

Climate change and land-use scenarios of crop pollination risk

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Abstract

Recent studies have highlighted rapid ongoing changes in insect biodiversity, in part driven by the interaction of land-use and climate change. Insect pollinator biodiversity specifically has been shown to be undergoing rapid changes, with potential consequences for the provision of crop pollination. Here, we present a global assessment of the interactive effects of climate change and land use on pollinator abundance, and predictions of the risk to crop pollination worldwide from the inferred changes in pollinator abundance. We show that the combination of agricultural land use and recent climate change is associated with particularly large reductions in the abundance of insect pollinators. We further show that the tropics will likely experience the greatest risk of future pollination shortfalls, putting at risk the production of crops that depend on insect pollination. The risk to these crops is highest, and predicted to increase most rapidly, in regions of sub-Saharan Africa and northern South America, especially in areas used to produce cocoa, mango, pumpkin, melon, watermelon, and coffee. The health, well-being, and livelihoods of billions of people to some extent depends upon the availability and affordability of crops dependent on animal pollination. Climate change and agricultural land use could put these people's wellbeing at high risk.

One sentence summary

Climate change associated crop pollination risk is highest, and predicted to increase most rapidly, in regions of sub-Saharan Africa and northern South America, especially in areas used to produce cocoa, mango, pumpkin, melon, watermelon, and coffee.

Introduction

Recent studies have highlighted rapid ongoing changes in insect biodiversity (1–3). Some of these studies have reported net declines (1, 3), at least for terrestrial insects (1), while others have shown mixtures of

gains and losses (2). Pollinator biodiversity specifically has been shown to be undergoing rapid changes (4–8), with potential consequences for the provision of pollination to wild plants and crops. Importantly, evidence for insect biodiversity trends has been biased toward Western Europe and North America, with little coverage of tropical and subtropical regions (9, 10), although see (11, 12) for two recent examples. Although a few studies have shown steep declines of insects in the tropics (3), evidence about insect biodiversity trends there often remains anecdotal (13), with global syntheses (1, 2) having strong geographic biases towards non-tropical regions.

Among the drivers of insect and pollinator biodiversity changes, human land-use disturbance and climate change are prominent (5, 10, 14–17). Climate change in particular is emerging as an increasingly important driver (8, 14, 18–21), as well as the interactive effects of land-use and climate change (22–26). For interactions, a key mechanism underpinning this effect is the altered microclimatic conditions in areas where vegetation has been modified for human land use (22). This results in synergistic effects of climate change and land use, leading to greater declines in biodiversity than if the pressures acted alone (22, 24–26). Tropical insects are expected to be more susceptible to climate change and interactive effects, given their narrower physiological tolerance compared to non-tropical species (27). Although evidence about the response of tropical insects to climate change interactions remains scarce (21), recent studies show greater effects in tropical than non-tropical insect biodiversity (26).

Changes in the biodiversity and composition of pollinator communities will have large effects on the provision of pollination services. Pollen limitation from animal pollinator losses has already been shown to reduce the reproductive success of wild plants (4, 28), and the productivity of certain crops (29–32), although there is no evidence yet that pollen limitation is causing wholesale reductions in yields of crops that rely on animal pollination (33). Evidence that insect biodiversity responds to human pressures more strongly in the tropics than elsewhere (17, 26) is noteworthy, given that the majority of animal-pollination-dependent crops are grown there (34). However, it is not only tropical countries that will experience the effects of pollinator losses and subsequent pollen limitation, with highly developed countries benefiting from imports of animal-pollination-dependent foods from tropical areas (35). Abundance, species diversity and functional diversity of pollinators have all been implicated as determinants of the delivery of pollination service (32, 36–40). Previous attempts to model the provision of crop pollination service have been based on predictions of pollinator abundance, which bears a direct relation to pollination of plants, and has been shown to give a reasonable approximation of pollen deposition in at least some study systems (41).

