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**Understanding the relation between climate change risks  
and biodiversity disclosures: An international analysis**

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Manuscript Type:	Research Paper
Keywords:	Biodiversity disclosures, Climate change risks, Carbon emissions, Waste generation, Sustainable development, Cross-country



## Understanding the relation between climate change risks and biodiversity disclosures: An international analysis

### Abstract

#### Purpose

This study explores the relation between firm-level climate change risks, measured by carbon emissions and waste generation, and the level of biodiversity disclosures.

#### Design/methodology/approach

Drawing on an international sample from 2009 to 2021, our study employs panel regression models to assess the effects of climate change risks on biodiversity disclosures. We also conduct a range of sensitivity analyses, including additional proxies, endogeneity tests, and alternative samples to examine the robustness of our inferences.

#### Findings

We find that firms with higher carbon emissions and waste generation levels tend to disclose extensive biodiversity information. Furthermore, we provide evidence that the disaggregated components of carbon (Scope 1 and 2) emissions and waste (hazardous and non-hazardous) generation volumes are positively associated with biodiversity disclosures. Our results also reveal that the effects of climate change risks on biodiversity disclosures are stronger for firms from environmentally sensitive industries. Finally, our results show that climate and biodiversity protection regulations appear to be effective in limiting legitimization efforts.

#### Originality/value

Consistent with legitimacy theory, our findings suggest that high carbon and waste emitting firms tend to utilize increased biodiversity disclosures as a legitimizing tool to conform to societal expectations and protect their legitimacy.

**Keywords:** Biodiversity disclosures; Climate change risks; Carbon emissions; Waste generation, Sustainable development, Cross-country

## 1. Introduction

Along with global warming and climate change, biodiversity loss has been recognized as the most challenging and rapidly increasing threat to the planet, which is currently experiencing the sixth wave of mass species extinction (Adler *et al.*, 2018; Atkins and Maroun, 2018; Roberts *et al.*, 2021). Given that modern businesses and biodiversity systems have reciprocal relationships, including the impacts of business activities on biodiversity and the subsequent effects of deteriorating biodiversity on corporate sustainability (Boiral and Heras-Saizarbitoria, 2017; Hassan *et al.*, 2020), preserving biodiversity and ecosystems enables businesses to identify ecological risks/threats, enhance stakeholder relationships, and improve sustainability (Haque and Jones, 2020; Treepongkaruna, 2024). Therefore, global organizations whose survival and sustainability are dependent on critical ecosystems' services should treat biodiversity and species as their primary stakeholders (Roberts *et al.*, 2021).

There is a growing interest in an emerging body of biodiversity research, especially with regard to corporate accountability towards biodiversity conservation and reporting (Atkins and Maroun, 2018; Gaia and Jones, 2017). However, the majority of prior studies have been mostly qualitative and mainly focused on assessing biodiversity models (Siddiqui, 2013), analyzing biodiversity management (Boiral and Heras-Saizarbitoria, 2017), and exploring changes in biodiversity (Atkins and Maroun, 2020). Few empirical studies have explored whether biodiversity partnerships (Adler *et al.*, 2018; Hassan *et al.*, 2020), environmental NGOs (Gaia and Jones, 2020), and board gender diversity (Carvajal *et al.*, 2022; Haque and Jones, 2020) influence biodiversity reporting. While past research (Ding *et al.*, 2023; Giannarakis *et al.*, 2017) has explored the impacts of corporate carbon/environmental performance on other forms of environmental disclosures, it is still unclear how firm-level climate change risks (CCRs) influence biodiversity disclosures (BIODSL). This study, therefore, examines the effects of CCRs on BIODSL.

Meanwhile growing concerns over the rapidly increasing biodiversity crisis have led the global regulators and policymakers to undertake environmental initiatives/reforms aiming to protect biodiversity and ecosystems (Adler *et al.*, 2018; Gaia and Jones, 2017). For example, the United Nations (UN) introduced the Paris Agreement in 2015 to prevent the global temperature increase and achieve carbon neutrality worldwide by 2050 (Bilal *et al.*, 2024; Luo and Tang, 2021). In addition, the UN adopted the Convention on Biological Diversity (CBD) in 1992 at the Rio Earth

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3 Summit. Ratified by all UN members except the USA, the CBD is considered one of the main  
4 global efforts to combat climate change, halt biodiversity loss, and restore ecosystems (CBD,  
5 2020). Although corporate environmental disclosures are a function of global climate change  
6 initiatives and environmental regulatory frameworks of the country in which a firm operates (Luo,  
7 2019; Tauringana and Chithambo, 2015), the efficacy of such initiatives in improving corporate  
8 accountability with regard to biodiversity preservation and associated disclosures has not been  
9 adequately undertaken in previous studies.

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16 Past research (Bui *et al.*, 2020; Ding *et al.*, 2023; Giannarakis *et al.*, 2017; Luo, 2019) examining  
17 the link between climate change performance and disclosures has offered two opposing  
18 viewpoints. From the legitimacy theory perspective, one stream of literature (Cho and Patten,  
19 2007; Luo, 2019) suggests that firms with inferior carbon/environmental performance are exposed  
20 to increased stakeholder demands/pressures, and therefore issue more environmental disclosures  
21 to protect reputation/legitimacy. Consistent with signaling theory, another stream of literature (Bui  
22 *et al.*, 2020; Giannarakis *et al.*, 2017) argues that firms with better carbon/environmental  
23 performance provide extensive environmental information to signal their superior performance.  
24 However, these/past studies have primarily focused on the environmental performance–  
25 disclosures link, and there are no systematic studies within the biodiversity accounting literature  
26 that may explain the impacts of *CCRs* on *BIODSL*. Moreover, few related studies (Gaia and Jones,  
27 2020; Haque and Jones, 2020) assessing biodiversity reporting practices have mainly focused on  
28 particular countries (e.g., the UK) or regions (e.g., Europe).

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39 Therefore, drawing on global firms over the period 2009-2021, our study explores the effects of  
40 *CCRs* in terms of excessive carbon emissions and waste production volumes on the levels of  
41 *BIODSL* in a multi-country setting. Consistent with legitimacy theory, we find that firms with  
42 higher *CCRs* tend to issue more *BIODSL*. We also reveal that the disaggregated components of  
43 carbon (Scope 1 and 2) emissions and waste (hazardous and non-hazardous) levels increase  
44 *BIODSL*. Our analysis further shows that the positive effects of *CCRs* on *BIODSL* are stronger for  
45 firms from environmentally sensitive industries. Finally, our results demonstrate the effectiveness  
46 of global climate and biodiversity protection initiatives, such as the Paris Agreement and the CBD,  
47 in limiting corporate engagement in legitimation practices.  
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3 Our study contributes to the existing literature in a number of ways. First, our study extends the  
4 climate and biodiversity accounting literature (Carvajal *et al.*, 2022; Ding *et al.*, 2023; Haque and  
5 Jones, 2020) by examining whether *CCRs* influence *BIODSL*. Unlike past studies, which have  
6 mainly focused on process-based environmental scores, our study employs actual volumes of both  
7 carbon emissions and waste generation, which are comprehensive and relevant proxies for *CCRs*.  
8 Although corporate carbon is an important dimension of climate change performance (Luo, 2019),  
9 waste generation is another unique dimension and objective reflection of *CCRs* (Gull *et al.*, 2024).  
10 As noted by Gull *et al.* (2024), broad environmental performance indicators used in prior research  
11 have limitations due to subjective measurements and different estimation methods. We also  
12 construct the unique *BIODSL* index based on comprehensive biodiversity protection reporting  
13 indicators. Using objective and relevant proxies, we thus provide novel evidence on the *CCRs*–  
14 *BIODSL* link from a multi-country perspective. Our findings suggest that firms with increased  
15 carbon emissions and waste generation levels disclose extensive biodiversity-related information  
16 to protect their legitimacy/reputation.  
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28 Second, our study contributes to the biodiversity accounting literature (Carvajal *et al.*, 2022;  
29 Haque and Jones, 2020; Treepongkaruna, 2024) by analyzing the moderating effect of  
30 environmentally sensitive industries on the *CCRs*–*BIODSL* nexus. Prior literature (Luo, 2019;  
31 Orazalin *et al.*, 2024) argues that carbon/environmental performance may influence voluntary  
32 disclosures depending on institutional factors and regulatory contexts. Our findings offer new  
33 evidence that firms from environmentally sensitive industries **disclose more** biodiversity-related  
34 disclosures to respond to increased stakeholder pressures/demands for environmental  
35 accountability and protect legitimacy.  
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43 Finally, our study complements past research (Ding *et al.*, 2023; Orazalin *et al.*, 2024) by exploring  
44 how climate and biodiversity protection regulations, such as the Paris Agreement and the CBD,  
45 affect *CCRs* and *BIODSL*. While prior investigations (Freedman and Jaggi, 2011; Orazalin *et al.*,  
46 2024) have mainly focused on specific environmental initiatives (e.g., the Kyoto Protocol),  
47 empirical evidence on the effects of the Paris Agreement and the *CBD* on corporate accountability  
48 and practices associated with biodiversity conservation is almost non-existent. Our findings  
49 suggest that both the Paris Agreement and the *CBD* are effective in limiting legitimization efforts,  
50 thus supporting the view that global environmental regulations may enhance corporate  
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3 accountability towards climate mitigation (Haque and Ntim, 2020; Tauringana and Chithambo,  
4 2015).  
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## 8 9 10 **2. Literature review and hypothesis development**

