

Article

Assessing the Opportunity Cost of Carbon Stock Caused by Land-Use Changes in Taiwan

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Abstract: As compared with conventional approaches for reducing carbon emissions, the strategies of reducing emissions from deforestations and forest degradation (REDD) can greatly reduce costs. Hence, the United Nations Framework Convention on Climate Change regards the REDD strategies as a crucial approach to mitigate climate change. To respond to climate change, Taiwan passed the Greenhouse Gas Reduction and Management Act to control the emissions of greenhouse gases. In 2021, the Taiwan government has announced that it will achieve the carbon neutrality target by 2050. Accordingly, starting with focusing on the carbon sink, the REDD strategies have been considered a recognized and feasible strategy in Taiwan. This study analyzed the net present value and carbon storage for various land-use types to estimate the carbon stock and opportunity cost of land-use changes. When the change of agricultural land to artificial forests generated carbon stock, the opportunity cost of carbon stock was negative. Contrarily, restoring artificial forests (which refer to a kind of forest that is formed through artificial planting, cultivation, and conservation) to agricultural land would generate carbon emissions, but create additional income. Since the opportunity cost of carbon storage needs to be lower than the carbon market price so that landlords have incentives to conduct REDD+, the outcomes of this study can provide a reference for the government to set an appropriate subsidy or price for carbon sinks. It is suggested that the government should offer sufficient incentives to reforest collapsed land, and implement interventions, promote carbon trading policies, or regulate the development of agricultural land so as to maintain artificial broadleaf forests for increased carbon storage.



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Keywords: REDD+; net present value (NPV); land-use changes; carbon stock; opportunity cost

1. Introduction

During 1951–2010, greenhouse gases increased the average global surface temperature by 0.5 °C to 1.3 °C [1]. Among them, CO₂ is the major human-made greenhouse gas. Between 1980 and 2011, the average annual increase of global CO₂ concentration was 1.7 ppm [2]. After emissions generated by energy consumption, the second major causes of greenhouse gas emissions are deforestation and forest degradation, which account for 17% of the global human-made CO₂ emissions [3–5].

Compared with conventional approaches for reducing carbon emissions, the strategies for reducing emissions from deforestation and forest degradation (REDD) can address climate change issues with greatly reduced costs [6]. As the extension of REDD, the REDD+ strategies additionally include forest protection, carbon storage, and sustainable forest management. Both the REDD and REDD+ strategies are recognized as fundamental strategies for mitigating climate change by the United Nations Framework Convention on Climate Change [7].

Many countries have realized that the REDD and REDD+ strategies contribute considerable financial benefits [8]. In 2011, the value of global REDD projects exceeded USD

87 million and that of afforestation/reforestation (A/R) projects reached USD 65 million, whereas the trading volumes of REDD and A/R projects were 7.3 MtCO_{2e} and 7.6 MtCO_{2e}, respectively [9]. In the same year, the global carbon market grew by 11% in value, to USD 176 billion [10].

The REDD+ strategies are a solution for protecting tropical forests [11] and an essential and proactive incentive approach for carbon emission reduction and sustainable forest management [12]. A study on a dry forest in southwestern Madagascar indicated that potential REDD+ payments can provide sufficient financial incentives to protect forests with relatively low opportunity costs of carbon storage [13]. However, the benefits of REDD+ can only be received through meaningful investments, with the benefits surpassing the costs. Therefore, governments tend to evaluate the costs and benefits of REDD+ investments before making decisions. White and Minang [14] divided costs into the following three categories: opportunity cost, implementation cost, and transaction cost. They concluded that the opportunity cost is the most crucial part of the benefits of REDD+ because the opportunity cost reveals the influence of deforestation, and facilitates evaluations of the amount of fair compensation for land-use change.

Pagiola et al. [15] implied that when a country reduces its forest loss rate to fulfill the REDD obligations, the opportunity cost is the most essential cost among all of the incurred costs. In fact, the opportunity cost indicates the highest net income of the alternative land use regarding deforested areas [16]. Guo and Gong [17] used a partial equilibrium model based on intertemporal optimization to examine the potential and the cost of promoting forest carbon sequestration through taxes/subsidies to land owners for reducing/increasing carbon storage in their forests. Yang and Li [18] estimated the opportunity cost for REDD+ using the net present value (NPV), and discovered that the estimated opportunity costs are uncertain and differ greatly by region. Product price, carbon density, and time are key factors for changes in the opportunity cost. Kovacs et al. [19] used optimization of a structural dynamic model to evaluate the cost of carbon sequestration when afforestation subsidies are used to encourage private landowners to switch from cropland to forests.

Grieg-Gran [20] mentioned that the opportunity cost is (1) applicable for plantations and other types of land use, and is affected by (2) the output of different types of land use according to soil and climate conditions, (3) the scale of the business, (4) the investment and technology employed, and (5) the distance to market and the quality of basic transportation facilities. Other factors influencing the opportunity cost include the price of agricultural products, discount rate, and investment payback period. Moreover, Huettner et al. [11] accentuated that the opportunity cost might change over time. The opportunity cost of agricultural land was likely to increase in the future because of its scarcity caused by the implementation of REDD+ and growing demand for forestry and agricultural products [21]. In this case, large rewards must be provided to prevent deforestation [22].

In recognition of the United Nations Framework Convention on Climate Change and its protocols, as well as relevant international convention resolutions, the Taiwan government passed the Greenhouse Gas Reduction and Management Act in 2015 [23]. This Act is formulated to reduce greenhouse gas emissions, to protect the Earth's environment, and to ensure sustainable development in Taiwan. The long-term goal is to reduce greenhouse gas emissions to less than 50% of those in 2005 by 2050. However, Taiwan has no carbon trading policies, and thus landlords cannot receive REDD+ payments despite the fact that the implementation of afforestation, reforestation, and reductions in deforestation can increase the amount of carbon stock and protect the environment. Thus, if Taiwan had its own carbon offset market, landlords could receive additional income from the REDD+ strategies, which would further incentivize protecting and regenerating forests. This would facilitate achievement of the long-term goal established by the Greenhouse Gas Reduction and Management Act. However, when landlords decide whether to implement the REDD+ strategies, they usually consider whether the cost is lower than the benefit. If the cost is lower than the benefit, the REDD+ strategies can be implemented. Because the opportunity cost is the most important item in cost, it has been crucial to analyze the trade-off between

the NPVs of different land use types and carbon storage, and the opportunity cost of carbon sequestration of land use changes due to the development of REDD+. Accordingly, starting with focusing on the carbon sink, the REDD+ strategies have been considered to be a recognized and feasible approach in Taiwan. However, the forestlands of Taiwan are insufficient and have too high an opportunity cost for the REDD+ strategies. Although previous studies have proposed some approaches to calculate land-use income and estimate the opportunity cost, they rarely investigated how to use the REDD+ strategies to increase the amount of carbon sinks, nor did they consider the opportunity cost of REDD+ in Taiwan. To address these issues, this study examines the balance between the NPV and carbon storage regarding various land-use types to estimate the carbon stock and opportunity cost of land-use changes. The results can provide a reference for the government to set an appropriate subsidy or price for carbon sinks.