Here, we present a global assessment of the interactive effects of climate change and land use on pollinator abundance, and predictions of the risk to crop pollination worldwide from the inferred changes in pollinator abundance (see Shiny app for full extended predictions, https://joemillard.shinyapps.io/pollinator_dependence_visualisation/). Our analyses are based on the PREDICTS database of biodiversity comparisons in different land uses (42), together with a list of species identified in the literature as likely pollinators (17). We use mixed-effects models to fit total pollinator community abundance as a function of land use (primary vegetation versus cropland) and a measure of historical temperature change between a baseline period (1901–1930) and the year prior to biodiversity sampling, standardized by monthly temperature variability in the baseline period (26). We standardised this temperature anomaly to up-weight regions that are experiencing novel temperatures beyond ordinary seasonal variation, which has previously been identified as important to insects (26). We then apply these models to predict which locations and crops are likely to be at greatest risk from pollination shortfalls, using estimated proportional reduction of pollinator total abundance as a proxy for relative risk to pollination services. This relative risk is then moderated according to estimates of where crops are grown (43), how dependent these crops are on animal pollination (34), projections of historical and future climate change, and a set assumptions for the relationship between local pollinator abundance and crop pollination (e.g. from linear to highly convex or concave). Lastly, we then combine these projections with estimates of the trade in pollination dependent crop production (35), to predict regions of the world that may be vulnerable to the indirect consequences of crop pollination risk via trade routes.

83 Results and Discussion

84 The abundance of insect pollinators responded strongly to the interaction of recent climate change and land
85 use (Figure 1). Within cropland experiencing novel temperatures (Standardized Temperature Anomaly =
86 1), pollinator abundance is 38.9% of natural habitat that has not experienced temperature increases. The
87 causal mechanism underpinning this interaction is unclear, but potentially relates to the homogenisation of
88 microclimatic conditions on anthropogenic land (44). Our results are qualitatively consistent with recent
89 results for a sample of all functional groups of insects (26), but importantly we show that responses to the
90 interactive effects of climate change and land use are stronger for pollinating than non-pollinating insects
91 (Figure 1). We also show that the strength of the interactive effect varies among taxonomic groups, with
92 the strongest effects seen in dipteran and hymenopteran pollinators (Figure S05). Whether the sensitivity
93 of pollinating insects to the interaction of climate change and land use relates directly to their reliance on
94 floral resources, or to other correlated traits typical of pollinators, is unclear. Most likely some combination
95 of both is true. For example, selection of animal pollinated plants is thought to be highly sensitive to
96 climatic conditions (45), suggesting that plant-pollinator interactions are highly sensitive to thermal changes.
97 Pollinator pilosity (i.e. hairiness), on the other hand, likely affects an insect’s ability to adapt to changes
98 in climate (46), and given its nature as a trait typical of bees and hoverflies, tends to be correlated with a
99 reliance on floral resources (47).

