Agricultural Fast Food: Bats Feeding in Banana Monocultures Are Heavier but Have Less Diverse Gut Microbiota

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Habitat alteration for agriculture can negatively affect wildlife physiology and health by decreasing diet diversity and increasing exposure to agrochemicals for animals foraging in altered landscapes. Such negative effects may be mediated by the disruption of the gut microbiota (termed dysbiosis), yet evidence for associations between habitat alteration, wildlife health, and the gut microbiota remains scarce. We examine the association between management intensity of banana plantations and both the body condition and gut microbiota composition of nectar-feeding bats Glossophaga soricina, which commonly forage within banana plantations across Latin America. We captured and measured 196 bats across conventional monocultures, organic plantations, and natural forests in Costa Rica, and quantified gut microbiome bacterial phylogenetic diversity using 16S rRNA amplicon sequencing. We found that gut microbiota from bats foraging in conventional monocultures were overall less phylogenetically diverse than those from bats foraging in organic plantations or natural forests, both of which were characterized by diverse bacterial assemblages and individualized microbiota. Despite lower diversity, co-occurrence network complexity was higher in conventional monocultures, potentially indicating altered microbial interactions in agricultural landscapes. Bats from both organic and conventional plantations tended to be larger and heavier than their forest counterparts, reflecting the higher food supply. Overall, our study reveals that whilst both conventional monocultures and organic plantations provide a reliable food source for bats, conventional monocultures are associated with less diverse and potentially dysbiotic microbiota, whilst organic plantations promote diverse and individualized gut microbiota akin to their natural forest-foraging counterparts. Whilst the long-term negative effects of anthropogenically-altered microbiota are unclear, our study provides further evidence from a novel perspective that organic agricultural practices are beneficial for wildlife health.

Keywords: co-occurrence networks, Enterobacteriaceae, Glossophaga soricina, habitat alteration, management intensity, organic agriculture, pesticide use
INTRODUCTION

The extent to which agriculture affects the health and physiology of persisting species is currently poorly understood. With deforestation moving at alarming rates—10 million ha lost per year between 2015 and 2020—and agricultural expansion as the main cause behind native habitat destruction (FAO and UNEP, 2020), the effects of agriculture on biodiversity have been widely explored (Dudley and Alexander, 2017). However, species that manage to thrive in agricultural habitats are faced with new challenges, such as dietary changes and, sometimes, exposure to pesticides and hormones (Henriques et al., 1997), which may lead to non-lethal yet detrimental effects to their health and physiology (Mingo et al., 2017). However, recent years have encouraged mixed organic farming practices to increase the sustainability of food production (Eyhorn et al., 2019), which may mitigate the detrimental effects of habitat conversion on wildlife diversity and individual fitness (Stein-Bachinger et al., 2020).

Negative physiological effects of agricultural landscapes may be mediated not only by a general lack of resources or by agrochemicals, but also by the disruption of the gut microbiota, an integral part of an animal’s well-being providing numerous functional benefits to the host (Suzuki, 2017). Beyond essential nutritional services, the gut microbial community influences physiological processes and triggers the immune system, contributing to host health (Brestoff and Artis, 2013). Gut microbiome diversity is shaped to a large extent by the host’s diet (Ingala et al., 2019), in addition to biological (e.g., sex, age) and environmental factors (Amato et al., 2013), yet many beneficial microbes are passed from parents to offspring via vertical transmission, thereby ensuring that they are maintained across the host population (Moeller et al., 2018).

Increasing evidence suggests that anthropogenic changes to natural habitats and associated pollution exposure can reduce the abundance of beneficial microbes, leading to a state called microbial dysbiosis (Petersen and Round, 2014). Dysbiosis is characterized by increased pathogen susceptibility (Murray et al., 2020) and an impaired gut homeostasis that can lead to loss of body condition in adults and slowed development in juveniles (Videvall et al., 2019; Gillingham et al., 2020). This in turn may reduce host fitness (Suzuki, 2017), a mechanism by which human-altered landscapes may act as ecological traps (low quality habitats that are preferred over high-quality ones; Patten and Kelly, 2010). A well-studied example is the effect of industrialization on the human gut microbiome, whereby large-scale shifts in diet and antibiotic exposure over the past 100 years has led to a negative shift in gut microbial composition (dysbiosis; Petersen and Round, 2014) and is hypothesized to contribute to prevalent diseases in industrialized societies (Sonnenburg and Sonnenburg, 2019). The equivalent effects of management intensity on wildlife health are unclear, yet there is some evidence that it modifies gut microbiome composition in some avian species (San Juan et al., 2020) and that agricultural habitats lead to lower gut microbiome diversity (Fackelmann et al., 2021). However, the effects of land use change on the gut microbiota, how this is linked to wildlife health, and specifically the extent to which organic plantations buffer wildlife gut microbiota community changes compared to conventional monocultures, remains unknown.