2. Materials and Methods

2.1. Theoretical Model

This study mainly employed the method of estimating the opportunity costs of the REDD+ strategies. The procedure of this method is described as follows: (1) examining changes of land use, (2) estimating the carbon storage for major land-use types, (3) calculating the NPV of different land uses, and (4) using the REDD Abacus software to create opportunity cost curves. The opportunity cost curves depicted the opportunity cost of land-use changes among various lands. The measure NTD/ton CO₂e is used to demonstrate the potential amount of carbon stock for each type of land-use change. The forest income was used as the cost of changes in terms of land use, in which the forest income was generated from logging (see Figure 1).

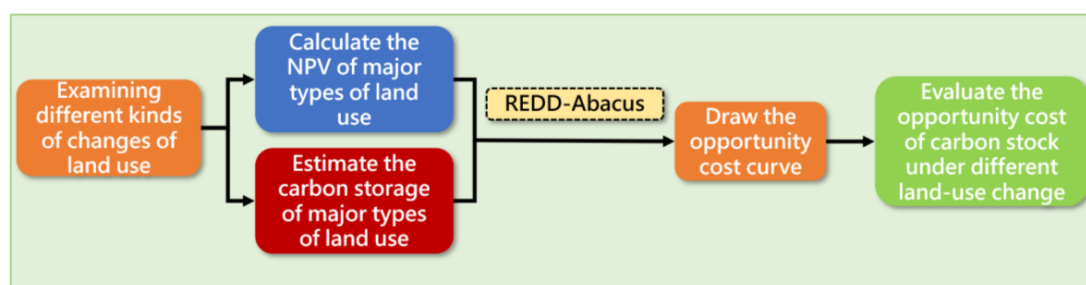


Figure 1. The workflow of this study.

The opportunity cost curves did not provide information on the amount of payments required for the increased amount of carbon stock, but only provided the estimated value of mean carbon stock and marginal opportunity cost [24].

The opportunity cost, using NTD/ton CO₂e as the unit, is calculated as follows:

$$\frac{NPV_{Time2} - NPV_{Time1}}{3.67 \times (Cstock_{Time1} - Cstock_{Time2})}$$

where NPV_{Time1} is the NPV of the original land-use type; NPV_{Time2} is the NPV of the later land-use type; $Cstock_{Time1}$ is the carbon storage of the original land-use type; and $Cstock_{Time2}$ is that of the later land-use type.

2.2. Materials

We divided land use into 10 types, i.e., natural coniferous forests, natural conifer–broadleaf forests, natural broadleaf forests, artificial coniferous forests, artificial conifer–broadleaf forests, artificial broadleaf forests, bamboo–tree mixed forests, bamboo forests, collapsed land, and other types of land (defined as agricultural land in this study) (see Table 1). We also referred to the National Greenhouse Gas Inventory Report released by the Environmental Protection Administration, Executive Yuan, Taiwan in 2019 [25] and

the Forestry Statistics Yearbook publicized by the Forestry Bureau, Council of Agriculture, Executive Yuan, Taiwan in 2019 [26] to obtain the areas nationwide where the use of forests was unchanged, where land was restored to forests during 1990–2017, and where plant cover was reduced by collapse or wind disasters during 2010–2017.

Table 1. Types of land-use changes during 2010–2017 in Taiwan.

Category	Forest Type	2010	2011	2012	2013	2014	2015	2016	2017
Unchanged forest (ha)	Natural coniferous forest	204,528	204,081	203,535	203,380	203,076	202,985	202,929	202,773
	Natural conifer–broadleaf forest	117,151	116,944	116,681	116,570	116,366	116,309	116,284	116,213
	Natural broadleaf forest	1,323,102	1,321,347	1,319,306	1,317,847	1,316,306	1,315,586	1,315,049	1,314,023
	Artificial coniferous forest	82,398	83,283	84,551	86,271	87,702	88,689	89,225	90,475
	Artificial conifer–broadleaf forest	52,625	52,643	52,647	52,647	52,603	52,589	51,360	52,522
	Artificial broadleaf forest	84,965	87,511	90,357	93,256	96,158	99,607	100,927	106,158
	Bamboo–tree mixed forest	114,382	114,296	114,259	114,218	114,183	114,161	114,146	114,128
	Bamboo forest	111,140	111,248	111,476	111,710	111,992	112,111	112,206	112,306
	Total	2,090,291	2,091,353	2,092,812	2,095,899	2,098,386	2,102,037	2,102,126	2,108,599
Planted forest (ha)	Artificial coniferous forest	250	144	150	226	155	139	112	68
	Artificial conifer–broadleaf forest	0	0	0	0	0	0	0	0
	Artificial broadleaf forest	2580	3399	3044	1179	1098	993	826	497
	Bamboo forest	0	1	0	0	0	0	0	0
	Total	2830	3544	3194	1406	1253	1132	938	565
Reduced forest cover caused by collapse or wind disasters (ha)	Natural coniferous forest	1184	447	545	192	341	127	94	193
	Natural conifer–broadleaf forest	835	207	263	148	241	94	62	108
	Natural broadleaf forest	7775	1755	2041	1496	1578	757	574	1062
	Artificial coniferous forest	253	74	83	60	50	18	19	29
	Artificial conifer–broadleaf forest	232	49	48	47	45	13	20	47
	Artificial broadleaf forest	913	150	157	75	97	39	34	147
	Bamboo–tree mixed forest	561	87	37	41	35	22	15	17
	Bamboo forest	357	53	24	45	21	10	17	7
	Total	12,109	2821	3197	2104	2408	1080	834	1610

Sources: National Greenhouse Gas Inventory Report released by the Environmental Protection Administration, Executive Yuan, Taiwan in 2019 [25] and Forestry Statistics Yearbook by the Forestry Bureau, Council of Agriculture, Executive Yuan, Taiwan in 2019 [26].