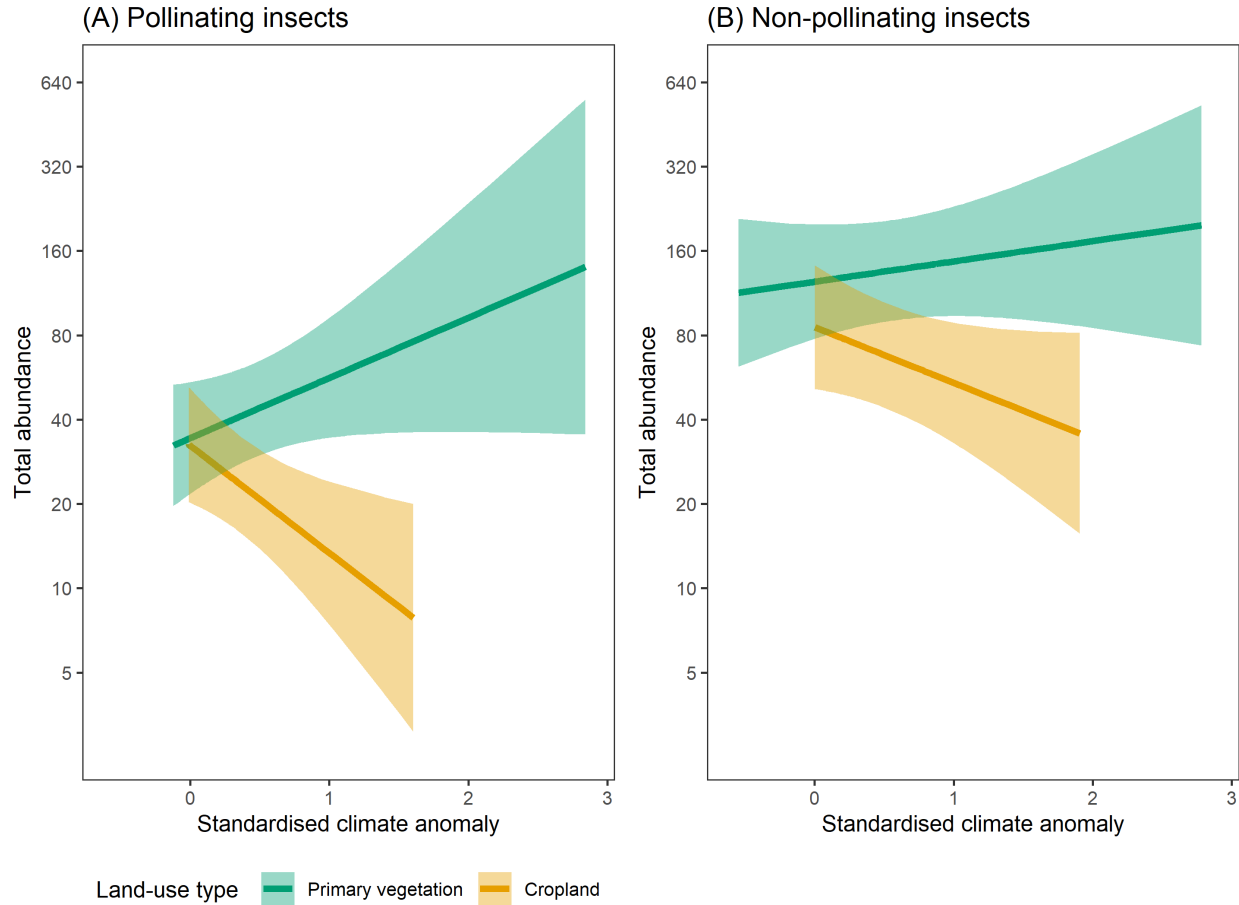


Figure 1: Response of pollinating (A) and non-pollinating (B) insect total abundance to the interactive effect of standardised temperature anomaly and land use (primary vegetation versus cropland) (A: $F = 23.3195$, $p < 0.001$; B: $F = 10.5764$, $p < 0.01$). Note that abundance is plotted on a log_e scale. Results are based on linear mixed-effects models. Site numbers are as follows (also see Supplementary Material Table 2): Pollinating insects (primary vegetation = 1166, cropland = 1507); likely non-pollinating insects (primary vegetation = 1747, cropland = 922). Shading represents 95% confidence intervals around the mean fitted effect. Green = primary vegetation; yellow = cropland.

100 We predict that total pollination production risk will increase under all climate scenarios (Figure 2). For all
 101 scenarios, our projections use change in pollinating insect abundance as a proxy for the relative risk to the
 102 production of crops dependent on insects, incorporating information on where crops are grown worldwide
 103 (43) as well as the fractional dependence of crops on animal pollination (34). Our estimates of risk are based
 104 on the distribution of crops as grown in 2000, meaning we do not account for changes in the distribution of
 105 crops over time, which are likely to occur as a result of the direct impacts of climate change, indirect effects
 106 through the loss of pollinator biodiversity, and socio-economic factors such as price changes in the global
 107 markets for particular crops. Moreover, our estimates of both relative risk as well as absolute production
 108 risk should be interpreted as indices of risk, rather than predictions of the actual amount of production
 109 likely to be lost, given a very high uncertainty in how pollinator abundance changes will translate into actual
 110 production losses.

