1   **Structure and function of the bacterial and fungal gut flora of Neotropical butterflies**

2

3   Alison Ravenscraft*[1,2], Michelle Berry[1], Tobin Hammer[3], Kabir Peay[1], Carol Boggs[4]

4

5   [1] Department of Biology, Stanford University, Stanford, CA 94305 USA

6   [2] Current address: Center for Insect Science, University of Arizona, Tucson, AZ 85721 US

7   [3] Ecology and Evolutionary Biology, University of Colorado Boulder, Boulder, CO 80309 USA

8   [4] Department of Biological Sciences, University of South Carolina, Columbia, SC 29208 USA

9

10   *Author for correspondence: aravenscraft@gmail.com; tel 314-791-0738

11

16   **Abstract**

17   The relationship between animals and their gut flora is simultaneously one of the most

18   common and most complex symbioses on Earth. Despite its ubiquity, our understanding of this

19   invisible but often critical relationship is still in its infancy. We employed adult Neotropical

20   butterflies as a study system to ask three questions: First, how does gut microbial community

21   composition vary across host individuals, species and dietary guilds? Second, how do gut flora

22   compare to food microbial communities? Finally, are gut flora functionally adapted to the

23   chemical makeup of host foods? To answer these questions we captured nearly 300 Costa Rican

24   butterflies representing over 50 species, six families and two feeding guilds: frugivores and

25   nectivores. We characterized the bacteria and fungi in guts, wild fruits and wild nectars via

26   amplicon sequencing and assessed the catabolic abilities of the gut flora via culture-based assays.

27   Gut communities were distinct from food communities, suggesting that the gut

28   environment acts as a strong filter on potential colonists. Nevertheless, gut flora varied widely

29   among individuals and species. On average, a pair of butterflies shared 21% of their bacterial

30   species and 6% of their fungi. Host species explained 25-30% of total variation in microbial

31   communities while host diet explained 4%. However, diet was still relevant at the individual

32   microbe level—half of the most abundant microbial species differed in abundance between

33   frugivores and nectivores. Diet was also related to the functional profile of gut flora: compared to

34   frugivores, nectivores' gut flora exhibited increased catabolism of sugars and sugar alcohols and

35   decreased catabolism of amino acids, carboxylic acids and dicarboxylic acids. Since fermented

36   juice contains more amino acids and less sugar than nectar, it appears that host diet filters the gut

37   flora by favoring microbes that digest compounds abundant in foods.

38   By quantifying the degree to which gut communities vary among host individuals,

39   species and dietary guilds and evaluating how gut microbial composition and catabolic potential

40   are related to host diet, this study deepens our understanding of the structure and function of one

41   of the most complex and ubiquitous symbioses in the animal kingdom.

42

43   **Key words:** microbiota, Lepidoptera, diet, feeding guild, symbiosis, catabolism

44 **Introduction**

45 Microbes have been detected in the gut of almost every animal studied to date. This ubiquity is

46 underpinned by the myriad functions these microbes serve: Gut microbes can assist animals with

47 the uptake, synthesis and recycling of nutrients, breakdown of toxic or recalcitrant chemicals,

48 and resistance to pathogens (Dillon and Dillon 2004). Despite their prevalence and importance,

49 our knowledge of how and why these symbiotic communities change across host individuals,

50 species, functional ecological groups, and geographical locations is still in its infancy.

51   At the most basic level, variation in the community composition of the gut flora can

52 result either from exposure of the host to different pools of potential microbial colonists

53 (including microbes transmitted vertically from parents to offspring), or from selective filtering

54 of microbes by the physical and chemical conditions of the gut. Differences in diet or gut

55 physiology between host dietary guilds, host species, and individual host organisms can affect

56 both colonization by, and survival of, microbes in the gut.

57   Across disparate animal groups, host dietary guild has frequently attracted interest as a

58 potential determinant of gut community composition. Different foods contain different microbial

59 flora, and it is often assumed that consumption of foods can directly alter the composition of the

60 gut community by introducing microbial colonists to the gut. This role appears to be most

61 important during initial colonization. Once established the gut community is normally resilient,

62 and most food-borne microbes pass through without becoming residents (e.g. Robinson et al

63 2010b; McNulty et al 2011). In addition to exposing hosts to alternative pools of microbes, diet

64 can alter the composition of both horizontally acquired and vertically transmitted gut microbes

65 by determining nutrient availability in, or affecting the chemical conditions of, the gut habitat

66 (Robinson et al 2010a). Ingestion of a particular nutrient will promote the growth of gut

67   microbes that digest that nutrient, leading to feedbacks between diet and microbial community

68   function. For example, anaerobic fungi use lignocellulose as a food source; consumption of high

69   fiber foods by ruminant species leads to greater abundance of these fungi than in non-ruminants

70   (Gordon and Phillips 1998; Solomon et al 2016). These two mechanisms—food-borne microbial

71   colonists and the chemical composition of the diet—are generally expected to result in similar

72   gut flora among hosts that eat similar foods. Indeed, in many systems, gut community

73   composition has been shown to be more similar between species that eat similar diets, even when

74   these species belong to evolutionarily divergent groups (Colman et al 2012; Delsuc et al 2014).

75   For example, herbivorous, omnivorous, and carnivorous mammals and fish host gut communities

76   that are more similar within than between these three diet categories (Ley et al 2008a; Muegge et

77   al 2011; Sullam et al 2012). In insects, species belonging to the wood-feeding and detritivorous

78   guilds both host convergent gut flora (Colman et al 2012).

79   But host dietary guild is only one of many factors that may affect gut microbial

80   community composition. Differences in colonization can also result from living in different

81   habitats (e.g. Xiang et al 2006; Belda et al 2011), which results in sampling of divergent

82   microbial pools by hosts. Additionally, the chemical conditions of the gut vary among species,

83   independent of diet, and may act as a filter that selects for a specific gut community (Rawls et al

84   2006). For example, the guts of small animals, such as insects, range from fully aerobic to

85   anaerobic and therefore favor different sets of microbes (Johnson and Barbehenn 2000).

86   Similarly, gut pH varies among species and likely determines which microbes can establish in

87   the gut (Beasley et al 2015).The host immune system can also regulate potential gut colonists

88   (McFall-Ngai 2007; Salzman et al 2009).

89   Gut flora vary among host species not only in the composition of the core community,

90   but also in the degree of variation among individuals. Such intraspecies variation can result from

91   inter-individual differences in all of the factors mentioned above, including individual hosts'

92   diets, microbial source pools, gut chemistries, and immune systems. Species with the least

93   variable communities are often those that depend on their gut flora to survive on particularly

94   poor or recalcitrant diets, such as termites or herbivorous ants; these species also often have

95   elaborately structured guts that both facilitate digestion and provide microenvironments that

96   support a stable gut community (Engel and Moran 2013). In contrast, omnivorous hosts with

97   simply-structured, tube-like guts, such as *Drosophila*, tend to have gut flora that vary more

98   among individuals (Engel and Moran 2013). In general, gut membership is expected to be more

99   stable when the gut flora serve a more crucial function for their host.

100   Insects offer a tractable system to understand the causes and consequences of variation in

101   gut microbial communities due to their variation in dietary guild, their abundance and species

102   diversity, and their often relatively simple gut flora. Further, insects are major primary

103   consumers, pollinators, and disease vectors in terrestrial ecosystems. Their gut flora may

104   therefore have a large, yet hidden, impact on ecosystem function (e.g. Nardi et al 2002). Indeed,

105   the ability of many insects to subsist on nutritionally unbalanced or recalcitrant foods often stems

106   directly from contributions of their gut flora (e.g. Warnecke et al 2007). Despite the importance

107   of, and increasing interest in, the insect gut flora in general, the adult lepidopteran microbiome

108   has been largely ignored. Although a few culture-based studies confirmed the presence of

109   bacteria in the adult gut decades ago (Steinhaus 1941; Kingsley 1972), culture-independent

110   profiling is only beginning. Such data exist for only a single species, *Heliconius erato*

111 (Nymphalidae), and indicate that about 12 bacterial species dominate the adult gut (Hammer et al

112 2014).

113   In contrast to adults, the gut flora of several larval lepidopterans—particularly those of

114 economic importance—have been characterized (Broderick et al 2004; Robinson et al 2010b;

115 Pinto-Tomás and Sittenfeld 2011). However, the adult gut flora differ from those of larvae

116 (Hammer et al 2014). This difference may derive from the fact that during pupation, the contents

117 of the guts are voided, antimicrobial peptides are secreted into the gut lumen, and gut itself is

118 replaced, presumably eliminating a large portion of the larval gut flora (Russell and Dunn 1996;

119 Hakim et al 2010; Johnston and Rolff 2015). Indeed, in a culture-based study, Kingsley (1972)

120 found that the density of colony-forming units derived from the guts of monarch butterflies

121 (*Danaus plexippus*) decreased 1000-fold between pupae and freshly emerged adults. Existing

122 data on larval lepidopterans therefore do not shed light on the gut flora of adult butterflies.

123 Butterflies are biologically and economically important as herbivores and pollinators, and in

124 areas outside of gut community ecology they are well-studied model organisms in the fields of

125 population and nutritional ecology (Boggs et al 2003). Characterization of the causes and

126 consequences of variation in the adult lepidopteran gut flora will therefore pave the way for

127 connections between microbiome research and classical questions in ecology and evolutionary

128 biology.

129   Here we use adult butterflies as a novel study system to characterize patterns of variation

130 in the gut flora at the level of host individuals, species, and feeding guilds. We focus on how host

131 nutritional ecology affects the community composition and functional capacity of the gut flora.

132 The Neotropical community of butterflies allows for a high degree of replication both within and

133 between species while controlling for geographic origin. Neotropical butterflies also exhibit

134   well-described ecological variation that likely structures the gut community, particularly in

135   feeding behavior: adults of some species feed on nectar, while others feed on the juice of rotting

136   fruits. Adults are easily captured in the wild, permitting examination of the natural gut flora and

137   avoiding the microbial community shifts frequently observed in captive animals (e.g. Chandler et

138   al 2011). These characteristics make Neotropical butterflies a useful system to elucidation

139   patterns in the composition and function of the gut microbial community.

140   We asked: (1) What bacteria and fungi are present in the guts of adult butterflies? (2)

141   How does gut microbial community composition vary among individuals, species, and dietary

142   guilds? (3) How does gut microbial composition compare to that of butterfly foods? (4) How

143   does community composition translate into functional abilities—specifically, catabolism of

144   various carbohydrates and amino acids?

145   Based on previous studies of other animal taxa (Ley et al 2008a; Colman et al 2012), we

146   expected butterfly gut communities to cluster according to both host diet and host species, such

147   that microbial beta diversity would be lowest between conspecific hosts, intermediate between

148   hosts that belong to different species but the same feeding guild, and highest between

149   heterospecific hosts that belong to different feeding guilds. We hypothesized that we would

150   observe signals of environmental acquisition of the gut flora. Specifically, we predicted that most

151   of the gut community would be a subset of the microbes present in the host's food, and that

152   frugivores, which feed on more "contaminated" foods (e.g. rotting fruit), would have more dense

153   gut communities than nectivores. Finally, we hypothesized that the digestive abilities of the gut

154   community would differ between feeding guilds, and that gut community catabolism should be

155   highest for nutrients abundant in the host's diet.

156

8

157   **Methods**

158   *Study site*

159   Butterflies were collected at La Selva Biological Research Station (10° 26'N, 83° 59' W). La

160   Selva is located in the lowlands on the Caribbean versant of the Cordillera Central in Costa Rica.

161   The station owns 1,600 hectares, 55% of which is primary tropical wet forest; the remainder

162   comprises secondary forest and disturbed habitats (e.g. abandoned pastures and plantations) in

163   various states of regrowth. This habitat diversity allowed us to collect a large diversity of

164   butterfly species while controlling for the geographic origin of our samples. La Selva also

165   provided the facilities necessary for both the culture-dependent and -independent components of

166   our project (e.g. a wet lab and laminar flow hood). From January to March 2013, we captured

167   butterflies and sampled butterfly foods along the trail system in primary forest, secondary forest

168   and recovering pastures. We returned to La Selva in March 2014 to collect additional food

169   samples (fruits and nectars).

170

171   *Study organisms*

172   Neotropical lepidopterans belong to two main feeding guilds: nectar feeders and fruit feeders.

173   The Papilionidae, Pieridae, Lycaenidae, Riodinidae, Hesperiidae, and some Nymphalidae feed

174   primarily on flower nectar, while several subfamilies of the Nymphalidae—Satyrinae,

175   Morphinae, Charaxinae, and some members of the Nymphalinae—feed primarily on rotting

176   fruits or other non-floral liquids (DeVries 1987, DeVries 1988). Our sampling included

177   representatives of all of the nectivorous families—though most individuals derived from the

178   Pieridae and Nymphalidae—as well as representatives from all the frugivorous nymphalid

9

179   subfamilies described above (Supplementary Table 1). Three *Heliconius* (Nymphalidae:

180   Heliconiinae) species were sampled; this genus supplements its nectivorous diet with pollen.

181

182   *Sample collection and processing*

183   We used aerial nets to catch both nectivorous and frugivorous species. To capture frugivores we

184   also used Van Someren-Rydon traps (Daily and Ehrlich 1995) baited with a fermented mixture

185   of fruit (primarily mango), molasses and rum. Traps were checked at least once every two days.

186   Upon capture, butterflies were placed in glassine envelopes and transported to the lab.

187   We identified butterflies to species using *The Butterflies of Costa Rica and Their Natural History*

188   (Devries 1987). Butterflies were fed a 5 to 10 microliter droplet of a filter-sterilized solution of

189   sugar and nigrosin dye (Sigma 198285) to assist with visualization of their guts during

190   dissection. Butterflies were euthanized with ethyl acetate, washed with 70% ethanol, and

191   dissected in a laminar flow hood to minimize contamination with environmental bacteria and

192   fungi. Guts were removed and homogenized in a small amount of inoculating fluid (described

193   below) in a sterile 2 mL tube. A portion of the homogenate was cultured to assess gut community

194   catabolic profile (described below). The remainder of the gut homogenate was suspended in cetyl

195   trimethyl ammonium bromide (CTAB) and refrigerated at 4°C until transportation to Stanford

196   University for DNA extraction. CTAB is an effective preservative for storage of insect samples

197   prior to microbiota analysis (Hammer et al 2015).

198   We also collected samples of the butterflies' foods to characterize potential source pools

199   of microbial gut communities. Trap baits were sampled in 2013 and 2014 by suspending

200   approximately 0.25 to 0.5 mL of rotten fruit slurry in CTAB. In 2014, nectar samples were also

201   collected. We used capillary tubes to extract nectar from flowers in the field, obtaining between

10

202   0.5 and 10 μL per flower. Nectars were also preserved in CTAB. We only collected nectar from

203   species at which we directly observed at least one instance of butterfly feeding. In 2013, we

204   sampled fruits of *Dipteryx oleifera* (Fabaceae) by scraping flesh from the surface of the fruit into

205   CTAB. In 2014, we rinsed the fruit with 5 mL of sterile water and preserved 0.25 to 0.5 mL of

206   the rinsate, since butterflies only feed on surface juices and do not consume the pulp. *Dipteryx*

207   *oleifera* is the second-most abundant leguminous tree in primary forest at La Selva (McDade et

208   al 1994) and was the only wild fruit we observed butterflies feeding upon during both visits to

209   the site.

210   Research was carried out under permit numbers 202-2012-SINAC and R-016-2014-OT-

211   CONAGEBIO.

212

213   *Community catabolic profiling*

214   A portion of each gut homogenate was seeded into 10 mL of inoculating fluid (IF-A, Biolog

215   Catalog No. 72401) and cultured in microbial catabolic phenotype plates (Gen III microplates,

216   Biolog Inc., Hayward, CA). The volume of gut homogenate used depended on the size of the

217   butterfly: we used 25 microliters for most animals, but took less volume from the largest species

218   (whose gut homogenate was more concentrated, since the guts themselves were large), and more

219   volume from the smallest species (whose gut homogenate was less concentrated, since the guts

220   were small). This was meant to correct for the grossest differences in inoculation density; further

221   corrections are described below.

222   The catabolic phenotyping plates allowed simultaneous assessment of catabolism of 71

223   carbon and nitrogen sources (e.g. glucose, fructose, ammonia, uric acid; Table S5) via a

224   colorimetric assay. Plates were incubated at ambient temperature. Absorbance at 590 nm was

11

225   measured twice per day using a 96-well plate reader (Chromate-4300, Awareness Technology

226   Inc., Palm City, FL) until color remained stable (5 days on average).

227   To reduce bias due to differences in inoculum density between the samples, the data were

228   standardized as in Garland et al (2001). Briefly, based on visual inspection of the distribution of

229   plate averages, we chose a target average absorbance value per plate of 190. For each plate, we

230   selected data from the timepoint at which the average absorbance was closest to this target.

