

# 1 The business case for investing in

## 2 biodiversity data

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51 **Author contributions**

52 FH conceptualised the paper. FH, JN, CA, APD, ME, AH, VJ, OVP, RMG, FR, FR and TR took  
53 part in workshop discussions that further developed the framework and ideas throughout the  
54 manuscript. FH created the financial model for private-sector investment towards biodiversity  
55 data. FH drafted the manuscript, and all authors contributed critically to writing all drafts.  
56

57 **Acknowledgements**

58 We thank Stiftelsen Oscar och Lili Lamms Minne for funding the “Emerging Methods in  
59 Business and Biodiversity: Data is the Shared Language” workshop at Ekenäs Herrgård, Sweden,  
60 on December 5-6, 2024.

61 **Conflict of Interest**

62 David M. Baker and Fabian Roger are affiliated with nature-tech companies (Archireef and  
63 DNAir AG, respectively) which are used as case studies in this article. These affiliations are  
64 disclosed in the interest of transparency. The authors declare that these relationships have not  
65 influenced the objectivity or integrity of the manuscript.

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87 **The business case for investing in biodiversity data**

88 **Abstract**

- 89 1. There is urgent demand for biodiversity data driven by the need to assess impacts, dependencies,  
90 risks, and to implement nature-based solutions. In a data-driven economy, without access to  
91 robust data and the tools built from it, public and private sector actors cannot reliably evaluate  
92 their relationships with biodiversity or the outcomes of any sustainable nature-positive  
93 intervention.
- 94 2. We identify three key barriers to effective biodiversity action: (1) the lack of biodiversity data;  
95 (2) limited biodiversity data literacy and the domain expertise required to apply data products in  
96 decision-making; and (3) the limited financing facilitation to channel capital, particularly from  
97 the private sector, toward reliable, high-impact open biodiversity data.
- 98 3. Building on this, we present a streamlined end-to-end framework of the key stages from  
99 biodiversity data to nature-positive action, mapping biodiversity data to data products and  
100 business use cases, establishing biodiversity data as a critical investment.
- 101 4. First, we explain the origins of primary biodiversity data and the interdependence of specimen-  
102 based primary biodiversity data with data generated from new technologies including  
103 environmental DNA, computer vision and acoustic monitoring. These, collectively feed open  
104 biodiversity infrastructures like the Global Biodiversity Information Facility (GBIF).
- 105 5. Then, we discuss biodiversity data products, focusing on the ability to interpret and effectively  
106 apply biodiversity models, metrics, and tools in relevant contexts. We address the challenges  
107 posed by the complexity of biodiversity, the importance of its definitions, and the use of  
108 aggregated metrics for biodiversity and ecosystem services in reporting, including the role of  
109 nature-tech. We show case studies from a finance-academia partnership, multinational industry, a  
110 tropical biodiversity hotspot and nature-tech to illustrate both progress, gaps and opportunities.
- 111 6. Finally, we propose an innovative blended financing model to incentivize and reward direct  
112 investments in biodiversity data from multiple sources, with specific attention to business and  
113 private capital funds. We conclude that investing in biodiversity data is the urgent step in  
114 enabling nature-positive action and driving scalable, data-driven solutions to the biodiversity  
115 crisis.

116 **Key words**

117 Biodiversity data, Business and biodiversity, blended finance, data mobilisation, financing biodiversity,  
118 Sustainable Finance, Global Biodiversity Information Facility (GBIF), Natural History Collections,  
119 Nature-tech, Nature-based Solutions

120

121 **Introduction**

122 Biodiversity underpins ecosystem functioning and the flow of nature's contributions to people, including  
123 climate regulation, clean air and water, food security, and disease control (Díaz et al., 2018; Mace et al.,  
124 2012). However, anthropogenic pressures including land-use intensification, pollution, invasive species,  
125 and climate change are driving rapid biodiversity loss, threatening planetary boundaries (Steffen et al.,  
126 2015). This degradation is not only ecological but economic. Despite mounting evidence, the 2024 Global  
127 Risks Report by the World Economic Forum suggests that biodiversity risks will only become significant  
128 in the next decade (WEF, 2024), downplaying the present-day severity of the crisis. Biodiversity loss is  
129 already costing the global economy over \$5 trillion per year (Ranger et al., 2023). Despite this, there is  
130 still persistent underinvestment in high-quality biodiversity data that is essential for developing urgent,  
131 data-driven solutions of national and international importance (Gerber & Iacona, 2024). This neglect  
132 reflects a fundamental failure of economic, environmental and social governance and long-term planning  
133 for future generations.

134

135 In parallel, global policy and regulatory landscapes are evolving to mainstream biodiversity into financial  
136 and business decision-making. The Kunming-Montreal Global Biodiversity Framework (GBF), agreed at  
137 COP15, marks a turning point, with Target 15 requiring large businesses and financial institutions to  
138 assess and disclose their biodiversity-related impacts, risks and dependencies across operations, supply  
139 chains, and portfolios (COP15, 2022). In the EU, the Corporate Sustainability Reporting Directive  
140 (CSRD) mandates environmental reporting for nearly 50,000 companies by 2025 (Faqih & Kramer,  
141 2024). Voluntary frameworks are also gaining momentum. The Taskforce on Nature-related Financial  
142 Disclosures (TNFD), now supported by over 500 organisations representing £17.7 trillion in assets, is  
143 developing guidance for integrating nature into financial decision-making (TNFD, 2024). Similarly, the  
144 Science Based Targets Network (SBTN), originally focused on climate, has expanded its scope to include  
145 biodiversity (SBTN, 2020). These frameworks aim to channel capital toward nature-positive solutions and  
146 encourage companies to embed biodiversity into sustainability strategies, just as the Paris Agreement  
147 catalysed corporate climate targets (Allen et al., 2025). Yet while these frameworks represent progress,

148 their implementation relies heavily on the availability, accessibility, and interpretability of biodiversity  
149 data.

150

151 Without high-quality biodiversity data, neither companies nor policymakers can assess impacts,  
152 dependencies, risks, or implement effective nature-based solutions. This challenge is particularly acute  
153 because, unlike carbon accounting, where standardised units like CO<sub>2</sub>-equivalents enable comparability,  
154 biodiversity lacks a unified metric (Jones & Solomon, 2013). Biodiversity data is multidimensional,  
155 spanning genetic, species, and ecosystem diversity, and is highly context-dependent (CBD, 2011;  
156 Exposito-Alonso et al., 2022). Essential Biodiversity Variables (EBVs) are an example of metrics  
157 developed to aid sustainability reporting on biodiversity (Pereira et al., 2013). Yet, these metrics are often  
158 derived from top-down approaches that rely on indirect data, which can be biased, incomplete, difficult to  
159 verify, and therefore misleading, highlighting the need for reliable, ground-truthed biodiversity data  
160 (Granqvist et al., 2025). Even seemingly simple metrics, such as species richness, can be misleading  
161 without appropriate ecological context (Hillebrand et al., 2018). It is little wonder that uncertainty persists  
162 as biodiversity models and metrics continue to evolve, with over 2,000 metrics currently available, their  
163 utility remains closely tied to the quality of underlying data (Burgess, 2024). Businesses, facing rising  
164 sustainability disclosure requirements, are increasingly turning to the Global Biodiversity Information  
165 Facility (GBIF), the largest biodiversity data infrastructure with over 2 billion records, to report on their  
166 biodiversity impacts, dependencies, risks, and implement effective nature-based solutions. This makes  
167 sense, as GBIF is the leading infrastructure for biodiversity data, and feeds into our global biosphere earth  
168 system models. However, it is important to have the domain knowledge to recognise that these data may  
169 not be suitable for use, particularly since they are unevenly distributed due to historical taxonomic and  
170 geographic biases in data collection, with 65% representing birds, which comprise only 0.5% of described  
171 species (GBIF, 2025; Troudet et al., 2017). Urgent investment towards biodiversity data collection and  
172 mobilisation is essential to expand the coverage and quality of open biodiversity data, enabling key  
173 stakeholders across both public and private sectors to make evidence-based decisions.