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12 Prior empirical studies (Bui *et al.*, 2020; Cho and Patten, 2007; Clarkson *et al.*, 2008; Ding *et al.*,  
13 2023; Giannarakis *et al.*, 2017; Luo, 2019; Siddique *et al.*, 2021) on the environmental  
14 performance–disclosure link have predominantly applied signaling and legitimacy theories.  
15 Signaling theory posits that businesses are likely to differentiate themselves by disclosing more  
16 information about their qualities and strengths in order to reduce information asymmetry and  
17 achieve a competitive advantage (Bui *et al.*, 2020; Del Gesso and Lodhi, 2024). Under this  
18 perspective, firms with superior climate change performance may use extensive environmental  
19 disclosures to signal to the market about their climate mitigation activities/practices and  
20 performance (Giannarakis *et al.*, 2017; Siddique *et al.*, 2021). Therefore, signaling theory predicts  
21 a negative association between *CCRs* and *BIODSL*, suggesting that firms with lower levels of  
22 carbon and waste are likely to disclose more information on their efforts to protect biodiversity.  
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32 By contrast, legitimacy theory (Deegan, 2002; Suchman, 1995) suggests that firms exposed to  
33 heightened stakeholder pressures/demands tend to issue extensive environmental information to  
34 protect reputation/legitimacy. High-polluting firms without effective climate mitigation  
35 activities/initiatives are generally perceived as socially irresponsible, and therefore face more  
36 rigorous environmental regulations, greater stakeholder pressures, and higher legitimacy risks  
37 (Bilal *et al.*, 2024; Luo, 2019). To protect legitimacy, such firms may engage in responsive  
38 strategies/reforms to show that they operate within the bounds of appropriate social behavior  
39 (Deegan and Rankin, 1997) and their activities are proper, desirable, and useful (Suchman, 1995).  
40 In this regard, extensive environmental disclosures may help high-polluting firms maintain  
41 legitimacy, protect reputation, and mitigate business risks (Cho and Patten, 2007; Ding *et al.*,  
42 2023). Nevertheless, such increased disclosures are viewed as a means of protecting image and  
43 legitimacy, and therefore, do not reflect actual sustainability performance (Soobaroyen and Ntim,  
44 2013). Thus, in light of biodiversity crisis, legitimacy theory predicts a positive association  
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3 between *CCRs* and *BIODSL*, indicating that firms with higher levels of carbon and waste disclose  
4 more biodiversity-related information to protect/maintain legitimacy.  
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7 Empirical studies exploring the association between environmental/carbon performance and  
8 disclosures have yielded mixed results (Bui *et al.*, 2020; Clarkson *et al.*, 2008; Dawkins and Fraas,  
9 2011; Giannarakis *et al.*, 2017; Luo and Tang, 2014; Qian and Schaltegger, 2017; Siddique *et al.*,  
10 2021). For example, among others, Bui *et al.* (2020) report a negative nexus between carbon  
11 emissions and disclosures, thus supporting the notion that firms disclose more carbon-related  
12 information to highlight their superior performance and differentiate themselves in the market,  
13 consistent with signaling theory. In contrast, Ding *et al.* (2023) document that high-polluting firms  
14 facing stakeholder pressures increase carbon disclosures to manage stakeholders' expectations and  
15 ultimately mitigate legitimacy risks. Other studies (Cho and Patten, 2007; Luo, 2019) have also  
16 revealed that poor climate change performance drives carbon/environmental disclosures, in line  
17 with legitimacy theory. However, and indicated before, none of the previous studies has examined  
18 the effects of *CCRs* on *BIODSL*. Thus, based on legitimacy and signaling theories, as well as the  
19 conflicting findings of past studies, we propose the following two contradictory hypotheses:  
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30 ***H1a:*** Carbon emissions and waste generation levels are negatively associated with the level of  
31 biodiversity disclosures, consistent with signaling theory.  
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34 ***H1b:*** Carbon emissions and waste generation levels are positively associated with the level of  
35 biodiversity disclosures, consistent with legitimacy theory.  
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### 41 **3. Data and methodology**

#### 42 *3.1. Sample and data*

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45 Our study focuses on all global firms with necessary data to construct the *BIODSL* index. Unlike  
46 previous biodiversity-related studies (Adler *et al.*, 2018; Carvajal *et al.*, 2022; Haque and Jones,  
47 2020), our work seeks to analyze a relatively large sample of firms originating from multiple  
48 industries and countries. We collected data from several sources. First, we obtained data on  
49 biodiversity indicators, carbon emissions, waste generation levels, and internal governance from  
50 ASSET4 ESG. Second, we collected data on accounting and financial controls from the  
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Worldscope database. Finally, we downloaded macroeconomic statistics, including GDP and inflation rates from the World Bank's database (World Bank, 2022).

The initial sample consisted of 13,933 firm-years from 45 countries from 2009 to 2021.<sup>1</sup> The sample was reduced by 5,809 firm-years without carbon, waste, and biodiversity data. Then, 498 firm-years were removed due to missing data on the control variables. Hence, our final sample consists of 7,626 firm-years from 892 firms operating in 10 industries and 44 countries from 2009 to 2021.<sup>2</sup> Appendix A reports the sample distributions by country (Panel A) and industry (Panel B). As shown in Panel A, the United States and the United Kingdom are the most represented countries, accounting for 17.22% (1,313 firm-years) and 13.59% (1,036 firm-years) of the sample, respectively. The remaining 5,277 firm-years (69.19%) are represented by other countries. Further, Panel B shows that the sample is based on 10 different industries, with the industrials representing the largest proportion of the sample (24.10%), followed by consumer discretionary (15.47%) and consumer staples (10.65%).