231   These measurements were standardized by subtracting the value of the negative control and

232   dividing by the plate's average absorbance.

233

234   *Microbial community characterization*

235   Samples were shipped to Stanford University, where they were homogenized via bead beating,

236   extracted with chloroform, and cleaned using the DNeasy Blood and Tissue Kit (Qiagen,

237   Germantown, MD). Our DNA extraction protocol is described in detail in Peay et al (2007).

238   Bacterial DNA was amplified with primer set 515f (5'-GTGCCAGCMGCCGCGGTAA-

239   3') and 806r (5'- GGACTACHVGGGTWTCTAAT-3'), which amplifies the V4 hypervariable

240   region of the 16S rRNA with few taxonomic biases (Bergmann et al 2011). The bacterial

241   amplicons were indexed with barcoded forward and reverse primers to (Caporaso et al 2012;

242   Kozich et al 2013). The PCR reaction contained 0.2 μM forward primer, 0.2 μM reverse primer,

243   0.2 mM dNTP, 0.65 U OneTaq HotStart (New England Biolabs) and 1X Thermopol buffer (New

244   England Biolabs) in a volume of 25 μL. The thermocycler program began with denaturation at

245   94 C for 3 minutes followed by 35 cycles of denaturation at 94 C for 45 seconds, annealing at 50

246   C for 60 seconds, and extension at 68 C for 90 seconds, with a final extension of 68 C for 10

247   minutes (adapted from the Earth Microbiome Project protocol; Gilbert et al 2014).

248   Fungal DNA was amplified with primers ITS1f (5'-

249   CTTGGTCATTTAGAGGAAGTAA-3') and ITS2 (5'-GCTGCGTTCTTCATCGATGC-3')

250   (White et al 1990; Gardes and Bruns 1993). The reverse primer was barcoded to index the

251   samples (Smith and Peay 2014). The PCR recipe was identical to the bacterial PCR recipe above.

252   Thermocycler conditions were denaturation at 95 C for 1 minute followed by 35 cycles of 94 C

253   for 30 seconds, 52 C for 60 seconds, and 68 C for 60 seconds, with a final extension of 68 C for

254   5 minutes. Each bacterial and fungal amplification was run in triplicate to minimize the effects of

255   stochastic amplification.

256   Bacterial and fungal PCR products were cleaned with AMPure XP magnetic beads

257   (Agencourt A63881) and the final concentration of each sample was quantified fluorometrically

258   with a 96-well plate reader (Qubit dsDNA HS kit, Thermo Fisher Q32854). An equal mass of

259   DNA from each sample was added to the bacterial or fungal library, respectively. Bacterial and

260   fungal libraries were sequenced in separate runs on an Illumina MiSeq platform at the Stanford

261   Functional Genomics Facility (Stanford, CA), with 2 x 250 chemistry for bacteria and 2 x 300

262   chemistry for fungi.

263

264   *Illumina sequence data processing and cleaning*

265   We used the program cutadapt (Martin 2011) to remove priming sites and poor quality bases at

266   the 5' and 3' ends of the sequences. Sequences were merged and clustered at a 97% similarity

267   cutoff with UPARSE (Edgar 2013). Both de-novo and reference-based chimera checking were

268   performed in UPARSE; bacterial reads were compared to the RDP Gold database and fungal

269   reads to the UNITE database. Bacterial taxonomy was initially assigned using the RDP classifier

270   (Wang et al 2007) with Greengenes (McDonald et al 2012) as the training set. We used the RDP

13

271   classifier with the Warcup (Deshpande et al 2016) training set to assign fungal taxonomy. We

272   checked and revised these assignments by aligning our representative sets of bacterial and fungal

273   sequences against the NCBI nucleotide collection using BLAST.

274   OTUs that were identified as lepidopteran 18S, archaeans, mitochondria, or chloroplasts

275   were excluded. (These accounted for less than 0.1%, less than 0.1%, 0.6%, and 4.8% of the raw

276   total reads, respectively, and 98% of the chloroplast reads derived from samples of butterflies'

277   foods.) Additionally, DNA extraction kits and other laboratory reagents are known to contain

278   microbial DNA that can contaminate microbiome analyses (Salter et al 2014). We found fourteen

279   bacterial OTUs that were present at higher abundance in the negative controls than the butterfly

280   gut samples (Supplementary Table 3); these were classified as contaminants and removed from

281   the dataset prior to statistical analyses. The fungal dataset (especially nectivore samples)

282   contained many ectomycorrhizal lineages that were unlikely to be members of the butterfly gut

283   community. This suggested that the butterfly gut samples (particularly those of nectivores) had

284   low fungal biomass, resulting in detection of ambient lab and environmental contaminants. We

285   removed lineages that were likely contaminants from the analysis: we omitted 135 OTUs

286   assigned to the genus *Mortierella*, which are decay fungi that inhabit soil. Wood decomposers

287   and ectomycorhizal, arbuscular or ericoid mycorrhizal fungi were also removed from the dataset,

288   as were lineages that were present at higher abundance in the negative controls than in the

289   samples (Supplementary Table 4). We retained lineages that were known phylloplane

290   inhabitants, plant pathogens, and fruit decomposers, as well as most ascomycete and

291   basidiomycete yeasts.

292   Frugivorous butterflies were captured either with traps baited with fermented fruit or with

293   aerial nets. For those captured in baited traps, a portion of the gut flora likely derived from

14

294   microbes in the trap baits. To identify these microbes, we tested for differential abundance of

295   bacterial and fungal OTUs between trapped frugivores and netted frugivores using the *mt*

296   function in the R package "phyloseq" (McMurdie and Holmes 2013), which includes an FDR

297   correction for multiple testing. No bacterial OTUs differed in abundance between netted and

298   trapped animals, but three fungal OTUs were significantly more abundant in the trapped

299   butterflies. These OTUs were all strains of the yeast *Kazachstania exigua*, which was also by far

300   the most abundant fungus detected in the traps. These three OTUs accounted for 43.7% of all

301   fungal reads from the fermented baits and only 0.3% of reads from wild fruits. We concluded

302   that the OTUs had been introduced into the trapped butterflies' guts via the fermented baits and

303   removed them from the data derived from butterfly samples.

304   To control for differences among samples in sequencing depth, we rarefied the

305   sequencing data using phyloseq. Bacteria were rarefied to 1000 and fungi to 200 sequences per

306   sample. These cutoffs allowed us to retain as many samples in the dataset as possible while still

307   profiling the dominant microbes. All sequencing results reported are based on rarefied data,

308   unless otherwise noted.

309

310   *Quantification of total bacterial abundance*

311   Bacterial DNA was quantified with qPCR using SYBR green fluorescent chemistry (iCycler IQ,

312   Bio-Rad, Hercules, CA). Since the 515f/806r primer set sometimes amplifies butterfly 18S in

313   addition to bacterial 16S rRNA, we designed a PNA clamp to block 18S amplification (Lundberg

314   et al 2013). The clamp sequence was GCCCGCTTTGAGCACTCT and it was synthesized by

315   PNA Bio (Thousand Oaks, CA). The reaction volume was 20 μL and the recipe was 7.5 μM

316   PNA clamp, 0.2 μM 515f primer, 0.2 μM 806r primer, and 1X PerfeCTa SYBR Green FastMix

15

317   for iQ (Quanta Biosciences, Gaithersburg, MD). Thermocycler settings were an initial 10-minute

318   denaturation at 95 C followed by 45 cycles of denaturation at 95 C for 15 sec, PNA clamp

319   annealing at 76 C for 10 sec, primer annealing at 50 C for 30 sec, and extension at 68 C for 30

320   sec. To check that amplified fragments were the expected length, we performed a melt curve

321   ramping from 55 C to 95 C in 0.5 C increments at 10 second intervals. PCR products were also

322   visualized using gel electrophoresis.

323   To calculate the starting number of 16S rRNA copies, each sample's threshold cycle was

324   compared to an internal standard curve ranging from an initial 10 to 10^7 copies per μL of *E. coli*

325   16S rRNA amplicons. (Generation of these standards is described below.) Each sample and

326   standard was run in triplicate and the results were averaged. The correlation coefficients of the

327   standard curves for all qPCR plates ranged from 0.973 to 0.988.

328   Since a fraction of each sample was removed for culture-based catabolic profiling prior to

329   DNA preservation and extraction, the qPCR estimates were corrected accordingly. For example,

330   if 50% of the gut homogenate from a given sample was removed prior to DNA preservation, that

331   sample's qPCR estimate was multiplied by two to obtain the estimated total 16S rRNA count for

332   the entire gut.

333   The 16S rRNA standard curve was generated as follows: 16S rRNA was amplified from

334   *E. coli* using primers 27f and 1492r, ligated into a plasmid vector, and cloned into chemically

335   competent cells. Colonies were screened for inserts of the expected size by PCR amplification

336   using M13f and M13r primers followed by gel electrophoresis. To further screen for the correct

337   insert, several colonies were bidirectionally sequenced. One colony that passed both screens was

338   selected for the standard curve. This clone was grown to saturation in LB + kanamycin selective

339   media. Plasmids were extracted using the Qiagen Plasmid Mini kit and linearized using the

340   restriction enzyme SpeI (FastDigest, Thermo Fisher FD1253). DNA was purified using a Qiagen

341   PCR cleanup column and run on a gel to verify complete linearization. DNA concentration was

342   quantified with PicoGreen and copy number per µL was calculated as the molecular weight of

343   plasmid plus insert (g/molecule, calculated as length in bp * 660 Da/bp / $6.02*10^{23}$) divided by

344   the measured DNA concentration (g/µL). Via serial dilution of the raw extract with sterile TE,

345   we generated a standard curve ranging from 10^7 copies/µL to 10 copies/µL.

346   We were not able to quantify total fungal abundance because our ITS primers amplified

347   fragments of vastly different lengths (ranging from 250 base pairs to over 1000 bp).

348

349   *Statistical analyses*

350   All models had Gaussian error structure and were fit in R version 3.2.2 (R Core Team 2015).

351   Models without random effects were fit using the *lm* command in the "stats" package. Models

352   with random effects were fit using the *lmer* command in the package "lme4" (Bates et al 2014).

353

354   *a. Variation in total bacterial load among host species and feeding guilds*

355   A log transformation was applied to the 16S rRNA counts prior to analysis. We used linear and

356   linear mixed effect models (Zuur et al 2009) to assess whether total bacterial load differed

357   among host species or between host feeding guilds. We used a likelihood ratio test to compare

358   models with and without a random effect of host species, including fixed effects of host diet,

359   wing length (as a proxy for the animal's size), and capture date. We then tested for significance

360   of the fixed effects using backwards model selection with likelihood ratio tests, including a

361   random effect of host species in all models.

362

17

363   *b. Variation in OTU richness*

364   To test whether observed bacterial and fungal richness differed among host species, we again

365   used a likelihood ratio test to compare models with and without a random effect of host species,

366   including a fixed effect of host diet to control for overall diet-based differences. We tested for

367   correlation between host diet and observed bacterial and fungal richness by comparing models

368   with a fixed effect of diet to models without the fixed effect; both models had a random effect of

369   host species. We omitted data from host species represented by a single individual from these

370   analyses.

371

372   *c. Variation in microbial community composition*

373   *Between feeding guilds:* To test for a relationship between butterfly feeding guild and gut

374   microbial community composition, we used ordination (NMDS) plots to visualize and

375   perMANOVAs to test for dissimilarity in microbial community composition (Anderson 2001).

376   Both operations were performed on the Bray-Curtis dissimilarities between rarefied samples. We

377   performed two perMANOVAs (the *adonis* test in the R package "vegan"; Oksanen et al 2015) to

378   test for dissimilarity in bacterial and fungal communities, respectively. These tests included

379   terms for both host diet and host species, thus simultaneously testing whether gut communities

380   differed between host species and feeding guilds. Host species represented by a single individual

381   were omitted from these analyses.

382   If communities differ in variance, perMANOVA results can be unreliable (Anderson and

383   Walsh 2013). To verify the results of the tests, we therefore tested for differences in dispersion

384   among feeding guilds using the *betadisper* function from the package "vegan." For bacteria, the

385   feeding guilds did not differ in dispersion (anova: *df* =1, F=0.12, p=0.73) therefore the data

18

386   conformed to the assumptions of perMANOVA. For fungi, the feeding guilds did differ in spread

387   (anova: df =1, F=4.4, p=0.04). However, after removing the five host species with extreme

388   dispersions, the feeding guilds no longer differed in spread (anova: df =1, F=1.3, p=0.21). The

389   perMANOVA results with these five species removed were qualitatively identical and

390   numerically similar to those obtained from all host species, therefore we report the test results for

391   all species (omitting those represented by a single individual).

392   In order to understand the effect of host diet on individual OTU abundances, we modeled

393   the relative abundance of a bacterial or fungal OTU (counts out of 1000 or 200, respectively) as a

394   function of the interaction between host diet and OTU identity, with a random effect of host

395   species. Since it was not feasible to model abundances of every OTU, we selected the 20

396   bacterial or 20 fungal OTUs that were present at the highest abundances in the pooled data and

397   were detected in at least 10% of either frugivore or nectivore samples. Prior to the analyses,

398   abundance data were transformed by log(x+1) to homogenize variance.

399   *Among host species:* We used perMANOVAs (described above) to test for dissimilarity

400   in the gut flora among host species while accounting for differences between feeding guilds.

401   Since perMANOVA assumes homogeneity of variance among communities, we again used the

402   *betadisper* function to test for differences in dispersion. Host species' bacterial and fungal

403   communities did differ in dispersion (bacterial anova: df =36, F=2.4, p<0.001; fungal anova: df

404   =28, F=11.4, p<0.001). For bacteria the difference was driven by five species that had unusually

405   low variance; when these were removed, the remaining 32 species no longer differed in spread

406   (anova: df =31, F=0.9, p=0.65). For fungi, the difference was also driven by five extreme

407   species. The remaining 24 species in the fungal dataset no longer differed in variance (anova: df

408   =23, F=1.3, p=0.16). For both bacteria and fungi, we re-ran the perMANOVA tests using the

19

409   subset host species that did not differ in dispersion. Results were again qualitatively the same as

410   those for all host species, so we report the results for all species.

411   *Among individuals:* To quantify beta diversity among individual butterflies' gut flora, we

412   calculated the median percentage of OTUs shared between individuals, where the percentage

413   shared between individuals A and B was calculated as 0.5 * (number of A's OTUs shared by B /

414   total OTUs in A + number of B's OTUs shared by A / total OTUs in B).

415   *Between food microbial communities and gut communities:* We used NMDS plots (on

416   Bray-Curtis distances between rarefied samples) to visualize differences between the butterfly

417   gut flora and the microbial composition of butterfly foods. To test whether gut flora were more

418   similar to the microbial communities of one food than to another food, we calculated the

419   pairwise Bray-Curtis distances between each butterfly gut community and each food community

420   and performed t-tests on the per-butterfly averages of these distances for the three foods (fruits,

421   nectars and baits). To correct the resulting p-values for multiple testing, we used the Benjamini-

422   Hochberg false discovery rate (FDR) method as implemented by the *p.adjust* function in R

423   (Benjamini and Hochberg 1995).

424

425   *d. Variation in microbial community function*

426   The Hellinger transformation was applied to the standardized catabolic data prior to all analyses.

427   *With gut community composition:* To test whether differences in microbial community

428   function were correlated with differences in microbial species composition, Mantel tests were

429   performed between the Bray-Curtis distances between bacterial or fungal microbial communities

430   and the Euclidean distances between catabolic profiles. Since trap bait-derived fungal OTUs

431   (three strains of *Kazachstania exigua*) were introduced into the Biolog plates via frugivores'

20

432   guts, these OTUs were included in the calculation of Bray-Curtis distances between fungal

433   communities. However, results were qualitatively equivalent when these three OTUs were

434   omitted from the Bray-Curtis calculations.