3. Results

3.1. The NPV of Forest Revenues

We classified forests into the following eight types: natural coniferous forest, natural conifer–broadleaf forest, natural broadleaf forest, artificial coniferous forest, artificial conifer–broadleaf forest, artificial broadleaf forest, bamboo–tree mixed forest, and bamboo forest. The actual revenues of the forests were unknown because they had not been deforested, but the wood had value. Accordingly, we multiplied the stock volume of forests by their wood price in order to obtain their stumpage value.

3.2. Forest Stock Volumes

Compiled according to the fourth forest resource survey conducted by the Forestry Bureau, Council of Agriculture, Executive Yuan, Taiwan in 2015 [27], the stock volume per hectare of each forest type is demonstrated in Table 2.

Table 2. Stock volume of current forest types in Taiwan.

Forest Type	Number of Plots	Stock Volume (m ³ /ha)	Standard Error (m ³ /ha)
Natural coniferous forest	241	499	31
Natural conifer–broadleaf forest	113	486	39
Natural broadleaf forest	2073	198	4
Artificial coniferous forest	153	318	18
Artificial conifer–broadleaf forest	75	287	25
Artificial broadleaf forest	135	147	15
Bamboo–tree mixed forest	314	67	14
Bamboo forests	526	12,773 (bamboo/ha)	3587 (bamboo/ha)

Source: the fourth forest resource survey conducted by the Forestry Bureau, Council of Agriculture, Executive Yuan, Taiwan in 2015 [25].

3.3. Wood Prices of Forests

3.3.1. Natural Coniferous Forests

According to the fourth forest resource survey conducted by the Forestry Bureau, Council of Agriculture, Executive Yuan, Taiwan in 2015 [27], natural coniferous forests can be categorized into hemlock, fir, spruce, natural cypress, natural pine, and other natural coniferous forests. We used the wood price information system established by the Forestry Bureau, Taiwan in 2020 [28], and applied the mean wood price of hemlock, fir, spruce, natural cypress (i.e., red, and yellow cypresses), and pine from April 2010 to April 2020 as the wood price of natural coniferous forests per cubic meter, and obtained a result of 14,561 NTD/m³.

3.3.2. Natural Broadleaf Forests

In the fourth forest resource survey conducted by the Forestry Bureau, Taiwan in 2015 [25], natural broadleaf forests had no subdivisions. Hence, we adopted the mean miscellaneous wood price between April 2010 and April 2020, based on the wood price information system established by the Forestry Bureau, Taiwan in 2020 [28] as the wood price of natural broadleaf forests per cubic meter, and obtained a result of 2537 NTD/m³.

3.3.3. Natural Conifer–Broadleaf Forests

Natural conifer–broadleaf forests were not subdivided by the fourth forest resource survey conducted by the Forestry Bureau, Taiwan in 2015 [27]. Therefore, we calculated the wood price of natural conifer–broadleaf forests per cubic meter using the mean wood price of natural coniferous and broadleaf forests and obtained a result of 8549 NTD/m³.

3.3.4. Artificial Coniferous Forests

Artificial coniferous forests were subdivided into artificial cypress, incense cedar, pine, fir, Japanese cedar, Taiwania, and other artificial coniferous forests in the fourth forest resource survey conducted by the Forestry Bureau, Taiwan in 2015 [27]. We used the aforementioned wood price information system, applied the mean wood price of cypress (i.e., red, and yellow cypresses), incense cedar, pine, fir, Japanese cedar, and Taiwania during April 2010–April 2020 as the wood price of artificial coniferous forests per cubic meter, and obtained a result of 4637 NTD/m³.

3.3.5. Artificial Broadleaf Forests

Artificial broadleaf forests were subdivided into artificial Taiwan zelkova, camphor tree, Taiwan acacia, Formosan ash, and other artificial broadleaf forests in said survey by the Forestry Bureau, Taiwan in 2015 [27]. We used the mean wood price of Taiwan zelkova, camphor tree, Taiwan acacia, and Formosan ash during April 2010–April 2020 as the wood price of artificial broadleaf forests per cubic meter, based on the aforementioned system, and obtained a result of 3794 NTD/m³.

3.3.6. Artificial Conifer–Broadleaf Forests

The artificial conifer–broadleaf forests had no subdivisions. Hence, the mean wood price of artificial coniferous and broadleaf forests was applied to calculate the wood price of artificial conifer–broadleaf forests per cubic meter, and a result of 4215 NTD/m³ was obtained.

3.3.7. Bamboo Forests

The bamboo forests were divided into two types: single-branch bamboo and clustered bamboo. The price of the two types of bamboo was calculated using the price in the Market Statistics of Taiwan Bamboo Utilization Market [29]. From 2003 to 2012, the total bamboo production in Taiwan was 15.84 million (45.175 mt), with a mean price of 2900 NTD/mt. Accordingly, the price of a bamboo was 8.27 NTD/m³.

3.3.8. Bamboo–Tree Mixed Forests

We applied the mean of the NPV of natural broadleaf forest per hectare (500,470 NTD/ha) and bamboo forest per hectare (100,560 NTD/ha) as the NPV of bamboo–tree mixed forest per hectare (300,520 NTD/ha), and obtained a wood price of bamboo–tree mixed forest of 4548 NTD/m³.

The stumpage value (million NTD/ha) of each forest type was obtained by multiplying the stock volume per hectare of forest by its wood price. The results are displayed in Table 3.

Table 3. Stock volume, wood price, and NPV of current forest types in Taiwan.

Forest Type	Stock Volume (m ³ /ha)	Wood Price (NTD/m ³)	NPV (Million NTD /ha)
Natural coniferous forest	499	14,561	727
Natural conifer–broadleaf forest	486	8549	415
Natural broadleaf forest	198	2537	50
Artificial coniferous forest	318	4637	147
Artificial conifer–broadleaf forest	287	4215	121
Artificial broadleaf forest	147	3794	56
Bamboo–tree mixed forest	67	4548	30
Bamboo forest	12,773 (bamboos/ha)	8 (NTD/bamboo)	10

Sources: the fourth forest resource survey conducted by the Forestry Bureau, Council of Agriculture, Executive Yuan, Taiwan in 2015 [27]. The wood price information system was established by the Forestry Bureau, Council of Agriculture, Executive Yuan, Taiwan in 2020 [28], and Lin et al. [27]. A discount rate for afforestation loans was from Bank of Taiwan in 2020 [30].