174

175 To achieve this, a concerted science communication effort is needed to clarify the biodiversity data  
176 pipeline: from data origins to usable products and business use cases (Figure 1). Yet, to our knowledge,  
177 no streamlined framework currently exists, risking the integrity of the pipeline. At the foundation of  
178 reliable biodiversity data lie often-overlooked natural history collections, including museums, herbaria,  
179 specimen repositories, and seed banks, that support species verification, biogeographic checklists, red-list  
180 assessments, and GBIF data uploads (Huybrechts et al., 2022; Mason Heberling et al., 2021; Davis,  
181 2023). These collections provide critical data for emerging technologies such as eDNA reference libraries

182 and training data for machine learning models, enabling advances like rapid eDNA assessment, remote  
183 sensing, acoustic monitoring, and computer vision to make biodiversity data collection faster, cheaper,  
184 and easier (Beery, 2023; Buxton et al., 2018; Deiner et al., 2021; van Klink et al., 2022). Another major  
185 barrier to progress remains the long-standing taxonomic impediment: a global shortage of taxonomists  
186 due to lack of funding prioritization and a critical bottleneck that limits our ability to understand and  
187 monitor biodiversity (Engel, 2021; Löbl et al., 2023). As demand for biodiversity data grows,  
188 organisations like the Nature Tech Collective, TNFD, and WWF have started to draw attention to the  
189 “biodiversity data gap,” highlighting the importance of careful interpretation of biodiversity data (Goran,  
190 2024; TNFD, 2024; WWF, 2024). Nature-tech experienced rapid growth as a response, and platforms  
191 have also proliferated, repackaging open-access data from GBIF behind paywalls, which attracted over  
192 USD 2 billion in investment in 2022, growing 52% annually since 2018 (Evison et al., 2022; Goren,  
193 2024), reflecting strong investor interest. Yet, without stewarding the very foundation of the biodiversity  
194 data pipeline and channeling our resources to financing of biodiversity data (Figure 1), discovery of  
195 solutions to the biodiversity crisis stands on a crumbling foundation, leaving us blind to impacts, risks,  
196 and the success of nature-based solutions.

197

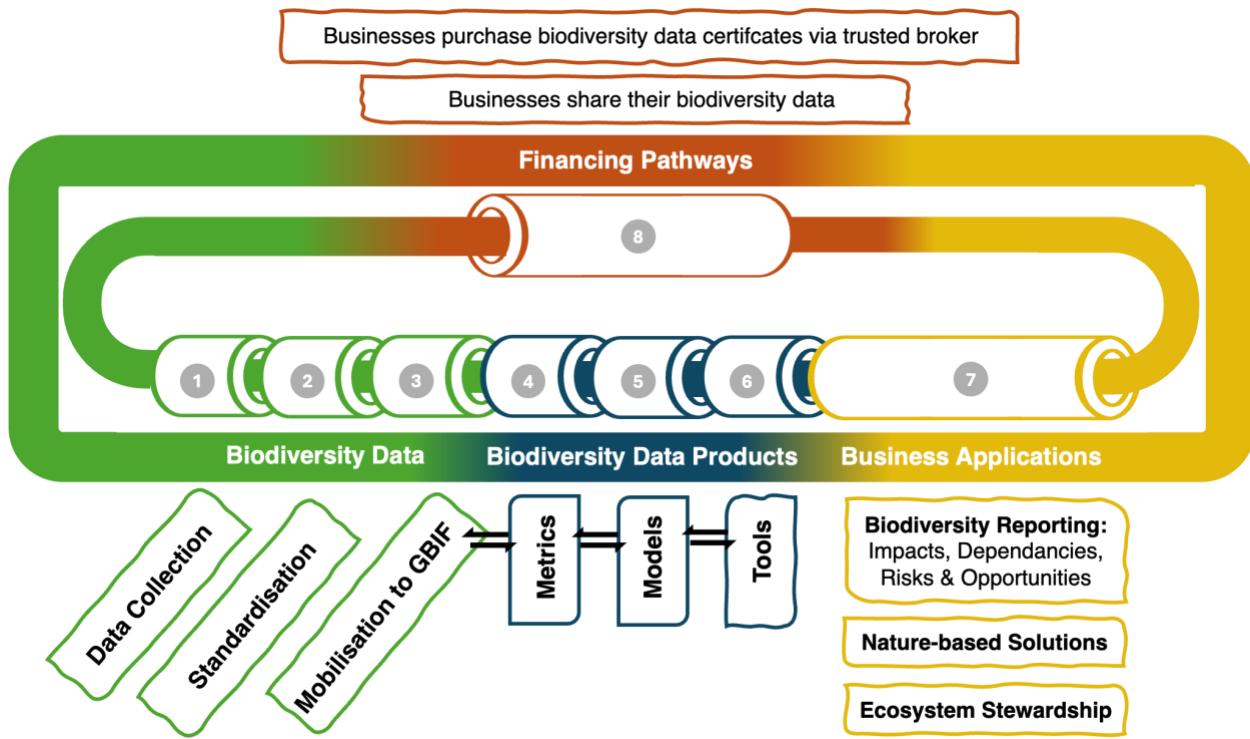
198 Limited biodiversity data remains a critical barrier to effective conservation, constraining both policy and  
199 investment decisions. This data gap parallels a persistent financing shortfall: only one-sixth of the  
200 required annual funding is currently met. While most biodiversity finance comes from public sources,  
201 private capital, accounting for less than 20% holds significant untapped potential to help bridge both the  
202 data and financing gaps essential for sustaining biodiversity (Beverdam et al., 2025). Addressing the  
203 biodiversity data gap requires channelling financial resources towards strengthening existing  
204 infrastructures like GBIF rather than duplicating efforts and fragmenting data. Without transparent  
205 biodiversity data and metadata, clear provenance, and proper validation, businesses that use these  
206 platforms risk producing unmeaningful analyses that jeopardize their operations. Biodiversity data and  
207 infrastructure are not just tools but valuable, investable assets with demonstrated returns: every €1  
208 invested in GBIF generates an estimated €3 in direct user benefits and up to €12 in broader societal  
209 biodiversity impact, with clear implications for business value (Deloitte, 2023). Despite advances  
210 ushering in a new era of tools to collect big biodiversity data (Musvuugwa, 2021), financing models to  
211 support the large-scale collection and mobilisation of biodiversity data essential for generating high-  
212 quality, findable, accessible, interoperable, and reusable (FAIR) data (Wilkinson et al., 2016) remain  
213 scarce. Mobilising financial resources across sectors, particularly private finance, and developing blended  
214 finance models are urgent to enable evidence-based, sustainable solutions and to avoid greenwashing,  
215 reputational harm, and ineffective strategies for climate mitigation and adaptation (Mair et al., 2024;

216 Ingram et al., 2024; Smith et al., 2019; White et al., 2023). The urgency is amplified by emerging markets  
217 developing for nature-based credits, which must be data-driven to avoid irreversible harm (Swinfield,  
218 2024). Recognising biodiversity data as a vital, investable public–private good will enable more accurate  
219 biosphere earth system modelling, forecasting biodiversity trends, reveal ecological and economic links,  
220 and drive transformative business action towards sustainability.