### 3.2. Measurement of biodiversity disclosures

We developed the *BIODSL* index based on **environmental data** from Asset4 ESG, following three main steps. First, we reviewed related studies (Adler *et al.*, 2018; Haque and Jones, 2020; Maroun and Atkins, 2018) to source relevant biodiversity disclosures. Second, **we referred to the GRI reporting guidelines to determine pertinent biodiversity protection indicators**. Third, we performed an in-depth analysis of all **environmental categories** provided by ASSET4 ESG to identify biodiversity-related reporting indicators. Consequently, we **detected nine specific** biodiversity-related indicators. **They are biodiversity loss reduction, sulfur and nitrogen oxides emissions reduction, waste reduction, e-waste reduction, toxic chemicals and substances reduction, environmental restoration, eco-designed products/processes, environmental innovation/product impacts minimization, and land environmental impact reduction**. Related research (Atkins and Maroun, 2018; Haque and Jones, 2020) suggests that firms should disclose the impacts of their

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<sup>1</sup>The dataset starts from 2009 because the majority of firms did not disclose biodiversity-related information prior to this year.

<sup>2</sup>To mitigate the effects of potential outliers, we winsorized all the continuous variables at the 1<sup>st</sup> and 99<sup>th</sup> percentiles.

business activities at the ecosystem and planet levels by performing a broader assessment of local and global ecosystems. Therefore, we used these reporting indicators to develop the *BIODSL* index because they better reflect firms' impacts on biodiversity and threatened species, along with other corporate initiatives/actions aimed at enhancing the positive impacts and mitigating the adverse effects of their business operations on the environment. A value of one was assigned if an indicator is disclosed and zero was assigned if that indicator is not disclosed. We then added all the indicators to obtain the overall disclosure score. This measurement approach is similar to previous studies assessing voluntary disclosures (Orazalin *et al.*, 2024).<sup>3</sup> Finally, we calculated the *BIODSL* index for each firm by comparing its reporting items to those of other firms from the same industry. Therefore, the index adjusted within each industry ranges between 0% and 100%, where larger values indicate higher levels of *BIODSL*.

### 3.3. Measurement of climate change risks

Corporate carbon emissions and waste generation levels are the major and most significant causes of climate change, and therefore their actual levels are recognized as objective and relevant proxies for assessing corporate climate-related risks (Gull *et al.*, 2024; Luo, 2019; Luo and Tang, 2021). Thus, we measure *CCRs* based on corporate carbon and waste generation data from Asset4 ESG. Consistent with related research (Orazalin *et al.*, 2024; Uyar *et al.*, 2023), we evaluate firm-level carbon emissions (*CARBON*) as the natural logarithm of total carbon (both Scope 1 and 2) emissions<sup>4</sup> and assess waste generation (*WASTE*) as the natural logarithm of total waste (both hazardous and non-hazardous) production. These proxies reflect a firm's climate risks and its actual environmental impacts in the form of carbon emissions and waste levels and enable to assess whether a firm has managed to reduce its carbon and waste levels compared to its past outcomes and industry peers (Luo and Tang, 2021). As such, lower values of *CARBON* and *WASTE* indicate

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<sup>3</sup>To ensure the internal consistency of the score, we estimated the Cronbach's alpha. The estimated value is higher than the minimum threshold value of 0.700, thus confirming the reliability and validity of the score.

<sup>4</sup>Scope 1 emissions are direct emissions from sources controlled/owned by a firm, whereas Scope 2 emissions are indirect ones from sources not controlled/owned by a firm. Since Scope 3 emissions (another form of indirect emissions) are missing for the majority of firms, we did not include them in our study.

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3 lower climate risks and better climate change performance relative to the firm's prior years and its  
4 industry benchmarks in a current year.  
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### 10 3.4. Control variables

11 Prior empirical studies (Bui *et al.*, 2020; Carvajal *et al.*, 2022; Clarkson *et al.*, 2008; Ding *et al.*,  
12 2023; Giannarakis *et al.*, 2017; Haque and Jones, 2020; Hassan *et al.*, 2020; Luo, 2019; Orazalin  
13 and Mahmood, 2018; Tauringana and Chithambo, 2015; Treepongkaruna, 2024) have shown that  
14 there are certain firm-specific and country-level characteristics that may affect environmental  
15 disclosures. In particular, we control firm-specific characteristics, including corporate governance  
16 (*CGOVR*), firm size (*FSIZE*), profitability (*FPROF*), financial risk (*FLVRG*), slack (*FSLACK*),  
17 and capital intensity (*CAPEX*) as potential drivers of biodiversity information. In addition, we  
18 control the country-level variables, such as inflation rates (*INFLN*) and *GDP* growth (*GDPGR*).  
19 The descriptions/measurements of the variables are summarized in Appendix B.  
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### 30 3.5. Empirical model

31 We employ the following models to test the impacts of *CCRs* on *BIODSL*:  
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$$33 \quad BIODSL = \beta_0 + \beta_1 CARBON + \beta_2 Controls + country, industry, and year fixed effects + \varepsilon$$

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37 (1)

$$38 \quad BIODSL = \beta_0 + \beta_1 WASTE + \beta_2 Controls + country, industry, and year fixed effects + \varepsilon$$

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42 (2)

43 where, *BIODSL* represents biodiversity disclosures, *CARBON* represents carbon emissions,  
44 *WASTE* represents waste generation, *Controls* represent the firm and country-level variables  
45 discussed above, and  $\varepsilon$  represents the error term. To account for systematic variations in the main  
46 variables across periods, industries, and countries, we also included year, industry, and country  
47 fixed effects.  
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## 4. Results and discussion

### 4.1. Descriptive analysis

Table 1 reports the descriptive statistics of the variables for the whole sample period spanning from 2009 to 2021. The *BIODSL* index has a mean of 50.00% and varies between 0.11% and 99.92%. The mean *CARBON* and *WASTE* are 13.03 and 11.17, respectively and these values are generally consistent with those of past research (Gull *et al.*, 2024; Orazalin *et al.*, 2024). Regarding the corporate governance variable, *CGOVQ* has a mean of 62.89. Finally, the mean values of *FSIZE*, *FPROF*, *FLVRG*, *FSLCK*, *CAPEX*, *INFLN*, and *GDPRG* are 16.18, 7.75%, 29.07%, 0.03, 0.05, 1.94%, and 1.63%, respectively.

Table 2 presents correlation measures among the variables. The coefficients show that *CARBON* and *WASTE* are positively correlated with *BIODSL*, thus providing preliminary evidence that firms with higher levels of carbon emissions and waste are more likely to issue extensive biodiversity information. The matrix also shows that the coefficients among the independent variables do not exceed a maximum threshold value of 0.80, indicating that serious multicollinearity is not expected.<sup>5</sup>

### 4.2. Regression results

Table 3 reports the regression results of the effects of *CARBON* and *WASTE* on *BIODSL*. As shown in Column (1), *CARBON* is significantly and positively related to *BIODSL*, supporting *H1b*. This evidence indicates that high-polluting firms issue extensive biodiversity-related information, consistent with the legitimacy viewpoint. Similarly, Column (2) specifies that the coefficient of *WASTE* is significant and positive, implying that higher waste levels may lead to more biodiversity-related information. This evidence validates *H1b* and suggests that firms with increased levels of waste tend to issue extensive biodiversity-related information. **It is evident from the coefficients that a one-unit increase in *CARBON* leads to a 1.523 percentage point increase in**

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<sup>5</sup>The last column (*VIF*) of the matrix also shows that none of the variance inflation factors exceeds 3.00, thus verifying the absence of multicollinearity.