Several bat species appear to adapt well to agricultural land use changes (Aziz et al., 2015). Pallas’s Long-tongued Bat (Glossophaga soricina, Glossophaginae: Phyllostomidae) is one of the most widespread species in the Caribbean lowland rainforests of Costa Rica, an area that has been, to a large extent, historically converted to banana monocultures. This nectar-feeding species is highly tolerant to anthropogenically modified habitats, inhabiting agricultural areas while maintaining stable populations in the remaining natural forests (Barquez et al., 2015), so it is an ideal choice for exploring the association between the intensity of agricultural habitat alteration, and the bats’ gut microbiota and associated health effects. High food availability and the presence of structures that act as roosts have allowed G. soricina to seemingly thrive in banana plantations, where it feeds mainly from banana nectar. Both conventional and organic plantations provide them with similar banana-dominated diets, yet bats foraging in organic plantations have higher protein intake than conventional monocultures (Alpízar et al., 2020), a diet slightly more like their typical one in natural forests—nectar-carbohydrates complemented with protein from pollen and insects (Alvarez et al., 1978).

In this study, we examine the effects of habitat conversion to conventional banana monocultures and organic banana plantations on the gut microbiota of G. soricina and investigate the relation between management intensity, gut microbiota composition, and body condition. Since diet (Ingala et al., 2019) and pesticides (Kittle et al., 2018) have been shown to affect microbiomes, we expected that individuals with homogeneous diets and exposed to agrochemicals would have altered gut microbiota. Accordingly, we expected: (1) lower alpha diversity indices in conventional monocultures; (2) more similar bacterial composition (beta diversity) between bats foraging in the same habitat; (3) high levels of microbiota homogenization in bats foraging in conventional monocultures compared to their natural forests or organic plantations-foraging counterparts; (4) co-occurrence networks with lower connectivity in bats from agricultural habitats; and (5) that exposure to pesticides in conventional monocultures would lead to lower body condition indicators (smaller, heavier bats) in individuals foraging in conventional monocultures.

MATERIALS AND METHODS

Focal Species, Study Area, and Sample Collection

Our focal species, Palla’s Long-tongued Bat (Glossophaga soricina, Phyllostomidae: Glossophaginae), is the most common nectar-feeding bat in our study areas in the Costa Rican Caribbean lowlands. It is a rather tolerant species to anthropogenic habitat modifications that is easily found in gardens and some plantations, thus classified as least concern by the IUCN (Barquez et al., 2015). Their main dietary items are nectar and pollen, but they complement their diet with insects, fruits, and floral parts. They roosts in tunnels, caves, buildings, hollow trees,
bridges, and sewers (Alvarez et al., 1978). The genus *Glossophaga* tends to have small foraging ranges (3.0 ± 1.0 ha) when food availability is high (Rothenwöhrer et al., 2011). The Costa Rican Caribbean lowlands are characterized by the presence of tropical moist and wet forests (Holdridge et al., 1966). Seasonality is low, with heavy annual rainfall patterns (average 2370–3710 mm) and warm temperatures (average 25–27°C) throughout the entire year and short dry periods (under a month, February and October) (Porras-Peñaranda et al., 2004; Solano and Villalobos, 2012).

