Gut microbes contribute to variation in foraging intensity in the honey

bee, *Apis mellifera*.

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Competing Interests

The authors declare no competing interests.

Abstract

Gut microbiomes are increasingly recognized for mediating diverse biological aspects of their hosts, including complex behavioral phenotypes. While many studies have reported that experimental disruptions to the gut microbiome result in atypical host behavior, studies that address how gut microbes contribute to adaptive behavioral trait variation are rare. Eusocial insects represent a powerful model to test this, due to their simple microbiomes and complex division of labor characterized by colony-level variation in behavioral phenotypes. While previous studies report correlational differences in gut microbiome associated with division of labor, here, we provide evidence that gut microbes play a causal role in defining differences in foraging behavior between honey bees. Gut microbial community structure consistently differed between hive-based nurse bees and bees that leave the hive to forage for floral resources. These differences were associated with variation in the abundance of individual microbes, including *Bifidobacterium asteroides, Bombilactobacillus mellis,* and *Lactobacillus melliventris.* Manipulations of colony demography and individual foraging experience suggested that differences in microbiome composition were associated with task experience. Moreover, single microbe inoculations with *B. asteroides*, *B. mellis,* and *L. melliventris* caused changes in foraging intensity. These results demonstrate that gut microbes contribute to division of labor in a social insect, and support a role of gut microbes in modulating host behavioral phenotypic variation.

Introduction

Gut microbiomes are emerging as important drivers and modulators of host phenotype [1], with evidence supporting the role of microbiomes in host digestion and nutrition, immune health, development, and, more recently, behavior [2–7]. Over the past decade, the "microbiota-gut-brain axis" [7], which describes bidirectional interactions between the gut microbiome and host brain and behavior, has developed as an increasingly important field of study with implications for

understanding the ecology and evolution of host-microbe interactions and animal behaviors.

However, current studies of the microbiota-gut-brain axis across animal taxa are largely

correlational, and causal studies exploring the relationship between gut microbes and host

behavior are rare [7].

The functional relationship between hosts and microbiomes varies across host traits. In

many cases, this relationship is obligatory, where the microbiome is necessary for normal

functioning of the host [7–10]. Recently, it has been proposed that gut microbiomes play a

facultative role in host phenotype, such that gut microbiomes contribute to phenotypic variation

between individuals [11, 12]. Results in support of a facultative relationship come from studies

across taxonomic groups. These include variations in non-behavioral and behavioral phenotypes

[11, 12], such as behaviors associated with aging and senescence [13–18] and severity of

neurodevelopmental disorders [19–21], as well as adaptive variations in behavioral traits both

within [22, 23] and between species [24]. Eusocial insects represent a powerful model for

understanding how gut microbes contribute to adaptive behavioral trait variation due to their

relatively simple and stable gut microbiomes and their complex division of labor [25–27]. This

division of labor is characterized by polyphenism between reproductive and non-reproductive

individuals, as well as colony-level behavioral trait variation between non-reproductive individuals

performing different tasks [25, 28]. In eusocial insects, gut microbes have been shown to

contribute to natural variation in memory [22] and social interactions [29]. Likewise, studies

indicate an association between gut microbiome and division of labor in eusocial insects [25, 30–

33]. However, whether gut microbes play a causal role in any aspect of eusocial insect division of

labor remains unknown.

We use the European honey bee, *Apis mellifera,* to investigate the causal relationship

between the gut microbiome and division of labor. Honey bees are a highly eusocial insect with

well-characterized and tractable social behaviors and gut microbiomes. In typical honey bee

colonies, sterile worker bees exhibit age-related division of labor, which is based on a pattern of

individual behavioral maturation. Typically, adult worker bees perform brood care ("nursing") and

other activities in the hive during the first 1-3 weeks of adult life and then transition to foraging

behaviors outside of the hive for the final 1-2 weeks of their life [28, 34]. However, honey bee

division of labor is flexible and responsive to changing colony needs, and thus does not

exclusively rely on the typical age patterns of behavioral maturation [35–39]. In addition, honey

bee foragers exhibit considerable individual variation in preference for nectar or pollen [40, 41],

exploratory behavior [42], and general foraging intensity [43]. Therefore, honey bee division of

labor involves individual variation in both behavioral maturation and task performance [44].