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3 *BIODSL*, whereas a one-unit increase in *WASTE* leads to a 1.714 percentage point increase. More  
4 specifically, an increase in one standard deviation of carbon emissions (in Table 1) increases the  
5 level of biodiversity disclosures by about 12.58% [ $(CARBON)2.34*1.523/(BIODSL)28.31$ ], while  
6 an increase in one standard deviation of waste generation increases disclosures by 16.59%  
7 [ $(WASTE)2.74*1.714/(BIODSL)28.31$ ], thus indicating that the results are also economically  
8 significant. Overall, the results in Table 3 support the notion that poor environmental performers  
9 have incentives to increase climate-related disclosures in order to protect their legitimacy and  
10 maintain support from stakeholders (Cho and Patten, 2007; Luo, 2019). These findings are also  
11 consistent with recent research (Ding *et al.*, 2023; Orazalin *et al.*, 2024) that poor climate change  
12 performance drives environmental initiatives and disclosures. Theoretically, these results are in  
13 line with legitimacy theory (Deegan, 2002; Deegan and Rankin, 1997; Suchman, 1995) and  
14 suggest that firms exposed to greater CCRs are likely to utilize increased *BIODSL* as a tool of  
15 legitimacy to conform to societal expectations, protect reputation, and enhance/maintain  
16 legitimacy.

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18 To test the effects of specific components of CCRs, we further utilize the disaggregated  
19 components of carbon emissions in terms of Scope 1 (*CARB\_SCI*) and Scope 2 (*CARB\_SC2*) and  
20 the disaggregated components of waste generation in terms of hazardous (*WAST\_HAZ*) and non-  
21 hazardous (*WAST\_NHAZ*) waste. To measure these variables, we take the natural logarithm of the  
22 disaggregated components, consistent with our *CARBON* and *WASTE* measures. Table 4 shows  
23 that *CARB\_SCI* and *CARB\_SC2* are positively connected to *BIOSCL*, implying that both Scope 1  
24 and 2 emissions drive biodiversity-related information. Similarly, the coefficients of *WAST\_HAZ*  
25 and *WAST\_NHAZ* are positive and significant, indicating that firms generating higher levels of  
26 both hazardous and non-hazardous waste issue more biodiversity-related information. Overall, the  
27 results in Table 4 support legitimacy theory and verify that firms with higher CCRs, i.e., those with  
28 higher levels of emissions and waste disclose extensive biodiversity-related information.

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30 For the control variables, *CGOVR* is positively associated with *BIODSL*, thus indicating that  
31 improved corporate governance leads to extensive biodiversity disclosures. Further, *FSIZE* is  
32 statistically positive with *BIODSL*, supporting the argument that large firms facing greater  
33 stakeholder pressures/demands for environmental transparency disclose more environmental  
34 information (Haque and Jones, 2020). Similarly, the coefficients of *FSLCK* and *CAPEX* are  
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3 positive, verifying that increased financial slack and capital expenditures may lead to extensive  
4 environmental information. Finally, *FLVRG* is negatively associated with *BIODSL*, supporting the  
5 argument that highly leveraged firms are less likely to demonstrate greater commitment to climate  
6 change initiatives and environmental reporting practices due to their focus on operating activities  
7 and short-term debts (Haque and Jones, 2020). The coefficients for other control variables are  
8 insignificant.  
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#### 10 11 12 13 14 15 16 17 4.3. Robustness tests

18 We conduct a range of sensitivity analyses to examine the robustness of our inferences. First, we  
19 employ alternative proxies for *BIODSL*, *CARBON*, and *WASTE*. In particular, following related  
20 research (Issa and Zaid, 2023; Treepongkaruna, 2024), we use the ASSET4 ESG's biodiversity  
21 protection disclosure score (*BIOSCR*), which reflects a firm's biodiversity reporting practices to  
22 protect biodiversity and species and reduce its impacts on the ecosystems. Further, we utilize two  
23 carbon intensity variables, measured as the ratio of total carbon emissions to total assets (*CARAST*)  
24 and the ratio of total carbon emissions to revenues (*CARREV*). Using the same approach, we  
25 estimate two measures of waste generation, calculated as the ratio of total waste production  
26 volumes to total assets (*WASAST*) and the ratio of total waste to revenues (*WASREV*).<sup>6</sup> As reported  
27 in Table 5, *CARBON* and *WASTE* have a positive association with *BIOSCR*, while *CARAST*,  
28 *CARREV*, *WASAST*, and *WASREV* are positively related to *BIODSL*, thus verifying that corporate  
29 climate risks increase biodiversity-related information.  
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40 Second, to address potential endogeneity that may arise from omitted variables, we use a two-stage  
41 least squares (2SLS) approach. Following past research (Gull *et al.*, 2024; Orazalin *et al.*, 2024),  
42 we utilize two instrumental variables: (1) the median values of *CCRs* (*CARBON\_HEAD* and  
43 *WASTE\_HEAD*) at the firm's headquarters and (2) their mean values (*CARBON\_IND* and  
44 *WASTE\_IND*) within the industry. While these instruments are likely to influence *CCRs*, they are  
45 less likely to correlate with *BIODSL*.<sup>7</sup> As reported in Table 6, the coefficients of *CARBON* and  
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52 <sup>6</sup>These alternative proxies have been widely utilized in past investigations (Gull *et al.*, 2024; Uyar *et al.*, 2023).

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54 <sup>7</sup>The first-stage results, Cragg-Donald Wald F-statistics, Kleibergen-Paap rk LM statistics, and Stock-Yogo critical  
55 values demonstrate the relevance and validity of our instrumental variables.  
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3 *WASTE* are significant and positive, indicating that the original models and main results do not  
4 suffer from omitted variables bias.  
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8 Third, we conduct a two-stage Heckman (Heckman, 1979) procedure to address potential sample  
9 selection issues. Since this analysis requires the identification of a proper exclusion restriction in  
10 the first stage, we use the industry averages of *CARBON* (*CARBON\_IND*) and *WASTE*  
11 (*WASTE\_IND*) as exclusion restrictions. In the first stages, the dependent variables are *CAR\_DUM*  
12 and *WAS\_DUM*, which are coded one if the *CARBON* and *WASTE* variables exceed the median  
13 values, and zero otherwise. Further, the Mills ratios are estimated based on the parameters in the  
14 first stages and then included as additional variables in the second stage regressions. The results  
15 in Table 7 are generally consistent with the main results, confirming that the inferences are robust  
16 to sample selection bias.  
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24 Fourth, we regress *BIODSL* on the first and second lagged explanatory variables. Although carbon  
25 emissions and waste generation are likely to affect biodiversity reporting, it is possible that firms  
26 with increased *BIODSL* may have higher *CCRs*, causing a reverse causality in this nexus. Using  
27 the lagged independent variables mitigates serious reverse causality problems in the baseline  
28 model (Uyar *et al.*, 2023), as *CCRs* in the preceding year are unlikely to be affected by *BIODSL*  
29 of the current year. As shown in Table 8, the first and second lag values of *CARBON* (*CARBON<sub>t-1</sub>*,  
30 and *CARBON<sub>t-2</sub>*) exhibit a positive association with *BIODSL*. Similarly, the first and second lags  
31 of *WASTE* (*WASTE<sub>t-1</sub>*, and *WASTE<sub>t-2</sub>*) are positively related to *BIODSL*. Overall, these estimates  
32 verify that the main findings are less likely to be determined by reverse causality concerns.  
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41 Finally, we re-estimate our analysis by excluding US and UK firm-years from the sample.  
42 Appendix A shows that the US accounts for 17.22% of the firm-years, while the UK accounts for  
43 13.59%. Because these two dominant economies may drive our results, we generate separate sub-  
44 samples and repeat our analysis. As shown in Table 9, we first exclude US firms in Columns (1-  
45 2), then UK firms in Columns (3-4), and then both US and UK firms in Columns (5-6). The results  
46 demonstrate that the effects of *CARBON* and *WASTE* on *BIODSL* are still significant and positive.  
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50 Overall, this analysis suggests that our main findings are not affected by the alternative samples.  
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#### 4.4. Additional analyses