We sampled bats in three foraging habitats with different degrees of alterations: organic banana plantations, conventional banana monocultures, and natural forests serving as controls, with two sites per habitat (Supplementary Figure 1). All sampling sites ranged between 50 and 350 ha and had a distance of at least 10 km between each other. The two banana plantation habitats differed in management intensity, agrochemical use, homogeneity level, and size. Organic plantations were mixed with other crops and included native plant species in edges or as corridors, relied on biological pest control methods (i.e., used no pesticides), and had an average size of 50–120 ha; while conventional banana monocultures comprised more than 200 ha of only banana plants and pesticides were applied regularly (27 active ingredients, 47.29 kg/ia/ha/year; Bravo et al., 2013). We sampled bats during 22 months between 2015 and 2018, covering all 12 months of the year to consider seasonal changes. We captured bats using three to eight ground-level mist nets of 6–12 m (Ecotone, Poland) that were set up 1 h after sunset to guarantee bats had fed before capture and the successful acquisition of fecal matter. We collected basic data from each individual (weight, forearm, sex, age, and reproductive state), and marked them using wing punches to avoid resampling. We captured and handled bats following the guidelines of the American Society of Mammalogists (Sikes and Gannon, 2011), with the compulsory Costa Rican permits (#SINAC-SE-CUS-PI-R-095-2016 and #R-003-2017-OT-CONAGEBIO). After capture, we placed bats in sterile cloth bags until they defecated or for a maximum of 1 h. Fecal material was stored in sterile vials with 96% ethanol in a 4°C cooler until transport to a −20°C freezer, where they were stored until processing. Each bat was sampled only once.

### Bacterial DNA Extraction and 16S rRNA Gene Amplicon Sequencing

We followed the protocol detailed by Wasimuddin et al. (2018), to homogenize and extract bacterial DNA using a Nucleo7 Spin 96 Soil Kit (Macherey-Nagel, Germany). During extraction, we included seven extraction controls with only the extraction reagents. We amplified the hypervariable V4 region of the 16S rRNA gene (291 bp) using the primer pair 515F (5'-GTG CCA GCM GCC GCG GTA AA-3') and 806R (5'-GACTACHVGGGTWTCTAAAT-3'). We followed a standardized Fluidigm protocol (Access Array System for Illumina Sequencing Systems, Fluidigm Corporation), in which individual PCR reactions were tagged with a 10-base pair identifier. The PCR (15 µl of volume) was performed as described in detail by Menke et al. (2014). Barcoded samples were purified (NucleoMag bead-based size selection, Macherey-Nagel, Germany) and quantified (DropSense, Trinean, United States) prior to pooling a sample library that was paired-end sequenced in a single run on an Illumina MiSeq platform (2 × 250 bp). We sequenced a total of 269 samples (one per bat individual), seven extraction blanks, and 10 PCR blanks.

### Bioinformatics

Initial sequencing read processing was done using QIME2 (version 2020.6) (Bolyen et al., 2019). We removed low quality sequences and trimmed primers. We truncated our forward and reverse reads to 190 bp due to the decrease of the average quality scores at the end of our sequences. Clustering into amplicon sequence variants (ASVs) and denoising was performed with the DADA2 algorithm (Callahan et al., 2016). We generated and constructed a phylogenetic tree with MAFFT (Katoh and Standley, 2013) and FastTree (Price et al., 2009), and rooted it using an archaeal sequence. Taxonomy was assigned to the ASVs using a pre-trained classifier (McDonald et al., 2012; Bokulich et al., 2018; Robeson et al., 2020) for the Greengenes database version 13.8 as a reference (Second Genome, Inc.). We removed all sequences classified as chloroplast, mitochondria, archaea, Eukaryota, or unclassified at the phylum level. After filtering, 9,763,418 reads from 11,951 ASVs remained.

### Data Analysis

We retained only samples that included over 5000 reads for the analysis, for a final sample size of 196 samples (conventional monocultures: 78, organic plantations: 70, and natural forests: 48). We carried out all analyses in R using the packages phyloseq 1.34.0 (McMurdie and Holmes, 2013), vegan 2.5-6 (Oksanen et al., 2019), and NetCoMi 1.0.2.9000 (Peschel et al., 2021).

### Effects of Foraging Habitat on Microbiota Alpha and Beta Diversity

To estimate alpha diversity within individuals, we rarefied counts to 5,000 reads based on rarefaction curve examination (Supplementary Figure 2). We calculated three alpha diversity indices: number of observed ASVs, Shannon index (weights ASVs by their relative evenness across a community) (Shannon, 1948), and Faith’s Phylogenetic Diversity index (accounts for the ASVs’ phylogenetic diversity) (Faith, 1992). To test the effect of the foraging habitat type whilst controlling for other confounding variables, we built three generalized linear models (GLMMs) with a gamma distribution for each alpha diversity index, using the glm 1.4.2 package (Krudnud, 2017). Initial models included foraging habitat, sex, age, reproductive state, sampling year, and sequencing depth. Model comparison and selection based on AIC was performed using the dredge function from the MuMIn package 1.4.2.9 package (Bartón, 2020).