The honey bee gut microbial community, most of which resides in the hindgut, is well-

characterized and relatively simple, composed of ~10-20 different species of facultatively

anaerobic and microaerophilic host-adapted bacteria within the taxonomic groups of

Actinomycetes (*Bifidobacterium*), Lactobacillaceae (*Bombilactobacillus* Firm-4 and *Lactobacillus*

Firm-5), Gammaproteobacteria (*Gilliamella* and *Friscella*), Alphaproteobacteria (*Bartonella* and

Bombella), and Betaproteobacteria (*Snodgrassella*) [27]. Furthermore, the individual members of

the honey bee gut microbiome are consistently present, but differ in abundance across different

individuals and populations of honey bees [27]. Of particular interest, the composition of honey

bee gut microbial communities can be experimentally manipulated. All species can be cultured

[45] and used to inoculate young bees [29, 46, 47], who must acquire their microbiome from older

bees or hive materials [48]. Furthermore, previous studies indicate an association between the

gut microbiome and various aspects of division of labor in the honey bee [30–32, 49, 50]. Due to

these attributes of the honey bee gut microbiome, it is a great model for understanding the causal

effects of gut microbes on complex host behaviors. We took advantage of these behavioral and

microbiome features to determine whether gut microbes play a causal role in honey bee division

of labor, with a specific focus on behavioral maturation and variation in foraging intensity.

Materials and Methods

Animal Husbandry, treatments, and collections

Honey bee colonies were managed using standard beekeeping techniques at the University of
Illinois Bee Research Facility in Urbana, IL. Honey bees in this area are a genetic mixture of
subspecies, primarily *Apis mellifera ligustica* and *carnica* subspecies. For collection of nurses and
foragers, nurses were identified as those actively feeding brood on a brood frame, and foragers
were identified as those returning to the hive with pollen loads on their hind legs or having a
distended abdomen due to nectar loads [37, 51]. To reduce genetic variation between workers in
the colonies used for the single-cohort colony, big-back colony and single microbe inoculation
studies (described below), we used bees derived from queens that were each instrumentally
inseminated with sperm from a different single drone (SDI) (queen rearing and inseminations
were performed by Sue Cobey, Honey Bee Insemination Service, Washington State University
and Dr. Osman Kaftanoglu, Apimaye USA). Sample size for 16S rRNA sequencing studies was
10 bees per group as in [52]. All bees used in 16S rRNA sequencing analyses were washed once
with 12.5% bleach in water and twice with double deionized water and flash frozen. All samples
were stored at -80°C until further analysis.

Single- and double-cohort colonies

Single-cohort colonies (SCC) were established in the summer of 2020 as previously described
[36, 37, 51]. Honeycomb frames with pupal brood were taken from one or more source colonies
derived from the above-mentioned queens and placed in a 34°C incubator at 60-65% relative
humidity. About 1000 newly eclosed bees (<24 hours old) were gently brushed from these frames
<24 hours later and placed in a small five-frame hive box with a new, unrelated mated queen, one
honey frame, an empty honeycomb frame, and three new frames with wax-covered plastic
foundation upon which bees can build new wax honeycomb. SCCs were kept outdoors, and bees

were collected as typical-age nurses and precocious foragers at approximately one week of age

(7-9 days of age, variation due to rainy weather), and as over-age nurses and typical-age foragers

at three weeks of age. To obtain over-age nurses, frames containing brood were removed and

replaced with empty comb two weeks after queen introduction to ensure that a new cohort of bees

did not eclose before collections. For the SCC depicted in Fig. 2 and Supplementary Fig. 1, bees

from four distinct SDI source colonies (unrelated queens and drones) were used to construct a

single SCC. Four SDI source colonies were used to obtain the necessary 1000 newly eclosed

bees, and allowed us to test for the role of source colony in gut microbiome development

(Supplementary Fig. 1). For each collection (Day 9 and Day 21), 10 bees of each source colony

were collected for each behavioral task group. On Day 9, nurses were only able to be collected

for two source colonies. For the SCC depicted in Supplementary Fig. 4, bees from a single SDI

source colony were used, and groups of 10 bees were collected at each collection (Day 7 and

Day 21) for each behavioral task group.

Double-cohort colonies (DCC) were used to assess the effects of different inoculation

treatments (see below) on foraging behaviors in a common colony environment. DCCs were

established similarly to SCCs with a new, mated queen, 1 plastic honeycomb frame provisioned

with honey and pollen paste, 1 empty plastic honeycomb frame, ~200 7-day old barcoded

treatment bees (100 of bees from a single microbe inoculated treatment group and 100 from a

corresponding microbiota-depleted group) and ~800 newly eclosed unbarcoded "background"

bees (mixed colony backgrounds), whose addition contributed to accelerated behavioral

maturation of the older cohort [51, 53]. In the presence of the older cohort, younger "background"

bees are not expected to contribute to the foraging force [54] and did not throughout the course

of the experiment.