435   *With host species and diet:* First, we tested for correlation between gut community

436   catabolic profile and both host diet and host species by performing a perMANOVA on the

437   Euclidean distances between the gut catabolic profiles. (Singleton host species were omitted

438   from this analysis.) To investigate differences between frugivores and nectivores in more detail,

439   we next modeled substrate catabolism as a function of the interaction between host diet and

440   substrate identity, including a fixed effect for the log of the number of 16S rRNA copies in the

441   host's gut to control for differences in bacterial density, and a random effect for the identity of

442   the host butterfly to control for overall differences in catabolic activity of the gut community

443   among individuals. Finally, to test whether catabolism of certain classes of substrate (for

444   example, amino acids or sugars) was consistently up- or down-regulated with host diet, we

445   modeled substrate catabolism as a function of the interaction between host diet and substrate

446   class, with a fixed effect for log(16S rRNA counts) and a random intercepts of host identity as

447   before, plus an additional random intercept and random diet effect (slope term) for each

448   substrate. Thirteen substrates belonged to classes with three or fewer substrates and were

449   therefore omitted from the class-level analysis. For both models, least-square means and the

450   contrasts between them were calculated using the *lsm* command in the "lsmeans" package (Lenth

451   2016) to assess the whether a substrate or substrate class was catabolized differently between

452   frugivores and nectivores. The resulting p-values for each substrate or substrate type,

453   respectively, were FDR corrected to control for multiple testing.

454

21

Results

Below we first describe the overall composition of the adult butterfly gut flora, then address patterns of variation in microbial community composition at the level of host feeding guild, host species, and host individual. Gut communities are then compared to the microbes present in butterfly foods. Functional differences between gut microbial communities are described last.

1. Bacteria and fungi present in the butterfly gut

We sequenced the bacterial flora of 306 butterflies, of which 290 were retained after rarefaction, and the fungal gut communities of 247 butterflies, 161 of which were retained after rarefaction. (See Table 1 for a full sample size breakdown and Supplementary Table 1 for counts per butterfly host species). After sequence processing, quality filtering, and removal of contaminant OTUs, 7.2 million bacterial sequences and 1.5 million fungal sequences were obtained from butterfly guts. After rarefying, a total of 958 bacterial OTUs and 880 fungal OTUs were observed across all butterflies. (Note that numbers of bacterial and fungal OTUs cannot be compared because bacteria and fungi were rarefied to different read depths.) Individual butterflies hosted a mean of 30 (interquartile range:18-37) bacterial OTUs (number observed at a 1000 read per sample cutoff) and 21 (IQR: 12-28) fungal OTUs (number observed at a 200 read per sample cutoff).

The most prevalent bacterial phyla and classes across all butterflies were Proteobacteria:Gammaproteobacteria (40% of all butterfly-derived bacterial sequences), Proteobacteria:Alphaproteobacteria (28%), Firmicutes:Bacilli (12%), Bacteroidetes:Flavobacteriia (7%), and Tenericutes:Mollicutes (4%). The dominant fungal class was Saccharomycetes (62% of all butterfly-derived fungal sequences) in the phylum

22

478   Ascomycota. All of these Saccharomycetes belonged to the order Saccharomycetales, which

479   defines the "budding yeasts" or "true yeasts." Other common fungal clades were

480   Basidiomycota:Tremellomycetes (8%), Zygomycota:Mucoromycotina (6%),

481   Ascomycota:Dothideomycetes (6%) and Basidiomycota:Microbotryomycetes (5%).

482   The 20 most abundant bacterial and fungal OTUs across all butterflies (excluding those

483   detected in less than 10% of frugivores or nectivores) are listed in Tables 2 and 3. The

484   predominant bacteria included known associates of butterflies (*Orbus* spp.; Kim et al 2013;

485   Hammer et al 2014) and other insects with sugar-rich diets (e.g. *Erwinia* sp., *Asaia* sp.*,*

486   *Commensalibacter intestini;* Crotti et al 2010), putative insect parasites and pathogens (e.g.

487   *Spiroplasma* sp., *Serratia* spp., *Wolbachia* sp.), common gut inhabitants (e.g.

488   *Klebsiella*/*Enterobacter, Vagococcus/Enterococcus*), and common fermenting bacteria (e.g.

489   *Lactococcus* sp., *Acetobacter* sp.). The most abundant fungi were largely ascomycotous yeasts

490   (especially members of the genera *Hanseniaspora*, *Pichia*, *Kazachstania* and *Candida*), many of

491   which are known to associate with insects and/or plants— particularly nectars, decaying fruits,

492   and beetle guts (Suh et al 2005; Kurtzman et al 2011). One OTU was most closely identified as a

493   *Rhizopus* species; this genus is commonly associated with decaying vegetable matter (Kirk et al

494   2008). Likely plant pathogens (e.g. *Clavariopsis* sp. and a Letiomycetes species) and plant-

495   associated basidiomycotous yeasts (OTUs that aligned most closely to the genera *Bensingtonia*)

496   were also common.

497   Several of the most abundant OTUs (those with the greatest number of counts across all

498   butterflies) were also the most prevalent (detected in the largest number of butterflies). The most

499   frequently detected bacterial OTUs were a *Vagococcus/Enterococcus* species found in 83% of

500   the butterflies, a *Klebsiella/Enterobacter* species found in 82%, and an *Orbus* species found in

23

501   73%. The most prevalent fungal OTUs were *Hanseniaspora uvarum* found in 45%,

502   *Hanseniaspora opuntiae* found in 42%, and *Hanseniaspora guilliermondii* found in 40%.

503

504   *1a. Total bacterial load in the butterfly gut*

505   We quantified the total abundance of bacteria in the guts of 261 butterflies. The total number of

506   16S rRNA copies in a butterfly's gut ranged from 5×10^5 to 1×10^{11}, with a median of 7.5×10^8

507   (interquartile range: $1.2\times10^8 – 2.7\times10^9$). Larger butterfly individuals hosted greater numbers of

508   bacteria (Supplementary Figure 1). Bacterial load differed among host species (Figure 1; random

509   host species intercept term: $df = 1$, $\chi^2 = 17.5$, $p < 0.001$) but did not differ between feeding guilds

510   after accounting for host species and size (fixed diet term: $df = 1$, $\chi^2 = 1.8$, $p = 0.18$).

511

512   *2. Variation in gut microbial community composition between host dietary guilds, species, and*

513   *individuals.*

514   *2a. Variation between the feeding guilds*

515   The gut flora of frugivores and nectivores did not differ from each other in observed OTU

516   richness (Figure 2; bacteria fixed feeding guild term: $df = 1$, $\chi^2 = 0.98$, p = 0.32; fungi fixed

517   feeding guild term: $df = 1$, $\chi^2 = 0.09$, p = 0.76). However, gut microbial community composition

518   did differ between the feeding guilds (bacteria: permanova, F= 13.2, $R^2 = 0.040$, p=0.001, Figure

519   3a; fungi: permanova, F= 7.4, $R^2 = 0.038$, p=0.001, Figure 3b). Host feeding guild explained 4%

520   of the variation in both bacterial and fungal community compositions.

521   Despite substantial inter-individual and inter-specific variation in microbial community

522   composition (described below), average relative abundances of 10 of the 20 most abundant

523   bacteria (Figure 4) and fungi (Figure 5) systematically differed between the feeding guilds. Of

24

524 the bacteria, *Swaminanthia/Asaia* sp., *Bartonella* sp. and a strain of *Commensalibacter intestini*

525 were at higher relative abundance in nectivores than frugivores. Another strain of

526 *Commensalibacter intestini*, *Wolbachia* sp., a Porphyromonadaceae species, *Gilliamella/Orbus*

527 sp., *Orbus* sp. and *Acetobacter* sp. were at higher relative abundance in frugivores than

528 nectivores. Of the fungi, an unidentified fungus (OTU 7), a Letiomycetes species and a

529 Pleosporales species were more abundant in nectivores, while an unidentified ascomycete (OTU

530 20), two strains of *Kazachstania exigua*, *Hanseniaspora guilliermondii*, *H. uvarum*, *H. opuntiae*,

531 and *Pichia fermentans* associated more strongly with frugivores.

532

533 *2b. Variation among host species*

534 Observed OTU richness differed among butterfly species (Supplementary Figure 2; bacteria

535 random host species intercept term: $df = 1$, $\chi^2 = 43.3$, p < 0.001; fungi random host species

536 intercept term: $df = 1$, $\chi^2 = 52.3$, p < 0.001). Gut microbial community composition also differed

537 among host species (bacteria: permanova host species term, $df = 35$, F= 2.3, $R^2 = 0.244$, p<0.001;

538 fungi: permanova host species term, F= 2.3, $R^2 = 0.319$, p<0.001). Host species explained 24%

539 and 32% of the variation in bacterial and fungal community composition, respectively. Gut flora

540 were more similar within than between species (Figure 6).

541

542 *2c. Variation among individuals*

543 The vast majority of variation in community composition was expressed among individuals:

544 After accounting for species- and guild-level differences, residual variation in community

545 composition was 71.7% for bacteria and 64.4% for fungi (permanova). On average, a pair of

546 butterflies shared 23% (IQR 13%-31%) of their bacterial OTUs and 10% (IQR 0%-16%) of their

25

547   fungal OTUs. No OTU was present in all butterfly individuals. Each bacterial OTU was found in

548   a mean of 9 individuals (IQR: 1-6) and each fungal OTU was found in a mean of 4 individuals

549   (IQR: 1-3).

550

551   *3. Comparison of the gut flora to food microbial communities*

552   We sequenced the bacterial flora of 87 food samples, 83 of which were retained after

553   rarefaction, and the fungal flora of 75 foods, 66 of which were retained after rarefaction (Table 1;

554   Supplementary Table 2). After sequence processing and quality filtering we obtained 7.7 million

555   bacterial and 3.2 fungal sequences from potential food sources (fruits, nectars, and trap baits).

556   After rarefying, a total of 1205 bacterial OTUs and 529 fungal OTUs were observed across all

557   the food samples. (Again note that numbers of bacterial and fungal OTUs cannot be compared

558   due to differences in rarefaction depth.)

559   The microbial communities in wild fruit juice, the fermented trap baits, and wild nectars

560   were distinct from each other and from butterfly gut communities (Figure 7; Figure 8). In fact,

561   butterfly gut communities were highly dissimilar to the microbial communities on their food

562   sources. For example, frugivores and wild fruits shared 7% of their bacterial OTUs on average

563   (IQR 3%-9%), and the median Bray-Curtis dissimilarity between their bacterial communities

564   was 0.91 (Figure 8a). Despite these pronounced differences, frugivores' gut bacteria more

565   closely resembled the microbial composition of the trap baits than nectivores' gut bacteria did,

566   and similarly, nectivores' gut communities were more similar to nectar bacteria than frugivores'

567   gut flora were (Figure 8a). Frugivores' and nectivores' bacterial communities, however, were

568   equivalently distinct from those inhabiting the surfaces of wild fruits (Figure 8a).

569   Fungal communities in frugivore guts more closely resembled those of wild fruits and

26

570   trap baits than did those of nectivore guts, as expected (Figure 8b). Surprisingly, frugivores' gut

571   fungal communities were more similar, on average, to fungi in nectar than nectivores' gut

572   communities were (Figure 8b), though the effect size was extremely small.

573   While gut communities were distinct from food microbial flora in terms of species

574   composition and relative abundances, the percentage of bacterial gut reads assigned to OTUs that

575   were also present in the host's diet was generally high. On average, about 85% percent of gut

576   bacterial reads derived from OTUs present in their host's food. Fungal OTUs present in food,

577   however, accounted for a comparatively small 24% to 40% of butterfly gut fungal reads.

578

579   *4. Functional abilities: gut microbial community catabolism*

580   We measured the catabolic profiles of 248 gut communities via culture-based assays. Gut

581   communities digested an average of 51 substrates (IQR: 41-63) out of 71 total. As two

582   communities diverged in bacterial species composition, their catabolic profiles also became more

583   different; however, the effect size was very small (Mantel test: $r = 0.067$, $p = 0.013$). Catabolism

584   was not correlated with fungal community composition (Mantel test: $p = 0.50$).

585   Across the entire dataset, host diet explained just 3% of the total variation in community

586   catabolic profile (permanova on Euclidean distances between catabolic profiles; $df = 1$, F= 7.6,

587     $R^2=0.029$, p=0.001), while host species identity explained 18% of the variation (df= 33, F= 1.4,

588     $R^2=0.180$, p=0.001). After accounting for diet- and species-level differences, residual variation

589   among individuals in catabolic profile was 79.1%.

590   Lack of strong correlation between catabolic profile and host diet could be due to the fact

591   that not all catabolic functions are relevant for bacteria in butterfly guts. To investigate diet-

592   based differences in catabolic profile in more detail, we modeled catabolism as a function of host

593   diet, substrate identity, and substrate class, controlling for butterfly identity and number of 16S

594   rRNA copies in the butterfly. We also tested an alternate random effects structure, nesting

595   butterfly individuals within butterfly species, but this did not improve model fit (anova: $df = 1$, χ^2

596   = 0.258 p =0.61). Catabolism of several individual substrates differed markedly between the

597   feeding guilds (feeding guild by substrate interaction term: $df = 70$, $\chi^2 = 428.9$ p <<0.001).

598   Frugivorous gut flora catabolized 15 substrates more actively than nectivorous gut communities

599   did, with the most marked difference being in catabolism of D-serine, saccharic acid, mucic acid,

600   L-serine and lactic acid (Supplementary Table 5). Nectivorous gut flora digested 10 substrates

601   more actively, with the most extreme differences in catabolism of mannitol, sucrose, glucose,

602   fructose, and maltose (Supplementary Table 5). Entire classes of nutrients were also catabolized

603   differently between the guilds (feeding guild by substrate class interaction term: $df = 6$, $\chi^2 = 61.7$,

604   p <<0.001). Specifically, nectivores' gut flora catabolized sugars and sugar alcohols more

605   actively than frugivorous gut flora, while frugivores' gut flora were more successful at digesting

606   amino acids, carboxylic acids, and dicarboxylic acids (Figure 9; Supplementary Table 6).

607   Total number of 16S rRNA copies was not significantly correlated with catabolism (p=

608   0.09 for the model of substrate identity and p=0.07 for the model of substrate class), suggesting

609   that our standardization procedure adequately controlled for any initial differences in cell

610   inoculation densities.

611

612   **Discussion**

613   We evaluated how microbial community structure and catabolic function varied among

614   host feeding guilds, species and individuals within a community of over 50 species of

615   Neotropical butterflies. On average, 30 bacterial OTUs and 21 fungal OTUs were detected per

616   gut. Although the bacterial and fungal communities in adult butterfly guts varied substantially

617   among individual hosts, they did differentiate among host species and between host feeding

618   guilds. Gut community composition varied less between host feeding guilds than among host

619   species, suggesting that other non-dietary aspects of host biology play a large role in structuring

620   the butterfly gut flora. Gut communities were distinct in composition from the microbial flora

621   found in butterfly foods, indicating that the adult butterfly gut environment strongly filters

622   potential colonists. Despite high variability in the gut microbial community, host dietary guild

623   was nevertheless associated with consistent changes in both the relative abundances of dominant

624   microbes and the catabolic capacities of the gut flora. Frugivorous gut communities were better

625   at digesting amino acids and carboxylic acids, while nectivorous gut flora outperformed

626   frugivorous communities in catabolism of sugars and sugar alcohols. These catabolic patterns are

627   congruent with the relative nutritional makeup of butterfly foods (Ravenscraft and Boggs 2016),

628   which suggests that gut flora specialize in digestion of compounds abundant in the host's diet.

629

630   1. Microbes present in the guts of adult butterflies

631   *1a. Bacteria*

632   The composition of the butterfly's bacterial flora is broadly similar to that of other

633   insects: at the phylum-level, Proteobacteria comprised 68% and Firmicutes 12% of all reads

634   while in comparison, a meta-analysis of the insect gut flora reported that 57% were

635   Proteobacteria and 22% were Firmicutes (Colman et al 2012). At finer taxonomic resolution (see

636   Table 2), the most abundant bacterial OTUs were mostly known gut colonists, particularly those

637   associated with sugar-rich diets, and OTUs related to insect parasites and pathogens. (Individual

638   OTUs are discussed in more detail below.) Total bacterial load varied among species, but there

639   were no consistent differences between the feeding guilds after controlling for differences in size

640   among individual hosts. Larger butterfly individuals hosted greater numbers of bacteria. This

641   conforms previous work which demonstrated that total microbial load scales with host size (Kieft

642   and Simmons 2015).