111 Increases in risk are seen for all assumed relationships between abundance loss and production risk (linear
 112 and varying degrees of convexity), although the magnitude of changes in relative risk and especially absolute
 113 production risk varied widely. The predicted rate of increase in average production risk was substantially

114 higher under RCP (Representative Concentration Pathway) 6.0 than RCP 2.6, suggesting that efforts to
115 mitigate climate change will reduce the risk to future crop production, alongside the many other bene-
116 fits (48). Relative production risk varied strongly between years under the convex abundance-production
117 relationships (Figure 2 and S13). This volatility may be explained by way in which the non-linearity of
118 abundance/production relationships interacts with inter-annual climate variability caused by the El Nino
119 Southern Oscillation (49), although at present this is unclear. While increasingly convex assumed relation-
120 ships between insect abundance loss and crop production risk led to steeper increases in relative production
121 risk with future climate change, absolute production risk was markedly lower owing to a lower baseline in
122 the present day (Figure 2). Results were robust to variation in climate predictions under different individ-
123 ual climate models (Figure S06), do not change markedly when abundance loss is capped at the maximum
124 model-fitted value (Figure S02), do not change markedly when tested for sensitivity to data quality (Fig-
125 ure S04), and likely hold across a set of plausible active season temperature thresholds (fitted through two
126 different approaches, see Figures S11 and S12) and abundance change slopes (Figure S10).