3.4. Carbon Price

Because Taiwan does not have a carbon market, we referred to Article 28 of the Greenhouse Gas Reduction and Management Act (penalty of a maximum of NTD 1500/tCO₂e for excess emissions; Environmental Protection Administration, Executive Yuan, 2015) and defined the carbon price as NTD 1500/tCO₂e.

3.5. Calculation of the NPV of Collapsed Land

Collapsed land contained no forests, and could no longer sustain production. Therefore, their NPVs (revenues) were regarded as 0 NTD/ha.

3.6. Calculation of the NPV of Agricultural Land

Most land converted for agriculture was located on hillsides and used for crops including tea, temperate fruits, alpine vegetables, and betel nut trees. In Taiwan, the crops that are commercially cultivated at a considerable scale include temperate fruits, namely pears, persimmons, plums, peaches, and apples [31] as well as alpine vegetables such as cabbages and spinaches. Accordingly, we calculated the NPV of the other land-use types using the mean output value per hectare of these crops, a one-year harvesting period, and a discount rate of 1.089% [30]. A mean NPV of 1.1632 million NTD/ha was obtained (Table 4).

Table 4. The NPV of various types of agricultural land-use.

Agriculture Land-Use Type	Output Quantity (kg/ha/year)	Price (NTD/kg/year)	Output Value (10,000 NTD/ha/year)	NPV (10,000 NTD/ha/year)
Pears	17,300	49.03	84.82	83.91
Peaches	13,570	98.55	133.74	132.29
Apples	17,465	56.63	98.90	97.82
Green plums	4933	25.55	12.61	12.47
Cabbages	56,462	17.00	95.98	284.84
Spinaches	20,343	24.01	48.85	144.96
Betel nut trees	3221	80.13	25.82	25.54
Mean	19,042	50.13	71.53	116.32

Sources: Young Farmers Counseling Platform, Council of Agriculture, Executive Yuan, Taiwan in 2020 [31], Agricultural Product Origin and Price Report System in 2020 [30], System of Agricultural Products Price in 2020 [32], Agriculture Production and Marketing Information Platform in 2020 [33], Agricultural and Food Agency Statistics Database in 2020 [34], Production and Marketing Analyses of Pears [35], Lu et al. [36], and the national statistics released by the Directorate-General of Budget Accounting and Statistics, Executive Yuan in 2014 [37], Bank of Taiwan, Discount Rate in 2020 [30].

3.7. Calculation on Forest CO₂ Storage

Relying on the fourth forest resource survey conducted by the Forestry Bureau, Council of Agriculture, Executive Yuan, Taiwan in 2015 [25], we employed the equation proposed by the Intergovernmental Panel on Climate Change (IPCC) to obtain the CO₂ storage and areas of various forests in Taiwan. We divided the CO₂ storage of forests by areas to obtain their CO₂ storage per hectare (Table 5). The CO₂ storage per hectare of natural and artificial coniferous forests, natural and artificial conifer–broadleaf forests, natural and artificial broadleaf forests, bamboo–tree mixed forests, and bamboo forests was 522.99 tCO₂/ha, 604.77 tCO₂/ha, 328.87 tCO₂/ha, 103.07 tCO₂/ha, and 110.36 tCO₂/ha, respectively.

Table 5. CO₂ storage of various types of forest land-use.

Forest Land-Use Type	CO ₂ Storage (10,000 mt)	Areas (ha)	CO ₂ Storage per Hectare (tCO ₂ /ha)
Coniferous forest	15,627	298,803	522.99
Conifer–broadleaf forest	10,361	171,321	604.77
Broadleaf forest	46,899	1,426,058	328.87
Bamboo–tree mixed forest	1278	123,995	103.07
Bamboo forest	1463	132,572	110.36

Source: the fourth forest survey released by Forestry Bureau, Council of Agriculture, Executive Yuan, Taiwan in 2015 [25].

3.8. CO₂ Storage in Collapsed Land

We assumed that collapsed land contained no forest or vegetation, and thus we defined its CO₂ storage as 0 tCO₂/ha.

3.9. CO₂ Storage in Agricultural Land

We assumed that hillside farming and pastoral land were used mainly as fruit tree farms, betel nut farms, and vegetable gardens. According to [38], the CO₂ storage of fruit tree farms and vegetable gardens are 452.1 tCO₂/ha and 126.5 tCO₂/ha, respectively. Based on [36], the CO₂ storage of betel nut farms is 154.0 tCO₂/ha. The mean CO₂ storage of fruit tree farms, betel nut farms, and vegetable gardens (244.2 tCO₂/ha) was applied to agricultural land (Table 6), obtaining a result of 244.2 tCO₂/ha.

Table 6. CO₂ storage of various types of agricultural land-use.

Agricultural Land-Use Type	CO ₂ Storage per Hectare (tCO ₂ /ha)
Betel nut trees	154.00
Fruits	452.10
Vegetables	126.50
Mean	244.20

Sources: Compiled from [38,39].

4. Discussion

4.1. The NPV and CO₂ Storage Regarding Land-Use Categories

Different land-use classifications have different NPV and CO₂ storage. The NPV indicates the discount of land output value (i.e., land profitability) and CO₂ storage refers to the amount of aboveground CO₂ fixation. Based on the data for the two indicators (Figure 2 and Table 7), we categorized land use into the following four classes.