643   The most abundant gut residents in our data were similar to those found in adult

644   *Heliconius erato* butterflies in Panama (Hammer et al 2014). The genera *Orbus*, *Enterobacter*,

645   *Asaia, Enterococcus*, *Lactococcus* and *Commensalibacter* were abundant in both studies, and

646   *Orbus* has also been isolated from the gut of the butterfly *Sasakia charonda* in South Korea (Kim

647   et al 2013). This suggests that adult butterflies do form consistent associations with some gut

648   microbes. What drives these associations is uncertain. Butterflies might maintain specific

649   relationships with some microbes via mechanisms such as vertical transmission. Gut flora can be

650   transmitted from mother to offspring via internal migration of gut microbes into the eggs, as

651   observed in the moth *Galleria mellonella*, or via the smearing of gut microbes on egg surfaces,

652   which has been hypothesized to occur in tobacco hornworm (*Manduca sexta*) (Brinkmann et al

653   2008; Freitak et al 2014). The relative frequency of such inter-generational transmission is

654   unclear, however, because the majority of the larval gut flora are purged during metamorphosis

655   (Kingsley 1972; Hammer et al 2014; but see Johnston and Rolff 2015). Other members of the gut

656   flora may be common in both butterflies and other insects because they are generally adapted to

657   the insect gut environment (e.g. Chouaia et al 2012) and opportunistically colonize many species.

658   Many of the most abundant genera we detected may fall into this category.

659

660   *1b. Gut fungi were primarily yeasts and common plant associates.*

661   The most abundant fungal OTUs were common plant and insect associates. Two thirds of

662   all fungal sequences belonged to the order Saccharomycetales, the "budding yeasts" or "true

663   yeasts." These single-celled fungi are often associated with sugar-rich environments and

664   decaying vegetable matter. Most of the remaining OTUs with reliable taxonomic assignments

665   were likely plant associates or pathogens. Studies characterizing the gut flora often focus on

666   bacteria and overlook fungi, but available evidence suggests that some insect guts host a rich

667   diversity of fungi, especially yeasts (Suh et al 2005). Some of these fungi may play an active role

668   in the gut, while others may simply use insects as a means of dispersal (Blackwell and Jones

669   1997; Starmer and Lachance 2011).

670

671   2. Variation in gut microbial community composition between host dietary guilds, among
672   species, and among individuals
673

674   *2a. A smaller degree of variation was attributable to differences between feeding guilds.*

675   Frugivores and nectivores did not differ in total bacterial load nor in bacterial or fungal

676   species richness, but did differ in the composition of their microbial flora. The degree of

677   variation explained by dietary guild was relatively small (about 4% for bacteria and fungi),

678   however it was comparable to that observed in many studies of microbial communities and is

679   likely biologically relevant: relative abundances of half of the most common OTUs differed

680   systematically between the feeding guilds. In general, the fungi more abundant in nectivores

681   were closely related to plant pathogens, while the fungi more abundant in frugivores were yeasts

682   associated with fruit decay, suggesting that this variation likely derived from exposure to

683   different microbial source pools. In contrast, there was no pattern in the types of bacteria that

684   differed in abundance between frugivores and nectivores.

685

31

686   *2b. A moderate degree of variation was attributable to differences among host species.*

687   About 24% and 32% percent of the variation in bacterial and fungal OTU compositions,

688   respectively, was attributed to differences among host species. Total bacterial load and bacterial

689   and fungal OTU richness also differed among butterfly species. Interspecies variation in the

690   butterfly gut flora could result from behavioral differences that expose butterflies to different

691   microbial pools. Such species traits could include habitat preferences. For example, species that

692   favor the forest understory could be exposed to more soil and wood decay fungi than those that

693   prefer the canopy. Additionally, many, but not all species drink from mud, dung, or carrion, in a

694   behavior known as "puddling" (Adler and Pearson 1982; Sculley and Boggs 1996), and variation

695   in these auxiliary feeding substrates could introduce different microbes to the gut. Differences

696   could also arise from variation in gut chemistry: for example, the guts of larger butterflies might

697   be more anoxic, or there could be interspecies differences in gut pH.

698   At least some of these factors could exhibit evolutionary signal in butterflies, resulting in

699   correlation between the butterfly phylogeny and gut community composition. Alternatively or

700   additionaly, vertical transmission of a fraction of the gut community from parent to offspring

701   could result in parallel divergence of microbiota and hosts. The potential for such evolutionary

702   patterns to contribute to interspecies differences in gut community composition is beyond the

703   scope of the current study, but will be addressed in a future paper (Ravenscraft et al, in prep).

704

705   *2c. Variation in microbial community composition among individual butterflies is high, but not*

706   *unusual.*

707   There was a great deal of variation among individual butterflies in microbial community

708   composition: the most prevalent bacterial and fungal OTUs were detected in only 83% and of

709 45% butterflies, respectively, and half of all OTUs were found in only one individual. Although

710 this degree of variation may seem extreme, it is comparable to that observed in other animals:

711 human colon microbiota were 34% similar among individuals in one study (Green et al 2006),

712 and in another, no OTUs were universally shared and 79% were found in only a single individual

713 (Tap et al 2009). A study of mammalian gut flora found that, on average, 56% of OTUs in an

714 individual mammal were unique to that individual (Ley et al 2008a). It has been proposed that

715 the high degree of variation observed in gut communities is the result of functional redundancy

716 at the OTU level: many microbes may require similar gut environmental conditions and/or serve

717 similar functional roles, leading to communities that differ in OTU membership but not in

718 overall function (Ley et al 2006; Peay et al 2016). However, it is also possible that variation in

719 membership *does* lead to significant variation in function (Peay et al 2016) and that the butterfly

720 gut flora do not play a consistent role across individuals.

721

722 *2d. The effect sizes of variables related to gut community composition are rarely reported or*

723 *compared.*

724 We had expected that a butterfly's feeding guild would explain a larger amount of

725 variation in gut community membership than host species identity, but we found the opposite.

726 The relative importance of dietary guild as a determinant of gut community composition has

727 been obscured in the literature by a trend of reporting only the significance (p-value) and not the

728 strength (either the effect size or R-squared) of these relationships. Diet has been shown to

729 significantly affect the community composition of the gut flora in many systems, including

730 mammals (Ley et al 2008a), fish (Sullam et al 2012), and insects (Colman et al 2012), but the

731 strength of this correlation was only reported for the last case. In a broad survey of the insect gut

732   flora, feeding guild (e.g. herbivore, omnivore, detritivore, pollenivore) explained 23% of

733   variation in the gut flora. The comparatively small amount of variation which diet explains in the

734   butterfly gut flora could be due in part to the broadly similar nature of butterfly foods (both being

735   sugary liquids).

736

737   3. Comparison of gut flora to food microbial communities.

738   *3a. Butterfly gut communities were highly dissimilar to fruit and nectar communities, but*

739   *generally shared more OTUs with the diet of their host than the diet of the other feeding guild.*

740   Host diet may shape the gut flora by serving as a source pool of microbes. Since it is

741   likely that adult butterflies acquire a substantial fraction of their gut flora from the environment

742   after emergence from the pupa (Hammer et al 2014), the microbial composition of adult butterfly

743   foods could be especially relevant. We asked whether the gut flora of Neotropical butterflies

744   might derive from their foods, and tested whether differences between frugivore and nectivore

745   gut communities were driven, in part, by differences in the microbes present in fruits and nectars.

746   Butterfly gut communities were more dissimilar to food microbial communities than we had

747   anticipated, with median percentage OTUs shared between a gut community and the microbes

748   residing in its host's food ranging from 5 to 16% for bacteria and 0 to 14% for fungi. However,

749   gut flora were generally more similar to the community on their host's food than to microbes on

750   the food of the other feeding guild. This suggests that a limited number of food-derived bacteria

751   and fungi either pass through or colonize the gut. Indeed, comparison of frugivores caught with

752   aerial nets to those caught via baited traps indicated that the yeast *Kazachstania exigua* was

753   transferred from the fermented baits into trapped butterflies' guts.

34

754   The high degree of differentiation between gut and food communities was surprising

755   because the guts of many nonsocial insects do seem to be colonized by environmental bacteria

756   (Ley et al 2008b; Boissière et al 2012). Indeed, prior studies of lepidopterans have found that

757   lab-reared larvae have a depauperate gut community compared to wild individuals, suggesting

758   that the larval gut *is* populated by microbes acquired from the environment (Xiang et al 2006;

759   Pinto-Tomás and Sittenfeld 2011; Belda et al 2011). Although the adult butterfly gut probably is

760   colonized at least in part by environmental microbes—as suggested by the high percentage of

761   bacterial reads derived from OTUs also found in butterfly foods—the vast majority of ingested

762   microbial species may fail to establish residence. The fact that the majority of the OTUs in

763   butterfly foods were not detected in adult guts emphasizes the strength of the filter the gut exerts

764   on potential microbial colonists, even in non-social hosts such as butterflies. Gut chemistry likely

765   filters potential colonists: most of the microbes in the gut (excluding transients) will belong to

766   the subset that can tolerate the gut's pH and oxygenation conditions. Competition or priority

767   effects from initial gut residents (those carried over from the larval host or established from early

768   adult meals) could also prevent food-derived microbes from successful colonization. Both areas

769   are fruitful avenues for future research aimed at understanding the community assembly of

770   animal gut flora.

771

772   *3b. Known habitat requirements of dominant gut microbes suggest that the adult butterfly gut is*

773   *acidic and potentially aerobic.*

774   The environmental requirements of the gut flora can help us infer chemical conditions in

775   the adult butterfly gut, which can be difficult to measure using traditional means. Acetic acid

776   bacteria (Acetobacteraceae) are obligately aerobic (Komagata et al 2014). The presence of many

35

777   members of the Acetobacteraceae in the adult gut suggests that it is not entirely anoxic; oxygen

778   may reach the outer layer of the gut via diffusion from gas in the tracheoles. Acetobacteraceae

779   also prefer acidic conditions, as do the yeasts *Hanseniaspora*, *Candida* and *Kazachstania* (Rosa

780   and Peter 2006; Komagata et al 2014). We could find no information about the pH of the adult

781   butterfly's gut, but in adults of the moth *Manduca sexta* it is between 5 and 6 (T. Hammer,

782   unpublished data). The larval gut is often alkaline, ranging from close to neutral pH to over 12

783   (Gross et al 2008), but the prevalence of acetic acid bacteria and acid-tolerant yeasts in our data

784   suggests that the adult gut is in fact acidic across most, if not all, butterfly species.

785

786   4. Functional potential of the butterfly gut flora.

787   *4a. The catabolic potential of the gut flora varies with host diet.*

788   The catabolic capacity of the gut flora was related to host feeding ecology. The gut flora

789   of frugivores exhibited increased catabolism of amino acids, carboxylic acids and dicarboxylic

790   acids compared to nectivores' gut flora, and decreased catabolism of sugars and sugar alcohols.

791   These differences in function likely result from differences in the chemical composition of the

792   diet. Nectars are 90% sugar by dry weight (Luttge 1977). Fruit generally contains more nitrogen

793   than nectar: at our field site, the juice of rotting *Dipteryx oleifera*, the dominant food available to

794   frugivorous butterflies during portions of the year, contains 33 times more essential amino acids

795   and 19 times more non-essential amino acids per unit sugar than the flower nectars fed upon by

796   butterflies (Ravenscraft and Boggs 2016). Frugivorous butterflies therefore provide a

797   comparatively nitrogen-rich gut environment that appears to support microbes capable of

798   catabolizing amino acids, while nectivores provide a more sugar-rich environment that favors

799   microbes specialized on sugar catabolism.

800   A 96-well plate is a very different environment from a butterfly gut, therefore we cannot

801   conclude that the functions we observed necessarily operate in vivo. However, our results

802   indicate that host species and host diet are related not only to gut community composition, but

803   also the functional capacities of the gut flora, and they suggest that the chemical makeup of the

804   diet selects for microbes that are specialized to digest that diet.

805

806   5. Butterfly gut flora include potential mutualists, commensals and pathogens.

807   A full understanding of the microbial gut community ultimately requires specific knowledge of

808   each of its members. From what is known about the microbial genera present in the butterfly gut,

809   we can begin to speculate whether they may be beneficial, commensal, or detrimental to butterfly

810   hosts.

811

812   *5a. Bacteria*

813   The genera *Enterobacter*, *Enterococcus*, and *Lactococcus* are commonly present in the

814   digestive tracts of a wide range of animals, from mammals to insects, and likely participate in

815   mutualistic or commensal relationships with their hosts (e.g. Robinson et al 2010; Engel and

816   Moran 2013; Delsuc et al 2014). *Acetobacter, Asaia*, and *Pantoea* are frequent residents of insect

817   guts (Crotti et al 2010; Engel and Moran 2013) and have been shown to directly benefit some

818   hosts. For example, *Acetobacter pomorum* and *Asaia* promote dipteran larval development,

819   probably through nutrient supplementation (Shin et al 2011; Mitraka et al 2013). *Pantoea*

820   *agglomerans* helps the swarming locust, *Schistocerca gregaria*, synthesize aggregation

821   pheromone (Dillon et al 2002). What these species may do for butterflies is currently unknown.

822   We also detected several genera that might be detrimental to butterflies. *Serratia* species

823   are often insect pathogens (Grimont and Grimont 2006). The genus *Bartonella* contains

824   opportunistic animal pathogens that can infect insects and mammals. Many are vectored by

825   insects— primarily flies (Minnick and Anderson 2000)— but the genus has also been detected in

826   the guts of other non-biting insects including honey bees, ants, and carrion beetles (Jeyaprakash

827   et al 2003; Stoll et al 2007; Kaltenpoth and Steiger 2014). *Spiroplasma* and *Wolbachia* are well-

828   known as reproductive parasites of insects; some kill males to promote their own spread via the

829   female line. *Spiroplasma* are known to colonize the gut lumen. *Wolbachia* are generally

830   associated with the reproductive organs, but have also been found in insect guts (Frost et al 2014;

831   Berasategui et al 2016). We dissected out the guts of the butterflies in our study, but we were not

832   always able to completely remove attached fat, Malphigian tubules, and reproductive tissue.

833   Although we are confident that most of the OTUs reported here were gut-derived, the origin of

834   the *Wolbachia* in our study is uncertain.

835   Although several of the most abundant bacterial genera we detected are often assumed to

836   be harmful to their hosts and could well be detrimental to butterflies, evidence suggests these

837   genera can sometimes be beneficial. *Spiroplasma* and *Wolbachia* can both protect their hosts

838   from pathogens or parasites (e.g. Hedges et al 2008; Jaenike et al 2010) and *Wolbachia* has been

839   shown to engage in nutritional symbioses, providing B vitamins to bedbugs (Hosokawa et al

840   2010). The high prevalence and vertical transmission of *Bartonella* in several fly species suggest

841   that this genus could also be beneficial to some insect hosts (Halos et al 2004). Overall,

842   butterflies host both potential mutualists and pathogens, but the balance seems shifted towards

843   more beneficial than detrimental bacteria.

844   The data that do exist regarding the community composition of the adult butterfly gut

845   flora are largely congruent with our findings. *Enterobacter*, *Enterococcus*, *Lactococcus,*

846   *Commensalibacter* and *Orbus* were dominant members of the adult butterfly gut flora in both our

847   data and in a study of the butterfly *Heliconius erato* in Panama (Hammer et al 2014).

848   *Commensalibacter* has also been detected in *Drosophila* guts, where it may protect the flies from

849   infection by pathogens (Roh et al 2008; Ryu et al 2008). It may play a beneficial role in

850   butterflies, as well: in a related study, we found that abundance of *Commensalibacter* was

851   positively correlated with increased adult lifespan in the temperate butterfly *Speyeria mormonia*

852   (Ravenscraft et al, in review). Little is known about the biology of *Orbus*, but the genus or its

853   close relatives have been detected in the guts of flies (Chandler et al 2011), bees (Kwong and

854   Moran 2013), and geographically disparate butterflies including *Speyeria mormonia* in Colorado

855   and *Sasakia charonda* in South Korea (Kim et al 2013). The apparently widespread associations

856   between butterflies and bacteria in the genera *Commensalibacter* and *Orbus* suggest that these

857   microbes may have a generalized relationship with lepidopterans or, more broadly, with sugar-

858   feeding insects.

859

860   *5b. Fungi*

861   Most of the fungal inhabitants of butterfly guts were saccharomycete yeasts. Members of

862   the genus *Hanseniaspora* are early colonizers of decaying fruits and are also commonly isolated

863   from drosophilid flies. They are dispersed by the flies, which also use them as a food source

864   (Kurtzman et al 2011). The genera *Candida* and *Pichia* are globally widespread and present in

865   many different environments. Some are pathogenic, but many are animal commensals or

866   mutualists. *Candida* and *Pichia* species have been detected in the guts of both vertebrates and

867   invertebrates including humans, gorillas, beetles, and larval lepidopterans (Rosa et al 1992; Suh

868   et al 2005; Marchesi 2010; Dematheis et al 2012; Hamad et al 2014). All reported isolates of

869   *Candida quercitrusa*, the species most closely related to the OTUs in our dataset, derive from

870   flowers or insects (Kurtzman et al 2011). Not much is known about the ecology of *Kazachstania*

871   *exigua*, but most strains have been isolated from human food products (Kurtzman et al 2011).