We assessed beta diversity between individuals through a MDS ordination using rarefied counts and applying three distance measures: unweighted UniFrac (accounts for phylogenetic distance, but only considers presence-absence) (Lozupone et al., 2011), weighted UniFrac (accounts for phylogenetic distance and abundance and is more appropriate to show core microbiota)
(Lozupone et al., 2011; Risely et al., 2020), and Bray-Curtis index (accounts for abundance, but not for phylogenetic distance) (Real and Vargas, 1996). The trend observed in the unweighted Unifrac ordination was similar to that of weighted Unifrac, therefore, we did not consider unweighted Unifrac further in our results. We tested for the effect of foraging habitat type, alongside other potential confounding variables (sex, age, reproductive state, sampling year, and sequencing depth) with permutational multivariate analyses of variance (PERMANOVA, adonis function); and for multivariate homogeneity of habitat dispersion (betadisper function), both from the vegan 2.5-6 package (Oksanen et al., 2019).

**Differential Abundance Analysis of Bacterial Taxa Using TOPICS**

We explored differential abundance using TOPICS, a family of analyses that models differential abundance for co-occurring clusters (“topics”) in sparse and compositional data (Woloszynek et al., 2019). First, ASVs were agglomerated using phyloseq’s tip_glom function ($h = 0.2$) in order to reduce dimensionality of the data, and only agglomerated taxa with over 20% prevalence were retained. We did not use rarefaction because the TOPICS function normalizes data internally. TOPICS analysis was performed using the themetagenomics 1.0.2 package (Woloszynek et al., 2019). Our model looked for structure according to the foraging habitat, with samples from natural forests used as a reference, and corrected according to sequencing depth.

**Co-occurrence Networks**

To construct co-occurrence networks, we applied a method similar to that of Calatayud et al. (2020) which compared bacterial taxa correlations with a null model distribution. Consistent with the TOPICS analysis, ASVs from the rarefied dataset were again agglomerated using phyloseq’s tip_glom function ($h = 0.2$), and only bacterial taxa with over 20% prevalence were retained to ensure statistical robustness ($n = 47$ taxa). We built co-occurrence networks for each foraging habitat type using the NetCoMi 1.0.2.9000 package (Peschel et al., 2021) and assessed significance of each association by comparing the strength of the association (Spearman’s $r$) to a null distribution based on 999 random permutations. Null distributions were generated by constructing 999 random adjacency matrices with the same row and column totals as the real data, using vegan’s nullmodel function with the “quasiswap_count” method to retain zero inflation. We calculated an association matrix based on Spearman’s correlation for each of the 999 random adjacency matrices and generated a null distribution of Spearman’s correlation coefficients for each potential edge in a fully connected network. Edges were considered significant if the original Spearman’s correlation fell outside three standard deviations of the null distribution mean. Edge weight represented the effect size of a one-tailed $t$-test of Spearman’s $r$ against the null distribution. This method is superior to those based solely on correlations, because it can identify and exclude correlations that result from structure alone (i.e., common taxa are statistically more likely to have more significant associations than rarer taxa) than those generated by chance. As such, it also more sensitive to weak but real associations, which would otherwise be discarded if applying just correlations alone to identify associations. We tested the robustness of our results by repeating these analyses when data was normalized by Centered Log Ratio (CLR), and, because sample sizes were unequal across habitat types, we also randomly sub-sampled 40 samples per group and repeated the analysis. Results from both sensitivity analyses were almost identical to those presented.

**Relationship Between Foraging Habitat Type, Body Condition, and the Microbiota**

To determine if an individual bat’s body size and mass are associated with foraging habitat and changes in microbiota alpha diversity and beta dissimilarity, we created two Gaussian-distributed GLMMs predicting forarem size and scale body mass (SBM = mass/forearm), respectively, using the package lme4 1.1-25 (Bates et al., 2019). Initial models included foraging habitat, sex, age, reproductive state, sampling year, sequencing depth, alpha diversity (Shannon index), and beta diversity (first axis for Weighted Unifrac and Bray-Curtis indices). Model comparison and selection based on AIC was performed using the dredge function from the MuMin 1.43.17 package (Bartón, 2020).