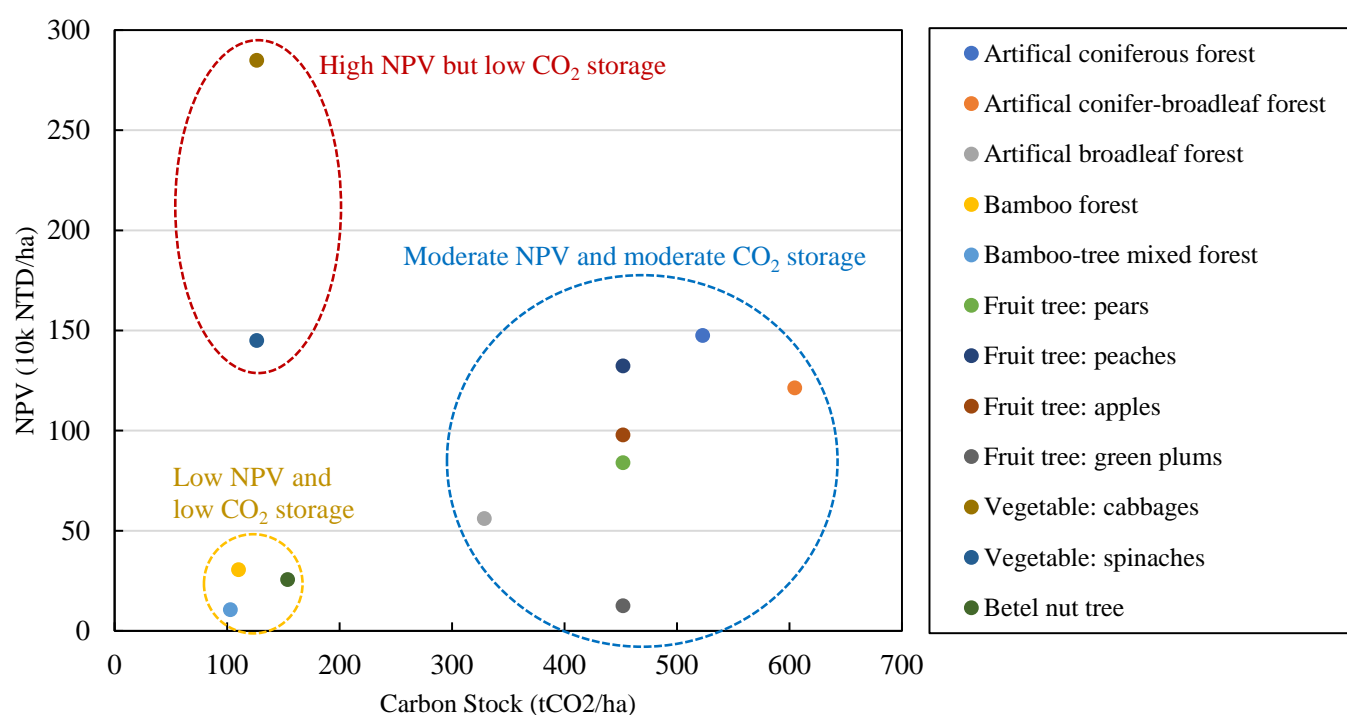


Figure 2. Plots of the NPV versus the CO₂ storage of various land-use types, which are divided into three categories: (1) high NPV but low CO₂ storage, (2) moderate NPV and moderate CO₂ storage, and (3) low NPV and low CO₂ storage.

Table 7. NPV and CO₂ storage of various land-use types.

Land-Use Type	NPV (10,000 NTD/ha)	CO ₂ Storage (tCO ₂ /ha)
Natural coniferous forest	727	522
Natural conifer-broadleaf forest	415	604
Natural broadleaf forest	50	328
Artificial coniferous forest	147	522
Artificial conifer-broadleaf forest	121	604
Artificial broadleaf forest	56	328
Bamboo-tree mixed forest	30	103
Bamboo forest	10	110
Fruit tree: pears	83	452
Fruit tree: peaches	132	452
Fruit tree: apples	97	452
Fruit tree: green plums	12	452
Vegetable: cabbages	284	126
Vegetable: spinaches	144	126
Betel nut tree	25	154
Agricultural land	111	244
Collapsed land	0	0

4.1.1. High NPV and High CO₂ Storage

Natural coniferous and conifer-broadleaf forests belonged to this classification. Natural trees have long growth times, larger diameters and volumes, and thus higher stumpage value. Furthermore, larger volumes suggest higher CO₂ storage. However, we did not include the data of natural forests in Figure 2 for comparison because natural forests in Taiwan are under logging bans.

4.1.2. High NPV but Low CO₂ Storage

Vegetables in agricultural land (i.e., cabbages and spinaches) belonged to this classification. Vegetables allow profits within a short time, and have a high unit price, as well as

preferable profitability and, thus, a high NPV. However, land planted with vegetables has low CO₂ storage, leading to low carbon fixation.

4.1.3. Moderate NPV and Moderate CO₂ Storage

Artificial forests (i.e., artificial coniferous, conifer–broadleaf, and broadleaf forests) and fruit trees in agricultural land (i.e., pears, peaches, and apples) belonged to this classification. These two forest types have modest levels of profitability and CO₂ storage ability.

4.1.4. Low NPV and Low CO₂ Storage

Bamboo–tree mixed forests, bamboo forests, and betel nut trees in agricultural land belonged to this classification. These forest types have relatively low profitability, and have undesirable CO₂ fixation capacity.

The results showed that a single land-use type was unlikely to have preferable probability (NPV) and carbon fixation ability (CO₂ storage) simultaneously. Natural forests had high NPV and CO₂ storage, but they required a long time to grow and could not be logged at random. Moreover, a high NPV could only be realized where large amounts of actual revenue could be obtained through harvesting. Accordingly, artificial forests and agricultural land with fruit trees were better choices to obtain both profitability and carbon fixation ability.

4.2. Opportunity Cost of Carbon Stock among Different Land-Use Changes

The opportunity cost curve of carbon stock was calculated according to land-use changes, the CO₂ storage, and NPVs of different land-use types. In Figure 3, the opportunity cost curve represents a comparison of the potential amount of carbon stock and its cost, in which the potential amount of carbon stock refers to the carbon emissions resulting from land-use changes. The horizontal axis depicts the amount of carbon emissions (i.e., the potential amount of carbon stock), while the vertical axis indicates the opportunity cost of carbon stock.

