

1 Global evidence synthesis on land use change and zoonotic risks

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24 Abstract

25 The COVID-19 pandemic has spotlighted the growing threat of zoonotic diseases, often exacerbated by
26 land-use changes (LUC) such as deforestation and habitat fragmentation. We conducted a systematic
27 literature review (2000–2024) to assess how different types of LUC affect zoonotic disease
28 transmission, summarizing key findings and trends in geographic focus on the vectorshosts/reservoirs
29 and pathogens studied, in addition to identifying research gaps. We also evaluated the potential of
30 restoration interventions to mitigate disease risks. Our analysis shows that LUC such as deforestation
31 and urbanisation often increase transmission risks, particularly for disease transmitted by mosquitos and
32 rodents, while some restoration strategies (e.g. reforestation, wetland conservation) can reduce these
33 risks. However, effects vary by disease group and region. The literature remains geographically biased,
34 with most studies concentrated in wealthier regions despite higher disease burdens in low-income areas.
35 We propose 50 high-priority locations, primarily in Western Africa and Southeast Asia, for future
36 research. Findings are available through an open-access online atlas, which includes supporting case
37 studies and policy briefs to inform One Health-oriented restoration planning.

38 Key words / Key phrases:

39 *Zoonotic disease, Ecosystem restoration, Land-use change, Biodiversity, Public health, Landscape*
40 *degradation*

41 The emergence of SARS-CoV-2 and the subsequent COVID-19 pandemic has brought global attention
42 to the serious global health threat posed by zoonotic diseases^{1,2} — diseases transmitted from animals to
43 humans. However, while COVID-19 captured worldwide attention, zoonotic spillover events have been
44 occurring throughout history with significant implications for global health³. Emerging zoonotic
45 diseases account for over 60% of infectious diseases globally⁴, posing substantial implications to public
46 health, economies and ecosystems⁵. While many have historically resulted in localised outbreaks, the
47 risk of zoonotic pathogen spillover from wildlife is increasing as humans and wildlife share
48 environments more frequently⁶. The drivers behind this include increased human encroachment into
49 natural environments, deforestation, agricultural expansion, and urbanisation. As human activity
50 reshapes natural landscapes, it increases human-wildlife interactions and creates ecological disruptions,
51 such as nutritional stress and altered host immunity, which can elevate the risk of zoonotic spillover^{7,8}.

52 Unprecedented rates of anthropogenic land-use change (LUC) – which we define as the transformation
53 of natural landscapes for urban, agricultural, or other human uses – have replaced natural environments
54 with human-dominated landscapes through urbanisation, agricultural intensification, deforestation, and
55 habitat fragmentation^{9,10}. These changes disrupt ecosystems and alter human–wildlife interactions,
56 creating conditions conducive to the emergence and transmission of zoonotic diseases¹¹. By reducing
57 natural habitats and home ranges, LUC forces many species into closer proximity to humans, increasing
58 the human-wildlife interactions and the likelihood of disease spillover¹². Land-use change, a major
59 driver of emerging infectious diseases, has contributed to over 30% of new disease emergences since
60 1960, including Nipah virus in Malaysia and Ebola spillovers in West and Central Africa^{8,13}. Human-
61 induced habitat loss and environmental degradation reduce biodiversity, which can have profound
62 effects on disease ecology^{14,15}. Biodiversity loss has been linked to increased disease transmission, as it
63 disrupts ecological processes that regulate disease dynamics, such as reductions in higher trophic-level
64 predators, which help control populations of disease-carrying species like rodents¹⁶. The "dilution
65 effect" theory suggests that biodiversity loss can increase disease risk because incompetent host species
66 are often lost first, leaving behind abundant, highly competent reservoirs. This community shift favours
67 efficient pathogen transmission, thereby amplifying outbreak potential^{17,18}.

68 In response to the challenges of accelerating land-use change, biodiversity loss, and increased zoonotic
69 disease emergence, global initiatives like the United Nations Decade of Ecosystem Restoration
70 (<https://www.decadeonrestoration.org/what-decade>), the European Green Deal
71 (https://www.esdn.eu/fileadmin/ESDN_Reports/ESDN_Report_2_2020.pdf), and the UN's 2030
72 Sustainable Development Goals (<https://sdgs.un.org/>) emphasise the critical need to restore degraded
73 landscapes, which we define as areas where ecological integrity, biodiversity, or ecosystem services
74 have been diminished due to human activities such as deforestation, pollution, or land conversion.
75 Restoration – the process of assisting recovery of degraded, damaged, or destroyed land – is increasingly
76 recognised as a strategy to reduce zoonotic disease risks by promoting more resilient and functional
77 ecosystems with higher biodiversity and improved ecological processes^{19,20}. Global restoration
78 initiatives, such as the New York Declaration on Forests, aim to restore 350 million hectares of degraded
79 land to conserve biodiversity and address climate change, food security, and public health challenges,
80 highlighting the scale and ambition of current restoration efforts
81 (<https://forestdeclaration.org/about/new-york-declaration-on-forests/>). Such efforts can improve
82 ecosystem services, including disease regulation, by supporting the maintenance of diverse species and
83 promoting healthier human-nature interactions^{21,22}.

84 While restoration holds promise for mitigating zoonotic risks, it also presents challenges. Increased
85 biodiversity during early recovery stages may temporarily heighten disease risks as ecosystems
86 recover^{23,24}. The assumption that reversing landscape degradation and increasing biodiversity will fully
87 restore disease regulation to its original state is overly simplistic. Restoration often occurs where human
88 interaction with the environment persists or increases, allowing continued exploitation of the landscape,
89 which can heighten the risk of pathogen exposure and spillover. For instance, agroforestry integrating
90 native trees into farmland can raise contact between wildlife, livestock, and humans²⁵, while urban green
91 spaces may create habitats for hosts/reservoirs such as rodents or vectors like ticks, potentially
92 facilitating the spread of diseases such as Bartonella or Lyme disease^{26,27}. Moreover, restored
93 ecosystems do not return to their original state, but can attain a level of complexity comparable to intact

94 habitats, with successful outcomes heavily dependent on the ecological, spatial, and social contexts of
95 implementation^{28,29}.

96 This study provides a comprehensive analysis of existing literature on how landscape changes influence
97 zoonotic disease transmission risks. Drawing on global studies, policy briefs and case studies from 2000
98 to 2024, we summarise the key findings, identifying trends in geographical focus,
99 vectors/host/reservoirs, pathogens, and the themes most explored. This study aims to systematically
100 assess: (i) whether specific geographical regions, vectorshosts/reservoirs, or pathogen groups are over-
101 or under-represented in the literature, (ii) how various pathogen groups respond to different types of
102 landscape changes, and (iii) the extent to which different themes and topics are covered, identifying gaps
103 and offering insights into potential impacts on local populations through socioeconomic factors and
104 human-wildlife interactions. A central contribution of this study is the identification of regions and
105 research areas most in need of further investigation, alongside the development of an interactive
106 evidence atlas. The atlas is a tool designed to make our findings accessible and actionable for
107 researchers, policymakers and public health officials, enabling dynamic exploration of the global
108 landscape of restoration and zoonotic disease transmission research. By bridging ecological,
109 epidemiological, and socioeconomic perspectives, this study aligns with the One Health framework,
110 emphasising the interconnectedness of human, animal, and environmental health in shaping zoonotic
111 disease risks.

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118 Results:

119 Literature Search:

120 We collected 49,753 studies from the Web of Science database and an additional 4,451 from citation and
121 manual searching. After duplicate removal, 52,713 articles identified via keyword search proceeded to
122 screening. Following a comprehensive review by multiple investigators, 173 peer-reviewed studies
123 were included for data extraction alongside 30 policy briefs and 58 case studies from grey literature. Of
124 the 173 peer-reviewed studies included, 39 focused specifically on restoration, while the remainder
125 addressed various forms of LUC, reflecting a marked research imbalance that underscores the need for
126 further investigation into restoration-related disease outcomes. This is detailed in the PRISMA
127 (Preferred Reporting Items for Systematic reviews and Meta-Analyses) flowchart (*Supplementary*
128 *Figure 1*).

129 The number of studies investigating the impact of restoration on zoonotic disease transmission risk has
130 remained consistently low since 2000, with fewer than four studies published annually until 2023, when
131 six studies were published (Figure 1a). In total, only 39 restoration-focused studies were conducted over
132 the last 24 years, averaging 1.2 studies per year from 2000-2019 and increasing to 3.0 studies per year
133 from 2020-2024. In contrast, research on landscape degradation and zoonotic disease transmission has
134 been more robust, with an average of 3.95 studies per year from 2000-2019 that increased substantially
135 to 11.0 studies per year from 2020-2024. This represents a 150% increase in restoration-focused
136 research and a 178% increase in degradation-focused research when comparing pre-2020 and post-2020
137 periods.

138 Twenty percent of the selected studies were either conducted globally or across multiple UN Geoscheme
139 Regions (United Nations Statistics Division. Standard country or area codes for statistical use.
140 <https://unstats.un.org/unsd/methodology/m49/>). The remaining 138 studies were carried out in 93
141 different countries. However, the distribution of studies across these countries was uneven, with 74

142 studies (53.6% of the 138 studies conducted nationally/regionally) concentrated in just four countries:
143 Brazil (28; 20.3%), the USA (26; 18.8%), Kenya (10; 7.2%), and Malaysia (10; 7.2%) (Figure 1b).

144 Figure 1.

145 Study Themes, Vectors/Hosts/Reservoirs and Pathogens:

146 Almost 75 percent of the research comprised field-based studies, while 36 studies focused on other
147 themes such as experimental work, hypothesis/theoretical frameworks, reviews, and descriptive
148 analyses. Field studies were the most common type of study across the majority of regions, excluding
149 Northern Africa, Western Africa and Western Asia (Figure 2a).

150 Twelve common disease vectors/hosts/reservoirs were identified across the studies. Mosquitoes (46
151 studies; 35%), rodents (25 studies; 19%), and ticks (18 studies; 13.6%) were the most frequently
152 studied, comprising 68% of the total. Ticks were the most studied in Northern America, while
153 mosquitoes were the primary focus in almost half of the other regions (East Africa, Middle Africa,
154 South-eastern Asia, South America & Oceania). Bats accounted for a significant portion of the studies
155 conducted in Middle Africa (33%), Southern Asia (33%) and Oceania (33%) (Figure 2b).

156 Mosquito-borne diseases were the most studied across the majority of regions, with exceptions in
157 Europe and Northern America, where tick-borne and rodent-borne diseases were more prevalent, and
158 Southern Asia, where zoonoses (bat- and rodent-borne) were the primary focus (Figure 2b). As shown in
159 Figure 2c. viruses were the most studied pathogens overall with 31%, though this varied by region.
160 Bacteria were the primary focus in Northern America, making up 66% of the studies, while protozoan
161 diseases were most studied in Eastern Africa (53%), South America (52%), Central America (40%) and
162 South-eastern Asia (40%) (Figure 2c). Detailed information on the specific diseases, pathogens, and
163 vectors/hosts/reservoirs, for each of the 173 studies is provided in *Supplementary Table 8*.