We also explored the relationship between bat body condition and common gut microbiota families using TOPICS analysis. For this, we first created a model predicting mass against foraging habitat, reproductive state, and size (forearm), and calculated a residual body mass (RBM) index. Using RBM instead of SBM allowed us to consider the variability from all variables that predicted body mass when comparing it to the gut microbiota composition. We divided the RBM values in three categories (terciles): low $[−3.13, −0.423]$, normal $[−0.423, 0.413]$, and high $[0.413, 5.71]$. The TOPICS analyses looked for structure according to RBM, and we used the normal category as a reference.

**RESULTS**

**Bacterial Composition**

The top ten most abundant bacterial families made up over 65% of the relative abundance of all sampled individuals (Figure 1A). Enterobacteriaceae (phylum Proteobacteria) was the dominant family across combined samples and was especially abundant in bats from conventional monocultures, alongside Burkholderiaceae (phylum Proteobacteria). Samples from organic plantations and natural forests showed higher proportions of Mycoplasmataceae, Moraxellaceae, Streptococcaceae, and Pseudonocardiaceae (phylum Actinobacteria; Figure 1A) and were less dominated by Enterobacteriaceae (Supplementary Table 1).

**Relationship Between Foraging Habitat and Alpha and Beta Diversity**

All alpha diversity indices were lower for bats foraging in conventional monocultures compared to other habitats (Figures 1B–D). Estimated mean observed ASV richness was
FIGURE 1 | Gut microbiome composition and diversity of Glossophaga soricina individuals using different foraging habitats (natural forests, organic banana plantations or conventional banana monocultures). (A) Relative abundance of bacterial families. Bars show the top 20 bacterial families found in the overall microbiome; remaining families are lumped into "Other families." Estimated alpha diversity accounting for sequencing depth, based on (B) number of observed ASVs, (C) Shannon index, and (D) Faith’s Phylogenetic Diversity. Multi-Dimensional Scaling to visualize beta diversity using (E) Weighted Unifrac index and (F) Bray-Curtis dissimilarity index colored by foraging habitat, and (G) Weighted Unifrac index and (H) Bray-Curtis dissimilarity index colored by dominant bacterial family (the most abundant bacterial family in each sample).

50.94 (95% CI [39.72, 65.33]), 53.12 (95% CI [32.80, 86.02]), and 32.76 (95% CI [20.32, 52.82]) for natural forests, organic plantations, and conventional monocultures, respectively, with significant differences in observed richness between conventional monocultures and the two other foraging habitats (Conventional-Organic: $t = 4.50, p < 0.001$; Conventional-Forest: $t = 3.09, p < 0.001$; Supplementary Table 2). Sequencing depth, sampling year, and reproductive state also explained a small amount of variation in some or all alpha diversity indices (Supplementary Table 2).

We next examined the relationship between foraging habitat and microbial beta diversity. Gut microbiota from individuals foraging in natural forests and organic plantations were more similar to each other in overall composition than those from conventional monocultures (Figure 1E) and Bray-Curtis (Figure 1F). The foraging habitat explained approximately 7% of variation in beta diversity across all axes (Supplementary Table 2), yet the major differences between samples from different habitat types lay in their dispersion. Gut microbiota from bats inhabiting conventional monocultures had highly similar phylogenetic composition to each other (average distance to centroid = 0.16, $F = 24.881, p < 0.0001$), yet were highly dispersed when measured by Bray-Curtis distance (average distance to centroid = 0.58, $F = 35.564, df = 2, p < 0.0001$), which does not account for phylogenetic relationships between taxa. Thus, samples from conventional monocultures are heavily dominated by a small number of families (Figures 1G,H and Supplementary Table 1), but these families consist of a large number of disparate ASVs, potentially representing different bacterial species. The separate clusters in the bats from conventional monocultures represented by
Bray Curtis (Figure 1F) represented microbiota dominated by either Enterobacteriaceae or Burkholderiaceae (Figure 1G), and because both belong to Proteobacteria, they cluster together when phylogeny is considered (Figure 1G), but separately when just ASV abundance is considered (Figure 1H). Sequencing depth and year also explained differences in beta diversity (Supplementary Table 2).