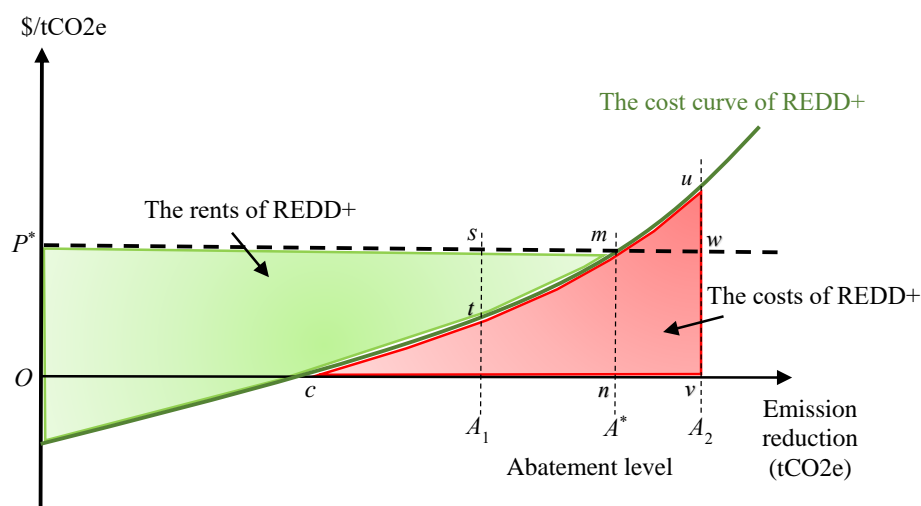


Figure 3. The opportunity cost curve of REDD+. Note that the opportunity cost curve represents a comparison of the potential amount of carbon stock and its cost, in which the potential amount of carbon stock refers to the carbon emissions resulting from land-use changes. The REDD+ payment is the rectangular area under carbon price ($\square OP^*mn$), the cost is the area under the opportunity cost curve (Δcmn), and the difference between the two is the net income (trapezoid OP^*mc). A region with an emission reduction level lower than A^* (e.g., A_1) relinquishes a certain amount of net income (Δtsm), whereas a region with an emission reduction level higher than A^* (e.g., A_2) has additional expenditures (trapezoid $nmuv$), and additional REDD+ revenue ($\square nmuv$) cannot compensate for the loss (Δmuw).

Figure 3 also shows that the carbon price is equivalent to the cost of the carbon stock when the emission reduction level and the carbon price are A^* and P^* , respectively. Under such an emission reduction level, the REDD+ payment is the rectangular area under the carbon price ($\square OP^*mn$), the cost is the area under the opportunity cost curve (Δcmn), and the difference between the two is the net income (trapezoid OP^*mc). A region with an emission reduction level lower than A^* (e.g., A_1) relinquishes a certain amount of net income (Δtsm), whereas a region with an emission reduction level higher than A^* (e.g., A_2) has additional expenditures (trapezoid $nmuv$), and additional REDD+ revenue ($\square nmuv$) cannot compensate for the loss (Δmuw).

Land-use change can result in two situations: carbon emissions and carbon stock. In the first situation, the opportunity cost of carbon would be negative. In the second situation, the opportunity cost of carbon would be positive. To examine the opportunity cost curves, we analyzed the NPV and CO₂ storage for 2010, 2012, 2015, and 2017, respectively (Figures 4–7 and Table 8). Data for areas in which forests remained unchanged, in which land was converted to forests, in which forest cover was reduced by collapse or wind disaster, and different forest types and other types of land (defined as agricultural land in this study) were analyzed.

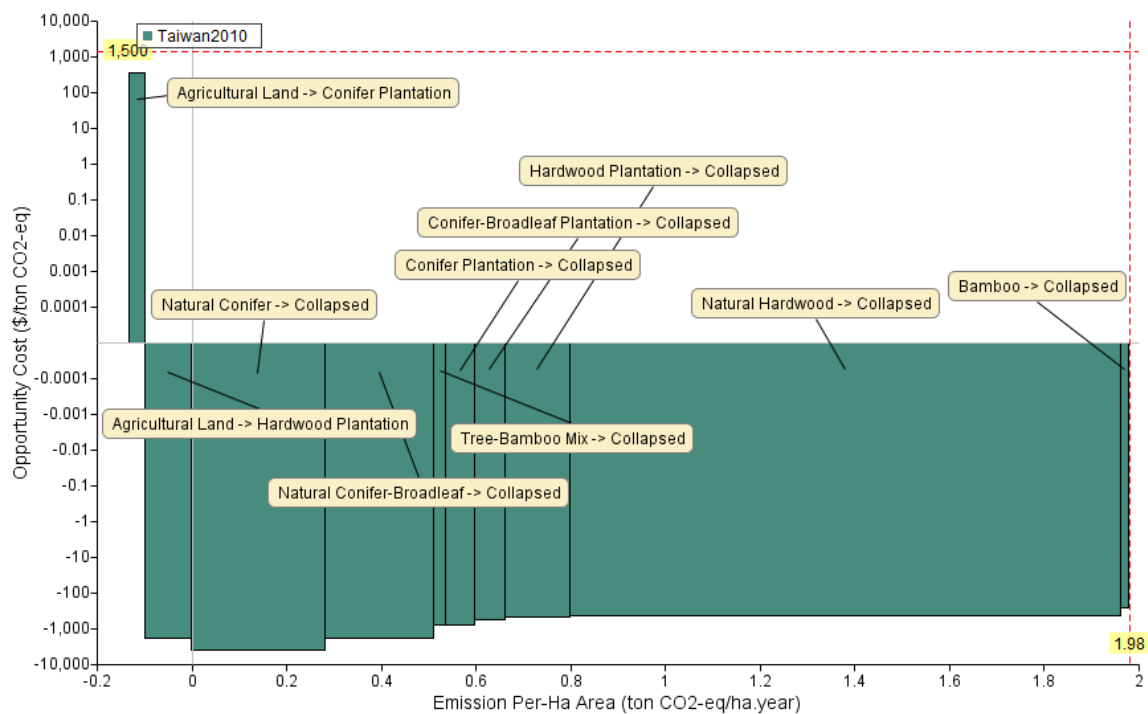


Figure 4. Opportunity costs of various kinds of land-use change in 2010. Note that the opportunity cost refers to the unit cost sacrificed by one kind of land-use change. In addition, the carbon emissions are mostly caused by changes from forests to collapsed land.