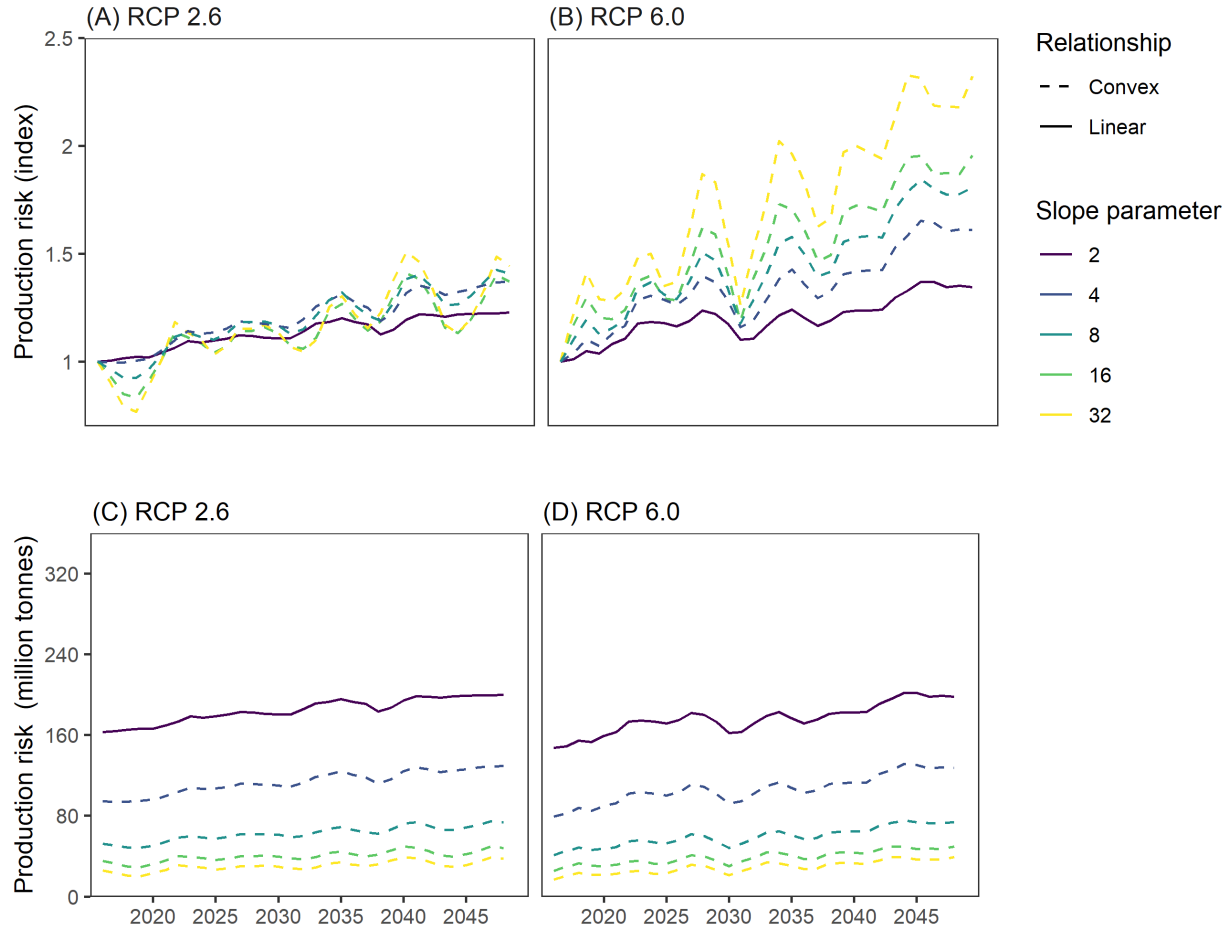


Figure 2: Projected change in total production risk under two RCP scenarios (2.6 and 6.0) and a set of different pollinator abundance-crop production relationships (linear and varying degrees of convexity). Results are shown both for an index of change in relative risk (A and B) and for absolute total production at risk (in tonnes) (C and D). For each year into the future, the standardised temperature anomaly was projected globally for all cells with production of crops dependent on animal pollination, using a 3-year rolling average of temperature anomaly estimates in each cell, from an ensemble of different climate models. We used data on crop production from the year 2000 (the latest year when such data are available for all crops). For each annual projection of standardised temperature anomaly, insect pollinator abundance on cropland was predicted according to the model shown in Figure 1A, and then expressed as proportional abundance loss compared to cropland that has experienced no warming (i.e. standardised temperature anomaly of 0). In each cell, the total production of each crop (43) was multiplied by dependence on animal pollination (34), and then adjusted for the predicted percentage reduction in insect pollinator abundance in that cell. These estimates of crop production at risk were summed across crops, and then across all terrestrial cells globally. Lines show different assumed relationships between insect pollinator abundance and crop production: dashed = convex relationships (of differing degrees, indicated by different colours: yellow = most convex; purple/blue = least convex); and solid = linear relationship.

127 Production risk from insect pollinator abundance losses is highest, and predicted to increase most rapidly,
 128 in regions of sub-Saharan Africa, northern South America, and southeast Asia (Figure 3A and Figure 3C).
 129 Somalia, Guinea Bissau, and Ivory Coast emerge as particularly vulnerable, given their high and increasing

130 production risk, and the high dependence of their economy on agriculture (Figure 3C). A number of countries
131 in Asia, South America, and the Caribbean have high and increasing levels of production risk, but lower levels
132 of agricultural dependence, making their economies less vulnerable overall. For example, the Philippines,
133 Indonesia, Papua New Guinea, Puerto Rico, Haiti, and Suriname all have high and increasing levels of
134 production risk (Figure 3C), but relatively low GDP adjusted pollination dependent production. Among
135 crops, cocoa is estimated to be at highest risk by a large margin, followed by mango and watermelon (Figure
136 3B). Coffee is also expected to have a combination of relatively high production risk and high value, suggesting
137 that regions in which it is grown may experience economic difficulties, unless the pollination service can be
138 replaced cost effectively. Projections of local crop production risk are sensitive to the assumed abundance-
139 production relationship (Figure S08), with the exception of South East Asia which is consistently resolved
140 as high risk, and the temperate realm which is consistently predicted as low risk (Figure S09).