**Differential Abundance Analysis Using TOPICS**

To understand which bacterial taxa may be driving differences in beta diversity between foraging habitats, we performed a differential abundance analysis using TOPICS analysis, which identifies groups of co-occurring bacteria associated with different treatment groups (Woloszynek et al., 2019). Co-occurring taxa were grouped to form 15 topics, of which seven were significantly associated with foraging habitat. The largest effect sizes represented the inflation of Enterobacteriaceae, Burkholderiaceae, Xanthomonadaceae, and Chitinophagaceae in conventional monocultures (Figure 2). Whilst the latter three families tended to co-occur together, the abundance of Enterobacteriaceae (the most common family across habitats) was independent of other taxa and formed its own topic.

The microbiota of bats foraging in natural forests, however, were characterized by a far more diverse suite of rare taxa, including Sphingobacteriaceae, Caulobacteriaceae, Actinomycetales (topic 11); Alcaligenaceae, Lachnospiraceae, Porphyromonadaceae (topic 5); and Chlamydiaceae (topic 1) all of which tended to be found at low abundances. Organic plantations were characterized by inflation of Moraxellaceae. Overall, differences between natural forests and organic plantation microbiota were much smaller in their effect sizes than those between conventional monocultures and either natural forests or organic plantations, supporting the beta diversity finding that the gut microbiota of bats from natural forest and organic plantation cluster apart from those of bats foraging in conventional monocultures.

**Relationship Between Foraging Habitat and Co-occurrence Network Structure**

Co-occurrence network analyses were used to assess the effect of foraging habitat on the gut microbiota ecological interactions. According to the number of edges and the average degree, network complexity increased with management intensity (Figure 3), and networks from conventional monocultures also have more negative interactions. The most highly connected taxa were different between each network, with no single node being important across networks, suggesting that whilst microbiota from natural forests and organic plantations are similar in phylogenetic composition, their taxon interactions differ. For the natural forest network, not one taxon was disproportionately connected, with Ruminococcaceae (Firmicutes), Porphyromonadaceae (Bacteroidetes), and Rhizobiales (Proteobacteria) taxa representing the best-connected nodes. In the organic plantation, Actinomycetales (Actinobacteria), Solirubrobacterales (Actinobacteria), and Lachnospiraceae (Firmicutes) were disproportionately important.

For the conventional monoculture network, negative interactions were largely mediated by Chitinophagaceae and Streptococcaceae (both Bacteroidetes), and positive interactions by Lactobacillales, Leuconostocaceae, Ruminococcaceae, and Lachnospiraceae (all Firmicutes). Other than Chitinophagaceae, taxa that were highly abundant in conventional monocultures, such as Enterobacteriaceae, Burkholderiaceae, and Xanthomonadaceae, did not appear to be important in networks.
Relations Between Foraging Habitat, Body Condition, and the Microbiota

Bats foraging in banana plantations had higher SBM and were larger than those foraging in natural forests, with organic plantations supporting particularly large individuals (Figures 4A,B and Supplementary Table 3). To examine associations between body mass and the microbiota, we accounted for foraging habitat by calculating RBM in a model predicting mass against habitat, reproductive state, and size (forearm), and then examined whether any taxa were associated with high or low RBM using TOPICS analysis. From the 15 topics created, only three were significantly related to the bats’ RBM (Figure 4C). We found no taxa associated to bats with low RBM (Supplementary Figure 3), whilst Chlamydiaceae, Xanthomonadaceae, Chitinophagaceae, and Weeksellaceae were associated with normal RBM. Bats with high RBM were associated with Mycoplasmataceae, Lactobacillales, and Campylobacterales. Interestingly, these taxa were more abundant in bats foraging in natural forest than in either banana plantation, even though the first tended to be smaller and lighter.