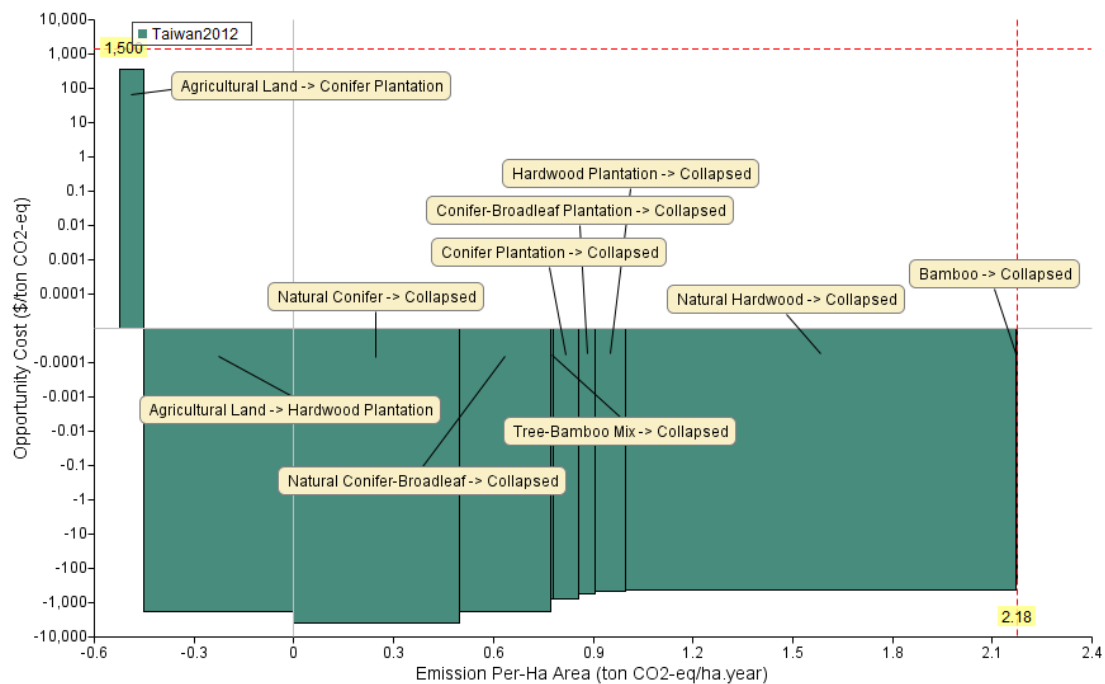


Figure 5. Opportunity costs of various kinds of land-use change in 2012. Note that the opportunity cost refers to the unit cost sacrificed by one kind of land-use change. In addition, the carbon emissions are mostly caused by changes from forests to collapsed land.

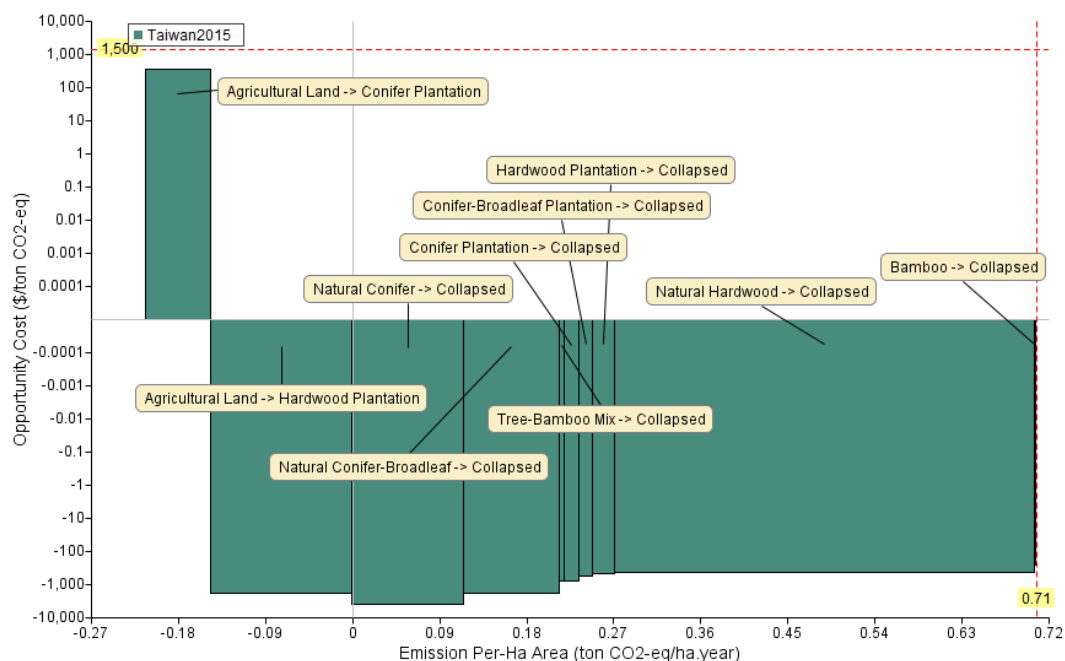


Figure 6. Opportunity costs of various kinds of land-use change in 2015. Note that the opportunity cost refers to the unit cost sacrificed by one kind of land-use change. In addition, the carbon emissions are mostly caused by changes from forests to collapsed land.

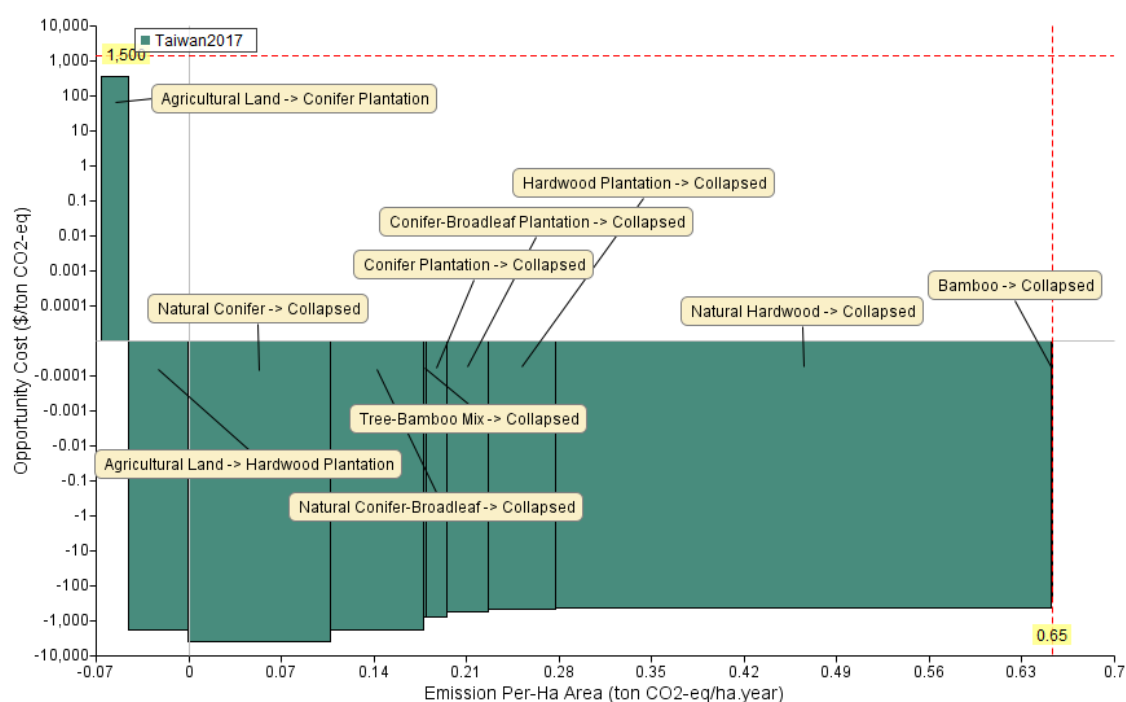


Figure 7. Opportunity costs of various kinds of land-use change in 2017. Note that the opportunity cost refers to the unit cost sacrificed by one kind of land-use change. In addition, the carbon emissions are mostly caused by changes from forests to collapsed land.