221

222 In this paper, we (1) present a framework for integrating biodiversity data into business use cases, (2)  
223 clarify the origins of primary biodiversity data, and (3) propose a blended innovative financig model that  
224 connects private capital to high-impact biodiversity data generation and mobilisation via a biodiversity  
225 data facilitator through open infrastructures such as GBIF. In doing so, we argue that the business case for  
226 investing in biodiversity data lies in its ability to unlock the value of information, enabling private and  
227 public sector actors to better address biodiversity impacts, dependencies, and risks, and to implement  
228 data-driven nature-based solutions for long term sustainability.

229 **An end-to-end framework from biodiversity data to business use cases**



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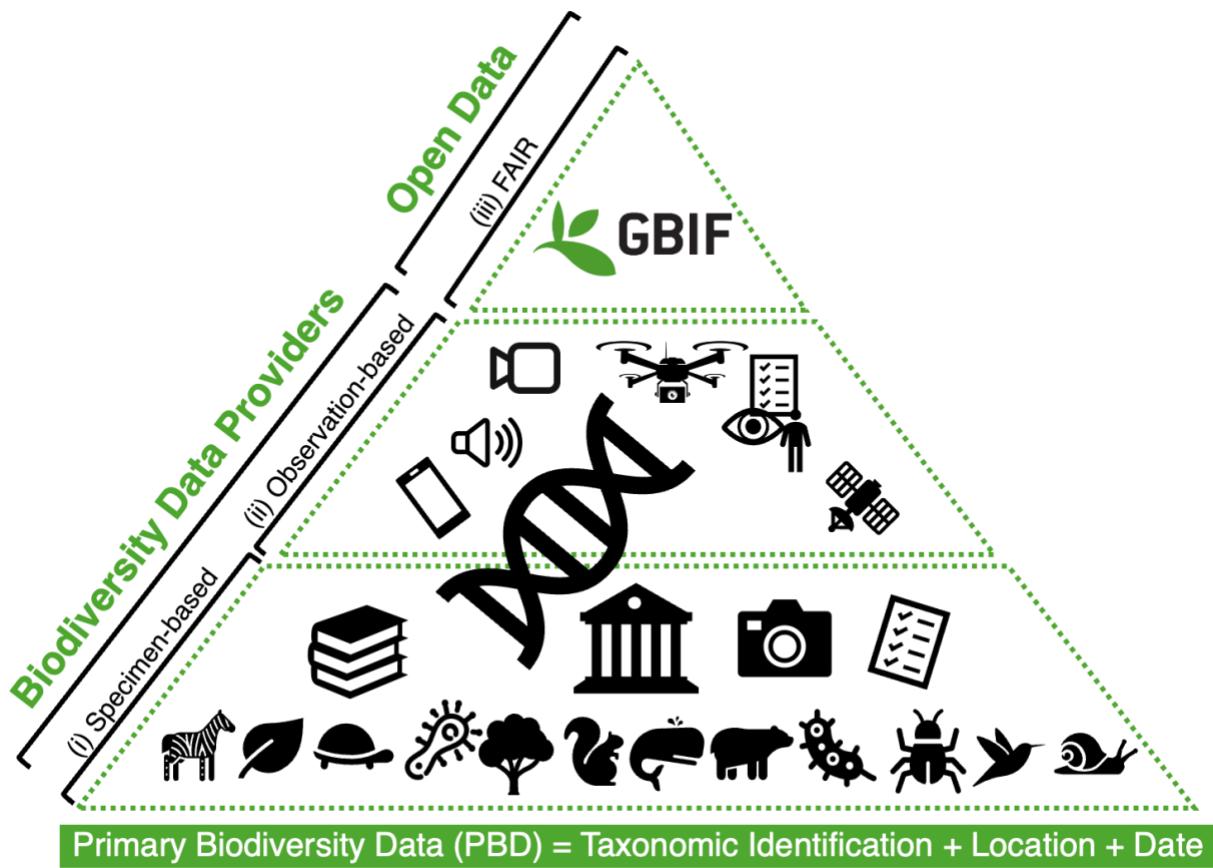
231

232 **Figure 1.** Framework for integrating biodiversity to business use-cases, grounded in data. This pipeline  
233 clarifies the steps to get from biodiversity data to biodiversity data products to business use cases. The process  
234 begins with biodiversity data (green), divided into three key components: (1) collection of biodiversity data  
235 from both the field and natural history collections using traditional and new technologies; (2) standardisation  
236 of this data in alignment with FAIR data principles and the biodiversity information standards (TDWG); and  
237 (3) mobilisation of these data into the Global Biodiversity Information Facility (GBIF), the world's largest  
238 biodiversity data repository. The next step involves biodiversity data products (blue), where biodiversity  
239 domain knowledge is needed to translate raw data into (4) biodiversity metrics (e.g., EBVs); (5) models and  
240 predictive models; and (6) data tools for users, which are iteratively updated as new data become available.  
241 This biodiversity expertise has traditionally been represented by academic research, consultancies, and public  
242 environmental agencies, but is increasingly adopted by the rapidly evolving nature-tech sector. Finally,  
243 business (yellow) represents the end-users of the data products (7). Important use cases for biodiversity data  
244 products include impacts, dependencies and risk reporting, investments in nature-based solutions (NbS) and  
245 monitoring of their outcomes, as well as better management practices through ecosystem stewardship. There  
246 are two important feedback loops in the framework. First, businesses are encouraged to invest in data  
247 collection and mobilisation of these data to the public domain, to improve their reporting and public image (8).  
248 Second, we propose a mechanism to incentivise direct investment in biodiversity data mobilization via a

249 biodiversity data facilitator (9). This flow emphasises that biodiversity data is a central priority of the entire  
250 pipeline. Without high-quality biodiversity data, none of the subsequent steps are possible.

## 251 **1 Biodiversity data**

252 To address the biodiversity data gap, we must understand biodiversity data origins and data types.  
253 Primary biodiversity data, or occurrence data, constitutes the majority of data published through GBIF  
254 (GBIF, 2024) and includes three key components: taxonomic level (e.g., species, genus), location, and  
255 date (Spear et al., 2023). Observation- and specimen-based biodiversity data are interdependent and must  
256 be prioritized together to address the biodiversity data gaps (Figure 2).



257  
258 **Figure 2.** Pyramid diagram illustrating sources of primary biodiversity data showing that specimen-based  
259 data are foundational to observation-based data, which are foundational to integrated primary biodiversity  
260 data. (i) Specimen-based data: Derived from physical specimens in natural history collections, including  
261 image files, checklists, and archival materials; (ii) Observation-based data: Derived from traditional  
262 species inventories and technologies such as DNA methods, camera traps, audio recordings, and citizen  
263 science. DNA methods (eDNA and metabarcoding, metagenomics) overlap with both specimen and

264 observation data, as they require physical sample collection to generate verifiable species names; and (iii)  
265 Mobilised primary biodiversity data: Integrated specimen- and observation-based data on open access  
266 biodiversity infrastructure, such as GBIF.