In this section, we conduct additional analyses to explore the effects of industry and regulatory factors on *CCRs* and *BIODSL*. In particular, we assess whether the link between *CCRs* and *BIODSL* differs between environmentally sensitive and non-sensitive firms. Available research (Boiral and Heras-Saizarbitoria, 2017; Orazalin et al., 2024) argues that firms in sensitive industries face more stringent regulations and higher stakeholder pressures due to their negative impacts on the environment, and therefore differ from those in non-sensitive industries in terms of their responses to climate change and reporting behavior. Thus, following Shrestha et al. (2023), we classify the energy, materials, and utilities industries as environmentally sensitive and all other industries as non-sensitive industries. We then test the moderating effect of the sensitive industries (*INDUM*), which equals to one if a firm is from an environmental sensitive industry, and zero otherwise. As reported in Columns (1) and (2) of Table 10, the interaction terms *CARBON\*INDUM* and *WASTE\*INDUM* are significantly positive with *BIODSL*. These results align with the legitimacy view and suggest that firms from sensitive industries **disclose more biodiversity-related information** to respond to increased stakeholder pressures/demands and protect legitimacy.

Further, we examine how the Paris Agreement influences the link between *CCRs* and *BIODSL*. Prior literature (Luo, 2019; Shrestha et al., 2023) suggests that environmental disclosures are a function of global environmental initiatives/reforms and regulations of the country in which a firm operates. Specifically, firms facing rigorous climate regulations reflect their reactions through environmental disclosures (Taurigana and Chithambo, 2015). Thus, we create the *PARIS* variable, which equals to one if a firm-year is from 2016 and onwards, and zero otherwise. Columns (3) and (4) of Table 10 display that the interactions *CARBON\*PARIS* and *WASTE\*PARIS* are significant and negative, indicating that the effects of *CARBON* and *WASTE* are weaker after the introduction of the Agreement. These results support the viewpoint that global climate regulations are effective in restricting legitimation efforts (Luo, 2019).

Finally, we test the moderating effect of the *CBD* on the *CCRs*–*BIODSL* link. As mentioned before, the USA is the only country that declined to ratify the *CBD*. Arguably, this is an interesting context to test whether the effects of *CCRs* on *BIODSL* differ between ratifying and non-ratifying nations. Thus, we introduce the *CBD* variable, which equals to one if firm-years belong to the UN members



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3 that ratified the *CBD*, and zero otherwise. Columns (5) and (6) of Table 10 show that although the  
4 interaction *WASTE*\**CBD* is negative and insignificant, the coefficient of *CARBON*\**CBD* is  
5 negatively significant, indicating that the effects of *CCRs* on *BIODSL* are weaker for firms from  
6 the ratifying countries. These results indicate that firms with higher *CCRs* from the ratifying  
7 nations are less inclined to utilize increased *BIODSL* as a legitimizing tool, thus demonstrating the  
8 effectiveness of global environmental regulations, such as the *CBD*, in limiting corporate  
9 engagement in symbolic efforts/practices.  
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## 20 5. Conclusion

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22 This study examines whether climate change risks in terms of increased carbon emissions and  
23 waste generation levels influence biodiversity disclosures. Focusing on global firms over the  
24 period from 2009 to 2021, our results support legitimacy theory (Deegan, 2002; Suchman, 1995)  
25 and suggest that firms with higher levels of carbon and waste disclose extensive biodiversity  
26 information. Furthermore, the disaggregated components of carbon (Scope 1 and 2 emissions) and  
27 waste (hazardous and non-hazardous waste) also drive biodiversity disclosures. The results also  
28 demonstrate that the effects of climate risks on biodiversity disclosure are stronger for firms from  
29 environmentally sensitive industries, suggesting that industry-related factors/regulations have an  
30 impact on corporate incentives to utilize extensive biodiversity reporting as a legitimizing tool to  
31 conform to societal expectations and protect reputation/legitimacy (Ding et al., 2023; Luo, 2019;  
32 Orazalin et al., 2024). Finally, the results reveal that global biodiversity protection initiatives,  
33 namely the Paris agreement and the *CBD*, appear to be more effective in limiting corporate  
34 engagement in utilizing increased biodiversity disclosures as a legitimizing tool. Overall, our  
35 findings offer economically significant evidence that firms' climate change risks in terms of  
36 excessive carbon emissions and waste production lead to higher biodiversity disclosure levels.  
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49 This work offers a number of important theoretical, policy, and practical implications. In  
50 particular, our results verify the legitimacy theory's (Deegan, 2002; Suchman, 1995) proposition  
51 that firms facing increased stakeholder pressure/demands disclose more environmental  
52 information in order to achieve social recognition and protect reputation/legitimacy. Although past  
53 studies (Cho and Patten, 2007; Ding et al., 2023; Luo, 2019) have applied this theoretical approach  
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3 to explore the environmental performance–disclosure link, our study provides novel evidence that  
4 high carbon and waste emitting firms may utilize extensive biodiversity disclosures as a  
5 legitimizing tool to maintain legitimacy, protect reputation, and mitigate business risks. With  
6 regard to policy implications, our results further suggest that regulators should design explicit  
7 initiatives and guidelines to monitor, regulate, and rate firm-level climate change risks and  
8 disclosure practices and report to all stakeholders accordingly. Furthermore, policymakers and  
9 regulators can promote biodiversity reporting by enhancing alignment among climate-related  
10 reforms, reporting guidelines, and environmental policies for sustainable development, and  
11 biodiversity conservation initiatives at the national and global levels. Policymakers also need to  
12 enact enforceable regulations with verifiable targets for carbon emissions, waste generation, and  
13 biodiversity conservation, especially in weakly regulated environmental settings. Our findings also  
14 suggest that financial analysts and investors should assess not only environmental transparency,  
15 but also corporate climate risks to make informed investment decisions. Finally, board members  
16 and managers should refrain from issuing biodiversity disclosures that do not reflect actual and  
17 verifiable corporate efforts/practices and performance.  
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30 Our study is subject to some limitations. For example, we focus on large firms, and as such, the  
31 observed relationships cannot be generalized to small and medium-sized firms. Hence, future  
32 research could extend our study by analyzing smaller firms. Furthermore, we explore the effects  
33 of carbon emissions and waste production on biodiversity disclosures. Future research, therefore,  
34 could consider other firm-specific and external factors that may influence disclosures. In addition,  
35 the indicators for measuring biodiversity disclosures, although used in prior research (Haque and  
36 Jones, 2020), assess the quantity rather than the quality of disclosures. Therefore, future research  
37 could shed further light on biodiversity transparency by employing alternative proxies.  
38 Nevertheless, despite these limitations, our work offers new insights into the relation between  
39 climate change risks and biodiversity disclosures.  
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#### 47 **Declaration of interest:**

48 The authors declare that they have no conflict of interest.  
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**Table 1. Descriptive Statistics**

<b>Variable</b>	<b>Obs.</b>	<b>Mean</b>	<b>Std. Dev.</b>	<b>Min.</b>	<b>Max.</b>
BIODSL (%)	7626	50.00	28.31	0.11	99.92
CARBON (ln)	7626	13.03	2.34	7.25	18.18
WASTE (ln)	6018	11.17	2.74	4.94	20.36
CGOVR (%)	7626	62.89	20.00	0.41	99.41
FSIZE (ln)	7626	16.18	1.38	12.98	19.24
FPROF (%)	7626	7.75	7.43	-16.95	32.79
FLVRG (%)	7626	29.07	15.91	0.00	73.80
FSLCK (ratio)	7626	0.03	0.03	0.00	0.17
CAPEX (ratio)	7626	0.05	0.04	0.00	0.20
INFLN (%)	7626	1.94	1.95	-4.48	19.60
GDPGR (%)	7626	1.63	3.21	-11.17	24.48

Notes: The descriptions/measurements of the variables are summarized in Appendix B.