Big-back colonies

Big-back colonies were established as previously described [36, 51, 53] and kept outdoors in the

summer of 2021. For each colony, ~800-900 newly eclosed bees from a single SDI source colony

were used; half with a paint mark on the thorax, and half with a plastic tag attached to the thorax

(~3 mm diameter, ~1 mm thick; "big-back" bees). Four days later, ~400 newly eclosed bees from

~10 mixed source colonies (headed by naturally mated queens) were introduced to increase the

proportion of precocious foragers in the first cohort [36, 51, 53]. The entrance to the hive was

blocked by a piece of Plexiglas with holes in it to prevent the big-back bees from leaving the hive,

but to allow paint-marked bees to come and go freely. We observed that guard bees patrolled the

hive entrance on the inside of the Plexiglas piece.

Bees were collected at 10 days of age: returning active foragers and nurses collected as

described above, and big-back/inactive forager bees collected as they were attempting to leave

the hive via the holes in the plastic, indicating they too were in the behavioral state of foraging but

had never left the hive. 10 bees per group were collected from all three colonies, with the

exception that only seven nurses from the focal group in colony 2 were found and collected.

Gut microbiome DNA extraction, 16S rRNA sequencing and analysis

Frozen honey bee guts were dissected under sterile conditions on dry ice. Combined mid- and

hind- guts of individual bees were homogenized by maceration with a disposable sterile pestle

(VWR, Radnor, PA, USA). The homogenate was added to a PowerSoil Pro Bead Solution tube

(Qiagen, Germantown, MD, USA), and DNA was extracted using a DNeasy PowerSoil Pro DNA

isolation kit (Qiagen), following manufacturer's instructions. The hypervariable V4 region of the

16S rRNA gene was amplified by PCR in triplicates, with a negative control (no DNA), using

Platinum Hot Start PCR Master Mix (Invitrogen, Waltham, MA, USA), primers and barcodes

designed in [55] with a final concentration 0.25 μM, and the following cycling conditions: 94° 3

min, 35x[94° 45s/50° 60s/72° 90s], 72° 10 min. Before sequencing, PCR products were visualized

on agarose gels to confirm negative controls did not have amplification, while all samples had

expected amplification. All samples met these criteria, indicating no contamination. Samples were

pooled based upon concentrations and sequenced on an Illumina MiSeq with 2x250bp paired-

end reads. The samples were split among three sequencing runs: #1: SCCs (Dec 2020), #2: Big-

back colonies (July 2021), #3: Typical colony nurses and foragers, single microbe inoculated bees

(Dec 2021). Reads from these sequencing runs can be found in NCBI's Sequence Read Archive

PRJNA958253. Sequences for typical colonies from naturally mated queens were retained from

[49].

Samples were demultiplexed using QIIME2, and paired end reads were truncated at the

first base with a quality score of <Q3 using DADA2 [56]. Paired-end reads were merged and

amplicon sequence variants (ASV) were identified using DADA2 [57]. Chimeric ASVs were

removed and the remaining ASVs were taxonomically classified using the BEExact database [58].

ASVs that were taxonomically identified as a bee specific genus by the BEExact database, but

were unclassified at the species level, were subsequently classified to species level if possible,

using NCBI megaBLAST. To do this, representative 16S rRNA gene V4 sequences for each ASV

obtained from the QIIME2 rep-seqs.qza object were BLASTed against the entire NCBI nucleotide

database, and unclassified ASVs were called as the first identified full species with the lowest E-

score, if query cover > 80% and percent identity > 92%. Prior to statistical analyses, ASVs that

were identified as mitochondrial or chloroplast were removed from the data. Mitochondrial and

chloroplast reads were retained for visualization of single microbe inoculated and microbiota-

depleted bee gut microbiomes.

For sequencing run #1, 3,220,034 (1,610,017 pairs) total sequence reads were obtained,

1,431,503 pairs were filtered, merged and identified as non-chimeric (90%), and 627 ASVs were

identified. For sequencing run #2, 945,310 (472,655 pairs) total sequence reads were obtained,

424,785 pairs were filtered, denoised, merged and identified as non-chimeric (90%), and 110

ASVs were identified. For sequencing run #3, 2,538,722 (1,269,361 pairs) total sequence reads

were obtained, 1,141,046 pairs were filtered, denoised, merged and identified as non-chimeric

(90%), and 832 ASVs were identified.