267  
268 The big data revolution has already expanded biodiversity data, and we are in an excellent point in history  
269 to start prioritising resources to fill data gaps in taxa and geographic coverage (Bayraktarov et al., 2019;  
270 Musvuugwa et al., 2021; Troudet et al., 2017). Digitisation and AI are accelerating data extraction from  
271 specimens in natural history collections, providing temporal biodiversity data. High-throughput systems  
272 like ALICE can rapidly digitise insect specimens for upload to GBIF (Dupont & Price, 2019; Garner et  
273 al., 2024). Techniques such as Named Entity Recognition (NER) and Optical Character Recognition  
274 (OCR) convert label text to digital format, enabling rapid processing of tens of thousands of specimens  
275 (Takano et al., 2024). Digitised herbarium specimens provide additional biodiversity data, such as trait  
276 data comparable to fresh tissues (Davis, 2023). While significant progress has been made, billions of  
277 specimens remain undigitised from European collections alone, requiring continued mobilisation efforts  
278 (Huybrechts et al., 2022). New scalable observation-based technologies are transforming biodiversity  
279 monitoring. eDNA and metabarcoding enable rapid species detection from water, soil, or air samples are  
280 useful for hidden taxa like fungi and insects, potentially completing centuries of manual inventory work  
281 in a single year (Deiner et al., 2021; Ronquist et al., 2020). Passive acoustic monitoring captures species  
282 sounds, enabling AI-driven species identification across taxa including birds, primates, and even soil  
283 organisms (Hildebrand et al., 2024; Buxton et al., 2018). Camera traps and global camera networks offer  
284 real-time species monitoring (Bjerge et al., 2021; Steenweg et al., 2017, Høye et al 2021), while citizen  
285 science platforms like iNaturalist leverage smartphones to gather data and train deep learning models  
286 (August et al., 2015; Beery, 2023). Advanced sensors on drones and UAVs, including thermal, LIDAR,  
287 hyperspectral, and RGB, further enhance species tracking and vegetation analysis. These tools enable  
288 species-level plant identification and habitat mapping (Larsen et al., 2023; Mäyrä et al., 2021). However,  
289 consistent data standardisation is necessary to ensure biodiversity data from diverse sources are actually  
290 useful to science and society. Biodiversity data must be curated to meet FAIR principles (Findable,  
291 Accessible, Interoperable, Reusable) and align with Open Science practices (Carroll et al., 2021;  
292 Wilkinson et al., 2016).

293  
294 The Biodiversity Information Standards group (TDWG) maintains standards like Darwin Core (DwC),  
295 which enables interoperability through standardised terms and vocabularies. Extensions such as the  
296 Humboldt Ecological Inventory and networks like DiSSCo further support standardised ecological and  
297 specimen-based data sharing. However, integrating big biodiversity data faces challenges with metadata

298 standards for cross-scale analysis (Maldonado et al., 2015; Hardisty et al., 2022). Metadata defines data  
299 context, without which we would not make full use of stitching together data from diverse sources. For  
300 instance, dnaDerivedData with MIxS provides structured metadata on DNA sampling and sequencing  
301 (Abarenkov et al., 2023). The Camera Trap Data Package (Camtrap DP) standardises image data sharing  
302 and is being expanded to include insects (Bubnicki et al., 2023). Ecological Metadata Language (EML)  
303 also supports detailed documentation across biodiversity datasets. Despite progress, challenges remain:  
304 such as data duplication, inconsistent quality, and interoperability issues (Pyle et al., 2021). Taxonomic  
305 inconsistencies and errors can be managed through automation and expert validation (ChecklistBank,  
306 2025; Whitley et al., 2024). These advances support platforms like GBIF, which mobilise open-access  
307 biodiversity data and bridge data gaps. Motivating data sharing remains difficult. Academic incentives  
308 include DOI citations and data papers, while businesses are beginning to see strategic value in sharing  
309 data. Companies can publish data via GBIF's Integrated Publishing Toolkit (IPT), usually free, and track  
310 impact through assigned DOIs and UUIDs (Case Study S2). Nonetheless, the business sector currently  
311 contributes only 0.3% of GBIF records, indicating a opportunity for greater engagement.

## 312 **2 Biodiversity data products**

313 There is a growing demand to transform raw biodiversity data into metrics and data products that can  
314 cater to diverse use cases and needs across different industry sectors (Burgess et al., 2024). This task  
315 requires reducing the complexity of biodiversity into manageable metrics, which arguably is an exercise  
316 of great oversimplification, yet a necessary one. With this inherent constraint in mind, we reflect on  
317 several issues in the current state-of-the-art of biodiversity reporting and the underlying data-products.  
318

319 In the context of biodiversity impact reporting, data products that provide regional or global heatmaps of  
320 biodiversity metrics are in high demand, as they allow easy area-based calculation of biodiversity value  
321 and impact. One example of such a data product is the Biodiversity Intactness Index (BII, Newbold et al.,  
322 2015; Phillips et al., 2021), which is proposed as a component indicator in the COP 16 draft of the GBF  
323 monitoring framework (CBD, 2024). Another example of a biodiversity model used in business context is  
324 GLOBIO (Schipper et al, 2020), also proposed as a GBF indicator (CBD, 2024). However, many of such  
325 global heat maps generated (Myers et al., 2000) are only weakly linked to the evaluation of the  
326 biodiversity impact of *specific decisions and actions*. For biodiversity data products to be actionable in a  
327 corporate setting, they need to relate biodiversity impacts and risks to operational and financial decisions  
328 taken by companies, so that impact tradeoff analysis can be performed. Examples include spatial planning  
329 for forest and agricultural land management, deciding from which countries and regions to source

330 materials and products, and investments into new factories and logistics facilities. A common  
331 denominator for many use cases is the urgent need for regional and local data and models (as opposed to  
332 global) to ensure high-quality analysis and drawing the right conclusions.

333

334 While the BII and other similar data-products are being used for company impact assessment and  
335 reporting, a concern raised is that the underlying models are largely untested for their predictive  
336 performance and their agreement with other indicators of biodiversity impact (Martin et al., 2019,  
337 Nyström, 2024). We see a big risk that insufficiently tested data products provide the foundations for  
338 company impact reporting and nature investments, with potentially negative consequences. This problem  
339 is further exacerbated by the quickly developing nature tech market, driven by the demand for attractive,  
340 ready-to-use biodiversity solutions and data. The absence of a thorough quality-checking and peer-review  
341 process in this context lends reason for concern and makes it difficult for customers to distinguish  
342 between “snake-oil salesmen” with questionable data products and those built on solid foundations.  
343 However, as outlined above, even models and data products that have been reviewed by the academic  
344 peer-review process, risk being mis-applied for purposes they were not designed for. Part of the reason for  
345 this misapplication is a lack of guidance on the use of existing and emerging biodiversity data products  
346 and metrics, which we identify as the challenge of biodiversity data literacy.

347

348 Despite best efforts regarding methodological considerations and quality assessments, data products are  
349 only as reliable as the data they are derived from. At present, the biggest bottleneck to better biodiversity  
350 models is arguably the lack of contextualized data, particularly in view of the vast taxonomic and spatial  
351 biases that exist. Closing the biodiversity data gaps is critical for enhancing the accuracy and reliability of  
352 biodiversity metrics for business use cases.