**Source(s):** Authors' own work

Table 2. Pairwise correlations

Variables	BIODSL	CARBON	WASTE	CGOVR	FSIZE	FPROF	FLVRG	FSLCK	CAPEX	INFLN	GDPGR	VIF
BIODSL	1.00											
CARBON	0.31***	1.00										2.32
WASTE	0.25***	0.60***	1.00									1.65
CGOVR	0.16***	0.20***	0.17***	1.00								1.07
FSIZE	0.41***	0.63***	0.43***	0.23***	1.00							1.78
FPROF	-0.05***	-0.11***	-0.10***	0.00	-0.13***	1.00						1.59
FLVRG	-0.02*	0.10***	-0.06***	0.00	0.14***	-0.15***	1.00					1.11
FSLCK	-0.08***	-0.13***	-0.12***	0.02*	-0.16***	0.59***	-0.06***	1.00				1.53
CAPEX	0.03***	0.27***	0.09***	0.05***	0.03**	0.12***	0.03***	0.10***	1.00			1.13
INFLN	-0.11***	-0.05***	-0.01	-0.03**	-0.18***	0.11***	0.07***	0.05***	0.05***	1.00		1.11
GDPGR	-0.03**	0.00	-0.02*	-0.01	-0.03***	0.15***	-0.05***	0.06***	0.04***	0.20***	1.00	1.07

Notes: \*\*\*, \*\*, and \* correlation is significant at the 0.01, 0.05, and 0.10 levels (2-tailed), respectively.

Source(s): Authors' own work

**Table 3. The association between climate change risks and biodiversity disclosures**

	(1) BIODSL	(2) BIODSL
CARBON	1.523*** (0.212)	
WASTE		1.714*** (0.163)
CGOVR	0.098*** (0.015)	0.086*** (0.016)
FSIZE	6.383*** (0.328)	5.437*** (0.321)
FPROF	0.017 (0.047)	0.044 (0.053)
FLVRG	-0.098*** (0.019)	-0.080*** (0.022)
FSLCK	35.259*** (11.481)	25.300* (13.182)
CAPEX	21.337** (8.609)	31.481*** (9.429)
INFLN	-0.270 (0.294)	-0.139 (0.318)
GDPGR	-0.232 (0.160)	-0.020 (0.167)
Constant	-86.19*** (4.753)	-73.07*** (5.348)
Country fixed effects	Yes	Yes
Industry fixed effects	Yes	Yes
Year fixed effects	Yes	Yes
Observations	7626	6018
Adj. R <sup>2</sup>	0.308	0.285

Notes: This table presents the results from regressing biodiversity disclosures (BIODSL) on corporate carbon emissions (CARBON), waste levels (WASTE), and other covariates. Robust standard errors below estimated coefficients are reported in parentheses. \*, \*\*, and \*\*\* denote significance of the estimated coefficients at the 10%, 5%, and 1% levels, respectively.

Source(s): Authors' own work



**Table 4. Disaggregated components of climate change risks and biodiversity disclosures**

	(1) BIODSL	(2) BIODSL	(3) BIODSL	(4) BIODSL
CARB_SC1	1.132*** (0.165)			
CARB_SC2		1.713*** (0.204)		
WAST_HAZ			1.439*** (0.167)	
WAST_NHAZ				1.870*** (0.176)
CGOVR	0.102*** (0.015)	0.092*** (0.015)	0.051*** (0.020)	0.056*** (0.019)
FSIZE	6.632*** (0.311)	6.353*** (0.307)	5.557*** (0.360)	5.025*** (0.359)
FPROF	0.016 (0.047)	0.005 (0.047)	-0.079 (0.062)	0.010 (0.061)
FLVRG	-0.095*** (0.019)	-0.082*** (0.019)	-0.079*** (0.026)	-0.085*** (0.027)
FSLCK	36.070*** (11.478)	26.478** (11.538)	50.094*** (14.991)	33.113** (14.915)
CAPEX	24.103*** (8.601)	31.432*** (8.294)	27.797** (11.552)	34.811*** (10.953)
INFLN	-0.286 (0.294)	-0.276 (0.294)	-0.683* (0.355)	-0.555 (0.351)
GDPGR	-0.242 (0.160)	-0.223 (0.160)	-0.258 (0.195)	-0.273 (0.190)
Constant	-83.85*** (4.826)	-86.83*** (4.716)	-69.06*** (6.372)	-73.17*** (6.229)
Country fixed effects	Yes	Yes	Yes	Yes
Industry fixed effects	Yes	Yes	Yes	Yes
Year fixed effects	Yes	Yes	Yes	Yes
Observations	7626	7626	3975	4105
Adj. R <sup>2</sup>	0.308	0.310	0.309	0.305

Notes: This table presents the results from regressing biodiversity disclosures (BIODSL) on disaggregated corporate carbon emissions in terms of Scope 1 (CARB\_SC1) and Scope 2 (CARB\_SC2) components, disaggregated waste levels in terms of hazardous (WAST\_HAZ) and non-hazardous (WAST\_NHAZ) components, and other covariates. Robust standard errors below estimated coefficients are reported in parentheses. \*, \*\*, and \*\*\* denote significance of the estimated coefficients at the 10%, 5%, and 1% levels, respectively.

**Source(s):** Authors' own work

**Table 5. Robustness test: alternative measures**

	(1)	(2)	(3)	(4)	(5)	(6)
	BIOSCR	BIOSCR	BIODSL	BIODSL	BIODSL	BIODSL
CARBON	1.282*** (0.335)					
WASTE		2.494*** (0.241)				
CARAST			4.291*** (0.934)			
WASAST				1.009*** (0.119)		
CARREV					3.177*** (0.444)	
WASREV						0.334*** (0.045)
CGOVR	0.060** (0.023)	0.043 (0.026)	0.101*** (0.015)	0.092*** (0.016)	0.098*** (0.015)	0.090*** (0.016)
FSIZE	6.789*** (0.516)	4.569*** (0.514)	7.953*** (0.237)	7.042*** (0.267)	7.952*** (0.236)	7.093*** (0.267)
FPROF	-0.187*** (0.071)	-0.127 (0.079)	0.009 (0.047)	0.024 (0.053)	0.023 (0.047)	0.029 (0.053)
FLVRG	-0.007 (0.030)	0.012 (0.035)	-0.091*** (0.019)	-0.088*** (0.022)	-0.095*** (0.019)	-0.089*** (0.022)
FSLCK	21.782 (17.399)	17.620 (20.319)	35.714*** (11.547)	26.000** (13.217)	37.293*** (11.514)	24.820* (13.221)
CAPEX	73.801*** (14.314)	86.790*** (15.869)	33.872*** (8.452)	38.812*** (9.486)	31.442*** (8.409)	40.283*** (9.499)
INFLN	0.054 (0.459)	0.251 (0.501)	-0.265 (0.295)	-0.110 (0.322)	-0.249 (0.295)	-0.106 (0.322)
GDPGR	-0.482** (0.245)	-0.384 (0.268)	-0.265 (0.162)	-0.046 (0.168)	-0.268* (0.161)	-0.043 (0.168)
Constant	-69.98*** (7.625)	-54.53*** (8.682)	-91.77*** (4.720)	-78.47*** (5.349)	-91.92*** (4.717)	-78.52*** (5.364)
Country fixed effects	Yes	Yes	Yes	Yes	Yes	Yes
Industry fixed effects	Yes	Yes	Yes	Yes	Yes	Yes
Year fixed effects	Yes	Yes	Yes	Yes	Yes	Yes
Observations	7626	6018	7626	6018	7626	6018
Adj. R <sup>2</sup>	0.232	0.220	0.305	0.279	0.308	0.277

Notes: This table presents the results from regressing biodiversity disclosures (BIODSL) on climate change risks using the alternative proxies for carbon emissions (CARAST and CARREV) and waste levels (WASAST and WASREV) and the alternative measure of biodiversity disclosure (BIOSCR). Robust standard errors below estimated coefficients are reported in parentheses. \*, \*\*, and \*\*\* denote significance of the estimated coefficients at the 10%, 5%, and 1% levels, respectively.