141 For both countries and crops it is difficult to predict how the consequences of our production risk measure will
142 play out in the real world. In part this is due to the push and pull of multiple uncertainties, some of which
143 we do explore here (e.g. the relationship between pollinator abundance and crop production), but many of
144 which we do not or cannot (e.g. the changing distribution of crops, the economic viability of hand pollination,
145 the buffering effects of managed pollinators, the effects of climate change alone, and other technological
146 solutions such as the breeding or engineering of pollinator independent cultivars). Regardless, at the global
147 scale independent of changes in agricultural intensity, it's not unreasonable to state that the distribution
148 of standardised temperature anomaly and animal-pollinator dependent crop production should predict a
149 substantial portion of the variation in pollinator-mediated losses to crop production. For cocoa, recent
150 research has focused on the direct effects of climate change on crop production (50, 51), often overlooking
151 those that might be pollinator-mediated. Given the difficulties in teasing these mechanisms apart, and that
152 the direct effects of climate change are easier to intuit, it is probable that this trend continues. However, and
153 importantly, solutions to direct and pollinator-mediated effects will differ. For example, shade trees might
154 protect crops from extreme temperatures (50), but might not for the ceratopogonid fly pollinators on which
155 cocoa pollination depends. This is particularly pertinent given that some cocoa varieties are limited more
156 by pollination availability than resource availability (52).