353

### 354 **3 Business use cases**

#### 355 **Improved biodiversity impacts, dependencies and risk reporting**

356 Businesses face increasing demands to assess and disclose biodiversity impacts, dependencies, and risks  
357 under regulations such as the EU Corporate Sustainability Reporting Directive (CSRD) and frameworks  
358 like the Taskforce on Nature-related Financial Disclosures (TNFD) (Figure 1). Meeting these  
359 requirements needs long-term investment in high-quality biodiversity data to develop reliable data  
360 products and metrics that support science-based targets and withstand regulatory and investor scrutiny  
361 (ESRS E4, 2023). Current biodiversity data gaps and biases undermine the accuracy of risk assessments.  
362 The TNFD’s 2023 scoping study identified existing “nature data” as outdated, inconsistent, and

363 insufficiently detailed for confident decision-making (TNFD, 2023). Addressing these deficiencies  
364 requires sustained investment in biodiversity data collection, standardisation, and mobilisation to improve  
365 data accessibility, comparability, verifiability, and assurability. Initiatives like the proposed “Nature Data  
366 Public Facility” aim to leverage platforms such as GBIF to provide decision-useful data products essential  
367 for corporate reporting and risk management (TNFD, 2024). Companies proactively investing in  
368 biodiversity data collection and publishing primary data on GBIF will enhance transparency and generate  
369 credible biodiversity metrics which may be used as key performance indicators in sustainability  
370 disclosures. For instance, TotalEnergies has shared over 51,000 biodiversity records from environmental  
371 impact assessments since 2019, supporting compliance with the Kunming-Montreal Global Biodiversity  
372 Framework (GBF Target 19) and bolstering accountability (Case Study S2; Figueira et al., 2023).

373

#### 374 **Nature-based solutions and sustainable ecosystem stewardship**

375 Nature-based solutions (NbS) require robust biodiversity data to quantify ecosystem outcomes (Seddon et  
376 al., 2020; Díaz et al., 2023). Longitudinal biodiversity monitoring underpins the development of data  
377 products and metrics that validate NbS effectiveness and enable adaptive management (Griscom et al.,  
378 2017; Pettorelli et al., 2021). Across industries, baseline biodiversity data are needed to quantify impacts  
379 and inform sustainable resource management. For example, in agriculture, investments in monitoring soil  
380 biodiversity, nutrient cycling, carbon sequestration, and water retention allow companies to substantiate  
381 claims of regenerative practices such as no-till farming and crop rotation (Case Study S1). Beyond supply  
382 chains, NbS projects, such as coral reef restoration (Case Study S4) also depend on high-resolution  
383 biodiversity data for baselines, to build on active biodiversity monitoring often obtained through eDNA,  
384 remote sensing, and manual efforts. In biodiversity hotspots such as Borneo (Case Study S3), there is  
385 significant potential if biodiversity data are effectively mobilised. This would enable the development of  
386 data-driven, nature-based interventions that can protect biodiversity, support local communities, and  
387 foster a blue-green economy (Ferraro & Hanauer, 2014; Struebig et al., 2015).

388

389 Nature-tech companies further leverage these biodiversity data products within innovative technologies to  
390 support sustainable development and deliver scalable conservation outcomes (Case Study S4; Watson et  
391 al., 2021). Moreover, investment in biodiversity data supports rigorous baseline data for, reporting, and  
392 verification frameworks essential for biodiversity credits, offsets, and other market-based instruments  
393 (Swinfield, 2024; Aide, 2024). Ensuring transparency and verifiability through data products enhances  
394 investor confidence and safeguards the integrity of biodiversity-positive projects (Faqih, 2024; TNFD,  
395 2023). Sustained investment in biodiversity data collection, standardisation, and mobilisation is therefore  
396 fundamental to producing credible data products and metrics that underpin NbS efficacy, ecosystem

397 stewardship, and long-term biodiversity conservation (Díaz et al., 2019; Mace et al., 2018; Pettorelli et al.,  
398 2021).

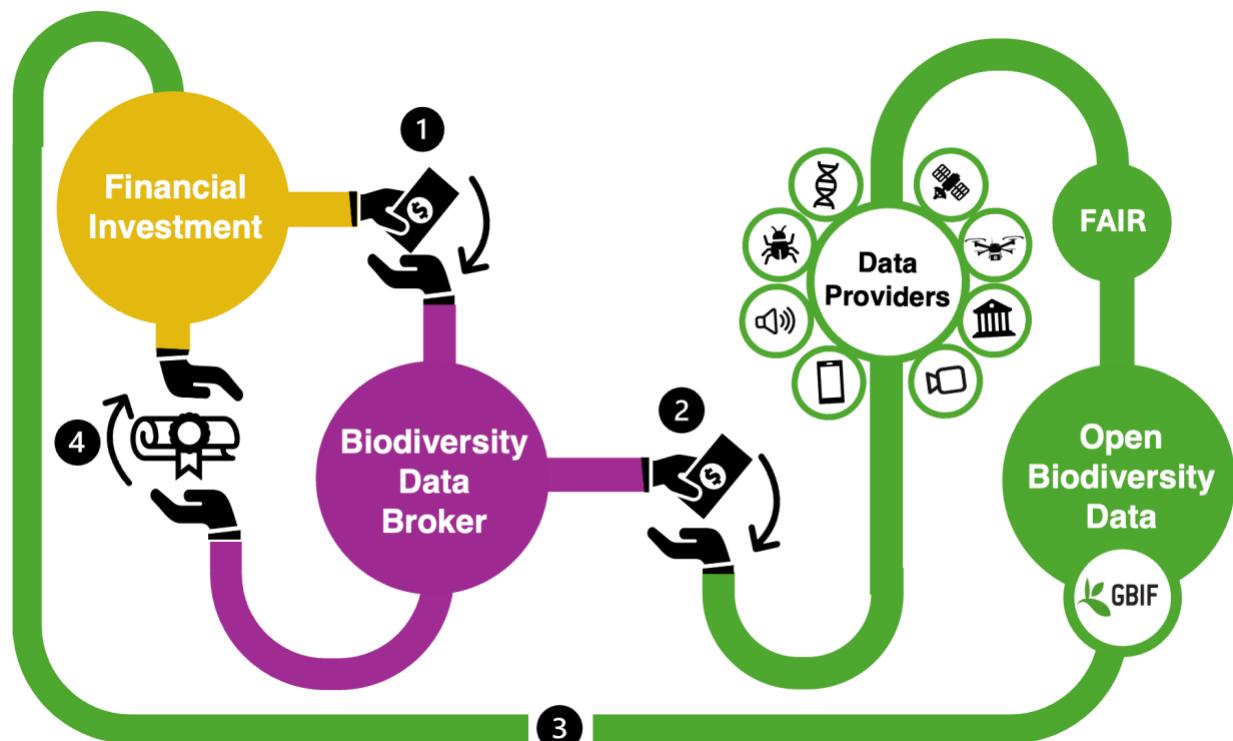
## 399 **4 An Innovative Financing Model to Address the Biodiversity Data Gap**

400 Addressing the global biodiversity data gap demands a sustainable financing model that incentivises the  
401 large-scale collection, standardisation, and mobilisation of biodiversity data. Biodiversity data  
402 certification offers such a solution by linking investment to measurable contributions in data generation  
403 and sharing. With this blended finance model, combining public, private, and philanthropic capital allows  
404 for targeted investment in biodiversity data and data infrastructures by reducing risk and aligning  
405 incentives across sectors (Flammer et al, 2025; Beverdam et al, 2025). It is particularly well-suited to  
406 financing the mobilisation of biodiversity data in regions and ecosystems where private capital alone is  
407 insufficient, such as the tropics (Case Study S3). This framework provides businesses and investors with  
408 certified recognition, while securing sustainable financing for data providers to contribute high-quality  
409 open-access biodiversity data via GBIF.