164 Figure 2.

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166 Hierarchical Cluster Analysis:

167 To compare how different types of landscape changes influence disease transmission across disease
168 groups, we calculated a directional consensus standardised score from all relevant studies and visualised
169 the results using a hierarchical cluster heatmap (Figure 3). The x-axis includes both land-use change
170 drivers (e.g., deforestation, fragmentation, urbanisation), restoration interventions (e.g., reforestation,
171 wetland restoration) and biodiversity–disease relationships such as “Biodiversity loss” (studies that
172 reported declines in species richness/abundance) and “Dilution effect” (studies that explicitly tested
173 whether higher biodiversity reduced transmission risk). For instance, mosquito-borne and bat-borne
174 diseases clustered with fragmentation and deforestation, indicating a consistent association across
175 studies. However, given the heterogeneity of study designs and contexts, these clusters should be
176 interpreted with caution and seen as suggestive rather than definitive. Directional consensus
177 standardised scores, representing the effect of landscape changes on disease transmission risk to
178 humans, ranged from -3.5 (indicating an increase in risk) to 3.5 (indicating a decrease in risk). Most
179 zoonoses, especially those transmitted by rodents and bats, showed negligible or increased transmission
180 risk with landscape degradation. Rodent- and bat-borne diseases were affected by urbanisation
181 (standardised score: -2.75 and -1.75, respectively), land use alterations (standardised score: -0.75 both),
182 and deforestation (-0.75 and -2.75, respectively), while bat-borne diseases also experienced significant
183 changes with habitat fragmentation (-2.75), and rodent-borne diseases with biodiversity loss, defined as
184 a broader decline in species richness and ecosystem diversity (-0.75). A similar pattern was observed for
185 vector-borne diseases, with the transmission risk to humans from mosquito-borne diseases being
186 particularly influenced by deforestation (-3.5), fragmentation (-3.29), anthropization (-2.75), and
187 urbanisation (-1.75). In contrast, tick-borne diseases exhibited a decrease in transmission risk with
188 urbanisation (+2.26). The raw aggregated effect values can be found in *Supplementary Table 6*.

189 Restoration interventions showed mixed effects on disease transmission risk. Support for the dilution
190 effect was strongest in tick-borne, mosquito-borne, and other vector-borne diseases (mean score:
191 +2.26), with consistent evidence also for rodent-borne diseases (+3.26). Mosquito-borne disease risk

192 improved through wetland restoration (+3.47) and broad conservation efforts aimed at maintaining or
193 enhancing existing ecosystems and wildlife populations (+1.26), reflecting a consistent positive effect
194 across a limited number of restoration studies. Conversely, reforestation, which restores tree cover in
195 previously deforested areas, was associated with an increased risk of zoonoses and tick-borne diseases
196 (-0.75), potentially due to altered breeding habitats. Similarly, species removal, referring to deliberate
197 reduction or eradication of certain species, often invasive or disease-host species, was linked to
198 heightened risk of rodent-borne diseases (-0.75), but a reduced transmission risk for tick-borne diseases
199 (+2.26). Afforestation, the planting of trees in areas that were not previously forested, and restoration
200 services, ecosystem management interventions aimed at rehabilitating degraded environments, had no
201 significant mean effect on disease transmission risk.

202 Figure 3.

203 The reviewed studies showed varying degrees of attention to different themes. Only 18.5% of studies
204 discussed socioeconomic factors. Similarly, evidence for human-wildlife interactions was explored in
205 34.7% of the publications. Beyond these thematic discussions, a sizeable number of studies (64.7%)
206 included recommendations or proposed solutions, reflecting a strong emphasis on providing actionable
207 insights to their conclusions. Additionally, limitations and challenges were acknowledged in 75.7% of
208 the studies (*Supplementary Figure 2*).

209 Necessity for Further Research:

210 To identify priority areas for future research, we modelled the current geographical distribution of
211 evidence using a hierarchical Bayesian spatial model and quantified the Need for Further Research
212 (NFR) as the product of local uncertainty and population density (Figure 4). This approach highlights
213 areas where new studies would yield the greatest informational value for understanding links between
214 landscape change and disease emergence. We selected 50 locations that would benefit from further
215 research on disease dynamics in areas undergoing landscape changes. Amongst the 50 locations, 84% of
216 sites were equally distributed between Western and Middle Africa (21 locations) and South-Eastern Asia

217 (21 locations). We identified 12% (6 locations) in Central and South America and only one site each in
218 Oceania and Northern America.

219 Figure 4.

220 Case Studies and Policy Reports:

221 Among the 58 landscape restoration case studies, 16 (28%) were conducted before 2000, 24 (41%)
222 between 2000 and 2010, and 18 (31%) from 2011 onward. Most studies were concentrated in Middle
223 Africa (29%), followed by Eastern Africa (14%) and Central America (10%). South America and South-
224 Eastern Asia contributed 7% each, while Europe, North America, Western Africa, and Southern Asia had
225 5% each. Minimal representation was noted in Central Asia, Northern Africa, and Western Asia (2–3%),
226 indicating a need for broader geographic coverage. Biodiversity conservation and natural resource
227 management (38%), agroforestry (21%), and afforestation (9%) were prominent themes, with smaller
228 efforts in wetland restoration (5%) and rewilding (3%). Unique approaches included controlled under-
229 burning and water supply management. However, investigating of zoonotic risk and infectious diseases
230 were absent as explicit objectives.

231 Among the 30 national and organisational policy briefs reviewed, 43% were published before 2020, and
232 57% afterward. Most were concentrated in Europe (33%) and global frameworks (30%), with limited
233 representation from regions experiencing higher degradation and zoonotic risk. Recommendations
234 emphasised community engagement (33%), governance for resilience (27%), gender equity (23%),
235 capacity-building (27%), and integrating One Health approaches (13%). Targeted actions, such as
236 regulating wildlife trade (7%) and promoting sustainable land use (27%), aim to address health and
237 environmental challenges, with a focus on long-term funding, adaptive management, and cross-sectoral
238 collaboration.

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241 *Online Evidence Atlas:*

242 Our online evidence atlas (<https://bradduthie.shinyapps.io/atlas/>) provides an interactive visualisation of
243 the geographic distribution of scientific literature, case reports, and policy briefs examining how
244 restoration and landscape degradation influence disease transmission. The tool enables users to explore
245 spatial and thematic patterns, identify research gaps, and access linked references through clickable map
246 markers. By integrating scientific and grey literature, the atlas supports researchers in identifying under-
247 studied regions and themes, while offering policymakers a means to visualise where evidence for
248 restoration–health linkages is strongest or lacking. A schematic of the atlas layout is presented in Figure
249 5, illustrating its main components and interactive features.

250 Further details on the user interface, including instructions on navigating layers, filtering studies, and
251 accessing regional summaries, are provided in the Supplementary Information (Figure 3).

252 Figure 5.

253 Discussion:

254 While research on landscape change and zoonotic spillover risk has increased since COVID-19, a key
255 gap remains: studies investigating restoration impacts are limited. This imbalance—far more studies
256 address degradation than restoration—underscores the need for empirical research explicitly linking
257 landscape restoration to disease transmission risk, particularly as global restoration efforts accelerate
258 without adequate disease monitoring frameworks in place. Our findings indicate a predominant focus on
259 mosquitoes and mosquito-borne diseases, which were the most frequently studied globally. However,
260 the attention given to specific pathogens, viruses, bacteria, and protozoans, varied across regions. Our
261 findings highlight a clear geographic and economic bias in the global research landscape, which
262 becomes even more evident when examining the disproportionate distribution of studies. We also
263 observed that different types of land-use change, including restoration activities, trigger diverse
264 responses depending on the disease studied, underscoring the need for more context-specific research.
265 Recent work by Mahon et al. (2024)³⁰ complements our synthesis by providing a quantitative baseline
266 and strengthening the contextualisation of biodiversity loss and land-use change as key drivers of
267 zoonotic risk.

268 Almost half (43%) of all studies were concentrated in four countries - USA, Brazil, Kenya, and Malaysia
269 - while 102 countries had none. Nearly 65% of studies occurred in four UN Geoscheme regions (Europe,
270 Northern America, South America, and South-Eastern Asia), with a clear dominance of high-income
271 countries, particularly the USA. According to the World Bank's 2024 classification, all but one study
272 from Europe and Northern America were conducted in high-income countries. The exception was a
273 study including four upper middle-income countries: Albania, Bosnia and Herzegovina, North
274 Macedonia, and Ukraine. Overall, nearly 80% of all studies originated from high- or upper-middle-
275 income countries, leaving less than 20% from lower-middle- or low-income regions. Geographic and
276 socio-economic biases in research effort are well documented, with high-income countries
277 overrepresented in ecological and health studies^{31,32}. Such imbalances risk overlooking key regions
278 where restoration-driven disease dynamics may be most relevant, underscoring the need for a more
279 globally inclusive research agenda. Grey literature searches reveal a major policy gap: restoration

280 policies rarely address potential impacts on disease transmission, especially in lower-income regions
281 with the highest disease burdens. Addressing these imbalances will require increased investment in
282 research and policy development tailored to the specific challenges of these regions and closer
283 integration of restoration, health, and community policies to ensure interventions are both socially and
284 epidemiologically sustainable³³.

285 Despite these patterns, few studies address socioeconomic factors; those that do emphasise a strong link
286 between lower GDP and higher disease risks. Lower-income countries tend to have higher disease
287 burdens, such as those caused by mosquito-borne diseases like malaria and dengue, or protozoan
288 infections like Chagas disease, yet these geographic areas are clearly underrepresented in the
289 research^{34,35}. Addressing this gap requires a One Health approach³⁶, integrating human, animal, and
290 environmental health, because interventions in LMICs must consider complex social, ecological, and
291 health factors driving disease risk, which may differ from patterns observed in more frequently studied
292 high-income regions.

293 The distribution of the identified NFR locations underscores the urgent need for more focused research
294 in lower-income regions, particularly in Western Africa, Middle Africa and South-eastern Asia, where
295 landscape restoration may have the greatest potential to influence public health outcomes, positively or
296 negatively, depending on intervention type, implementation context, and disease ecology. These regions
297 face higher disease burdens but remain critically understudied. Expanding research efforts in these areas
298 with large-scale, longitudinal studies^{37,38} could provide invaluable insights into how landscape changes
299 influence disease spillover, ultimately helping to inform more effective restoration strategies and public
300 health policies.

301 Socioeconomic disparities also influence human-wildlife interactions, a key factor in the transmission
302 zoonotic disease and vector-borne disease with zoonotic cycles³⁹. Rural communities engaged in
303 agriculture or resource extraction face elevated vulnerability to these diseases during periods of
304 landscape transformation for livelihood support^{40,41}. The process of converting natural landscapes into
305 agricultural or extraction areas inadvertently increases exposure to disease through more frequent
306 contact with vectors and wildlife hosts/reservoirs, whether through land degradation or certain types of

307 restoration interventions⁴². Additionally, reliance on bushmeat and illegal wildlife trade in low-income
308 regions further increases transmission risks⁴³. Despite the significant role of human–wildlife
309 interactions in zoonotic disease risk, only one-third of the studies explicitly acknowledged these factors,
310 partly reflecting that nearly half focused on vector-borne transmission where wildlife host interactions
311 are less directly addressed. Communities in these regions face a dual burden: the environmental impacts
312 of landscape changes and the economic costs of disease management, with limited access to healthcare
313 and other resources⁴⁴. These constraints exacerbate the risk of zoonotic and vector-borne diseases by
314 limiting capacity for disease prevention, surveillance, and response, particularly in low-income regions
315 where economic disparities affect long-term public health infrastructure and community
316 development^{45,46}.