Source(s): Authors' own work

**Table 6. Robustness test: two-stage least squares (2SLS)**

	First stage	Second stage	First stage	Second stage
	(1)	(2)	(3)	(4)
	CARBON	BIODSL	WASTE	BIODSL
CARBON_HEAD	0.545*** (0.044)			
CARBON_IND	0.234*** (0.083)			
CARBON		3.563** (1.387)		
WASTE_HEAD			0.576*** (0.056)	
WASTE_IND			0.816*** (0.137)	
WASTE				3.767*** (1.042)
CGOVR	0.005*** (0.001)	0.089*** (0.016)	0.004*** (0.001)	0.078*** (0.017)
FSIZE	1.011*** (0.013)	4.282*** (1.449)	1.019*** (0.023)	3.297*** (1.128)
FPROF	-0.013*** (0.003)	0.038 (0.050)	-0.024*** (0.005)	0.093 (0.059)
FLVRG	0.008*** (0.001)	-0.113*** (0.022)	-0.015*** (0.002)	-0.049* (0.027)
FSLCK	-2.383*** (0.681)	40.371*** (11.976)	0.526 (1.113)	24.021* (13.386)
CAPEX	12.983*** (0.580)	-5.246 (19.996)	8.030*** (0.886)	15.049 (12.593)
INFLN	0.008 (0.018)	-0.248 (0.295)	0.022 (0.027)	-0.195 (0.321)
GDPGR	-0.004 (0.010)	-0.185 (0.160)	0.007 (0.014)	0.000 (0.168)
Constant	-13.44*** (1.390)	-80.80*** (6.004)	-21.31*** (2.196)	-67.91*** (5.987)
Country fixed effects	Yes	Yes	Yes	Yes
Industry fixed effects	Yes	Yes	Yes	Yes
Year fixed effects	Yes	Yes	Yes	Yes
Observations	7626	7626	6018	6018
Cragg-Donald Wald F-stat.	67.692		87.525	
Kleibergen-Paap rk LM statistic	101.401		137.836	
Stock-Yogo critical value	19.93		19.93	
Adj. R <sup>2</sup>	0.705	0.299	0.525	0.264

This table reports the 2SLS results from regressing biodiversity disclosures (BIODSL) on climate change climate change risks and other covariates. Robust standard errors below estimated coefficients are reported in parentheses. \*, \*\*, and \*\*\* denote significance of the estimated coefficients at the 10%, 5%, and 1% levels, respectively.

**Source(s):** Authors' own work

**Table 7. Robustness test: a two-stage Heckman model**

	First stage	Second stage	First stage	Second stage
	(1)	(2)	(3)	(4)
	CAR_DUM	BIODSL	WAS_DUM	BIODSL
CARBON_IND	0.410*** (0.129)			
CARBON		0.836* (0.430)		
WASTE_IND			0.394*** (0.099)	
WASTE				1.444*** (0.279)
CGOVR	0.003*** (0.001)	0.085*** (0.021)	0.005*** (0.001)	0.138*** (0.029)
FSIZE	-0.043*** (0.016)	6.897*** (0.568)	0.000 (0.017)	6.184*** (0.478)
FPROF	-0.010*** (0.003)	-0.027 (0.062)	-0.006* (0.003)	0.013 (0.079)
FLVRG	0.004*** (0.001)	-0.158*** (0.028)	-0.006*** (0.001)	-0.119*** (0.038)
FSLCK	-2.004** (0.800)	-7.466 (17.203)	-1.433* (0.855)	-14.003 (20.913)
CAPEX	16.942*** (0.652)	23.796 (15.049)	4.450*** (0.594)	67.401*** (18.865)
INFLN	-0.003 (0.019)	-0.908** (0.379)	0.013 (0.020)	-0.543 (0.433)
GDPGR	-0.022** (0.011)	-0.338 (0.229)	-0.016 (0.011)	-0.153 (0.255)
Constant	-4.87** (1.983)	-88.62*** (6.886)	-5.22*** (1.489)	-89.50*** (8.550)
Mills ratio		3.535* (2.079)		7.610 (5.835)
Country fixed effects	Yes	Yes	Yes	Yes
Industry fixed effects	Yes	Yes	Yes	Yes
Year fixed effects	Yes	Yes	Yes	Yes
Observations	7626	7626	6018	6018
Wald chi2	0.000***	0.000***	0.000***	0.000***

This table reports the two-stage Heckman selection results from regressing biodiversity disclosures (BIODSL) on climate change risks and other covariates. Industry average values of carbon emissions (CARBON\_IND) and waste levels (WASTE\_IND) are used as exclusion restrictions in the first stage estimations. The Mills ratios are estimated based on the parameters in the first stage and are included in the second stage regressions. Robust standard errors below estimated coefficients are reported in parentheses. \*, \*\*, and \*\*\* denote significance of the estimated coefficients at the 10%, 5%, and 1% levels, respectively.

Source(s): Authors' own work

**Table 8. Robustness test: lag values of independent variables**

	(1) BIODSL	(2) BIODSL	(3) BIODSL	(4) BIODSL
CARBON <sub><i>t-1</i></sub>	1.280*** (0.226)			
CARBON <sub><i>t-2</i></sub>		1.017*** (0.241)		
WASTE <sub><i>t-1</i></sub>			1.689*** (0.173)	
WASTE <sub><i>t-2</i></sub>				1.681*** (0.187)
Controls	Included	Included	Included	Included
Constant	-83.37*** (5.011)	-81.49*** (5.299)	-69.29*** (5.645)	-64.96*** (5.986)
Country fixed effects	Yes	Yes	Yes	Yes
Industry fixed effects	Yes	Yes	Yes	Yes
Year fixed effects	Yes	Yes	Yes	Yes
Observations	6734	5842	5278	4545
Adj. R <sup>2</sup>	0.311	0.317	0.287	0.294

Notes: This table presents the results from regressing biodiversity disclosures (BIODSL) on the first and second lag values of corporate carbon emissions (CARBON<sub>*t-1*</sub> and CARBON<sub>*t-2*</sub>) and waste levels (WASTE<sub>*t-1*</sub> and WASTE<sub>*t-2*</sub>). Robust standard errors below estimated coefficients are reported in parentheses. \*, \*\*, and \*\*\* denote significance of the estimated coefficients at the 10%, 5%, and 1% levels, respectively.