410

411 For example, the proposed biodiversity data certification model will offer a verifiable, data-driven  
412 mechanism to assess and demonstrate positive biodiversity impact. By linking investment to certified  
413 biodiversity data contributions, it reduces greenwashing risks and enhances transparency, enabling  
414 investors to allocate capital more effectively towards nature-positive outcomes. For instance, the Swedish  
415 MISTRA FinBio programme is developing eDNA monitoring and standardised data protocols for reliable  
416 biodiversity metrics that inform financial decision-making in agriculture and land management (Case  
417 Study S1). The project will publish its data to GBIF, ensuring open access and long-term utility, but once  
418 these methods are used, the biodiversity data facilitator can support the long-term biodiversity data  
419 mobilisation of these sites and other farms that will participate in biodiversity monitoring. Embedding  
420 biodiversity data stewardship with ecosystem stewardship to investment frameworks can show how the  
421 certification model can strengthen accountability and drive credible, impact-focused in the finance sector.  
422 TotalEnergies, a multinational company, already publishes biodiversity data from their environmental  
423 impact assessments directly to GBIF (Case Study S2). By leveraging in-house expertise and adhering to  
424 FAIR principles, the company exemplifies how private sector actors can integrate biodiversity data  
425 mobilisation into corporate sustainability frameworks, producing verifiable key performance indicators  
426 (KPIs) (Case Study S2). However, many companies may lack the capacity for employing internal data  
427 management, highlighting the need for a biodiversity data facilitator to broaden and incentivise private  
428 sector contributions to open biodiversity data (Figure 3).

429 The innovative financing model is particularly crucial for Earth's biodiversity-rich regions such as the  
430 tropics, where data gaps remain vast despite immense ecological importance. In Sabah, Malaysian Borneo  
431 (Case Study S3), decades of ecological research and millions of dollars from private sector stakeholders,  
432 including SimeDarby and IKEA, have generated extensive biodiversity data that remain largely  
433 inaccessible or unpublished on GBIF. Mobilising these data through certified financing can unlock  
434 valuable insights for sustainable land management and conservation, supporting both biodiversity and  
435 employment in local communities. Similarly, in the blue-green economy, nature-based solutions like  
436 those developed by Archireef - a nature-tech startup focused on coral reef restoration (Case Study S4) -  
437 demonstrate the potential for data-driven approaches to drive scalable ecosystem restoration and  
438 biodiversity monitoring. Innovative financing models can also support early-stage companies, such as  
439 Archireef, in sustaining their nature-positive impacts by funding efforts to mobilise and enhance  
440 biodiversity data. Ultimately, this financing model protects the reputational integrity of both investors and  
441 data providers. It mitigates greenwashing, strengthens ESG action, and positions biodiversity data as a  
442 strategic asset for long-term sustainability. Together, these cases illustrate how targeted, verifiable  
443 sustainable blended financing can accelerate biodiversity data mobilisation and support nature-positive  
444 outcomes in critical terrestrial and marine ecosystems.



445  
446 **Figure 3.** Innovative financing model for bridging the biodiversity data gap: (1) Private sector and other  
447 financial investors (yellow) provide funding to the biodiversity data facilitator (purple) to support data  
448 mobilisation; (2) the biodiversity data facilitator allocates these funds to partners for biodiversity data

449 collection, advancement of data standards, and mobilisation of data onto GBIF; (3) biodiversity data  
450 mobilised from all investors is made openly available on GBIF, allowing private sector and other  
451 investors to download and monitor their biodiversity impact; and (4) in return, businesses receive  
452 biodiversity data certification, providing verification of their positive biodiversity impact.

453 **5 Summary and call-to-action**

454 The biodiversity crisis exposes material impacts, dependencies, and risks for the private sector,  
455 compounded by persistent failures of governance and underinvestment in biodiversity data. Despite  
456 advances in technology and open infrastructures like GBIF, critical data gaps undermine effective nature-  
457 based solutions. We call on businesses to lead in financing biodiversity data collection and mobilization  
458 through an innovative blended finance model. This approach reduces reputational risk, enables informed  
459 decision-making, and strengthens long-term sustainability for both business and nature.

460 **Supporting Information**

461 **Case study S1. Biodiversity data for financial metrics: MISTRA FinBio,**  
462 **Sweden**



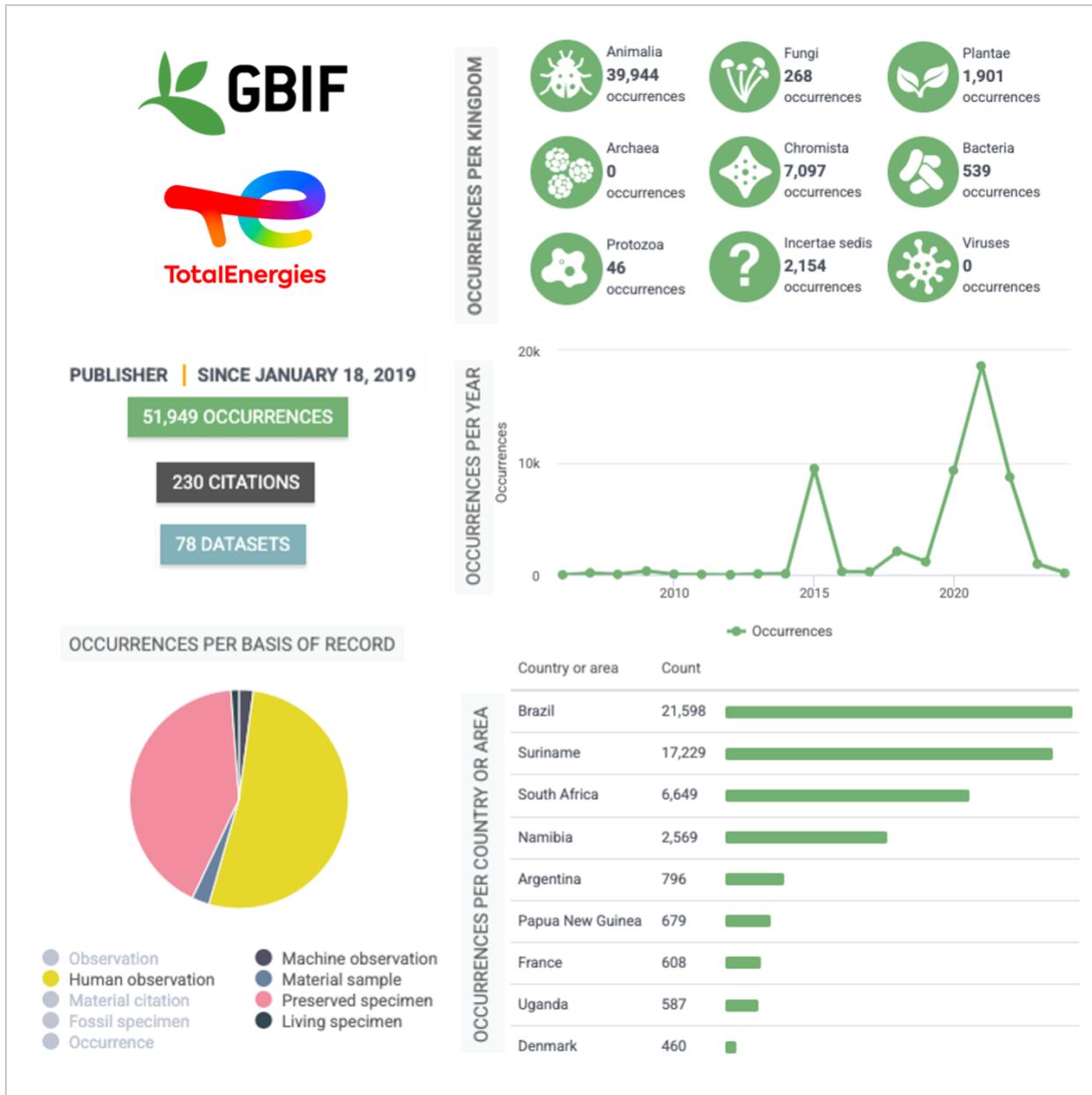
463  
464 **Figure S1.** Photo of Malaise traps in pilot fields, taken by E Granqvist May 2024.