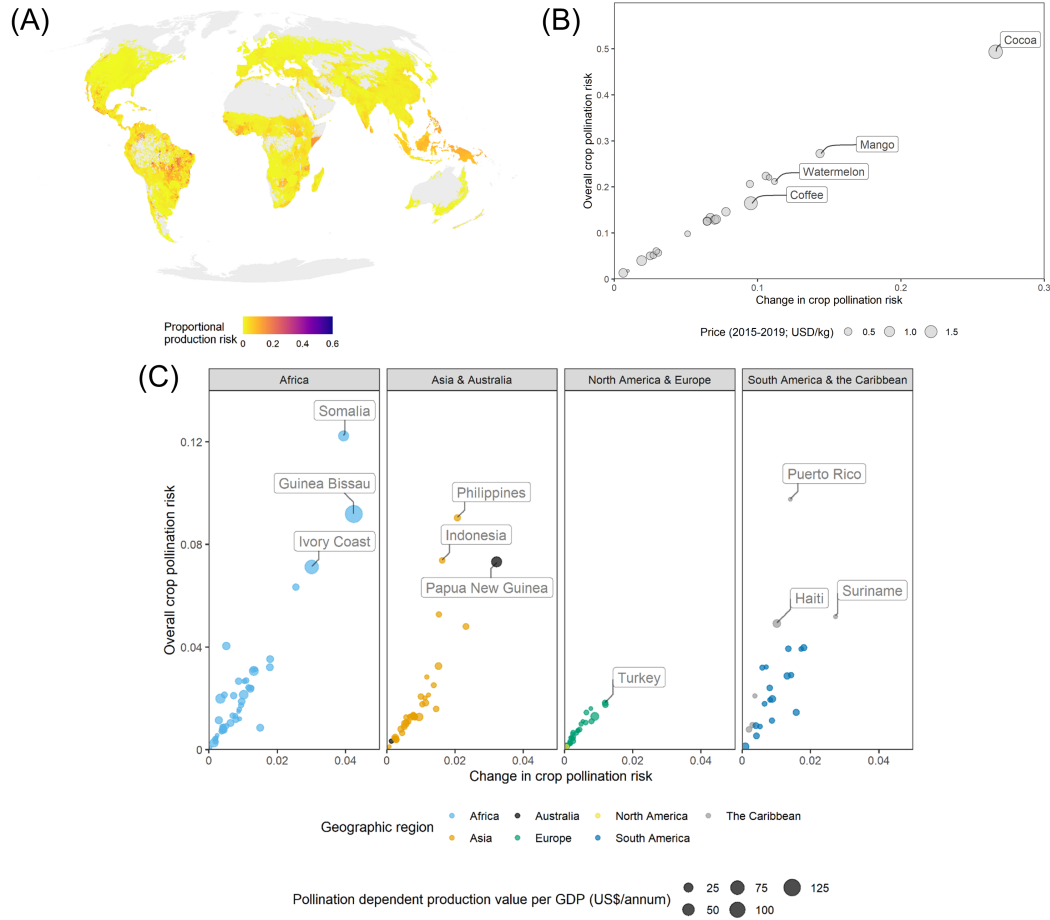


Figure 3: Projected change globally in the fraction of crop production estimated to be at risk in 2050 under the RCP 6.0 climate scenario, assuming a linear relationship between insect pollinator abundance loss and production loss for crops dependent on animal pollination. All projections are based on mean projections of the standardized temperature anomaly based on temperature estimates from an ensemble of individual climate models. (A) The sum of crop production at risk across all crops with some dependence on animal pollination (in each spatial cell), as a proportion of the production of all crops grown in that spatial cell (i.e. proportional production risk; see Figure S08 to see how cell-level risk differs among abundance-production scenarios). (B) Proportional production risk for the 20 crops with the highest total pollination dependent production globally (see Figure S01 for the top 20 crops). Overall risk here is the median of proportional production risk for all spatial cells in which that crop appears, whilst change in risk is the difference in overall risk between the start and the end of the series. Point size represents an estimation of the per kg value of each crop for the years 2015-2019, calculated from (53). (C) Proportional production risk at the level of each country. Here overall risk is the median of proportional production risk for all cells of that country, whilst change in risk is the difference in overall risk between the start and the end of the series. Point size here represents the total value of the pollination dependent production in that country adjusted for GDP, calculated from the product of total pollination dependent production per annum according to (43) and (34), and the per kg value of each crop (53). Colour represents the geographic region of each country, distinguishing between regions within a panel: Blue, Africa; orange, Asia; black, Australia; green, Europe; yellow, North America; dark blue, South America; grey, the Caribbean.

157 Countries besides those with high production risk will feel the impact of losses of pollinator biodiversity
158 and the crops that depend on them, through disruption to imports. Total import risk per capita tends to

159 be highest in northern countries, particularly continental western Europe, Canada, and parts of the Middle
 160 East. Some of the countries with the highest levels of import risk are those that have a high dependence
 161 on food imports, such as the Cayman Islands, Singapore, and Aruba. Interestingly, the Netherlands has
 162 the third highest per capita import risk, consistent with its status as the greatest importer of cocoa beans
 163 worldwide (54). Our predictions of import risk are based on trade patterns from the present day (35),
 164 meaning we do not account for changes in trade flows that will likely occur in the future.