317 Research understandably concentrates on high-burden disease hosts and vectors, such as mosquitoes,
318 rodents, and ticks. Mosquito-borne diseases including malaria, dengue fever, and lymphatic filariasis,
319 remain a significant burden globally, with an estimated 212 million malaria cases annually⁴⁷. In contrast,
320 Europe and Northern America see a higher prevalence of tick-borne diseases like Lyme disease⁴⁸, while
321 rodent-borne zoonoses linked to rapid urbanisation are the primary focus in regions like Central and
322 Southern Asia⁴⁹. However, this focus can inadvertently overshadow less-studied diseases and contexts,
323 particularly in regions or ecosystems undergoing active landscape restoration underscoring the need for
324 more geographically balanced and context-specific research. While there is extensive research in
325 Europe and Northern America on mosquito vectors and climate change impacts, relatively few studies
326 have examined how land-use change, or restoration may influence mosquito-borne disease transmission
327 in these regions⁵⁰. As rising temperatures enable vector range expansion to higher latitudes, filling this
328 knowledge gap is important to anticipate emerging risks.

329 Our cluster analysis shows varied effects of landscape changes on zoonotic diseases. Rodent- and bat-
330 borne diseases often increase with landscape degradation, while tick-borne diseases tend to decline with
331 urbanisation and landscape alteration. Restoration efforts generally reduce transmission risks for vector-
332 borne diseases, but exceptions exist—such as increased mosquito-borne disease risk during
333 afforestation⁵¹. Evidence for a dilution effect was strongest for vector-borne zoonoses (ticks,

334 mosquitoes), with more variable results for directly transmitted zoonoses⁵². Similarly, there was a
335 decrease in transmission risk for rodent-borne zoonoses with restoration efforts, but an increase in bat-
336 borne disease transmission⁵³.

337 The review of policy and case studies highlights the need for inclusive governance frameworks that
338 engage local communities, indigenous groups, and stakeholders to align restoration with environmental
339 and socio-economic goals^{54,55}. Strengthening community capacity, especially for women and
340 marginalised groups, is crucial for sustainable land management because it enhances local ownership,
341 improves long-term project viability, and ensures restoration efforts are adapted to local contexts⁵⁶.
342 Communities with greater capacity are better equipped to maintain restored landscapes beyond initial
343 implementation phases and can develop contextually appropriate solutions that address both ecological
344 and social needs⁵⁷. Integrating One Health solutions addresses biodiversity loss, land degradation, and
345 emerging diseases by bridging environmental, human, and animal health³⁶. The One Health Joint Plan of
346 Action (2022–2026), developed by the Quadripartite Alliance (WHO, FAO, WOAH, and UNEP),
347 provides a framework for coordinated responses to health threats at the human-animal-environment
348 interface⁵⁵. Landscape restoration should prioritise ecological resilience by combining scientific
349 knowledge, local expertise, and socio-economic considerations such as community livelihoods, land
350 tenure systems, and public health infrastructure. Engaging stakeholders - local communities, public
351 health officials, conservation agencies, and policymakers - is essential to ensure context-specific,
352 sustainable implementation⁵⁶.

353 This study has some limitations⁵⁸. Relying solely on Web of Science may have excluded studies from
354 other databases or non-English sources. Although grey literature was included, studies addressing
355 ecological and health outcomes separately may have been missed if not integrated in a single source. We
356 also excluded non-zoonotic animal diseases, which could offer insights into ecological mechanisms like
357 dilution or amplification. The included literature is geographically and socioeconomically biased, with a
358 focus on high-income countries, limiting applicability to low-income regions where zoonotic risks and
359 landscape dynamics differ. Additionally, the synthesis spans diverse definitions of “restoration” and

360 “degradation,” and heterogeneity in study design requires cautious interpretation. Still, by classifying
361 landscape change into subtypes, we enhance interpretability and uncover disease-specific patterns.
362 These findings highlight the need for more region-specific, interdisciplinary research to inform One
363 Health approaches to landscape management and zoonotic disease mitigation. The spatial scale of
364 included studies varied widely, from local site-level assessments to national-scale analyses, which may
365 influence disease outcomes in scale-dependent ways. Future work should account for this heterogeneity
366 to better interpret LUC disease relationships.

367 As the UN’s Decade for Restoration drives large-scale environmental restoration efforts, it is crucial to
368 understand how these initiatives impact disease spillover risks. While our review found a growing
369 number of studies examining the effects of land-use change on zoonotic risk, research specifically
370 evaluating restoration efforts remains limited, with only 39 studies identified over a 24-year period. This
371 highlights a critical evidence gap that needs to be addressed to ensure that restoration projects do not
372 inadvertently increase health risks. The RESTOREID project is an important initiative in this context, as
373 it aims to examine the intricate links between restoration efforts and infectious disease dynamics. The
374 development of an online interactive evidence atlas as part of this study offers a valuable tool for
375 stakeholders to explore and interpret the current knowledge base, facilitating more informed, One
376 Health-oriented restoration planning and future research targeting both ecological and public health
377 outcomes.

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384 Methods:

385 We employed a multistep approach to review and model the evidence on the relationship between
386 landscape alterations and infectious disease emergence. This project builds on the Eklipse Biodiversity
387 and Pandemics evidence project³⁸, funded by the European Commission Directorate-General for
388 Environment (DG Env), further advancing the understanding of the links between biodiversity, land use,
389 and zoonotic disease risk. For this study, we systematically mapped the evidence based on a recent
390 report⁵⁹ through the following steps: conducting a systematic literature search, screening and reviewing
391 articles, extracting data, synthesising and mapping the results, and modelling geographical areas in need
392 of further research⁶⁰.

393 Systematic literature search:

394 To address the limited evidence on landscape restoration and zoonotic disease transmission, we included
395 studies on both landscape degradation and restoration, allowing for a comprehensive assessment of how
396 different types of land-use change influence disease emergence. A comprehensive literature search was
397 conducted using the Web of Science search engine in April 2024. The search was performed using the
398 specific keywords listed in *Supplementary Table 2*. These keywords were derived from the previous
399 work done in the Eklipse Biodiversity and Pandemics report³⁸, a knowledge synthesis initiative
400 commissioned by several European Commission Directorates and international partners to identify
401 interdisciplinary research and action priorities for a strategic research agenda addressing the links
402 between biodiversity, ecosystem change, and pandemic risk. This ensured alignment with established
403 research in the field. Keywords were refined iteratively to capture a broad spectrum of relevant literature
404 while maintaining specificity to the study objectives. To supplement the search, forward and backward
405 citation chasing was conducted using the CitationChaser Shiny app
406 (<https://estech.shinyapps.io/citationchaser/>) to capture any additional relevant literature. Although the
407 search was limited to a single engine, the extensive results from both the initial and supplementary
408 searches provided thorough coverage due to the relatively small number of articles meeting the
409 inclusion criteria. Additionally, we reviewed grey literature related to current and planned restoration

410 projects. However, we opted not to include grey literature from national government policies or
411 organisational reports in this synthesis, as we found very few that specifically addressed zoonotic
412 diseases.

413 *Screening and review of articles:*

414 We implemented a single stage abstract screening process with two reviewers. The online software
415 Rayyan.ai (<https://www.rayyan.ai/>) facilitated the screening and review of the obtained publications⁶¹.
416 Each publication was classified as "Include," "Exclude," or "Maybe" based on the title, abstract, and
417 predefined inclusion criteria. For "Maybe" entries, both reviewers jointly screened the full text to reach
418 a consensus. Any studies that discussed the impact of landscape restoration, degradation or land use
419 changes on zoonotic disease transmission risks were included. Types of studies included: reviews,
420 descriptive studies, theoretical studies, experimental studies, policy frameworks, perspectives &
421 opinion pieces. Studies were classified as field studies when real-world evidence was collected through
422 environmental sampling or field surveys, and as theoretical/modelling studies when based solely on
423 passive data sources or hypothesised relationships without active data collection. Any studies that
424 focussed on diseases with no transmission potential to humans and any pharmaceutical and clinical trials
425 were excluded.

426 *Data extraction:*

427 Data were extracted from the articles using a predefined template (*Supplementary Table 3*). Along with
428 the article metadata, the additional information extracted included geographical location of the study,
429 study scale, thematic focus, transmission type, pathogen group, and species involved were recorded. In
430 this study we defined zoonotic diseases as those transmitted directly from animals (wildlife, livestock,
431 or companion animals) to humans (e.g., hantaviruses, Ebola Virus Disease). Vector-borne diseases were
432 defined as those transmitted by arthropods such as mosquitoes, ticks, or fleas, and may include zoonotic
433 (e.g., Lyme disease) or non-zoonotic diseases (e.g., dengue, some malaria species). Diseases with dual
434 zoonotic-anthroponotic cycles (e.g., dengue, yellow fever, Zika) and all malaria species, including

435 *Plasmodium knowlesi*, are classified as vector-borne to ensure consistency. Review papers covering
436 multiple diseases with heterogeneous pathogen or transmission pathways were coded as a single entry
437 under the category most representative of their primary thematic focus such as multiple zoonoses or
438 wildlife diseases, to avoid misclassification and over-inflation of disease counts. For transparency, we
439 also explicitly indicate in *Supplementary Table 7* which diseases were classified as zoonotic, vector-
440 borne or other according to these definitions. The key data included the type of restoration,
441 direct/indirect impacts of landscape restoration and degradation on disease and human-wildlife
442 interactions and extracted evidence on disease outbreaks and socioeconomic considerations. The
443 category "biodiversity loss" included studies reporting reduced biodiversity (quantitatively via richness,
444 abundance, or diversity indices, or qualitatively via species/habitat loss) linked to disease outcomes.
445 "Dilution effect" was used only for studies that explicitly examined epidemiological consequences of
446 biodiversity change and referred to the mechanism as a dilution effect. Validation criteria assessed bias,
447 imprecision, inconsistency, reporting bias, and the quality of evidence, along with reviewer
448 confidence⁶². In addition to peer-reviewed studies, we also reviewed grey literature, including national
449 government policies, case studies, and policy reports. We applied a systematic search strategy using
450 predefined keywords related to landscape restoration (e.g., ecosystem restoration, reforestation, land
451 rehabilitation) and zoonotic diseases (e.g., disease spillover, emerging infectious diseases, One Health)
452 to identify relevant studies and projects in Google Scholar and restoration project databases such as the
453 Global Restoration Projects database, FAO's Forest and Landscape Restoration Mechanism, and the UN
454 Decade on Ecosystem Restoration platform. For these documents, we extracted key results, lessons
455 learned from the case studies, and policy recommendations.