Source(s): Authors' own work

**Table 9. Robustness test: alternative samples**

	Excluding US firms		Excluding UK firms		Excluding US and UK firms	
	(1)	(2)	(3)	(4)	(5)	(6)
	BIODSL	BIODSL	BIODSL	BIODSL	BIODSL	BIODSL
CARBON	1.479*** (0.233)		1.746*** (0.228)		1.790*** (0.256)	
WASTE		1.888*** (0.176)		1.561*** (0.175)		1.727*** (0.191)
Controls	Included	Included	Included	Included	Included	Included
Constant	-87.55*** (5.292)	-72.44*** (5.814)	-84.54*** (5.214)	-71.42*** (5.872)	-84.70*** (5.976)	-69.43*** (6.529)
Country fixed effects	Yes	Yes	Yes	Yes	Yes	Yes
Industry fixed effects	Yes	Yes	Yes	Yes	Yes	Yes
Year fixed effects	Yes	Yes	Yes	Yes	Yes	Yes
Observations	6313	5103	6590	5349	5277	4434
Adj. R <sup>2</sup>	0.338	0.315	0.304	0.287	0.337	0.321

Notes: This table presents the results from regressing biodiversity disclosures (BIODSL) on corporate carbon emissions (CARBON), waste levels (WASTE), and other covariates excluding US and UK firms. Robust standard errors below estimated coefficients are reported in parentheses. \*, \*\*, and \*\*\* denote significance of the estimated coefficients at the 10%, 5%, and 1% levels, respectively.

Source(s): Authors' own work

**Table 10. Additional analyses: the effects of regulatory settings**

	(1)	(2)	(3)	(4)	(5)	(6)
	BIODSL	BIODSL	BIODSL	BIODSL	BIODSL	BIODSL
CARBON	1.251*** (0.231)		1.846*** (0.281)		2.756*** (0.389)	
WASTE		1.380*** (0.209)		2.125*** (0.226)		2.049*** (0.336)
CARBON*INDUM	0.896*** (0.305)					
WASTE*INDUM		0.689*** (0.258)				
CARBON*PARIS			-0.492** (0.250)			
WASTE*PARIS				-0.623*** (0.230)		
CARBON*CBD					-1.476*** (0.372)	
WASTE*CBD						-0.393 (0.345)
INDUM	-8.727* (4.470)	-11.813*** (3.525)				
PARIS			12.778*** (4.219)	11.127*** (3.894)		
CBD					15.741*** (5.396)	-1.073 (4.503)
Controls	Included	Included	Included	Included	Included	Included
Constant	-86.70*** (5.285)	-68.04*** (5.892)	-90.07*** (5.126)	-77.51*** (5.596)	-98.53*** (6.472)	-71.57*** (6.487)
Country fixed effects	Yes	Yes	Yes	Yes	Yes	Yes
Industry fixed effects	Yes	Yes	Yes	Yes	Yes	Yes
Year fixed effects	Yes	Yes	Yes	Yes	Yes	Yes
Observations	7626	6018	7626	6018	7626	6018
Adj. R <sup>2</sup>	0.309	0.285	0.308	0.285	0.310	0.285

Notes: This table presents the results from regressing biodiversity disclosures (BIODSL) on corporate carbon emissions (CARBON), waste levels (WASTE), and the moderating effects of environmentally sensitive industries (INDUM), the Paris Agreement (PARIS), and the Convention on Biological Diversity (CBD). Robust standard errors below estimated coefficients are reported in parentheses. \*, \*\*, and \*\*\* denote significance of the estimated coefficients at the 10%, 5%, and 1% levels, respectively.

**Source(s):** Authors' own work

## Appendix A: Sample breakdown by country and industry

<b>Panel A: Country-level sampling</b>				
<b>Country</b>	<b>No. of firms</b>	<b>Percentage</b>	<b>Firm-years</b>	<b>Percentage</b>
AUSTRALIA	27	3.03	239	3.13
AUSTRIA	4	0.45	44	0.58
BELGIUM	11	1.23	88	1.15
BRAZIL	41	4.60	377	4.94
CANADA	31	3.48	281	3.68
CHILE	12	1.35	81	1.06
CHINA	3	0.34	25	0.33
COLOMBIA	5	0.56	31	0.41
DENMARK	7	0.78	64	0.84
FINLAND	14	1.57	146	1.91
FRANCE	36	4.04	304	3.99
GERMANY	37	4.15	324	4.25
GREECE	3	0.34	26	0.34
HONG KONG	16	1.79	103	1.35
HUNGARY	2	0.22	18	0.24
INDIA	7	0.78	69	0.90
IRELAND	7	0.78	68	0.89
ISRAEL	1	0.11	11	0.14
ITALY	18	2.02	160	2.10
JAPAN	109	12.22	822	10.78
LUXEMBOURG	2	0.22	24	0.31
MALAYSIA	9	1.01	62	0.81
MEXICO	3	0.34	23	0.30
NETHERLANDS	19	2.13	168	2.20
NEW ZEALAND	9	1.01	58	0.76
NORWAY	6	0.67	56	0.73
PHILIPPINES	8	0.90	68	0.89
PORTUGAL	6	0.67	58	0.76
RUSSIAN FEDERATION	3	0.34	16	0.21
SINGAPORE	8	0.90	60	0.79
SOUTH AFRICA	40	4.48	334	4.38
SOUTH KOREA	21	2.35	201	2.64
SPAIN	17	1.91	154	2.02
SWEDEN	21	2.35	175	2.29
SWITZERLAND	26	2.91	241	3.16
TAIWAN	22	2.47	179	2.35
THAILAND	8	0.90	66	0.87
TURKEY	6	0.67	41	0.54
UNITED ARAB EMIRATES	2	0.22	12	0.16
UNITED KINGDOM	107	12.00	1036	13.59
UNITED STATES	158	17.71	1313	17.22
<b>Total</b>	<b>892</b>	<b>100.00</b>	<b>7626</b>	<b>100.00</b>
<b>Panel B: Industry-level sampling</b>				
<b>Industry</b>	<b>No. of firms</b>	<b>Percentage</b>	<b>Firm-years</b>	<b>Percentage</b>
Basic Materials	95	10.65	785	10.29
Consumer Discretionary	138	15.47	1182	15.50
Consumer Staples	81	9.08	669	8.77
Energy	56	6.28	464	6.08
Health Care	50	5.61	445	5.84
Industrials	215	24.10	1866	24.47
Real Estate	73	8.18	594	7.79
Technology	50	5.61	441	5.78
Telecommunications	53	5.94	504	6.61
Utilities	81	9.08	676	8.86
<b>Total</b>	<b>892</b>	<b>100.00</b>	<b>7626</b>	<b>100.00</b>

Source(s): Authors' own work



## Appendix B: Variable descriptions

Variable	Acronym	Description
Biodiversity disclosure	BIODSL	The index is calculated based on 9 disclosure indicators, which reflect a company's impacts on biodiversity and threatened species, together with corporate initiatives/policies/actions to enhance the positive impacts and mitigate the adverse effects of its business operations on the environment and ecosystems. These indicators include biodiversity loss reduction, sulfur and nitrogen oxides emissions reduction, waste reduction, e-waste reduction, environmental restoration, eco-design products/processes, toxic chemicals and substances reduction, environmental innovation/product impacts minimization, and land environmental impact reduction. Each indicator is assigned a value of one if reported, and zero otherwise. The weighted average index is then calculated within each industry and ranges between 0% and 100%.
Carbon emissions	CARBON	Natural logarithm of actual carbon emissions.
Waste levels	WASTE	Natural logarithm of actual waste levels
Corporate governance	CGOVR	The score assesses the quality of corporate governance related to board structures (diversity, independence, and committees), their functions, roles, responsibilities, and executive compensation. It varies between 0% and 100%.
Firm size	FSIZE	Natural logarithm of total assets.
Profitability	FPROF	Net earnings divided by total assets*100.
Financial risk	FLVRG	Total debts divided by total assets*100.
Slack	FSLCK	The ratio of cash and cash equivalents to total assets.
Capital intensity	CAPEX	The ratio of capital expenditures to total assets.
Inflation rate	INFLN	Percentage change in the retail prices of goods, products and services that can be changed or fixed between two consecutive years.
GDP growth	GDPGR	Percentage change in real GDP between two consecutive years.

Source(s): Authors' own work