872   While *Kazachstania* was present at artificially high abundance in our raw data due to its

873   dominance in the trap baits, it was also detected on wild fruits and in the guts of butterflies that

874   had not fed on trap baits, suggesting that it does naturally occur in butterfly guts at low

875   abundance.

876   Many of the abundant fungal OTUs outside of the Saccharomycetales were too divergent

877   from known species' sequences to identify them with fine taxonomic resolution. One OTU was a

878   member of the class Letiomycetes, a class of plant pathogens. Another belonged to the order

879   Pleosporales, which predominantly consists of plant-decaying fungi, though some species

880   associate with live plants (Zhang et al 2009). These fungi may have arrived in the adult gut via

881   spores in fruits and nectars, or they could have colonized the larval gut via ingestion of leaf

882   tissue and then persisted through pupation. In summary, the fungal flora of the butterfly gut

883   included many sugar-loving yeasts and plant associates, some of which may pass through insect

884   guts as accidental vagrants or hitchhikers, as well as potential insect gut-associated commensals

885   or mutualists.

886

887   **Conclusions**

888   Gut microbiota are a ubiquitous feature of animal life, yet we still understand little about

889   the structure and function of these symbiotic communities. Here we developed adult Neotropical

40

890   butterflies as a study system to partition variation in gut community membership at the levels of

891   the host individual, species, and feeding guild; to compare the microbial species in the gut to

892   those in host foods; and to investigate the catabolic potential of the gut flora. All samples were

893   collected from a single site, thus eliminating the potential for large-scale geographic variation

894   that is often present in, and can potentially confound, studies of similar scale.

895   We found that the majority of variation in the butterfly gut flora is expressed at the level

896   of the individual host, followed by host species, with the least amount of variation explained by

897   dietary guild. The consistency of these patterns across both bacterial and fungal communities

898   lends support to the observed pattern. The pattern itself demonstrates the need for future work to

899   expand from the current focus on broad diet categories and host taxonomy to additional, species-

900   level or individual-level traits— for example, gut chemistry— in order to better understand the

901   causes and consequences of the gut community structure in nature.

902   Butterfly gut flora had little resemblance to food microbial communities, suggesting that

903   conditions in the gut exert a strong ecological filter on colonists. The initial colonists of the

904   butterfly gut may be derived largely from butterfly foods, but once established, the butterfly gut

905   community appears to be resistant to perturbation via ingestion of microbes. However, foods do

906   influence the gut flora via their chemical composition: culture-based assays suggested that the

907   chemical composition of the diet selects for gut microbes that are good at digesting common

908   chemical components of that diet.

909   Butterfly guts host a rich community of bacteria, as well as yeasts, which are woefully

910   understudied in contemporary analyses of the gut flora. By characterizing variation of the gut

911   microbiota within and among butterfly species we have laid the foundation for a mechanistic

912   understanding of how this hidden symbiosis affects and is affected by its host.

913

Acknowledgments

924

Data Accessibility

926 Raw Illumina sequences will be made available in the NCBI Short Read Archive. All other data

927 (e.g. the OTU table, bacterial abundances calculated from qPCR, butterfly fecundity, egg

928 chemical composition, etc) will be uploaded to Dryad.

929

Literature Cited

Adler P, Pearson D (1982) Why do male butterflies visit mud puddles? Can J Zool 60:322–325.

Anderson MJ (2001) A new method for non-parametric multivariate analysis of variance. Austral Ecol 26:32–46. doi: 10.1111/j.1442-9993.2001.01070.pp.x

Anderson MJ, Walsh DCI (2013) PERMANOVA, ANOSIM, and the Mantel test in the face of heterogeneous dispersions: What null hypothesis are you testing? Ecol Monogr 83:557–574. doi: 10.1890/12-2010.1

Beasley DE, Koltz AM, Lambert JE, et al (2015) The evolution of stomach acidity and its relevance to the human microbiome. PLoS One 10:1–12. doi: 10.1371/journal.pone.0134116

Belda E, Pedrola L, Peretó J (2011) Microbial diversity in the midguts of field and lab-reared populations of the European Corn Borer *Ostrinia nubilalis*. PLoS One 6:e21751. doi: 10.1371/journal.pone.0021751

Benjamini Y, Hochberg Y (1995) Controlling the false discovery rate: A practical and powerful approach to multiple testing. J R Stat Soc Ser B 57:289–300. doi: 10.2307/2346101

Berasategui A, Axelsson K, Norlander G, et al (2016) The gut microbiota of the pine weevil (*Hylobius abietis*, Coleoptera: Curculionidae) is similar across Europe and resembles that of other confer-feeding beetles. Mol Ecol. doi: 10.1111/mec.13702

Bergmann GT, Bates ST, Eilers KG, et al (2011) The under-recognized dominance of Verrucomicrobia in soil bacterial communities. Soil Biol Biochem 43:1450–1455. doi: 10.1016/j.soilbio.2011.03.012

Blackwell M, Jones K (1997) Taxonomic diversity and interactions of insect-associated ascomycetes. Biodivers Conserv 6:689–699. doi: 10.1023/A:1018366203181

Boissière A, Tchioffo MT, Bachar D, et al (2012) Midgut microbiota of the malaria mosquito vector *Anopheles gambiae* and interactions with *Plasmodium falciparum* infection. PLoS Pathog 8:e1002742. doi: 10.1371/journal.ppat.1002742

Brinkmann N, Martens R, Tebbe CC (2008) Origin and diversity of metabolically active gut bacteria from laboratory-bred larvae of *Manduca sexta* (Sphingidae, Lepidoptera, Insecta). Appl Environ Microbiol 74:7189–7196. doi: 10.1128/AEM.01464-08

Broderick NA, Raffa KF, Goodman RM, Handelsman J (2004) Census of the bacterial community of the gypsy moth larval midgut by using culturing and culture-independent methods. Appl Environ Microbiol 70:293–300. doi: 10.1128/AEM.70.1.293

43

958   Caporaso JG, Lauber CL, Walters WA, et al (2012) Ultra-high-throughput microbial community analysis on the

959   Illumina HiSeq and MiSeq platforms. ISME J 6:1621–1624. doi: 10.1038/ismej.2012.8

960   Chandler JA, Lang J, Bhatnagar S, et al (2011) Bacterial communities of diverse *Drosophila* species: Ecological

961   context of a host-microbe model system. PLoS Genet 7:e1002272. doi: 10.1371/journal.pgen.1002272

962   Chouaia B, Rossi P, Epis S, et al (2012) Delayed larval development in *Anopheles* mosquitoes deprived of *Asaia*

963   bacterial symbionts. BMC Microbiol 12:S2. doi: 10.1186/1471-2180-12-S1-S2

964   Colman DR, Toolson EC, Takacs-Vesbach CD (2012) Do diet and taxonomy influence insect gut bacterial

965   communities? Mol Ecol 21:5124–5137. doi: 10.1111/j.1365-294X.2012.05752.x

966   Crotti E, Rizzi A, Chouaia B, et al (2010) Acetic acid bacteria, newly emerging symbionts of insects. Appl Environ

967   Microbiol 76:6963–6970. doi: 10.1128/AEM.01336-10

968   Daily GC, Ehrlich PR (1995) Preservation of biodiversity in small rainforest patches: rapid evaluations using

969   butterfly trapping. Biodivers Conserv 4:35–55. doi: 10.1007/BF00115313

970   Delsuc F, Metcalf JL, Wegener Parfrey L, et al (2014) Convergence of gut microbiomes in myrmecophagous

971   mammals. Mol Ecol 23:1301–1317. doi: 10.1111/mec.12501

972   Dematheis F, Kurtz B, Vidal S, Smalla K (2012) Microbial communities associated with the larval gut and eggs of

973   the western corn rootworm. PLoS One 7:e44685. doi: 10.1371/journal.pone.0044685

974   Deshpande V, Wang Q, Greenfield P, et al (2016) Fungal identification using a Bayesian classifier and the Warcup

975   training set of internal transcribed spacer sequences. Mycologia 108:1–5. doi: 10.3852/14-293

976   Dillon RJ, Dillon VM (2004) The gut bacteria of insects: nonpathogenic interactions. Annu Rev Entomol 49:71–92.

977   doi: 10.1146/annurev.ento.49.061802.123416

978   Dillon RJ, Vennard CT, Charnley AK (2002) A note: Gut bacteria produce components of a locust cohesion

979   pheromone. J Appl Microbiol 92:759–763. doi: 10.1046/j.1365-2672.2002.01581.x

980   Edgar RC (2013) UPARSE: highly accurate OTU sequences from microbial amplicon reads. Nat Methods 10:996–

981   8. doi: 10.1038/nmeth.2604

982   Engel P, Moran NA (2013) The gut microbiota of insects - diversity in structure and function. FEMS Microbiol Rev

983   37:699–735. doi: 10.1111/1574-6976.12025

984   Freitak D, Schmidtberg H, Dickel F, et al (2014) The maternal transfer of bacteria can mediate trans-generational

985   immune priming in insects. Virulence 5:547–554. doi: 10.4161/viru.28367

986 Frost CL, Pollock SW, Smith JE, Hughes WOH (2014) *Wolbachia* in the flesh: Symbiont intensities in germ-line

987   and somatic tissues challenge the conventional view of *Wolbachia* transmission routes. PLoS One 9:e95122.

988   doi: 10.1371/journal.pone.0095122

989 Gardes M, Bruns TD (1993) ITS primers with enhanced specificity for basidiomycetes- application to the

990   identification of mycorrhizae and rusts. Mol Ecol 2:113–118. doi: Doi 10.1111/J.1365-294x.1993.Tb00005.X

991 Garland JL, Mills AL, Young JS (2001) Relative effectiveness of kinetic analysis vs single point readings for

992   classifying environmental samples based on community-level physiological profiles (CLPP). Soil Biol

993   Biochem 33:1059–66.

994 Gilbert JA, Jansson JK, Knight R (2014) The Earth Microbiome project: successes and aspirations. BMC Biol

995   12:69. doi: 10.1186/s12915-014-0069-1

996 Gordon GL, Phillips MW (1998) The role of anaerobic gut fungi in ruminants. Nutr Res Rev 11:133–68. doi:

997   10.1079/NRR19980009

998 Green GL, Brostoff J, Hudspith B, et al (2006) Molecular characterization of the bacteria adherent to human

999   colorectal mucosa. J Appl Microbiol 100:460–469. doi: 10.1111/j.1365-2672.2005.02783.x

1000 Grimont F, Grimont PAD (2006) The Genus *Serratia*. In: Balows A, Trüper H, Dworkin M, et al (eds) Prokaryotes.

1001   Springer-Verlag, New York, pp 219–244

1002 Gross EM, Brune A, Walenciak O (2008) Gut pH, redox conditions and oxygen levels in an aquatic caterpillar:

1003   Potential effects on the fate of ingested tannins. J Insect Physiol 54:462–471. doi:

1004   10.1016/j.jinsphys.2007.11.005

1005 Hakim RS, Baldwin K, Smagghe G (2010) Regulation of midgut growth, development, and metamorphosis. Annu

1006   Rev Entomol 55:593–608. doi: 10.1146/annurev-ento-112408-085450

1007 Halos L, Jamal T, Maillard R, et al (2004) Role of Hippoboscidae flies as potential vectors of *Bartonella* spp.

1008   infecting wild and domestic ruminants. Appl Environ Microbiol 70:6302–6305. doi:

1009   10.1128/AEM.70.10.6302-6305.2004

1010 Hamad I, Keita MB, Peeters M, et al (2014) Pathogenic eukaryotes in gut microbiota of western lowland gorillas as

1011   revealed by molecular survey. Sci Rep 4:1–9. doi: 10.1038/srep06417

1012 Hammer TJ, Dickerson JC, Fierer N (2015) Evidence-based recommendations on storing and handling specimens

1013   for analyses of insect microbiota. PeerJ e1190. doi: 10.7717/peerj.1190

1014   Hammer TJ, McMillan WO, Fierer N (2014) Metamorphosis of a butterfly-associated bacterial community. PLoS

1015   One 9:e86995. doi: 10.1371/journal.pone.0086995

1016   Hedges LM, Brownlie JC, O'Neill SL, Johnson KN (2008) *Wolbachia* and virus protection in insects. Science

1017   322:702. doi: 10.1126/science.1162418

1018   Hosokawa T, Koga R, Kikuchi Y, et al (2010) *Wolbachia* as a bacteriocyte-associated nutritional mutualist. Proc

1019   Natl Acad Sci 107:769–774. doi: 10.1073/pnas.0911476107

1020   Jaenike J, Unckless R, Cockburn SN, et al (2010) Adaptation via symbiosis: recent spread of a *Drosophila* defensive

1021   symbiont. Science 329:212–215. doi: 10.1126/science.1188235

1022   Jeyaprakash A, Hoy MA, Allsopp MH (2003) Bacterial diversity in worker adults of *Apis mellifera capensis* and

1023   *Apis mellifera scutellata* (Insecta: Hymenoptera) assessed using 16S rRNA sequences. J Invertebr Pathol

1024   84:96–103. doi: 10.1016/j.jip.2003.08.007

1025   Johnson KS, Barbehenn R V. (2000) Oxygen levels in the gut lumens of herbivorous insects. J Insect Physiol

1026   46:897–903. doi: 10.1016/S0022-1910(99)00196-1

1027   Johnston PR, Rolff J (2015) Host and symbiont jointly control gut microbiota during complete metamorphosis.

1028   PLoS Pathog 11:e1005246. doi: 10.1371/journal.ppat.1005246

1029   Kaltenpoth M, Steiger S (2014) Unearthing carrion beetles' microbiome: Characterization of bacterial and fungal

1030   hindgut communities across the Silphidae. Mol Ecol 23:1251–1267. doi: 10.1111/mec.12469

1031   Kieft TL, Simmons KA (2015) Allometry of animal-microbe interactions and global census of animal-associated

1032   microbes. Proc R Soc B Biol Sci 282:20150702. doi: 10.1098/rspb.2015.0702

1033   Kim JY, Lee J, Shin N-R, et al (2013) *Orbus sasakiae* sp. nov., a bacterium isolated from the gut of the butterfly

1034   *Sasakia charonda*, and emended description of the genus *Orbus*. Int J Syst Evol Microbiol 63:1766–1770. doi:

1035   10.1099/ijs.0.041871-0

1036   Kingsley V (1972) Persistence of intestinal bacteria in the developmental stages of the Monarch butterfly (*Danaus*

1037   *plexippus*). J Invertebr Pathol 20:51–58.