456 Data synthesis and visualisation:

457 The data were synthesised into multiple evidence maps, categorising studies according to the UN
458 Geoscheme Regions (<https://unstats.un.org/unsd/methodology/m49/#ftn13>). Regional classifications
459 follow the United Nations Geoscheme; thus, "Middle Africa" is used to denote the region more
460 commonly referred to as "Central Africa". These maps detailed the number of studies and different

461 research themes, vector species, transmission types, and pathogen groups. To capture variation within
462 broad categories, we classified landscape changes into subtypes (e.g., fragmentation, reforestation,
463 wetland restoration) and assessed their directional effects on disease transmission risk across different
464 zoonotic groups. Visualisations were created using the *ggplot2* package in R Statistical Software
465 (v4.4.1; R Core Team, 2024). To illustrate the impact of various landscape changes on disease
466 transmission, a hierarchical cluster heatmap was generated using the *ComplexHeatmap* package⁶³. This
467 approach allowed us to summarise patterns across different disease groups and landscape change types.
468 Each study that reported a statistically significant relationship was scored as -1 if the landscape change
469 was associated with an increase in disease transmission risk, +1 if associated with a decrease in risk, and
470 0 if no statistically significant effect was found. Only studies presenting empirical evidence with defined
471 methods and statistical analyses were included in the scoring of landscape change effects. Theoretical or
472 conceptual articles without original data were excluded from the quantitative synthesis and heatmap
473 analysis to ensure that only well-supported empirical findings contributed to effect scoring. These
474 values were aggregated by disease type and landscape change category, and a mean score was calculated
475 for each combination, then standardised. The resulting standardised scores reflect the relative direction
476 and consistency of effect across studies, rather than formal statistical effect sizes. For example, a score
477 of -3.5 indicates strong and consistent evidence across multiple studies for increased transmission risk,
478 while +3.5 reflects consistent evidence of decreased risk. The heatmap was then clustered hierarchically
479 to highlight similarities and differences in how various disease types respond to different landscape
480 changes, both degradation and restoration. In this review, “risk of transmission” refers specifically to the
481 potential for zoonotic pathogens to spill over into human populations due to landscape changes. While
482 some studies assessed pathogen prevalence in animal hosts, only those discussing implications for
483 human health, directly or indirectly, were included in the final analysis.

484 Need for further research:

485 We extracted 271 geographical locations from 90 of the included studies and reports and georeferenced
486 location latitude and longitude coordinates in decimal degrees. We generated pseudo-absence points

stratified by human population density to avoid bias from uninhabited areas and ensure background sampling reflected the spatial distribution of human presence. We then extracted data from eight environmental and demographic covariates selected based on their relevance to landscape change and zoonotic disease emergence: mean temperature and total precipitation (climate variables influencing vector and host distributions), elevation (as a proxy for habitat variability), global human modification index (capturing anthropogenic pressure), mammal richness index (as a proxy for host biodiversity), forest cover loss (proxy for deforestation indicating habitat degradation), accessibility to the nearest city (reflecting human mobility and contact potential), and human population density (as a driver of landscape pressure and disease risk). These are described in further detail in *Supplementary Table 4*. We projected the covariate data in the World Geodetic System version 84 (GISGeography, <https://gisgeography.com>) and resampled them by using a pixel resolution of 5 minutes of arc (0.08333 degrees), equating to $\approx 10 \times 10$ km resolution. We then modelled the distribution of current evidence using a binomial logistic regression in a hierarchical Bayesian framework. The binary outcome variable was the presence (1) or absence (0) from 1000 generated pseudo-absence points of georeferenced evidence (study location) within each pixel. We used a binomial logistic regression model in a hierarchical Bayesian framework with an intrinsic conditional autoregressive model⁶⁴ to model the distribution of current evidence. To account for spatial autocorrelation between neighbouring pixels, which could otherwise bias the model, we included an intrinsic conditional autoregressive (iCAR) component. This spatial model structure improves accuracy by borrowing strength from nearby pixels with similar values. We calculated the uncertainty using the standard deviation of each pixel. We then identified 50 locations that would benefit from further research to maximise information gained from landscape alteration and disease emergence. We quantified the necessity for further research (NFR), a previously described measure⁶⁵, as the product of the uncertainty of the geographical distribution of current evidence and human population density. Population density was included as a covariate in the Bayesian model to capture its role as a driver of disease emergence, and as a weighting factor in the NFR index to prioritise areas with higher potential human exposure. These two uses serve distinct purposes, epidemiological modelling and research prioritisation. We identified and placed a hypothetical research study on the pixel with the highest NFR value, then gradually reduced NFR around this first hypothetical

515 study by a 50-km radius, applying spatial discounting within a 50 km radius around each selected point
516 to avoid overlap. We used the same procedure to add consecutive research by using the pixels with the
517 highest NFR until we identified 50 locations that could benefit from additional research. This approach
518 selects discrete pixel locations iteratively and applies a 50 km spatial discounting radius, producing
519 point-based rather than continuous outputs. The mapped priorities therefore represent specific modelled
520 locations, not broader spatial regions. The number 50 was chosen to balance global representation with
521 visual interpretability.

522 *Development of Online Evidence Atlas:*

523 To complement our systematic map, we developed an online evidence atlas hosted on the *Shinyapps*
524 platform⁶⁶. This atlas serves as a dynamic repository for the literature identified in our review, aimed at
525 facilitating future searches of studies related to the effects of restoration and landscape degradation on
526 zoonotic disease transmission, like the interactive visualising tool *EviAtlas*⁶⁶. The online platform will
527 allow users to access summarised data from our review, including geographic distributions, vector and
528 pathogen types, and thematic focuses, with options for downloading data for further analysis. By
529 making this information openly accessible, we aim to foster interdisciplinary collaboration, support data
530 sharing, and assist in bridging research gaps. The evidence atlas offers a centralised source for
531 researchers, governments, and policymakers to explore real-time data and better understand current
532 trends in LUC research, identifying where additional focus may be needed based on spatial or thematic
533 gaps.

534 The atlas was developed using R, with a combination of CSS (Cascading Style Sheets) for the user
535 interface (UI) design and utilises the *shiny* package to convert R code into HTML, creating an
536 interactive online platform accessible from all major web browsers⁶⁷. Once completed, the app was
537 deployed on Shiny's servers at <https://www.shinyapps.io>, ensuring global accessibility and ease of use.
538 To visually present the spatial distribution of studies, we used the R package *leaflet*, which allows for an
539 interactive map display of scientific literature locations, country case reports, policy briefs, and the 50
540 identified “Need for Research” locations. These points of interest are overlaid on a map, providing a

541 comprehensive view of where research has been conducted and where gaps remain. Users can interact
542 with the map to view detailed metadata about each study, including vector, disease, and theme
543 information, categorised by UN Geoscheme regions for consistency with the review. Additionally, the
544 platform hosts data summaries from the review, including proportional breakdowns of studied vectors,
545 disease types, and research themes. By structuring these summaries by geographic region, the atlas
546 enables users to quickly identify patterns and assess gaps in research across different regions. We are
547 exploring options for open contributions to the Online Evidence Atlas, enabling future inclusion of peer-
548 reviewed and practitioner-submitted case studies with appropriate quality controls and metadata
549 standards.

550 Data availability:

551 The data that supports the findings in this study can be found in *Supplementary tables 6 and 8*.

552 Code availability:

553 The code used for the analysis in this study can be found at
554 https://github.com/adamjohnfell/RestoreID_Project.

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560 Author Contributions:

561 All authors contributed to the conceptualisation of the research. AF and SJ collected the data and
562 conducted the analysis. AF and SJ lead the writing for the draft manuscript. All authors contributed
563 critically to the revisions of the manuscript and gave final approval for publication.

564

565 Competing Interests Statement:

566 The authors declare no competing interests.

567 Figure legends/captions and brief titles (for main text figures):

568 *Temporal & geographic distributions of published studies*

569 Figure 1. (A) Annual count of published studies from 2000 to 2024 investigating the impacts of
570 landscape degradation or restoration on zoonotic disease transmission risks. (B) The geographic
571 distribution of 138 studies focusing on the effects of landscape degradation and restoration on zoonotic
572 disease transmission risks during the same period. Colours represent the percentage of studies published
573 in each country.

574 *Geographic distributions of study themes, vectorshosts/reservoirs and pathogens*

575 Figure 2. Fifteen UN Geoscheme Regions (represented by different colours; see
576 <https://unstats.un.org/unsd/methodology/m49/> for regional definitions), with corresponding pie charts
577 illustrating the proportions of various 2a) research themes, 2b) vectorshosts/reservoirs species and 2c)
578 pathogen groups represented in a total of 138 studies. Some regions do not have corresponding pie
579 charts as these studies lacked that information. Mosquito, tick and bat silhouettes obtained from
580 PhyloPic (<https://www.phylopic.org/>).

581 *Effects of land-use change and biodiversity–disease relationships*

582 Figure 3. Heatmap showing the size and direction of reported effects of land-use change and
583 biodiversity–disease relationships on disease transmission risk across different disease groups (Vector
584 borne diseases in blue, Zoonoses in brown, General in purple, Neglected diseases in green, Wildlife
585 diseases in pink & Tuberculosis in orange). Each cell represents the mean directionality derived from
586 studies reporting statistically significant results: blue with downward arrow indicates a reported positive
587 association (e.g., decreased disease risk), red with upward arrow indicates a negative association (e.g.,
588 increased risk), and white indicates a reported no significant effect with a bidirectional marker.

589 *Priority sites with necessity for further research*

590 Figure 4. Fifteen UN Geoscheme Regions (represented by different colours; see
591 <https://unstats.un.org/unsd/methodology/m49/> for regional definitions) with the locations of 50 priority
592 sites with necessity for further research (NFR) on landscape alterations and disease emergence.

593 *Online Evidence atlas*

594 Figure 5. Overview of the Online Evidence Atlas (<https://bradduthie.shinyapps.io/atlas/>) interface
595 displaying the interactive map and filtering panels.

