex as it interlinks with other economic and social variables, and the demand for globally consistent and locally relevant data keeps growing, geospatial data and analyses offer an efficient and complementary approach to evaluators to explore new and increasingly complex questions and topics. Whether applied on its own or in combination with other complementary data and processes, geospatial approaches and methods have undoubtedly opened up new avenues for use in evaluation, and are here to stay.

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4The Liberian Government used the term “Voluntary departure” for the 2010–2011 removals.
Fig. 8  Forest Loss Within and Outside of Sapo National Park, Liberia
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