1038   Kirk P, Cannon P, Minter D, Stalpers J (eds) (2008) Dictionary of the Fungi, 10th edn. CABI, Wallingford, UK

1039   Komagata K, Iino T, Yamada Y (2014) The Family Acetobacteraceae. In: Rosenberg E, DeLong EF, Lory S, et al

1040   (eds) Prokaryotes Vol. 8 Alphaproteobacteria Betaproteobacteria, 4th edn. pp 3–78

1041   Kozich JJ, Westcott SL, Baxter NT, et al (2013) Development of a dual-index sequencing strategy and curation

1042   pipeline for analyzing amplicon sequence data on the miseq illumina sequencing platform. Appl Environ

1043   Microbiol 79:5112–5120. doi: 10.1128/AEM.01043-13

1044 Kurtzman CP, Fell JW, Boekhout T (eds) (2011) The Yeasts, A Taxonomic Study, Volume 2, 5th edn. Elsevier, San

1045   Diego, CA

1046 Kwong WK, Moran NA (2013) Cultivation and characterization of the gut symbionts of honey bees and bumble

1047   bees: Description of *Snodgrassella alvi* gen. nov., sp. nov., a member of the family Neisseriaceae of the

1048   betaproteobacteria, and *Gilliamella apicola* gen. nov., sp. Int J Syst Evol Microbiol 63:2008–2018. doi:

1049   10.1099/ijs.0.044875-0

1050 Lenth R V (2016) Least-Squares Means: The R Package lsmeans. J Stat Softw 69:1–33. doi: 10.18637/jss.v069.i01

1051 Ley RE, Hamady M, Lozupone C, et al (2008a) Evolution of mammals and their gut microbes. Science 320:1647–

1052   1651. doi: 10.1126/science.1155725

1053 Ley RE, Lozupone CA, Hamady M, et al (2008b) Worlds within worlds: evolution of the vertebrate gut microbiota.

1054   Nat Rev Microbiol 6:776–788. doi: 10.1038/nrmicro1978

1055 Ley RE, Peterson DA, Gordon JI (2006) Ecological and evolutionary forces shaping microbial diversity in the

1056   human intestine. Cell 124:837–848. doi: 10.1016/j.cell.2006.02.017

1057 Lundberg DS, Yourstone S, Mieczkowski P, et al (2013) Practical innovations for high-throughput amplicon

1058   sequencing. Nat Methods 10:999–1002. doi: 10.1038/nmeth.2634

1059 Luttge U (1977) Nectar composition and membrane transport of sugars and amino acids: a review on the present

1060   state of nectar research. Apidologie 8:305–319. doi: 10.1051/apido:19770402

1061 Marchesi JR (2010) Prokaryotic and eukaryotic diversity of the human gut. Adv. Appl. Microbiol., 1st edn. Elsevier

1062   Inc., pp 43–62

1063 Martin M (2011) Cutadapt removes adapter sequences from high-throughput sequencing reads. EMBnet.journal

1064   17:10. doi: 10.14806/ej.17.1.200

1065 McDade LA, Bawa KS, Hespenheide HA, Hartshorn GS (1994) La Selva: ecology and natural history of a

1066   Neotropical rain forest. University of Chicago Press, Chicago

1067 McDonald D, Price MN, Goodrich J, et al (2012) An improved Greengenes taxonomy with explicit ranks for

1068   ecological and evolutionary analyses of bacteria and archaea. ISME J 6:610–618. doi: 10.1038/ismej.2011.139

1069 McFall-Ngai M (2007) Adaptive immunity: care for the community. Nature 445:153. doi: 10.1038/445153a

1070   McMurdie PJ, Holmes S (2013) Phyloseq: An R package for reproducible interactive analysis and graphics of

1071   microbiome census data. PLoS One 8:e61217. doi: 10.1371/journal.pone.0061217

1072   McNulty NP, Yatsunenko T, Hsiao A, et al (2011) The impact of a consortium of fermented milk strains on the gut

1073   microbiome of gnotobiotic mice and monozygotic twins. Sci Transl Med 3:1–14. doi:

1074   10.1126/scitranslmed.3002701

1075   Minnick MF, Anderson BE (2000) *Bartonella* interactions with host cells. In: Oelschlaeger TA, Hacker JH (eds)

1076   Subcell. Biochem. Bact. Invasion into Eukaryot. Cells. Springer, New York, NY, pp 97–118

1077   Mitraka E, Stathopoulos S, Siden-Kiamos I, et al (2013) *Asaia* accelerates larval development of *Anopheles*

1078   *gambiae*. Pathog Glob Health 107:305–11. doi: 10.1179/2047773213Y.0000000106

1079   Muegge BD, Kuczynski J, Knights D, et al (2011) Diet drives convergence in gut microbiome functions across

1080   mammalian phylogeny and within humans. Science 332:970–4. doi: 10.1126/science.1198719

1081   Nardi JB, Mackie RI, Dawson JO (2002) Could microbial symbionts of arthropod guts contribute significantly to

1082   nitrogen fixation in terrestrial ecosystems? J Insect Physiol 48:751–763.

1083   Oksanen J, Blanchet FG, Kindt R, et al (2015) vegan: Community Ecology Package. R package version 2.3-0.

1084   Peay KG, Bruns TD, Kennedy PG, et al (2007) A strong species-area relationship for eukaryotic soil microbes:

1085   Island size matters for ectomycorrhizal fungi. Ecol Lett 10:470–480. doi: 10.1111/j.1461-0248.2007.01035.x

1086   Peay KG, Kennedy PG, Talbot JM (2016) Dimensions of biodiversity in the Earth mycobiome. Nat Rev Microbiol

1087   14:434–447. doi: 10.1038/nrmicro.2016.59

1088   Pinto-Tomás A, Sittenfeld A (2011) Comparison of midgut bacterial diversity in tropical caterpillars (Lepidoptera:

1089   Saturniidae) fed on different diets. Environ Entomol 40:1111–1122.

1090   R Core Team (2015) R: A language and environment for statistical computing. R Foundation for Statistical

1091   Computing, Vienna, Austria. URL http://www.R-project.org/.

1092   Ravenscraft A, Boggs CL (2016) Nutrient acquisition across a dietary shift: fruit feeding butterflies crave amino

1093   acids, nectivores seek salt. Oecologia 181:1–12. doi: 10.1007/s00442-015-3403-6

1094   Rawls JF, Mahowald MA, Ley RE, Gordon JI (2006) Reciprocal gut microbiota transplants from zebrafish and mice

1095   to germ-free recipients reveal host habitat selection. Cell 127:423–433. doi: 10.1016/j.cell.2006.08.043

1096   Robinson CJ, Bohannan BJM, Young VB (2010a) From structure to function: the ecology of host-associated

1097   microbial communities. Microbiol Mol Biol Rev 74:453–76. doi: 10.1128/MMBR.00014-10

1098   Robinson CJ, Schloss P, Ramos Y, et al (2010b) Robustness of the bacterial community in the cabbage white

1099   butterfly larval midgut. Microb Ecol 59:199–211. doi: 10.1007/s00248-009-9595-8

1100   Roh SW, Nam Y-D, Chang H-W, et al (2008) Phylogenetic characterization of two novel commensal bacteria

1101   involved with innate immune homeostasis in *Drosophila melanogaster*. Appl Environ Microbiol 74:6171–

1102   6177. doi: 10.1128/AEM.00301-08

1103   Rosa CA, Hagler AN, Mendonca-Hagler LCS, et al (1992) *Clavispora opuntiae* and other yeasts associated with the

1104   moth *Sigelgaita* sp. in the cactus *Pilosocereus arrabidae* of Rio de Janeiro, Brazil. Antonie Van Leeuwenhoek

1105   62:267–272. doi: 10.1007/BF00572594

1106   Rosa CA, Peter G (eds) (2006) Biodiversity and Ecophysiology of Yeasts. Springer-Verlag

1107   Russell V, Dunn PE (1996) Antibacterial proteins in the midgut of *Manduca sexta* during metamorphosis. J Insect

1108   Physiol 42:65–71. doi: 10.1016/0022-1910(95)00083-6

1109   Ryu JH, Kim SH, Lee HY, et al (2008) Innate immune homeostasis by the homeobox gene caudal and commensal-

1110   gut mutualism in *Drosophila*. Science 319:777–782. doi: 10.1126/science.1149357

1111   Salter SJ, Cox MJ, Turek EM, et al (2014) Reagent and laboratory contamination can critically impact sequence-

1112   based microbiome analyses. BMC Biol 12:87. doi: 10.1186/s12915-014-0087-z

1113   Salzman NH, Hung K, Haribhai D, et al (2009) Enteric defensins are essential regulators of intestinal microbial

1114   ecology. Nat Immunol 11:76–82. doi: 10.1038/ni.1825

1115   Sculley CE, Boggs CL (1996) Mating systems and sexual division of foraging effort affect puddling behaviour by

1116   butterflies. Ecol Entomol 21:193–197.

1117   Shin SC, Kim S-H, You H, et al (2011) *Drosophila* microbiome modulates host developmental and metabolic

1118   homeostasis via insulin signaling. Science 334:670–4. doi: 10.1126/science.1212782

1119   Smith DP, Peay KG (2014) Sequence depth, not PCR replication, improves ecological inference from next

1120   generation DNA sequencing. PLoS One 9:e90234. doi: 10.1371/journal.pone.0090234

1121   Solomon K V., Haitjema CH, Henske JK, et al (2016) Early-branching gut fungi possess a large, comprehensive

1122   array of biomass-degrading enzymes. Science 351:1192–1195. doi: 10.1126/science.aad1431

1123   Starmer WT, Lachance MA (2011) Yeast Ecology. In: Kurtzman CP, Fell JW, Boekhout T (eds) The Yeasts, 5th

1124   edn. Elsevier, San Diego, CA, pp 65–83

1125   Steinhaus E (1941) A study of the bacteria associated with thirty species of insects. J Bacteriol 42:757–90.

1126   Stoll S, Gadau J, Gross R, Feldhaar H (2007) Bacterial microbiota associated with ants of the genus *Tetraponera*.

1127   Biol J Linn Soc 90:399–412. doi: 10.1111/j.1095-8312.2006.00730.x

1128   Suh S-O, McHugh J V, Pollock DD, Blackwell M (2005) The beetle gut: a hyperdiverse source of novel yeasts.

1129   Mycol Res 109:261–265. doi: 10.1017/S0953756205002388

1130   Sullam KE, Essinger SD, Lozupone CA, et al (2012) Environmental and ecological factors that shape the gut

1131   bacterial communities of fish: a meta-analysis. Mol Ecol 21:3363–3378. doi: 10.1111/j.1365-

1132   294X.2012.05552.x

1133   Tap J, Mondot S, Levenez F, et al (2009) Towards the human intestinal microbiota phylogenetic core. Environ

1134   Microbiol 11:2574–2584. doi: 10.1111/j.1462-2920.2009.01982.x

1135   Wang Q, Garrity GM, Tiedje JM, Cole JR (2007) Naïve Bayesian classifier for rapid assignment of rRNA sequences

1136   into the new bacterial taxonomy. Appl Environ Microbiol 73:5261–5267. doi: 10.1128/AEM.00062-07

1137   Warnecke F, Luginbühl P, Ivanova N, et al (2007) Metagenomic and functional analysis of hindgut microbiota of a

1138   wood-feeding higher termite. Nature 450:560–565. doi: 10.1038/nature06269

1139   White TJ, Bruns T, Lee S, Taylor J (1990) Amplification and direct sequencing of fungal ribosomal RNA genes for

1140   phylogenetics. PCR Protoc A Guid to Methods Appl. doi: http://dx.doi.org/10.1016/B978-0-12-372180-

1141   8.50042-1

1142   Xiang H, Wei G, Jia S (2006) Microbial communities in the larval midgut of laboratory and field populations of

1143   cotton bollworm (*Helicoverpa armigera*). Can J Microbiol 52:1085–1092. doi: 10.1139/W06-064

1144   Zhang Y, Schoch CL, Fournier J, et al (2009) Multi-locus phylogeny of Pleosporales: A taxonomic, ecological and

1145   evolutionary re-evaluation. Stud Mycol 64:85–102. doi: 10.3114/sim.2009.64.04

1146   Zuur AF, Ieno EN, Walker NJ, et al (2009) Mixed Effects Models and Extensions in Ecology with R. Springer, New

1147   York

1148

1149 **Table 1.** Sample sizes

1150

1151 Number of butterfly or food samples ("individuals") and species remaining in the sequencing
1152 datasets after rarefying.

1153

	Bacterial dataset		Fungal dataset	
	Individuals	Species	Individuals	Species
Frugivores	148	24	92	15
Nectivores	142	28	69	19
Fruits	12	1	13	1
Nectars	46	8	39	8
Trap baits	25	n/a	14	n/a

1154

51

Table 2. Taxonomic identities of the 20 most abundant bacterial OTUs

The 20 most abundant bacterial OTUs. "% Total" is the percentage of sequences (pooled across all butterflies) that were assigned to the OTU. "% Butterflies", "% Frugivores", and "% Nectivores" are the percentages of all butterflies, frugivores, and nectivores in which the OTU was detected. Taxonomy was assigned by the RDP classifier using the Greengenes training set, except as otherwise noted. Confidences measure the degree of certainty of the taxonomic assignment.

OTU ID	% Reads	% Butterflies	% Frugivores	% Nectivores	Frug:Nect	Phylum	Class	Order	Family	Genus	Species	Confidence	Notes
OTU_1	9.8	73	84	63	1.3	Proteobacteria	Gammaproteobacteria	Orbales	Orbaceae	*Orbus*		100	a
OTU_4	8.7	82	81	84	1.0	Proteobacteria	Gammaproteobacteria	Enterobacteriales	Enterobacteriaceae	*Klebsiella/Enterobacter*		100	b
OTU_3	8.3	56	42	70	0.6	Proteobacteria	Alphaproteobacteria	Rhodospirillales	Acetobacteraceae	*Swaminathania/Asaia*		97	c
OTU_6	6.0	83	86	80	1.1	Firmicutes	Bacilli	Lactobacillales	Enterococcaceae	*Vagococcus/Enterococcus*		62	d
OTU_2	4.8	44	41	48	0.9	Bacteroidetes	Flavobacteriia	Flavobacteriales	[Weeksellaceae]			96	
OTU_19	3.2	46	34	59	0.6	Proteobacteria	Gammaproteobacteria	Enterobacteriales	Enterobacteriaceae	*Serratia*	*marcescens*	94	
OTU_5	2.7	28	27	29	0.9	Tenericutes	Mollicutes	Entomoplasmatales		*Spiroplasma/Entomoplasma*		100	e
OTU_10	2.6	28	36	20	1.8	Bacteroidetes	Bacteroidia	Bacteroidales	Porphyromonadaceae			68	
OTU_9	2.5	30	27	34	0.8	Proteobacteria	Alphaproteobacteria	Rhizobiales	Bartonellaceae	*Bartonella*		92	f
OTU_12	2.5	44	40	49	0.8	Firmicutes	Bacilli	Lactobacillales	Streptococcaceae	*Lactococcus*		100	
OTU_18	2.3	46	72	19	3.8	Proteobacteria	Alphaproteobacteria	Rhodospirillales	Acetobacteraceae	*Acetobacter*		100	g
OTU_11	2.2	34	46	22	2.1	Proteobacteria	Gammaproteobacteria	Orbales	Orbaceae	*Gilliamella/Orbus*		91	h
OTU_17	2.2	27	36	18	2.0	Proteobacteria	Alphaproteobacteria	Rhodospirillales	Acetobacteraceae	*Commensalibacter*	*intestini*	95	
OTU_14	1.7	19	25	13	1.9	Proteobacteria	Gammaproteobacteria	Enterobacteriales	Enterobacteriaceae	*Providencia/Serratia*		73	i
OTU_13	1.7	18	22	13	1.7	Proteobacteria	Alphaproteobacteria	Rickettsiales	Rickettsiaceae	*Wolbachia*		100	
OTU_15	1.6	26	31	20	1.5	Proteobacteria	Gammaproteobacteria	Pseudomonadales	Pseudomonadaceae			73	
OTU_23	1.4	23	26	19	1.4	Proteobacteria	Gammaproteobacteria	Pasteurellales				100	
OTU_569	1.4	27	18	37	0.5	Proteobacteria	Gammaproteobacteria	Enterobacteriales	Enterobacteriaceae	*Erwinia/Pantoea*		56	j
OTU_22	1.4	25	23	27	0.9	Proteobacteria	Alphaproteobacteria	Rhodospirillales	Acetobacteraceae	*Commensalibacter*	*intestini*	93	
OTU_1106	1.2	19	16	21	0.8	Bacteroidetes	Flavobacteriia	Flavobacteriales	[Weeksellaceae]			68	

Notes

a. RDP assigned to the order Pasteurellales with no finer resolution. Closest NCBI hit was Orbales:Orbaceae:*Orbus*.

b. In a blast search against NCBI, matched equally closely to the genera *Klebsiella* and *Enterobacter*.

c. RDP assigned to genus *Swaminanthia;* closest NCBI hit was *Asaia*.

d. Closest NCBI hit was *Enterococcus*

e. In the blast search against NCBI, matched equally closely to the genera *Spiroplasma* and *Entomoplasma*

f. RDP assigned to family with no finer resolution. Closest NCBI match was the genus *Bartonella*.

g. Closest NCBI species matches were *indonesiensis* or *okinawensis.*

h. RDP assigned to the order Pasteurellales with no finer resolution. Closest NCBI hit was Orbales:Orbaceae:*Gilliamella/Orbus*.

i. RDP assigned to family with no finer resolution. Closest NCBI matches were the genera *Providencia* and *Serratia*.

j. Closest NCBI match was the genus *Pantoea.*

1164 **Table 3.** Taxonomic identities of the 20 most abundant fungal OTUs

1165

1166 The 20 abundant fungal OTUs detected in at least 10% of the butterfly samples. "% Total" is the percentage of sequences (pooled
1167 across all butterflies) that were assigned to the OTU. "% Butterflies", "% Frugivores", and "% Nectivores" are the percentages of all
1168 butterflies, frugivores, and nectivores in which the OTU was detected. Taxonomy was assigned by the RDP classifier using the
1169 Warcup training set. Where a rank is *incertae sedis*, the closest available rank is written in brackets. Confidences measure the degree
1170 of certainty of the taxonomic assignment and are cumulative from higher to lower taxonomic ranks. Assignments with particularly low
1171 confidence are greyed out. (One additional fungal OTU, a *Torulaspora* species, was observed at high abundance in the pooled data but
1172 was detected in less than 10% of frugivores and less than 10% of nectivores, so is omitted here.)