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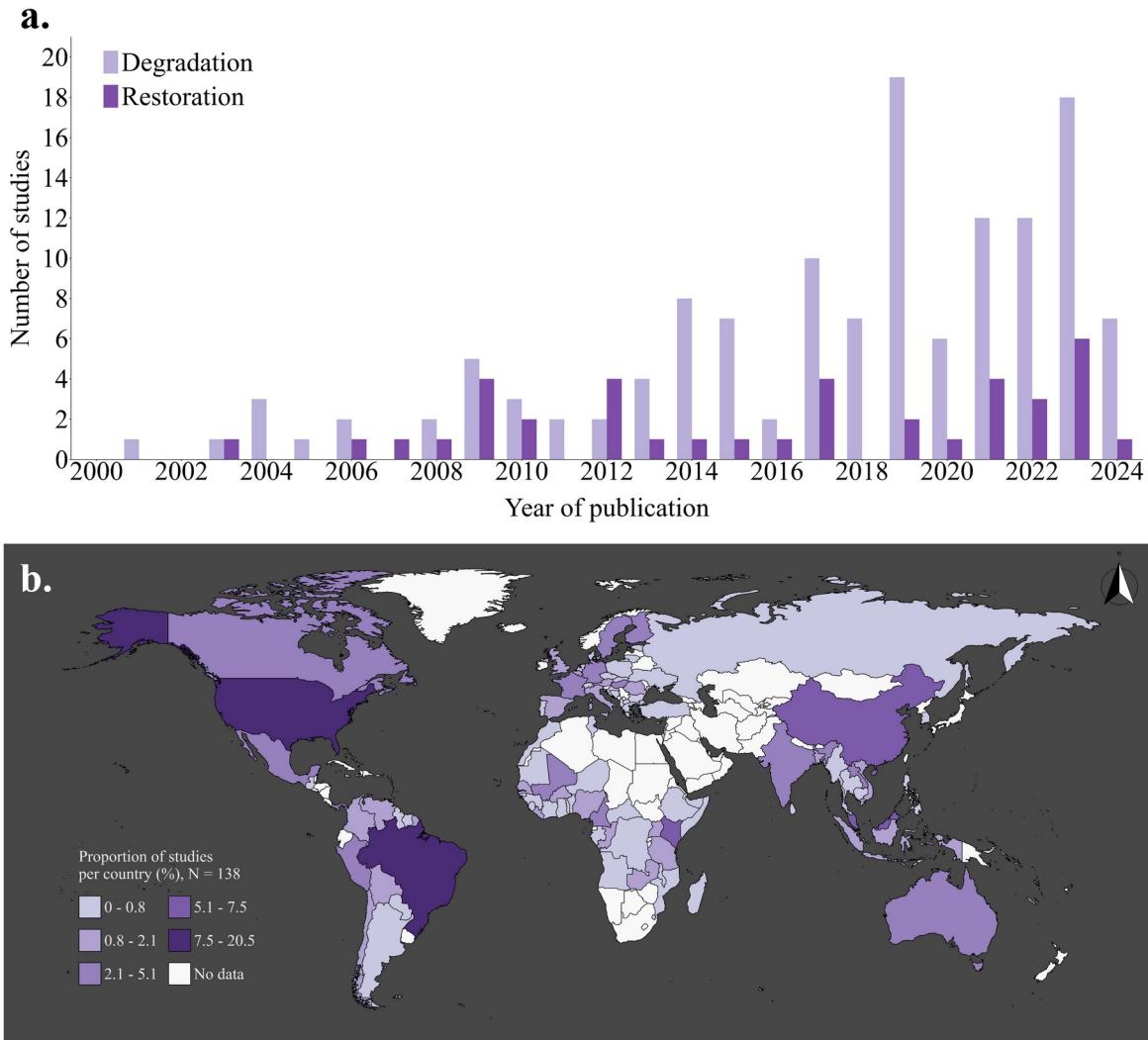
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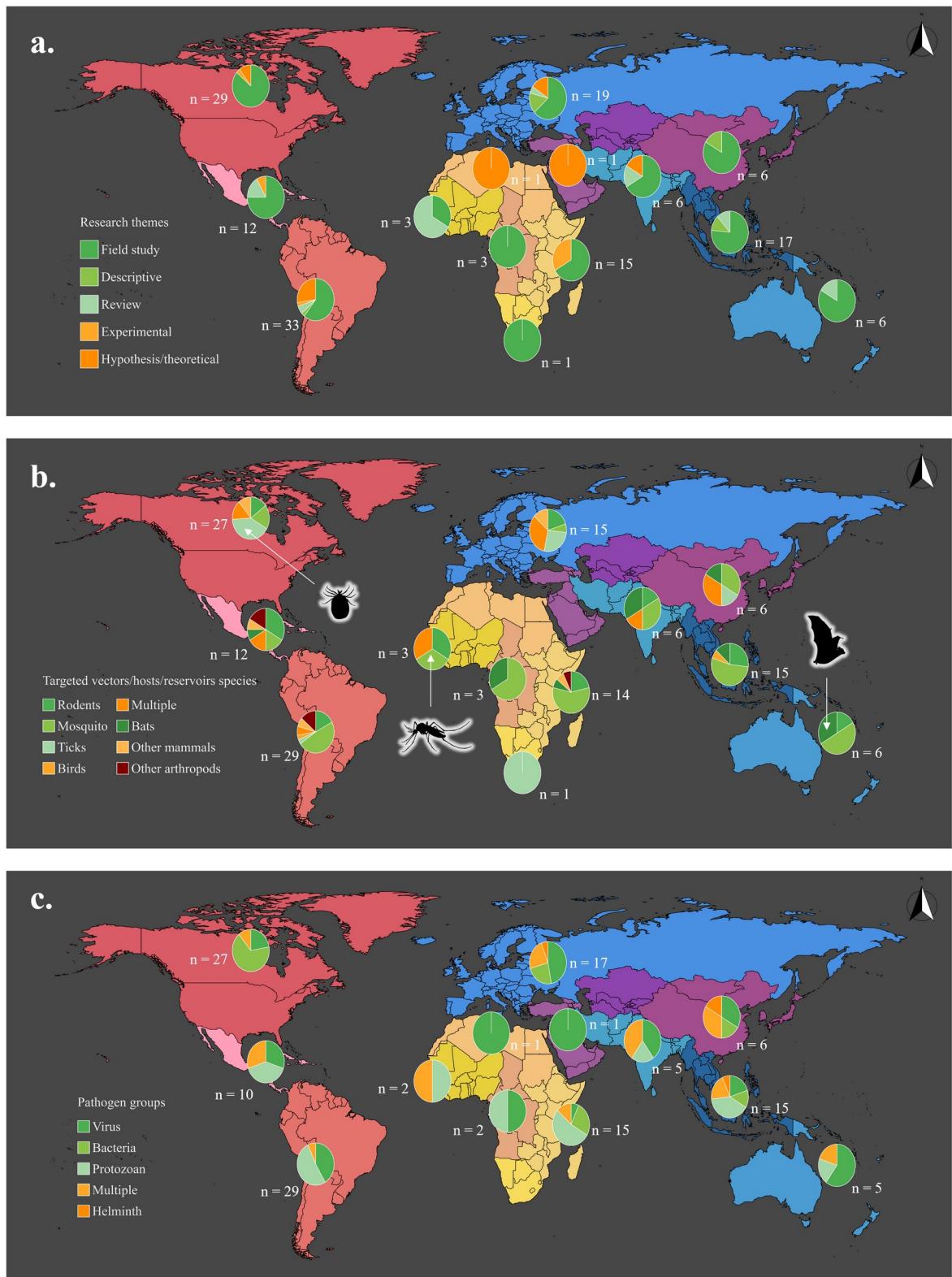
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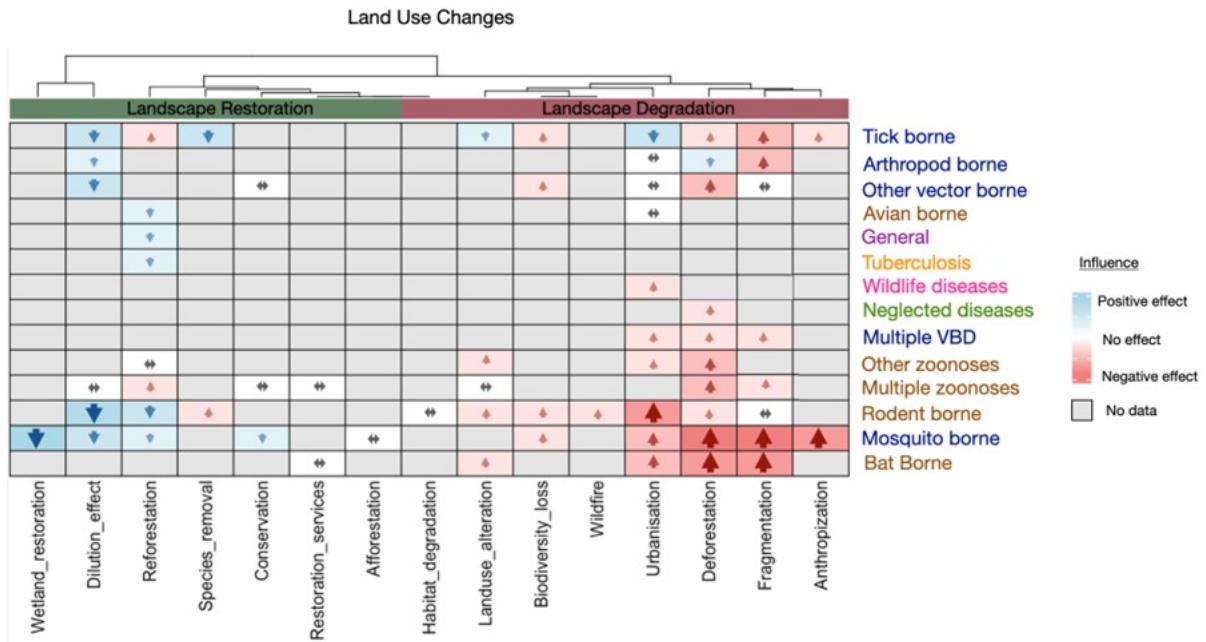


779 Figure 1. (A) Annual count of published studies from 2000 to 2024 investigating the impacts of
 780 landscape degradation or restoration on zoonotic disease transmission risks. (B) The geographic
 781 distribution of 138 studies focusing on the effects of landscape degradation and restoration on zoonotic
 782 disease transmission risks during the same period. Colours represent the percentage of studies published
 783 in each country.

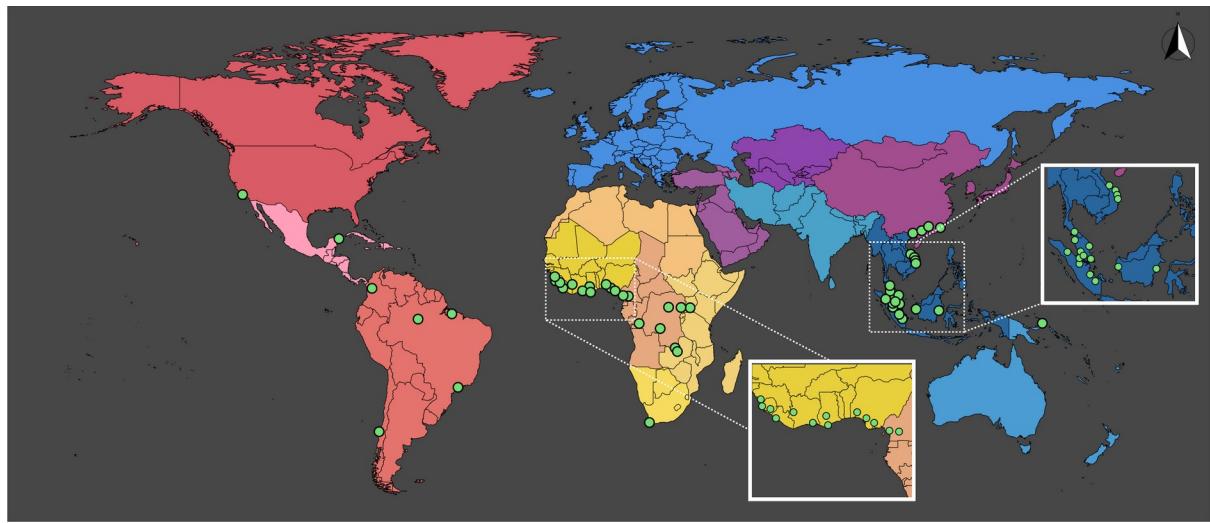


784 Figure 2. Fifteen UN Geoscheme Regions (represented by different colours; see
 785 <https://unstats.un.org/unsd/methodology/m49/> for regional definitions), with corresponding pie charts
 786 illustrating the proportions of various 2a) research themes, 2b) vectors/host/reservoirs species and 2c)

787 pathogen groups represented in a total of 138 studies. Some regions do not have corresponding pie
788 charts as these studies lacked that information. Mosquito, tick and bat silhouettes obtained from
789 PhyloPic (<https://www.phylopic.org/>).