465  
466 As part of the FinBio Research program hosted at the Stockholm Resilience Centre (FinBio, 2023), the  
467 “Biodiversity Data for Financial Metrics” work package aims to connect biodiversity data to financial  
468 decision-making to support financial institutions in contributing to biodiversity and nature-positive  
469 outcomes, particularly in Sweden and the Nordics. FinBio operates as a collaborative partnership between

470 academic and financial institutions, bringing together academic and impact partners to develop practical  
471 tools that guide investment decisions, promoting both the greening of finance and the financing of green  
472 initiatives that can be adopted throughout the financial sector. Specifically, the “Biodiversity Data for  
473 Financial Metrics” work package focuses on the application of modern monitoring technologies such as  
474 environmental DNA (eDNA) and Earth Observations for assessing biodiversity impact. This includes  
475 methods for assessing Essential Biodiversity Variables (EBVs) using standardised eDNA collection from  
476 Malaise traps and soil samples, with laboratory and bioinformatic protocols, accuracy measurements, and  
477 abundance estimation, with trend analyses covering a five-year period in Sweden. One pilot project, in  
478 collaboration with Svensk Kolinlagring - a non-profit organization - connects stakeholders to improve soil  
479 health and increase carbon storage in Swedish agricultural soils. The project works with approximately 40  
480 farms and focuses on measuring farmland biodiversity as part of its efforts to enhance carbon  
481 sequestration, which the IPCC recognizes as one of the most cost-effective and scalable climate solutions.  
482 The project aims to deliver several key outcomes, including biodiversity data from the agricultural sector  
483 using eDNA monitoring methods, analysis of biodiversity changes in carbon sequestration management  
484 systems, and the development of a standardized data-driven biodiversity index for farmers. This index  
485 will serve as both a measurement tool and a component of potential business cases to attract investment in  
486 sustainable agricultural practices. Open data and open methods are core principles within the Biodiversity  
487 Data for Financial Metrics work package, and the collected pilot data will be shared via GBIF upon  
488 completion of the project.

489

490 Case study S2. Total Energies share biodiversity data on GBIF



491

492 **Figure S2.** TotalEnergies data publisher metrics displayed on their GBIF publisher page, showcasing key  
 493 performance indicators (KPIs) for company reporting. Metrics include: occurrences per kingdom,  
 494 occurrences per year, occurrences per country or area, and occurrences per basis of record. These metrics  
 495 provide insights into data distribution and can be used to evaluate the company's contribution to  
 496 biodiversity monitoring.

497

498 Companies may publish biodiversity data directly to GBIF by establishing institutional agreements and

499 complying with GBIF's Data Publisher and Data User Agreements. Registration as a data publisher  
500 requires endorsement from a national GBIF node. The process typically involves collaboration with  
501 contractors and field technicians to ensure data and metadata quality. Companies must establish internal  
502 workflows, select and prepare biodiversity data according to the DarwinCore (DwC) standard, define  
503 access restrictions, and publish under a Creative Commons license.

504

505 TotalEnergies is a global energy company operating in 120 countries and became a biodiversity data  
506 publisher on GBIF in 2018 to strengthen their long-term biodiversity impact. The company committed to  
507 sharing biodiversity data collected through in-house environmental impact assessments, including field  
508 surveys in remote and offshore locations, with both the scientific community and the public. By  
509 publishing its data to GBIF, TotalEnergies considers it a valuable contribution to global scientific  
510 research and international conservation efforts. To date, they have published 51,949 occurrences on GBIF  
511 (Figure S2). The company employs a variety of data collection methods, such as sediment, soil, and water  
512 sampling, camera transects, and passive acoustic monitoring and opportunistic observations of marine  
513 megafauna and birds. These data also include hydrocarbons, metals, microbiology, and benthic fauna—  
514 environmental data that allow assessments of habitat sensitivity. TotalEnergies' biodiversity data adheres  
515 to GBIF's quality standards by following the DwC standard and FAIR principles. They have committed  
516 to contributing data annually from a minimum of five projects or sites to GBIF, with regular reporting on  
517 these contributions. By doing so, TotalEnergies enhances its reputation through collaboration with a  
518 reputable organization and gains measurable data from GBIF biodiversity occurrences, which they can  
519 reference in their sustainability reports.

520 **Case study S3. The untapped tropical biodiversity data potential in Sabah,**  
521 **Malaysian Borneo**



522  
523 **Figure S3.** Logos of selected conservation areas in Sabah, Malaysian Borneo (top row), with long-term  
524 biodiversity data not yet available on GBIF, and key research and conservation stakeholders (bottom  
525 row): Yayasan Sabah, SEARRP, Sabah Biodiversity Center, Universiti Malaysia Sabah, Shell, Sime  
526 Darby, IKEA and the SAFE Project. Aerial image: lowland rainforest near Lahad Datu, Sabah photo by  
527 Fevziye Hasan.

528

529 Tropical rainforests around the world are biodiversity hotspots but face intense pressure from logging and  
530 agricultural conversion. Sustainable management and informed decision-making using long term  
531 biodiversity data, are essential for both conservation areas and agricultural landscapes. Sabah, a  
532 Malaysian state on the island of Borneo, contains extensive areas of primary tropical rainforest with high  
533 levels of biodiversity and endemism. Over several decades, a combination of institutional, research, and

534 private-sector initiatives has generated substantial ecological and biodiversity data, much of which  
535 remains unpublished or inaccessible, especially through GBIF.

536 The Yayasan Sabah Group (Sabah Foundation), which manages large forest areas, has fostered long-term  
537 conservation and research partnerships, notably with the Southeast Asia Rainforest Research Partnership  
538 (SEARRP) (see Figure S3). Through these collaborations, they have established permanent research plots,  
539 compiled species inventories, and developed long-term ecological datasets — all providing valuable  
540 insights into biodiversity changes over time.

541 One key example is the Sow-A-Seed project, funded by IKEA and launched in 1998 in Kalabakan. This  
542 large-scale restoration effort aimed to rehabilitate 18,500 hectares of logged and fire-damaged forest and  
543 continued for 25 years. Data from the project demonstrated that tailored restoration techniques, ongoing  
544 biodiversity monitoring, and cooperation with forestry operations helped drive ecosystem recovery. The  
545 restored area was eventually designated as a Class 1 protected forest, highlighting the success of these  
546 efforts (Axelsson et al., 2024).

547 The *Stability of Altered Forest Ecosystems (SAFE) Project*, funded in part by Sime Darby, the world's  
548 largest producer of Certified Sustainable Palm Oil, was one of the largest ecological experiments in the  
549 world, lasting over a decade until funding constraints due to the COVID-19 pandemic. The project  
550 produced more than 150 datasets covering biodiversity and ecosystem processes across various land uses,  
551 including selectively logged forests and oil palm plantations. SAFE's research covers a wide range of  
552 organisms, from soil microbes and invertebrates to vertebrates, and includes detailed studies of ecological  
553 and biogeochemical processes. While these datasets are publicly accessible on Zenodo, they have yet to  
554 be integrated into GBIF.