1173

OTU ID	% Reads	% Butterflies	% Frugivores	% Nectivores	Frug:Nect	Phylum	Pconf	Class	Cconf	Order	Oconf	Family	Fconf	Genus	Gconf	Species	Sconf
OTU_7	5.0	22	7	37	0.2	Zygomycota	20	[Mucoromycotina]	15	Mucorales	15	Mucoraceae	7	Rhizopus	6	caespitosus	6
OTU_10	3.7	13	9	16	0.6	Basidiomycota	68	Agaricostilbomycetes	29	Agaricostilbales	29	Agaricostilbaceae	27	Bensingtonia	27	yuccicola	9
OTU_20	3.2	9	12	3	4.0	Ascomycota	68	Saccharomycetes	13	Saccharomycetales	13	Saccharomycodaceae	3	Hanseniaspora	3	uvarum	3
OTU_17	2.9	45	56	25	2.2	Ascomycota	100	Saccharomycetes	100	Saccharomycetales	100	Saccharomycodaceae	100	Hanseniaspora	100	uvarum	94
OTU_2469	2.6	42	57	16	3.6	Ascomycota	100	Saccharomycetes	100	Saccharomycetales	100	Saccharomycodaceae	100	Hanseniaspora	100	opuntiae	62
OTU_25	2.4	35	51	11	4.6	Ascomycota	100	Saccharomycetes	100	Saccharomycetales	100	Pichiaceae	99	Pichia	99	fermentans	70
OTU_2439	2.2	21	28	9	3.1	Ascomycota	100	Saccharomycetes	100	Saccharomycetales	100	Saccharomycetaceae	100	Kazachstania	100	exigua	100
OTU_11	1.9	29	34	18	1.9	Ascomycota	100	Saccharomycetes	100	Saccharomycetales	100	Saccharomycetales_Incertae sedis	97	Candida	97	quercitrusa	96
OTU_2306	1.9	40	53	18	2.9	Ascomycota	100	Saccharomycetes	100	Saccharomycetales	100	Saccharomycodaceae	100	Hanseniaspora	100	guilliermondii	42
OTU_2503	1.5	15	6	24	0.2	Ascomycota	35	Saccharomycetes	7	Saccharomycetales	7	Saccharomycodaceae	5	Hanseniaspora	5	uvarum	5
OTU_78	1.5	21	16	25	0.6	Ascomycota	97	Leotiomycetes	90	[Leotiomycetes]	32	[Leotiomycetes]	32	Leohumicola	17	levissima	14
OTU_24	1.5	13	15	9	1.7	Ascomycota	99	Saccharomycetes	98	Saccharomycetales	98	Saccharomycetaceae	98	Kazachstania	96	hellenica	96
OTU_41	1.4	8	3	13	0.2	Basidiomycota	46	Microbotryomycetes	8	Sporidiobolales	7	[Sporidiobolales]	7	Rhodotorula	1	cresolica	0
OTU_127	1.2	14	7	20	0.4	Ascomycota	94	Dothideomycetes	80	Pleosporales	75	[Pleosporales]	38	Clavariopsis	37	aquatica	37
OTU_997	1.2	13	22	0	n/a	Ascomycota	100	Saccharomycetes	100	Saccharomycetales	100	Saccharomycetaceae	100	Kazachstania	100	exigua	100
OTU_151	1.2	5	0	11	0	Ascomycota	43	Saccharomycetes	10	Saccharomycetales	10	Saccharomycetaceae	5	Kazachstania	2	hellenica	2
OTU_1735	1.1	9	10	7	1.4	Ascomycota	78	Saccharomycetes	40	Saccharomycetales	40	Pichiaceae	17	Pichia	10	haplophila	7
OTU_3	1.1	13	6	20	0.3	Ascomycota	22	Saccharomycetes	18	Saccharomycetales	18	[Saccharomycetales]	11	Cyberlindnera	5	fabianii	5
OTU_96	1.0	11	6	14	0.4	Ascomycota	100	Saccharomycetes	100	Saccharomycetales	100	Pichiaceae	100	Pichia	100	jadinii	100
OTU_73	1.0	16	11	20	0.5	Ascomycota	100	[Pezizomycotina]	19	[Pezizomycotina]	19	[Pezizomycotina]	19	Fontanospora	17	fusiramosa	17

1174

Figure 1. Total bacterial load per butterfly by species.

Total bacterial 16s counts per butterfly were estimated via qPCR. Host species are arranged in ascending size (wing length) along the x-axis. Dots indicate copy number for individual frugivores (red) or nectivores (blue), and boxplots depict medians and interquartile ranges of the data. Whiskers are placed at 1.5 times the interquartile range or, if all data fall within this range, they are placed at most extreme value measured. Note that although feeding guild is displayed, it was not a significant predictor of total bacterial load.
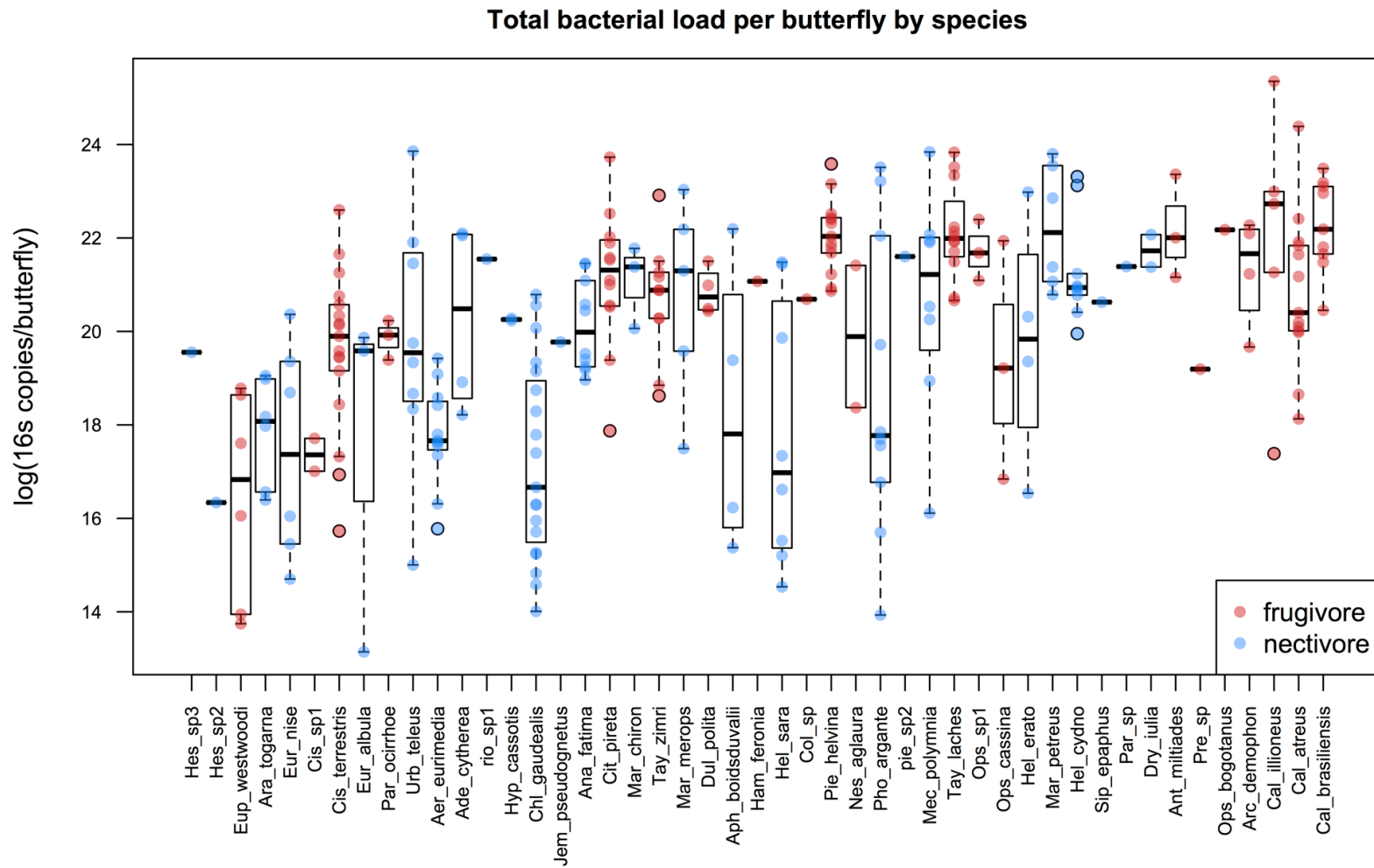

Total bacterial load per butterfly by species

Figure 2. Microbial species richness in frugivores and nectivores.

Bacterial (a) and fungal (b) observed species richness per butterfly did not differ between frugivores and nectivores. P-values are the result of model comparisons and control for host species (see text). Boxplot features are as described in Figure 1. One outlier of 162 bacterial OTUs in a nectivore is not shown in panel (a). Note that since bacteria and fungi were rarefied to different cutoffs, species richness estimates are not comparable between the two.
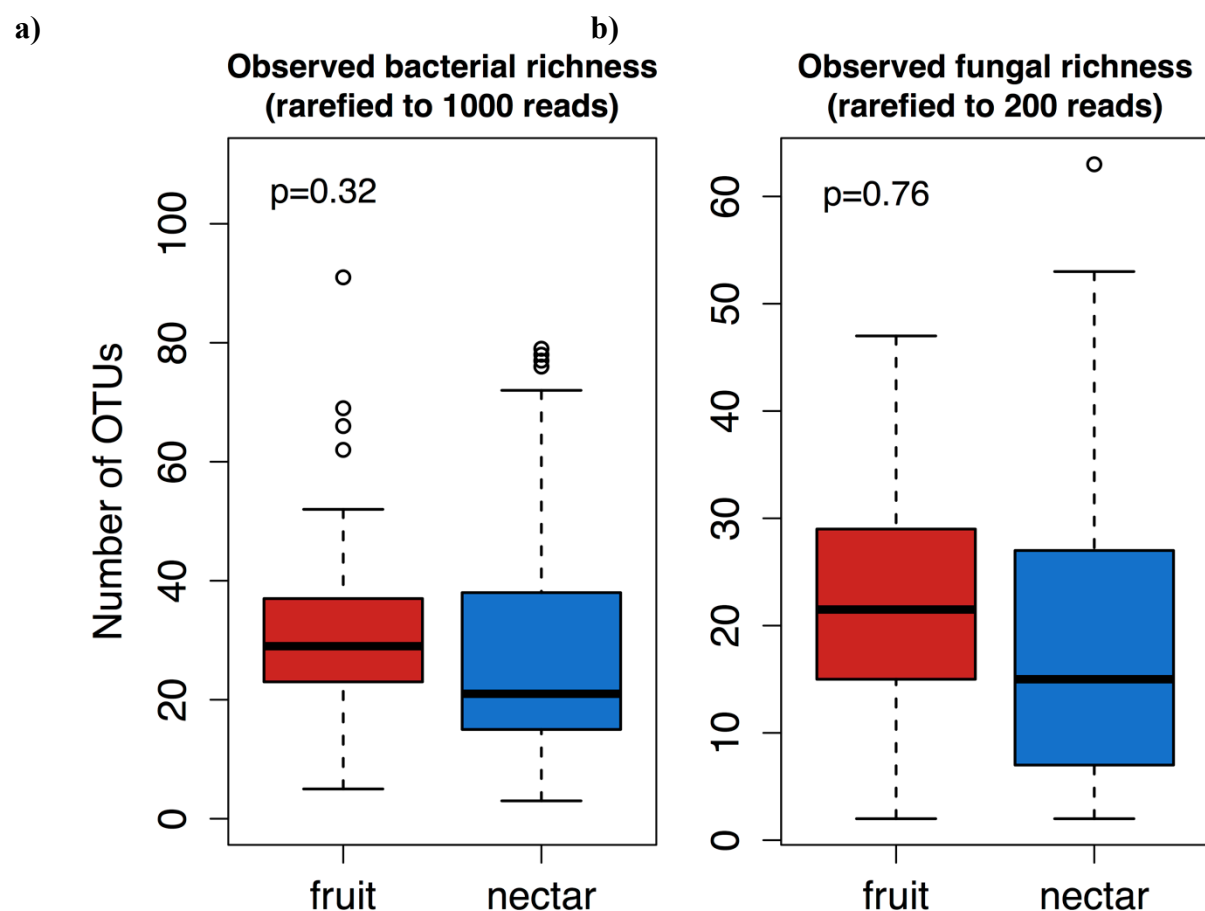
Figure 3. Ordinations of bacterial and fungal community composition.

NMDS plots of the Bray-Curtis dissimilarities between butterflies. PerMANOVA tests confirmed that the gut flora of frugivores and nectivores differed in bacterial and fungal OTU composition (p-values displayed on plots). (a) Bacterial gut flora of 148 frugivores and 142 nectivores. (b) Fungal gut flora of 92 frugivores and 69 nectivores. One outlier is not displayed in panel (b).
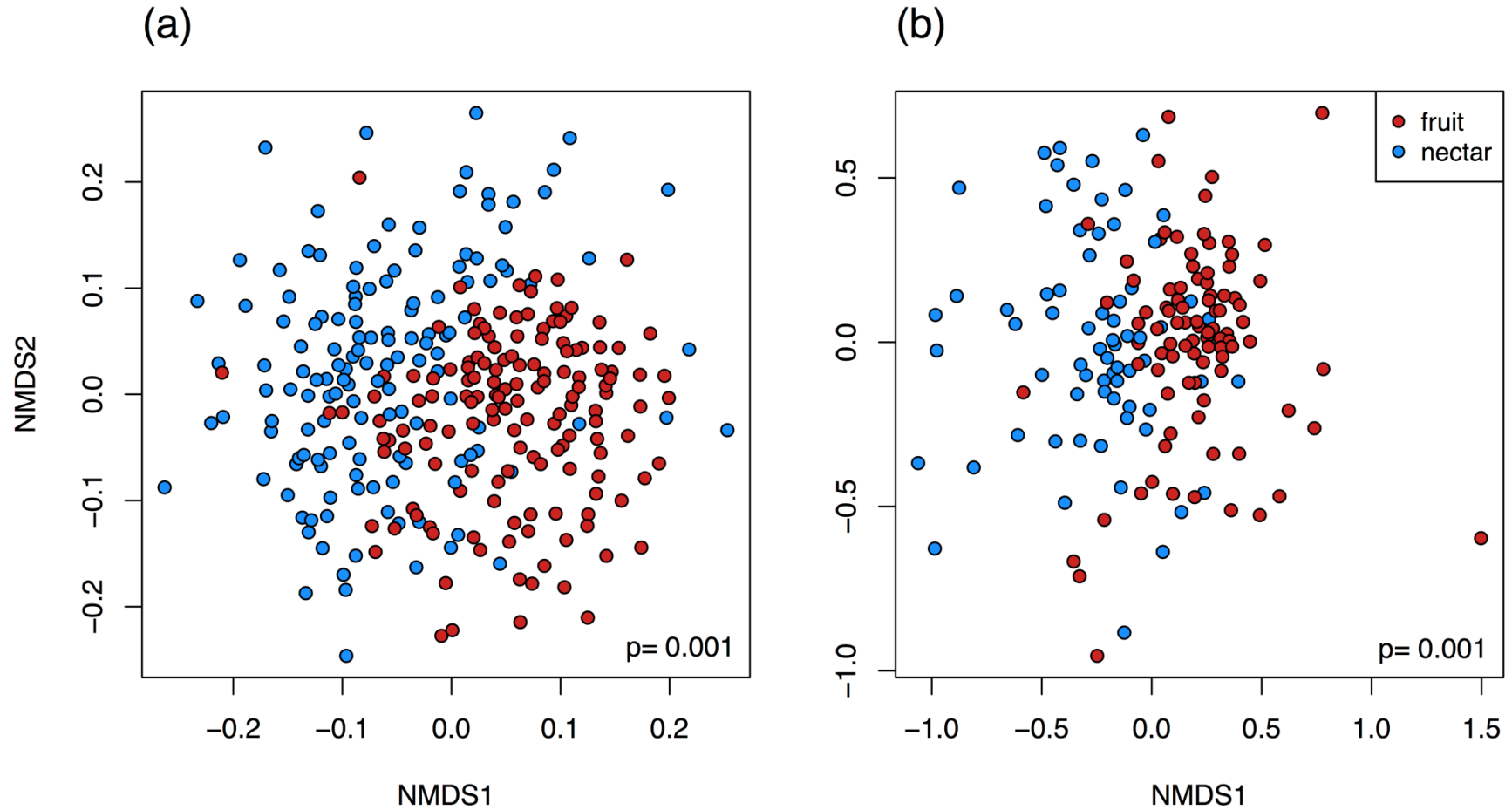
Figure 4. Effect of host diet (a) and host species (b) on bacterial community composition.