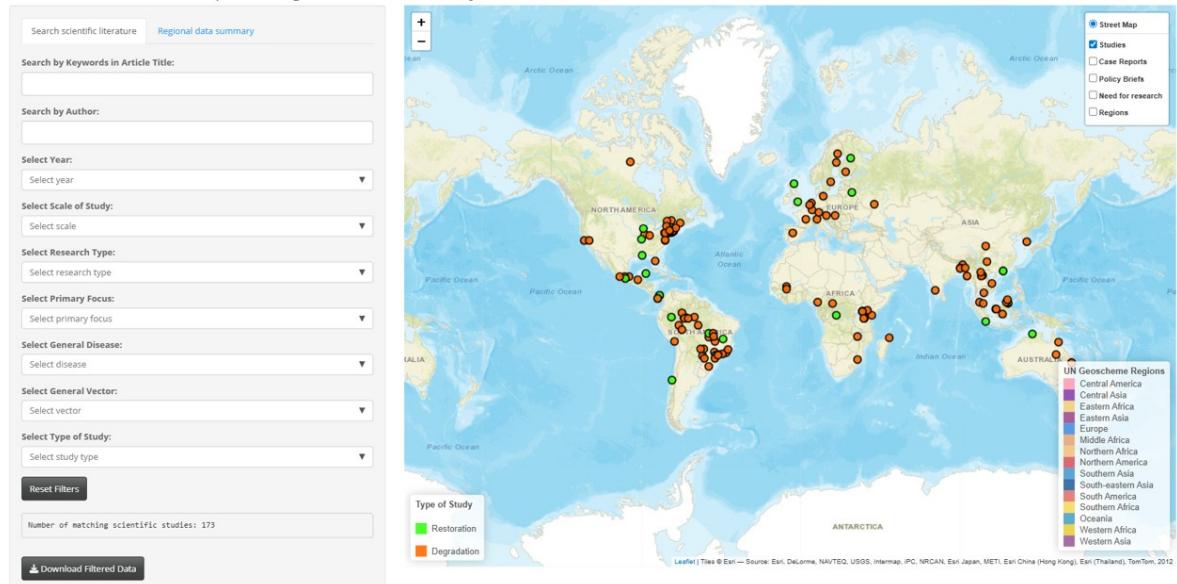


790 Figure 3. Heatmap showing the size and direction of reported effects of land-use change and
 791 biodiversity–disease relationships on disease transmission risk across different disease groups (Vector
 792 borne diseases in blue, Zoonoses in brown, General in purple, Neglected diseases in green, Wildlife
 793 diseases in pink & Tuberculosis in orange). Each cell represents the mean directionality derived from
 794 studies reporting statistically significant results: blue with downward arrow indicates a reported positive
 795 association (e.g., decreased disease risk), red with upward arrow indicates a negative association (e.g.,
 796 increased risk), and white indicates a reported no significant effect with a bidirectional marker.



797 Figure 4. Fifteen UN Geoscheme Regions (represented by different colours; see
798 <https://unstats.un.org/unsd/methodology/m49/> for regional definitions) with the locations of 50 priority
799 sites with necessity for further research (NFR) on landscape alterations and disease emergence.

Interactive Evidence Map and Regional Data Summary



800 Figure 5. Overview of the Online Evidence Atlas (<https://bradduthie.shinyapps.io/atlas/>) interface
801 displaying the interactive map and filtering panels.

Supplementary Material:

Supplementary table 1. Components of the research question using the *PerSPEcTiF* framework.

| <u>Perspective</u> | <u>Setting</u> | <u>Phenomenon of interest</u> | <u>Environment</u> | <u>Timing</u> | <u>Findings</u> |
|---|----------------|---|---|---|--|
| Journal articles, national policies and organisational reports relevant to restoration of degraded ecosystems | Global | <p>Impact of landscape degradation on zoonotic spillovers, disease outbreaks in animal and human populations, cross-species pathogen transmission, and pandemics.</p> <p>Restoration of degraded ecosystems reducing or increasing the disease risk.</p> <p>Dynamics of ecosystem restoration at the human/wildlife/ livestock/ pathogen interface.</p> <p>Consequences on local populations following ecosystem restoration.</p> | Terrestrial, freshwater & marine ecosystems | From, and including, 2000 to early 2024 | <p>Current state of evidence and knowledge gaps.</p> <p>Comparison of existing protocols and policies.</p> <p>Consequences of restoring ecosystems on the mechanisms of disease dynamics- local and international perspective.</p> |

Inclusion criteria:

The inclusion criteria ensured that the studies captured a comprehensive range of research focused on biodiversity, land-use changes including deforestation, intensive agriculture, and urbanisation, restoration efforts, human health impacts, and infectious disease transmission risks.

- Peer-reviewed literature published between 2000-2024
- English language
- Focus on land-use changes and infectious disease emergence or transmission.
- Focus on landscape restoration and infectious disease emergence or transmission.
- Focus on the influence of socioeconomic factors landscape alterations and infectious diseases and vice versa.

Supplementary table 2. Electronic database search keyword.

| <u>Term</u> | <u>Keywords</u> |
|------------------|--|
| Biodiversity | biodivers*, biodivers* loss*, communiti* composit*, divers* affect*, divers* decreas*, ecolog* factor*, ecolog* variabl*, increas* speci*, speci* divers*. |
| Land-use | communit* restor*, habitat* restor*, ecosystem* restor*, ecolog* restor*, landscape* ecology*, ecosystem* service*, habitat* fragment*, habitat* alterat*, human-wildlife* interact*, ecosystem* health*, communit* structure*, deforestation*, afforestation*, edge effect*, forest* cover*, forest* fragment*, forest* patch*. |
| Health | affect* human*, dilut* effect*, direct* transmit*, emerg* outbreak*, human* diseas*, human* exposur*, human* health*, human* popul*, infect* preval*, preval* increas*, public* health*, spillover* risk*, epidemic*, pandemic*. |
| Zoonotic disease | affect* diseas*, decreas* diseas*, increas* diseas*, diseas* emerg*, diseas* preval*, diseas* suppress*, infecti* diseas*, vector-born* diseas*, zoonot* diseas*, zoonot* releas*, diseas* spillover*, diseas* ecology*, diseas* transmiss*, host-pathog* inter*, diseas* outbreak*, diseas* reservoir*, diseas*, zoonos*, pathogen* trans*, infection*. |

Supplementary table 3. Data extraction form.

| Section | Attribute | Explanation |
|---------------------|---------------------------|--|
| Pre-filled metadata | Article type | Web of Science/Citation chasing |
| | Article source | Review/Original article/Book chapter/Comment |
| | Publication details | Title, authors, publication year, DOI |
| | Language of publication | The language the article is written in |
| Data extraction | Geographical location | Location or study area of research |
| | Location coordinates | Yes/No – if yes, lat and long were extracted |
| | Scale of the study | Global/Continental/Multi-regional/National/Regional/Local |
| | Theme | Dilution effect/Disease dynamics at the human/wildlife/livestock/ pathogen interface at degraded landscapes/Impact of landscape degradation on spillover risk/Impact of restoration degradation on spillover risk/Social, economic and public health impact on restoration of degraded landscapes/Existing policies on landscape restoration and disease emergence |
| | Infectious disease | Disease which is the main focus of the article |
| | Disease transmission type | Zoonoses/Vector-borne/Generalist |
| | Pathogen group | Virus/bacteria/protozoan/fungi/parasite/prion |
| | Vector/reservoir/host | Species which is the main focus of the article |
| | Method to assess risk | Historical data/Environmental sampling/Modelling/Population screening/Other |
| | Ecosystem | Terrestrial/Freshwater/Marine/Salt marsh |
| Research type | Recommendations | Yes/No |
| | | Hypothesis or theoretical/ Experimental/ Field study/ Descriptive/Review/Recommendation/Opinion or perspective |
| | Knowledge areas | Model/theory/Framework or protocol/Lessons learnt/Knowledge gaps/Tools/Other |

| | | |
|---|--|-----------------|
| Type of restoration | Afforestation/Reforestation/Species removal/Species introduction/Wetland restoration/Other | |
| Impact of landscape restoration on disease | Direct/Indirect/NA | |
| Impact of landscape degradation on disease | Direct/Indirect/NA | |
| Impact on human-wildlife interactions | Direct/Indirect/NA | |
| Evidence for Landscape restoration on disease outbreaks | “Extracted word by word from the article” | |
| Evidence for Landscape degradation on disease outbreaks | “Extracted word by word from the article” | |
| Evidence for effects on human-wildlife interactions | “Extracted word by word from the article” | |
| Are Socioeconomic factors considered? | “Extracted word by word from the article” | |
| Limitations & challenges | “Extracted word by word from the article” | |
| Knowledge gaps/future research | “Extracted word by word from the article” | |
| Recommendations & proposed solutions | “Extracted word by word from the article” | |
| Validation | Risk of bias* | High/Medium/Low |
| | Imprecision** | High/Medium/Low |
| | Inconsistency*** | High/Medium/Low |
| | Reporting bias**** | High/Medium/Low |
| | Quality of evidence***** | High/Medium/Low |
| | Reviewer confidence | High/Medium/Low |

Validation troubleshooting:

*Risk of bias - Bias occurs when the results of a study do not represent the truth because of inherent limitations in design or conduct of a study.

**Imprecision - For EXPERIMENTAL STUDIES ONLY. The GRADE approach to rating imprecision focuses on the 95% confidence interval around the best estimate of the absolute effect.

*** Inconsistency - For REVIEWS ONLY. Certainty in a body of evidence is highest when there are several studies that show consistent effects.

**** Reporting bias - Reporting bias occurs when the dissemination of research findings is influenced by the nature and direction of the results.

*****Quality of evidence - Final assessment of quality.

Supplementary table 4. Eight climatic, environmental and demographic covariates used to generate the “Necessity for Further Research” locations.

| <u>Covariate</u> | <u>Source</u> |
|--|---|
| Climate covariates | |
| Mean temperature | https://www.worldclim.org |
| Precipitation | https://www.worldclim.org |
| Environmental covariates | |
| Altitude | https://www.worldclim.org |
| Global human modification of terrestrial systems | Global Human Modification of Terrestrial Systems. Palisades, New York: NASA Socioeconomic Data and Applications Centre (SEDAC). https://doi.org/10.7927/edbc-3z60 . |
| Mammal richness index | https://www.iucnredlist.org/resources/other-spatial-downloads#SR_2023 |
| Forest cover loss proxy for deforestation | Hansen, M. C., et al. High-Resolution Global Maps of 21st-Century Forest Cover Change. <i>Science</i> 342 (2013): 850-53. https://glad.earthengine.app/view/global-forest-change . |
| Demographic covariates | |
| Accessibility to the nearest city | Nelson, A., et al. A suite of global accessibility indicators. <i>Sci Data</i> 6, 266 (2019). https://doi.org/10.1038/s41597-019-0265-5 |
| Population density | Centre for International Earth Science Information Network - CIESIN - Columbia University. 2018. Gridded Population of the |

World, Version 4 (GPWv4): Population Density, Revision 11.