555 With a huge value of information for key stakeholders in government, conservation, and business, these  
556 two projects have generated hundreds of thousands—possibly millions—of valuable biodiversity data  
557 points. This resource holds strong potential to support efforts to protect biodiversity on one of the world's  
558 most diverse islands.

559 Mobilizing Sabah's biodiversity data from past, current, and future research provides a cost-effective way  
560 to improve data access and inform conservation and sustainable land management.

561 Investing in the mobilisation of these data should go hand in hand with supporting conservation and  
562 sustainable solutions. Together, these efforts can help ensure long-term biodiversity conservation that  
563 benefits people, the environment, and local economies.

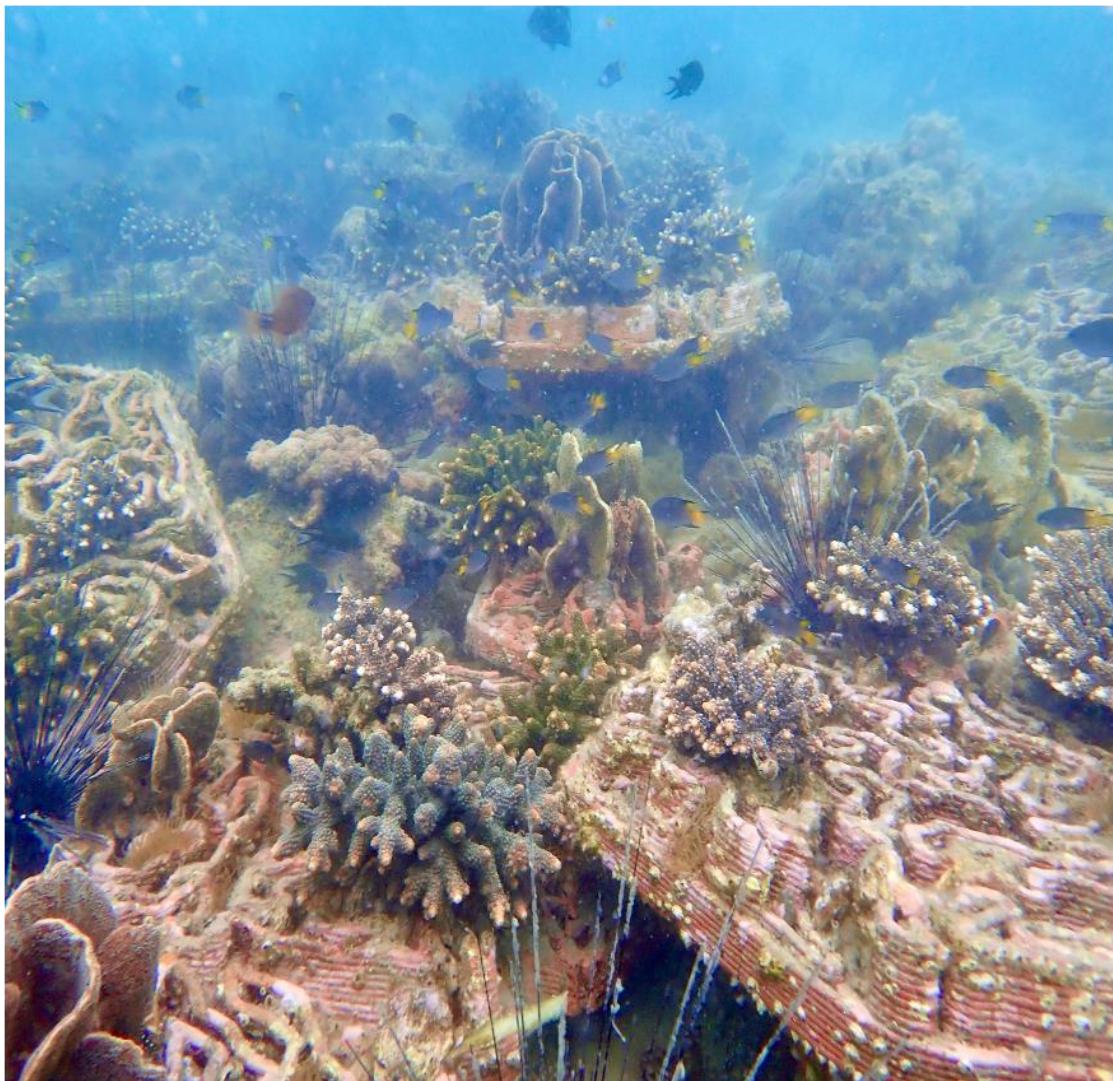
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565

566

567

568 **Case study S4. Biodiversity data-driven nature-tech for coral**  
569 **ecosystem restoration**



570  
571 **Figure S4.** Photo of Archireef 3D-printed ceramic Reef Tiles™, photo by David Baker.  
572 Archireef is a pioneering nature-tech startup that spun off from groundbreaking coral restoration research  
573 at the University of Hong Kong. As a new nature -tech company using GBIF data, Archireef is committed  
574 to sharing biodiversity data through GBIF and is currently exploring viable approaches to achieve this.

575 The company tackles the global crisis of degrading coral reefs by developing innovative, scalable  
576 solutions that combine marine biology with advanced engineering. Their flagship product – 3D-printed  
577 ceramic Reef Tiles™ – represents a major leap forward in restoration technology. These specially  
578 designed tiles stabilize degraded seafloors and mimic natural reef structures when assembled into modular  
579 "hives" underwater. The artificial reefs are then seeded with resilient coral fragments either salvaged from  
580 damaged areas or cultivated in nurseries. This approach has proven remarkably effective, with a Hong  
581 Kong government-funded pilot project achieving an exceptional 91% survival rate across three coral  
582 species over four years, far outperforming traditional restoration methods.

583

584 The company operates on a unique business-to-business model that helps corporations meet their ESG  
585 commitments and CSR goals through meaningful environmental action. Archireef works with clients  
586 across industries including real estate, maritime shipping, and financial services, enabling them to make  
587 measurable contributions to marine conservation. As an early adopter of the Taskforce on Nature-related  
588 Financial Disclosures (TNFD) framework, Archireef is helping shape how businesses report and account  
589 for their biodiversity impact. Their comprehensive "Reef as a Service" offering covers everything from  
590 tile production and installation to three years of intensive monitoring and maintenance at restoration sites.

591

592 To scientifically validate their impact, Archireef employs cutting-edge monitoring techniques including  
593 detailed coral health assessments, 3D photogrammetry to measure reef complexity, and environmental  
594 DNA (eDNA) analysis to track biodiversity changes. By cross-referencing their findings with global  
595 biodiversity databases like GBIF and Barcode of Life Data System (BOLD), the company has  
596 demonstrated that their artificial reefs increase local biodiversity by an impressive 25% compared to  
597 unrestored areas. This data-driven approach not only proves the effectiveness of their solution but also  
598 provides corporate partners with concrete metrics to demonstrate their environmental contributions.  
599 Through this combination of scientific rigor, technological innovation, and sustainable business practices,  
600 Archireef is setting new standards for large-scale marine ecosystem restoration while creating tangible  
601 opportunities for businesses to support ocean conservation.

602

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