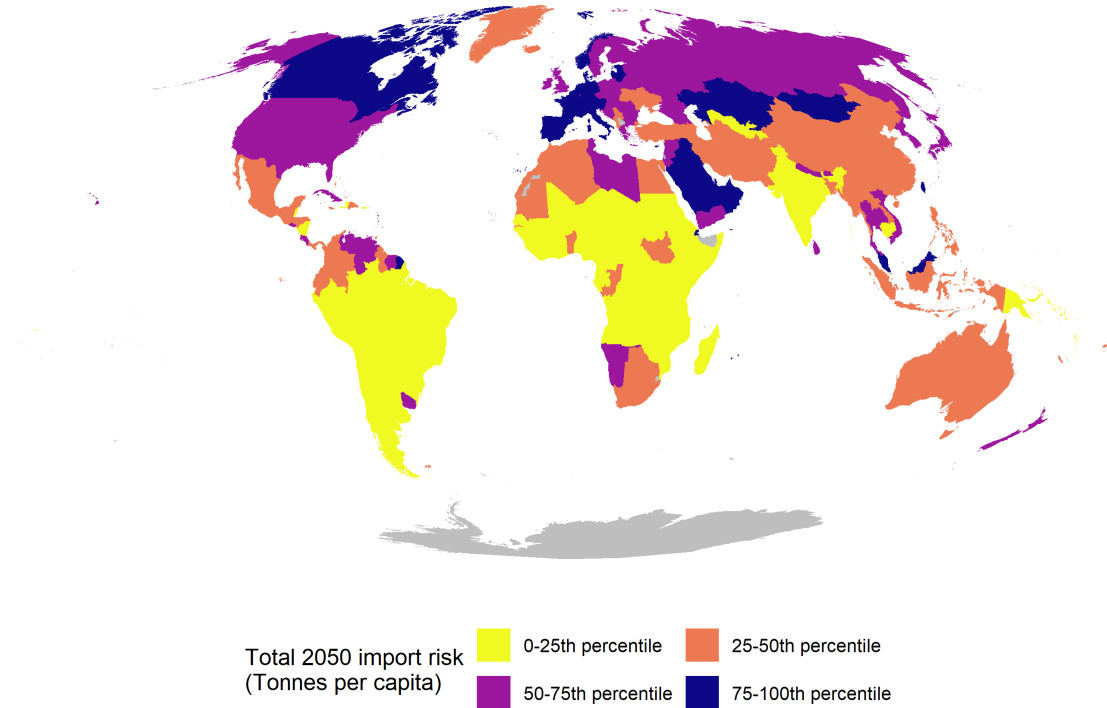


Figure 4: Projected total import risk for 2050 under the RCP 6.0 scenario, assuming a linear relationship between insect pollinator abundance loss and production loss. Projections are based on changes in the mean standardized temperature anomaly based on projections from an ensemble of individual climate models. Import risk is a measure of how the effects of localised production risk might be distributed among other countries, calculated using trade flow data from (35). For each yearly time step into the future, local production risk in each spatial cell is attributed to importers according to the quantity of pollination trade imported from each country. For example, if an importer is dependent on 3 countries for imports, at a proportion of 30%, 50%, and 20%, then any change in import risk should scale as a function of the local production risk aggregated at those same proportions. Total import risk here has been adjusted for country population size and then converted to a percentile. Colours represent each percentile grouping: yellow, 0-25th percentile; orange, 25-50th; purple, 50-75th; dark blue, 75-100th.

165 Interacting effects of climate change and anthropogenic land use are rapidly restructuring biodiversity (26,
 166 55). Here we show that the combination of agricultural land use and recent climate change is associated
 167 with particularly large reductions in the abundance of insect pollinators. We further show that the tropics

will likely experience the greatest risk of future pollination shortfalls, putting at risk the production of crops that depend on insect pollination. The risk to these crops is highest, and predicted to increase most rapidly, in regions of sub-Saharan Africa and northern South America, especially in areas used to produce cocoa, mango, pumpkin, melon, watermelon, and coffee. Given the many factors that determine crop production and crop price (56), the likely effects of insect pollinator losses on crop production are unclear, and even if they do occur, conclusive attribution is likely to be challenging. Such complications likely in part explain why identifying a strong effect of pollinator losses on global crop yield and price has thus far been so difficult (33, 57, 58). Regardless, there is sufficient evidence to suggest that declining insect pollinator abundance will influence crop production risk, particularly for those in the global south (59). Such risk could manifest in the form of direct and immediate losses to crop production through pollinator shortfall, but also through decreased resilience to changes that will happen in conjunction (e.g. the effects of extreme temperature and drought on crop growth). The health, well-being, and livelihoods of billions of people to some extent depends upon the availability and affordability of crops dependent on animal pollination, and therefore on those pollinators (60). Climate change and agricultural land use could put these people’s wellbeing at very high risk.

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313 **Author contributions**

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321 **Competing interests**

322 All authors declare that they have no competing interests.

323 **Data and materials availability**

324 All data will be made available on FigShare, and all code on both Zenodo and GitHub. We also provide
325 a Shiny app to visualise various dimensions of the projections we present in this paper ([https://joemillard.
326 shinyapps.io/pollinator_dependence_visualisation/](https://joemillard.shinyapps.io/pollinator_dependence_visualisation/)).

327 **Supplementary Materials**

328 Materials and Methods

329 Figs. S1-S13

330 Tables

331 References