Palisades, New York: NASA Socioeconomic Data and Applications Centre (SEDAC).

<https://doi.org/10.7927/H49C6VHW>

Supplementary table 5. The “Necessity for Further Research” (NFR) map, a product of the uncertainty, population density, and deforestation. We predicted locations in the study area such as to minimise the mean NFR across all the pixels in the map. The coordinates of the highest NFR i value was located on the NFR map and the NFR was reduced by a sequence of 75%, 50% , and 25% in a 50-km radius in concentric circles around the site.

| <u>Number</u> | <u>Latitude</u> | <u>Longitude</u> | <u>City/Town</u> | <u>Country</u> | <u>Region</u> |
|---------------|-----------------|------------------|---------------------|----------------------------------|--------------------|
| 1 | 3.875 | 11.54167 | Yaoundé | Cameroon | Middle Africa |
| 2 | 4.041667 | 9.791667 | Douala | Cameroon | Middle Africa |
| 3 | -2.958333 | 104.7083 | Palembang City | Indonesia | South-eastern Asia |
| 4 | 6.708333 | -1.625 | Kumasi | Ghana | Western Africa |
| 5 | -1.625 | 103.625 | Jambi | Indonesia | South-eastern Asia |
| 6 | 1.541667 | 103.7917 | Johor | Malaysia | South-eastern Asia |
| 7 | 5.458333 | -4.041667 | Abidjan | Côte d'Ivoire | Western Africa |
| 8 | -36.875 | -73.20833 | San Pedro de la Paz | Chile | South America |
| 9 | 0.375 | 32.54167 | Kampala | Uganda | Eastern Africa |
| 10 | -34.125 | 18.70833 | Cape Town | South Africa | Southern Africa |
| 11 | 0.5416667 | 101.4583 | Pekanbaru City | Indonesia | South-eastern Asia |
| 12 | -0.04166667 | 109.375 | Pontianak | Indonesia | South-eastern Asia |
| 13 | 0.4583333 | 29.45833 | Beni | Democratic Republic of the Congo | Middle Africa |
| 14 | 6.291667 | -10.70833 | Paynesville | Liberia | Western Africa |
| 15 | -3.041667 | -59.95833 | Manus | Brazil | South America |
| 16 | 3.041667 | 101.7083 | Seri Kembangan | Malaysia | South-eastern Asia |
| 17 | 4.125 | 103.4583 | Cherating | Malaysia | South-eastern Asia |
| 18 | 22.79167 | 108.2917 | Nanning | China | Eastern Asia |
| 19 | 0.5416667 | 25.20833 | Kisangani | Democratic Republic of the Congo | Middle Africa |
| 20 | 34.375 | -119.625 | Montecito | USA | North America |
| 21 | 16.45833 | 107.5417 | Hue | Vietnam | South-eastern Asia |
| 22 | 15.45833 | 108.7917 | Bình Thuận | Vietnam | South-eastern Asia |

| | | | | | |
|----|------------|-----------|-----------------------|----------------------------------|--------------------|
| 23 | -11.70833 | 27.45833 | Lubumbashi | Democratic Republic of the Congo | Middle Africa |
| 24 | 8.458333 | -13.20833 | Freetown | Sierra Leone | Western Africa |
| 25 | -4.375 | 15.20833 | Kinshasa | Democratic Republic of the Congo | Middle Africa |
| 26 | 6.208333 | -75.625 | Medellin | Colombia | South America |
| 27 | 5.041667 | -1.291667 | Elmina | Ghana | Western Africa |
| 28 | 23.54167 | 111.2917 | Wuzho | China | Eastern Asia |
| 29 | 1.708333 | 101.4583 | Dumai | Indonesia | South-eastern Asia |
| 30 | -23.54167 | -46.375 | Suzano | Brazil | South America |
| 31 | -12.79167 | 28.20833 | Kitwe | Zambia | Eastern Africa |
| 32 | -4.291667 | 152.2917 | Kokopo | Papua New Guinea | Oceania |
| 33 | 24.79167 | 113.625 | Shaoguan | China | Eastern Asia |
| 34 | 2.208333 | 102.2917 | Malacca | Malaysia | South-eastern Asia |
| 35 | -5.875 | 22.375 | Kananga | Democratic Republic of the Congo | Middle Africa |
| 36 | -0.4583333 | 117.125 | Samarinda City | Indonesia | South-eastern Asia |
| 37 | 21.125 | -86.875 | Cancun | Mexico | Central America |
| 38 | 5.375 | 100.5417 | Kulim | Malaysia | South-eastern Asia |
| 39 | 7.958333 | -11.70833 | Bo | Sierra Leone | Western Africa |
| 40 | 24.45833 | 117.625 | Longhai City | China | Eastern Asia |
| 41 | 14.54167 | 109.125 | Binh Dinh | Vietnam | South-eastern Asia |
| 42 | 6.291667 | 5.625 | Benin City | Nigeria | Western Africa |
| 43 | 7.041667 | 100.4583 | Songkhla | Thailand | South-eastern Asia |
| 44 | 2.958333 | 99.04167 | Pematang Siantar City | Indonesia | South-eastern Asia |
| 45 | -1.458333 | -48.45833 | Belem | Brazil | South America |
| 46 | 7.375 | -7.541667 | Man | Côte d'Ivoire | Western Africa |
| 47 | 13.70833 | 109.2917 | Quy Nhon | Vietnam | South-eastern Asia |
| 48 | 5.458333 | 7.041667 | Ihiagwa | Nigeria | Western Africa |
| 49 | 7.375 | 3.958333 | Ibadan | Nigeria | Western Africa |
| 50 | 9.708333 | -13.45833 | Coyah | Guinea | Western Africa |

Supplementary table 6. The aggregated effect values for the effects of land use change on disease transmission risk across different disease groups for our hierarchical cluster heatmap (Fig. 3). Values represent the cumulative number of studies reporting significant effects, where positive values indicate more studies found a positive association (e.g., decreased disease risk) and negative values indicate more studies found a negative association (e.g., increased risk). The total

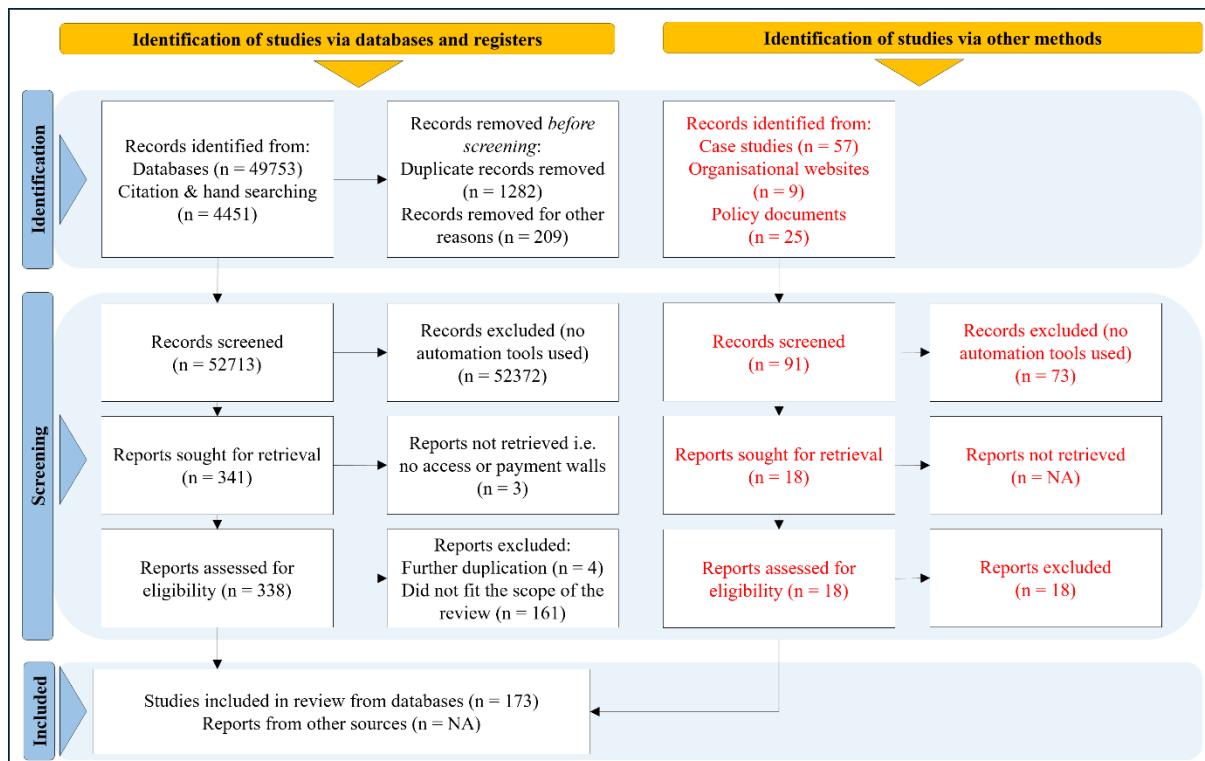
| Disease transmission | Afforestation | Anthropization | Biodiversity loss | Conservation | Deforestation | Dilution effect | Fragmentation | Habitat degradation | Landuse alteration | Reforestation | Restoration services | Species removal | Urbanisation | Wetland restoration | Wildfire | Total number of studies |
|--------------------------------|---------------|----------------|-------------------|--------------|---------------|-----------------|---------------|---------------------|--------------------|---------------|----------------------|-----------------|--------------|---------------------|----------|-------------------------|
| Arthropod-borne | NA | NA | NA | NA | 1 | 1 | -2 | NA | NA | NA | NA | NA | 0 | NA | NA | 7 |
| Avian-borne | NA | NA | NA | NA | NA | NA | NA | NA | NA | 1 | NA | NA | 0 | NA | NA | 2 |
| Bat-borne | NA | NA | NA | NA | -3 | NA | -3 | NA | -1 | NA | 0 | NA | -2 | NA | NA | 12 |
| General | NA | NA | NA | NA | NA | NA | NA | NA | NA | 1 | NA | NA | NA | NA | NA | 1 |
| Mosquito-borne | 0 | -3 | -1 | 1 | -17 | 2 | -5 | NA | NA | 1 | NA | NA | -2 | 4 | NA | 54 |
| Multiple VBD | NA | NA | NA | NA | -1 | NA | -1 | NA | NA | NA | NA | NA | -1 | NA | NA | 3 |
| Multiple Zoonoses | NA | NA | NA | 0 | -2 | 0 | -1 | NA | 0 | -1 | 0 | NA | NA | NA | NA | 9 |
| Neglected diseases | NA | NA | NA | NA | -1 | NA | NA | NA | NA | NA | NA | NA | NA | NA | NA | 1 |
| Other vector-borne | NA | NA | -1 | 0 | -2 | 2 | 0 | NA | NA | NA | NA | NA | 0 | NA | NA | 8 |
| Other zoonoses | NA | NA | NA | NA | -2 | NA | NA | NA | -1 | 0 | NA | NA | -1 | NA | NA | 7 |
| Rodent-borne | NA | NA | -1 | NA | -1 | 3 | 0 | 0 | -1 | 2 | NA | -1 | -3 | NA | -1 | 28 |
| Tick-borne | NA | -1 | -1 | NA | -1 | 2 | -2 | NA | 1 | -1 | NA | 2 | 2 | NA | NA | 36 |
| Tuberculosis | NA | NA | NA | NA | NA | NA | NA | NA | NA | 1 | NA | NA | NA | NA | NA | 1 |
| Wildlife diseases | NA | NA | NA | NA | NA | NA | NA | NA | NA | NA | NA | NA | -1 | NA | NA | 1 |
| <i>Total number of studies</i> | 3 | 4 | 4 | 3 | 41 | 15 | 36 | 1 | 8 | 16 | 2 | 3 | 27 | 4 | 1 | |

at the bottom reflects the number of studies evaluating the dilution effect, not the arithmetic sum of scores. Opposing results within a study can cancel each other numerically, so the column sum does not represent the study count