(a) Model-predicted estimates of the mean relative abundance per butterfly (counts out of 1000) of the 20 most abundant bacterial OTUs in the dataset. Relative abundance was modeled as a function of the interaction between host diet and OTU identity, with a random intercept of host species. Points indicate the model estimated mean and lines indicate one standard error. OTUs are arranged along the x-axis according to the magnitude and direction of the difference in their abundance between frugivores and nectivores. Each OTU is designated by genus or by the finest taxonomic resolution available, followed by OTU number in parentheses. The shaded gray box indicates OTUs that did not differ in abundance between the feeding guilds.
(b) Mean relative abundances of the 20 most abundant bacterial OTUs per host species. Darker shading indicates higher mean relative abundance.
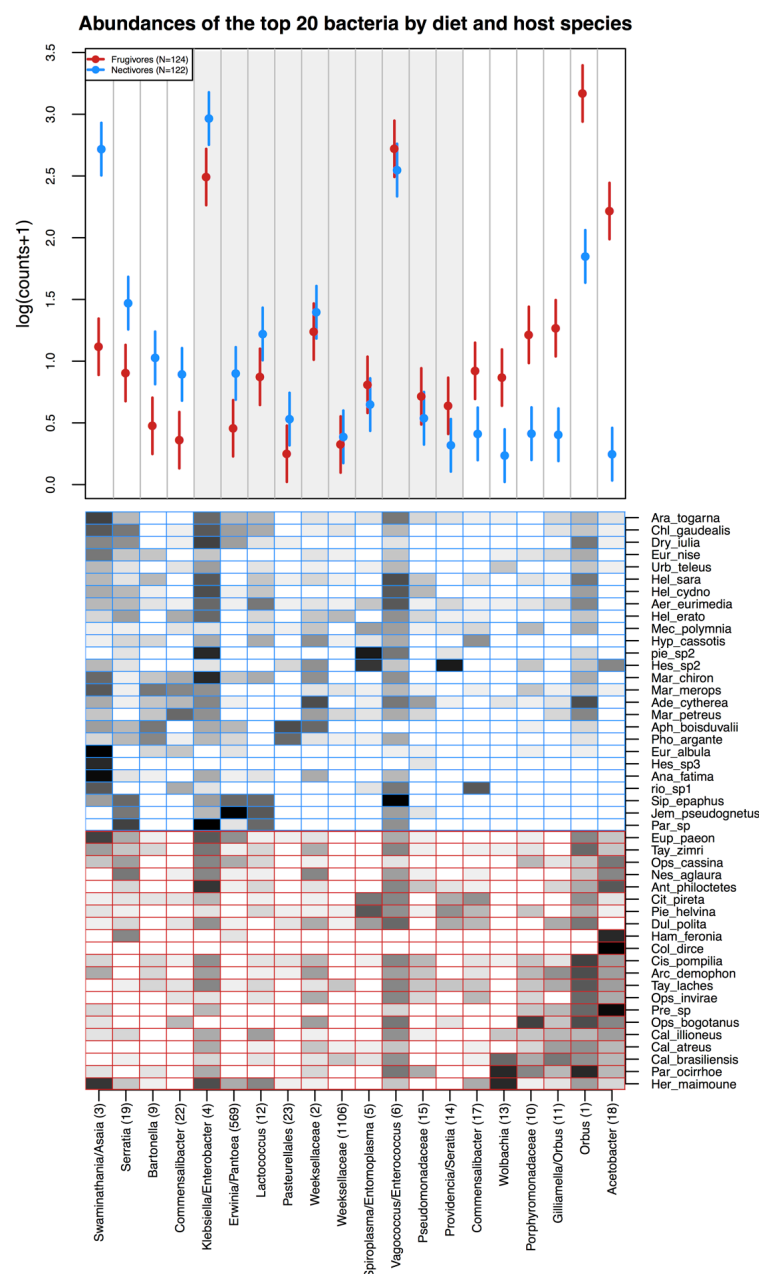
Figure 5. Effect of host diet (a) and host species (b) on fungal community composition.

(a) Model-predicted estimates of the mean relative abundance per butterfly (counts out of 1000) of the 20 most abundant fungal OTUs in the dataset. Relative abundance was modeled as a function of the interaction between host diet and OTU identity, with a random effect of host species. Points indicate the predicted mean and lines indicate one standard error. OTUs are arranged along the x-axis according to the magnitude and direction of the difference in their abundance between frugivores and nectivores. Each OTU is designated by genus or by the finest taxonomic resolution available, followed by OTU number in parentheses. The shaded gray box indicates OTUs that did not differ in abundance between the feeding guilds.
(b) Mean relative abundances of the 20 most abundant fungal OTUs per host species. Darker shading indicates higher mean relative abundance.
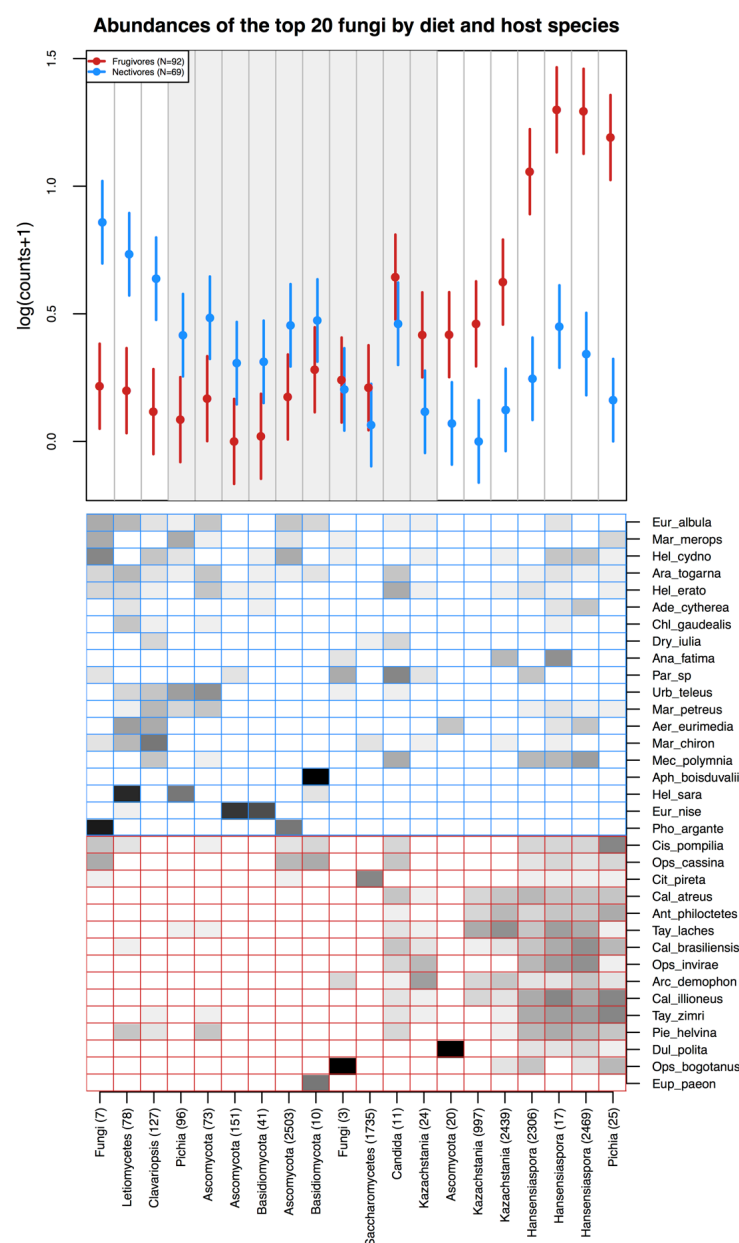
Figure 6. Dissimilarity of gut flora within versus between host species.

Every point represents the average Bray-Curtis dissimilarity between of a given individual's gut community and the gut communities of all other conspecific hosts (green) or all heterospecific hosts (orange). Boxplot features are as described in Figure 1. P-values are the results of t-tests for pairwise differences in dissimilarity.
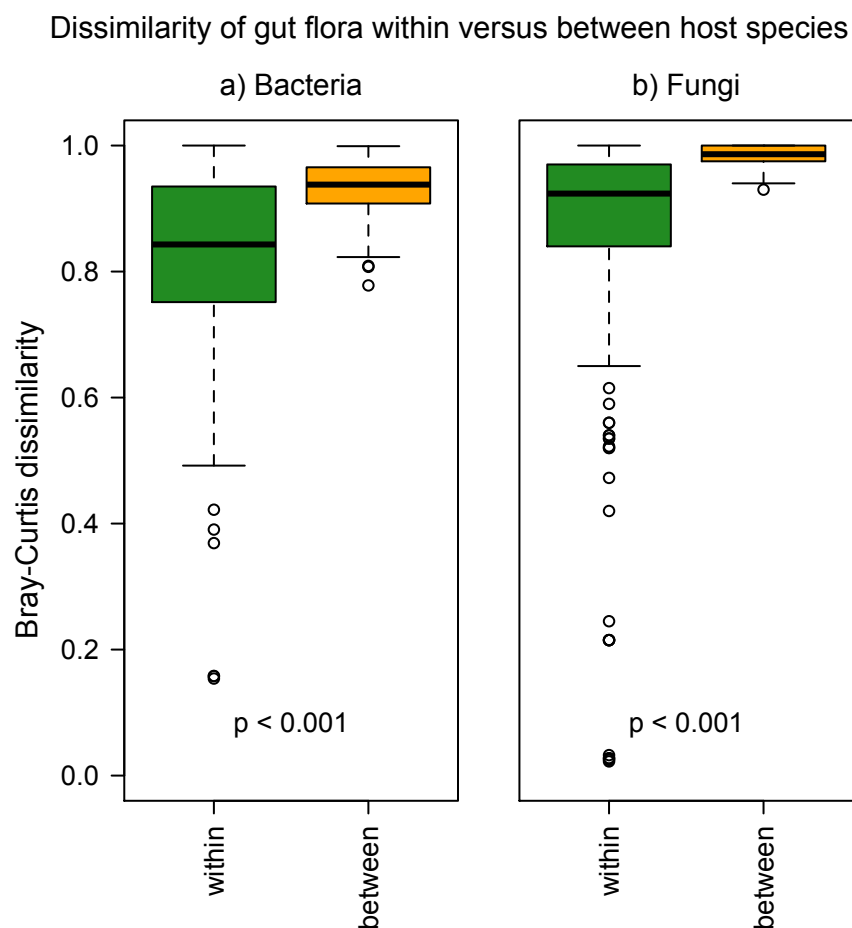

Dissimilarity of gut flora within versus between host species

Figure 7. Comparison of frugivore and nectivore gut communities to the microbial communities of butterfly foods.

NMDS plots of the Bray-Curtis dissimilarities between butterflies and their foods. (a) Bacterial flora of 149 frugivores, 149 nectivores, 46 nectars, 25 trap baits, and 12 fruits. (b) Fungal gut flora of 94 frugivores, 70 nectivores, 39 nectars, 14 trap baits, and 13 fruits. Two samples were fungal outliers and are not displayed in panel (b). (Sample sizes differ slightly from Figure 3 because in order to include as many food samples as possible, the combined gut and food sequence data were rarefied to different depths than the gut sequence data alone.)
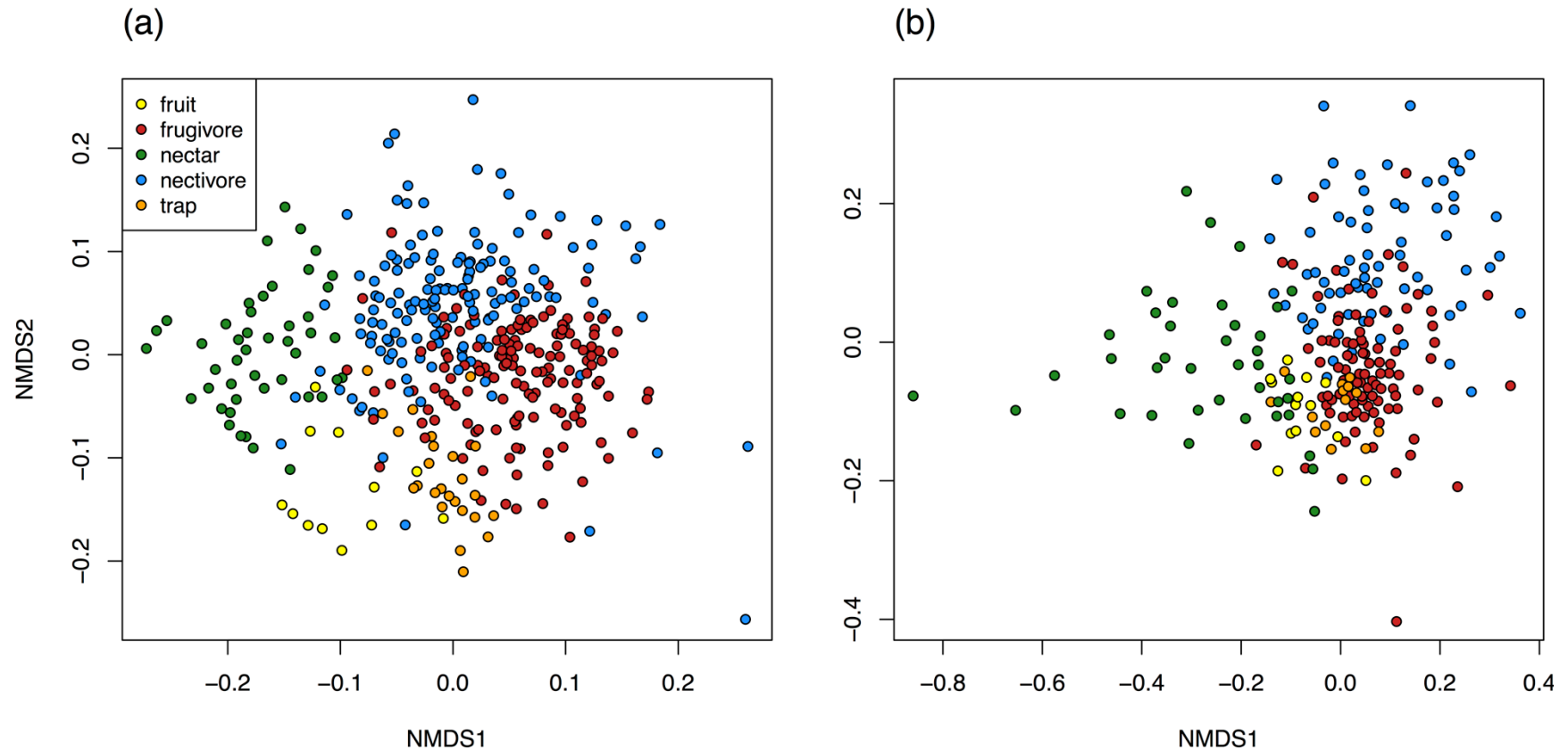
Figure 8. Dissimilarity between gut and food microbial communities**.**

Bray-Curtis dissimilarities between butterfly gut bacterial (a) and fungal (b) communities and the microbial communities in butterfly foods. Every point represents the average distance of a butterfly gut community to all food communities in a given category (fruit, trap bait, or nectar). Comparisons to frugivorous butterflies are colored red; comparisons to nectivores are colored blue. Boxplot features are as described in Figure 1. P-values are the results of t-tests for pairwise differences in dissimilarity; these have been FDR-corrected.

Note that three fungal strains of *Kazachstania exigua* were excluded from the butterfly data but not the food data, since these OTUs were introduced into frugivores' guts via the trap baits. (Described in the main text.) Frugivores would appear more similar to the trap baits if these OTUs had been included.

Dissimilarity between gut and food microbial communities

61

Figure 9. Differences in microbial community catabolism between frugivores and nectivores.

Points indicate the model-estimated mean difference in catabolism of a substrate (units of absorbance at 590 nm, standardized by plate and Hellinger transformed) between frugivores' and nectivores' gut microbial communities. Gut community catabolism of substrates in dark red and dark blue was significantly greater in frugivores or nectivores, respectively, after multiple test correction. Light red and light blue substrates were significantly different between guilds before, but not after multiple test correction. Thick lines and points indicate model-estimated means and standard errors for substrate classes. Red indicates classes that were catabolized more strongly by frugivore gut communities; blue indicates classes that were digested more actively by nectivore gut flora.