DISCUSSION

Our study explores differences in gut microbiota diversity, composition, and network complexity between three foraging habitats of the nectar-feeding bat G. soricina differing in the management intensity and examines their effects on gut microbiota composition and body condition. We found that management intensity had strong effects on microbiota diversity. Gut microbiota from bats foraging in conventional monocultures were highly homogenous both across and within samples and differed markedly from bats foraging in natural forest or organic plantation habitats, including in co-occurrence networks, potentially indicating microbial dysbiosis. Interestingly, bats foraging in both types of plantations were heavier and larger than those from natural forests, reflecting the reliable food supply provided by both, therefore bats from monocultures did not demonstrate lower condition indices, as we predicted. Nevertheless, less diverse gut microbiota in bats foraging in conventional monocultures may suggest that these habitats potentially have negative physiological consequences for the animals (e.g., gut inflammation and metabolic disease), and may act as ecological traps. Organic plantations, on the other hand, supported heavier bats that retained more diverse microbiota that were closer to their natural forest-foraging counterparts, indicating that organic practices may maintain high diversity of commensal microbiota.

Following typical community responses in human-modified landscapes (Teyssier et al., 2020) and agricultural habitats (Stein-Bachinger et al., 2020), the gut microbiota of bats foraging in conventional monocultures had lower alpha diversity and was mostly dominated by only few taxa, mainly Enterobacteriaceae and Burkholderiaceae (57.3% vs. 21.5% in organic and 19.06% in forests), we suggest this is caused by dietary changes and
Glossophaga soricina showed a less diverse diet in both types of banana plantations in comparison to natural forests (Alpízar et al., 2020), and such a simplified diet may result in a reduced microbiome diversity (Amato et al., 2013). However, no significant differences were found in the diet diversity between both banana plantation types (Alpízar et al., 2020), indicating that altered gut microbiotas in bats from conventional monocultures might be caused by other sources. Since our focal species typically supplements its diet with insects (Clare et al., 2014), an important protein source, lower insect availability in conventional monocultures (Matlock and de La Cruz, 2002; Markó et al., 2017) certainly leads to strong dietary changes in G. soricina, potentially modifying their gut microbiota composition. In addition, pesticides, especially glyphosate, inhibit several of the bacterial microbiome taxa gut microbiota composition. 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control for any seasonal shifts in composition. In bats, changes to flower and insect availability across the year may influence the gut microbiota. Even if G. soricina has been observed to shift diets between rainy and dry seasons (Willig et al., 1993), the Costa Rican Caribbean lowlands have reduced seasonality compared to temperate regions and even to more seasonal areas inside the tropics. Therefore, we believe that seasonal effects in this species would be small. Another factor that may affect our results are farming practices such as harvesting and levels of pesticide use, for which we did not have information. Variation in plantation harvesting schedules across farms and over the year may generate effects on bat foraging ecology that we cannot detect; for example, this may explain the large variation in individual microbiomes between bats (Figure 1F). In particular, measuring pesticide use and flower availability at sampling sites would help disentangle the mechanisms (diet vs. pesticide use) leading to simplified gut microbiotas in banana plantations.

Our study provides a first insight into the effects of organic and conventional agriculture on animal body condition and gut microbiota. Our results suggest that conventional banana plantations provide a reliable food supply for nectar-feeding bats, yet monocultures simplify the gut microbiota, which may have negative physiological consequences for the animals. In contrast, organic banana plantation management also provides reliable food sources while maintaining diversity within the gut microbiota of nectar-feeding bats. Our results suggest that organic practices represent a sustainable agricultural land use that maintain not only higher levels of biodiversity than monocultures but also microbial health for the bats.

**DATA AVAILABILITY STATEMENT**

The datasets presented in this study can be found in the OSF online repository under doi: 10.17605/OSF.IO/CMN5K.

**ETHICS STATEMENT**

The animal study was reviewed and approved by MINAE, Costa Rica.

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**AUTHOR CONTRIBUTIONS**

PA and MT conceived the objectives and study design with input from SS. PA coordinated the fieldwork, collected all samples, and carried out laboratory analyses in SS’ lab. PA with the supervision of AR, performed the bioinformatic and statistical analyses. PA took the lead in writing the manuscript in consultation with AR, and edited by AR, MT, and SS. All authors agreed on the final version of the manuscript, and different countries collaborated during this study, including a scientist from the country where the study was carried out.

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**SUPPLEMENTARY MATERIAL**

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fevo.2021.746783/full#supplementary-material
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