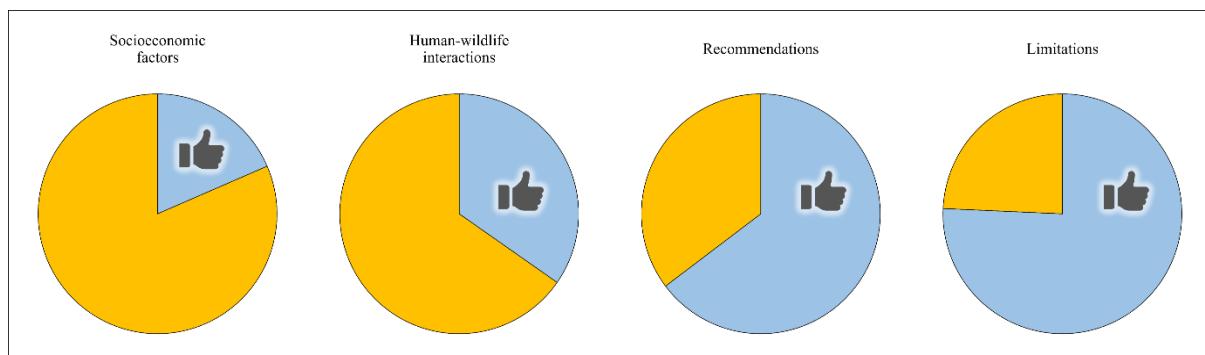
Supplementary table 7. Definition of disease categories used in the review.

| Category | Definition | Examples |
|---------------------------------------|---|--|
| Zoonotic (direct transmission) | Diseases transmitted from animals (wildlife, livestock, companion animals) to humans without an arthropod vector. | Puumala hantavirus, Leptospirosis, Rabies, Brucellosis |
| Vector-borne (zoonotic) | Diseases transmitted from animals to humans via arthropod vectors. | West Nile virus, Lyme disease, Chagas disease, <i>Plasmodium knowlesi</i> malaria, Cutaneous leishmaniasis |
| Vector-borne (non-zoonotic) | Diseases transmitted by arthropod vectors that can circulate exclusively in human–vector–human cycles (no animal reservoir required for maintenance although sylvatic cycles may or may not be present). | Dengue, Yellow fever, Zika, <i>Plasmodium falciparum</i> malaria, <i>P. vivax</i> malaria, <i>P. ovale</i> malaria, <i>P. malariae</i> malaria |
| General | Diseases primarily acquired from environmental sources. | Waterborne disease – cholera and soil borne disease like tetanus and soil helminths |
| Neglected diseases | Neglected tropical diseases (NTDs) are a diverse group of conditions caused by a variety of pathogens (including viruses, bacteria, parasites, fungi and toxins) and associated with devastating health, social and economic consequences in impoverished communities in tropical areas. Although dengue, chikungunya, and leishmaniasis are classified as NTDs, we categorized them as vector-borne diseases, giving priority to transmission mode. Diseases were classified as NTDs only when the term was used as an umbrella category rather than for specific pathogens. | Buruli ulcer |
| Wildlife diseases | Studies primarily focused on wildlife and diseases circulating in wild animal populations, with or without documented human spillover. | Avian malaria, chytridiomycosis |
| Multiple zoonoses/VBD | Studies that reported aggregated outcomes for more than one zoonotic disease or VBD without providing disease-specific estimates | Studies reporting combined risk patterns for several zoonoses |
| Other zoonoses | Includes zoonotic diseases that did not fall into any of the named categories (e.g., bat-borne, rodent-borne, mosquito-borne) due to limited data or heterogeneous transmission modes. | Brucellosis, Q fever |
| Other VBD | Includes VBD that did not fall into any of the named categories (e.g., arthropod-borne, | Tsutsugamushi disease |

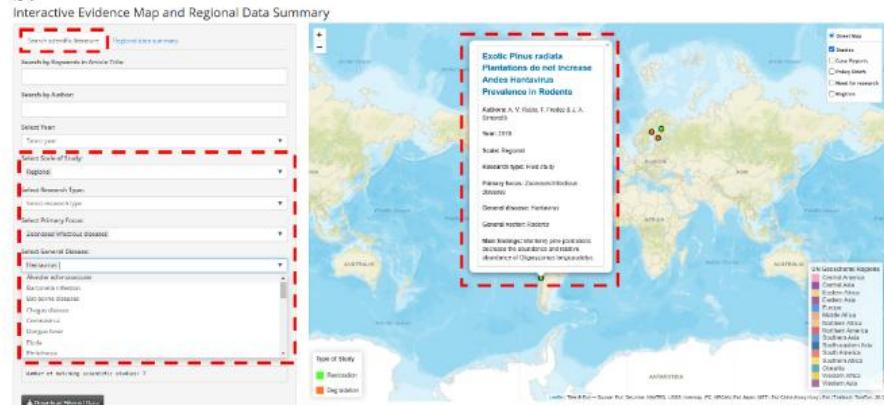
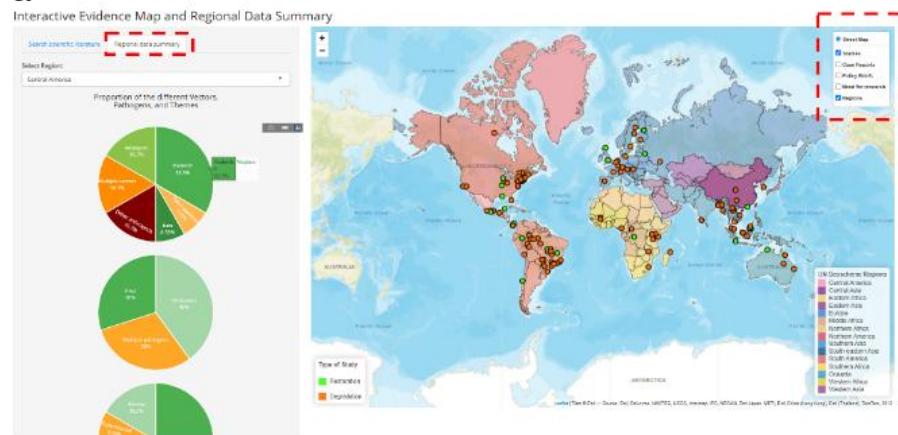
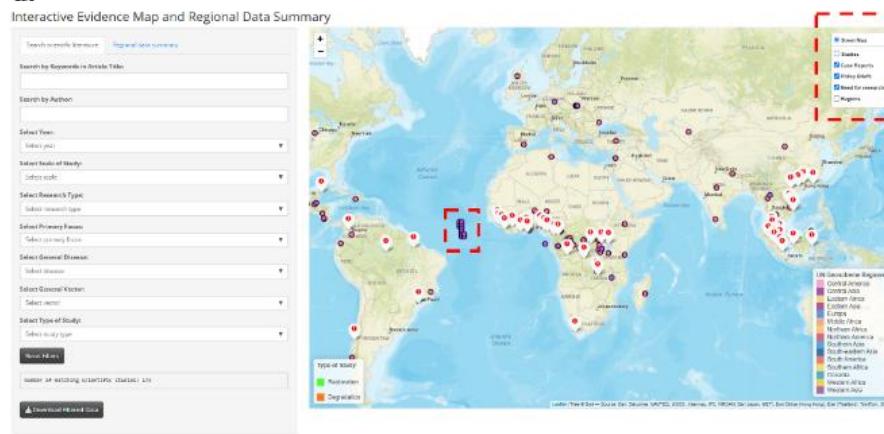
| Category | Definition | Examples |
|---------------------|---|--|
| | mosquito-borne, tick-borne) due to limited data or heterogeneous transmission modes. | |
| Tuberculosis | A directly transmitted bacterial disease caused by <i>Mycobacterium tuberculosis</i> , primarily affecting the lungs and spread via airborne droplets between humans. It was classified as a separate category given its unique transmission pathway, public health relevance, and the fact that it is not vector-borne or wildlife-associated. | <i>Mycobacterium tuberculosis</i> , Pulmonary TB |



Supplementary figure 1. Prisma flowchart detailing the articles screened and included for data extraction. Adapted from Page et al, 2021.



Supplementary figure 2. Pie charts illustrating the proportion of studies discussing various topics within their publications. Each pie chart displays the percentage of studies that included specific content, with blue indicating the presence of the content and yellow indicating its absence. The topics covered are as follows: socioeconomic factors, human-wildlife interactions, recommendations or proposed solutions, and limitations and challenges (n=173).

a.**b.****c.****d.**

Supplementary figure 3. Our online evidence map (<https://bradduthie.shinyapps.io/atlas/>) has a User Interface (UI) centred around an interactive map that displays the geographic locations of scientific literature, case reports, and policy briefs related to the effects of restoration and landscape degradation on disease transmission. Users toggle these layers on and off using the panel in the top-right corner (Supplementary figure 3a). Each marker on the map is clickable and opens a popup summary specific to the type of literature selected. For scientific literature, the summary includes the title, authors, year of publication, scale,

research type, primary focus, general disease, general vector, and a brief overview of the findings (Supplementary figure 3b). For grey literature, summaries provide the title, year, results, and lessons learned. Titles in each summary are hyperlinked to their source, allowing for easy access to full documents. Additionally, users can toggle the display of “Need for Research” locations and UN Geoscheme regional boundaries to visualise global research gaps.

The panel to the left-hand side allows users to filter through the scientific literature using the first tab labelled “Search scientific literature”. The user can use keywords to search for matches within the literature title or even search by author if known. The other options relate to the information in the summary popups, and all have pre-defined selection options from the drop-down menus (Supplementary figure 3b). After each selection, the studies are filtered displaying the ones with successful matches. Those studies that were global scale are present on the app but are not initially visible; this is because there is no generic location to add these to the map. Therefore, global studies are only visible once the scale selection of “Global” is selected. They are then visible in the Atlantic Ocean, with the global policy briefs too (Supplementary figure 3d). The filtered studies can be downloaded using the download button below these selection filters.

The second tab on the left-hand side panel, labelled “Regional data summary”, has a drop-down menu for the UN Geoscheme regions (Supplementary figure 3c). Once a region is selected, summary pie charts are displayed. These pie charts represent the summary of scientific literature from each of these regions for Vectors/Hosts/Reservoirs, Pathogens and Themes of the studies, and are the ones present in this review (Fig. 2a; 2b; 2c).