Table 8. Opportunity cost of carbon stock in various kinds of land-use change.

Initial Land-Use Type	Later Land-Use Type	Opportunity Cost of Carbon Stock (NTD/tCO _{2e})
Natural coniferous forest	Collapsed land	−3793.85
Natural conifer–broadleaf forest	Collapsed land	−1874.69
Natural broadleaf forest	Collapsed land	−418.57
Artificial coniferous forest	Collapsed land	−769.00
Artificial conifer–broadleaf forest	Collapsed land	−547.02
Artificial broadleaf forest	Collapsed land	−464.57
Bamboo–tree mixed forest	Collapsed land	−807.53
Bamboo forest	Collapsed land	−261.06
Agricultural land	Artificial coniferous forest	349.98
Agricultural land	Artificial coniferous forest	−1793.12

Forests that change into collapsed land resulted in carbon emissions, with changes from natural coniferous forest, natural conifer–broadleaf forest, and natural broadleaf forest into collapsed land being the most common land-use changes. Typhoons and heavy rains also lead to reductions of forest coverage. The carbon emissions at the front end of the curves in Figures 4–7 are mostly caused by changes from forests to collapsed land. The results demonstrate that the opportunity cost of carbon stock regarding the change from a forest to collapsed land is negative. This implies that carbon stock offers considerable benefits and contributes to income instead of cost. In addition, to prevent carbon emissions generated by the collapse of forests, the opportunity cost of carbon stock must be lower than carbon price to ensure that the investment in carbon stock is valuable. When the opportunity cost of carbon stock regarding the change from forest to collapsed land is negative and the carbon price is positive, the opportunity of carbon stock is lower than the carbon price, indicating that the carbon stock is profitable.

Moreover, agricultural land changes (i.e., changes from agricultural land to artificial coniferous forests and artificial broadleaf forests) also increase carbon stock, and the opportunity cost of carbon emissions resulting from change from agricultural land to

artificial broadleaf forest is negative. Although the conversion of artificial broadleaf forest to agricultural land generates carbon emissions, carbon stock does not result in loss but contributes additional income. Therefore, artificial broadleaf forests are likely to be restored to agricultural land. When the opportunity cost of carbon emissions regarding the change from agricultural land to artificial coniferous forest is positive, the profitability of carbon emissions depends on the carbon price. If the opportunity cost of carbon emissions is less than the carbon price, land use with more carbon emissions becomes more advantageous.

5. Conclusions and Suggestions

With a special geographical and climate environment, Taiwan has many high mountains (more than 3000 m²) and is affected by many typhoons, with the result that forestland collapse or landslides often occur. Therefore, it is very important to analyze the changes in carbon stock caused by land use changes (man-made or natural factors). We categorized hillside land in Taiwan into 10 land-use types, accessed the NPV and CO₂ storage of different land-use types, and further divided land types into four land-use scenarios: (1) high NPV and high CO₂ storage (natural forests), (2) high NPV but low CO₂ storage (vegetable gardens), (3) moderate NPV and moderate CO₂ storage (artificial forests and fruit trees), and (4) low NPV and low CO₂ storage (bamboo forests and betel nut farms). We concluded that choosing the land-use type with moderate NPV and moderate CO₂ storage can achieve the maximum amount of environmental and economic benefits.

Carbon emissions were mostly generated by collapsed forestland. The results indicate that the opportunity cost of carbon stock for the change from forests to collapsed land was negative. It implies that carbon stock offers sizeable benefits and contributes to income rather than cost. To ensure that an investment in carbon stock is beneficial, carbon emissions caused by collapsed land should be prevented. When the opportunity cost of carbon stock for the change from forest to collapsed land was negative, the carbon price became positive. This suggests that the opportunity cost of carbon stock was less than the carbon price, so that carbon stock was profitable and could attract investment in carbon stock from governments and forest owners. Furthermore, changing agricultural land into artificial forests generated carbon stock, and the opportunity cost of carbon emissions regarding the change from agricultural land to artificial broadleaf forest was negative. This indicates that restoring artificial broadleaf forests back to agricultural land generated carbon emissions, but carbon stock resulted in additional income rather than cost. Artificial broadleaf forests were, thus, likely to be restored to agricultural land.

The carbon emissions caused by the collapse of forests represented the loss of potential carbon stock in Taiwan, so that the negative opportunity cost of carbon stock represented additional income. Hence, we conclude that greater income could be obtained by offering enough incentive to encourage governments and forest owners to reforest collapsed land or prevent the reductions of forest coverage that are caused by collapse or wind disasters. Moreover, according to the results, among the changes of artificial forests to collapsed land, the opportunity cost of carbon stock for the change of artificial coniferous forest to collapsed land was the lowest. Therefore, reforesting collapsed land with coniferous trees would be the most beneficial course of action. However, suitable land and trees during reforestation should be identified as a first priority.

The opportunity cost of carbon emissions generated by the change from agricultural land to artificial broadleaf forest was negative, implying that carbon emissions increased in this case and contributed to additional income. Accordingly, artificial broadleaf forests present incentives for being converted to agricultural use. If governments hope to increase carbon storage in Taiwan and do not prefer to transform artificial broadleaf forests into agricultural land, they can consider the following measures: (1) strictly prohibiting overuse of forests and preventing owners from restoring artificial broadleaf forests to agricultural land, (2) implementing carbon trading or offset policies and increasing payment for carbon stock to maintain incentives to retaining artificial broadleaf forests, (3) increasing the operation costs of agricultural land, such as charging environmental protection fees for

hillside agricultural land use to reduce the incentive to change forests into agricultural land, and (4) increasing afforestation rewards and incentives.

We used the forest data of Taiwan to carry out the net present value and carbon storage in different land use types, analyzed the opportunity cost curve of carbon sequestration of land use changes, and further evaluated the opportunity cost of carbon sequestration under land use changes. Sensitivity analysis was not conducted to justify the parameters. Considering sensitivity analysis is important to simulation analysis, and as such it is suggested that subsequent researchers conduct sensitivity analysis around the different impacts of future climate change and different occurrence probability of forest collapsed.

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