To estimate the abundance of individual honey bee-associated microbial species in each

sample, the read counts for all ASVs that matched the same species were combined. We retained

the following honey bee-associated species: Acetobacteraceae related to *Commensalibacter*

[possibly Alpha-2.1 and Alpha-2.2], *Bartonella spp.*, *Bifidobacterium asteroides*, *Bifidobacterium*

coryneforme [syn. *Bifidobacterium indicum*], *Bombella apis* [previously *Parasaccharibacter*

apium], *Frischella perrara*, *Gilliamella apicola*, *Lactobacillus apis*, *Lactobacillus helsingborgensis*,

Lactobacillus kullabergensis, *Bombilactobacillus mellifer* [previously *Lactobacillus mellifer*],

Bombilactobacillus mellis [previously *Lactobacillus mellis*], *Lactobacillus melliventris*,

Apilactobacillus kunkeei, and *Snodgrassella alvi* [27, 59, 60]. We also retained other unclassified

Lactobacillus species, *Limosilactobacillus spp.*, *Arsenophonus spp.*, *Klebsiella spp.*, and *Mixta*

spp. because they made up a significant portion of reads (>500 reads in at least 1 sample). In the

case of single-microbe inoculated and microbiota-depleted bees (see below), *Enterococcus spp.*,

Staphylococcus spp., and Unassigned (not able to be classified below Kingdom of Bacteria) were

additionally retained for visualization. All other ASVs, each comprising 0-1% of total reads per

sample, and representing microbes that were not classified in the aforementioned groups and/or

are not typically associated with honey bees, were combined and labeled as "other." The variation

between samples in the abundance of "other" microbes is similar to previously published datasets

[29, 31]. To calculate the proportion of each species in each sample, that species' read count was

divided by the total read count for each sample. Raw read counts and/or proportion data were

used in analyses as described below. For the depiction of single microbe inoculated bee gut

microbiomes, "other" ASVs were combined based on genus (or the lowest taxonomic level

identified), and those representing >500 reads in at least 1 sample are depicted in a stacked

barplot, while all genera are included in supplementary data on Mendeley Data (DOI:

10.17632/f2s47y3nhn).

Due to the compositional nature of 16S rRNA sequencing data, typical measures of beta

diversity and differential abundance analyses have limitations [61–63]. Therefore we followed

analyses outlined in [61]. In short, for beta-diversity analyses, we used clr-transformations of raw

read counts for each sample in statistical tests and Aitchison distance for visualization (see

"Statistical analysis" section below). To estimate the differential relative abundance of individual

microbial species between samples through tools that have a compositional foundation (referred

to as "relative abundance" throughout the text), we used Analysis of Composition of Microbiomes

with Bias Correction (ANCOM-BC) [62, 64, 65] on raw read counts. This test relies on the total

sample read count and relative abundance of individual microbes in each sample to estimate the

true abundance of individual microbes in each gut ecosystem while accounting for the

compositional nature of sequencing data [64].

To estimate absolute bacterial species abundances in individual samples ("absolute

abundance"), we quantified the bacterial load in each sample using quantitative PCR (qPCR), as

in [29, 30]. We performed standard curves using serial dilutions of plasmids (TOPO pCR2.1

(Invitrogen)) containing the target sequence (10^8-10^3 copies per 3 μl), which were calculated from

the molecular weight of the plasmid and the DNA concentration of the plasmid. Primer efficiencies

were measured using: $E = 10^{(-1/slope)}$ [66], and the copy number of each target in 3 μl of DNA

sample was calculated from the sample's C_t score, primer efficiency, and standard curve using: n

= $E^{(intercept-Ct)}$x(DNA extraction elution volume/3) [29, 30]. These values were calculated for both

the 16S rRNA gene and the *Actin* gene. To account for differences in DNA extraction efficiency,

we calculated a normalized number of 16S rRNA gene copies by dividing the number of 16S

rRNA gene copies by that sample's number of *Actin* gene copies and multiplying this by the

median number of *Actin* copies for that experiment [29, 30]. To calculate the absolute abundance

of each microbial species in each sample, the relative abundance (proportion) of each species in

each sample was multiplied by the normalized number of 16S rRNA gene copies in that sample

[29, 30]. We then took the $\log_{10}$ value of each of these numbers and used these in further

analyses, replacing 0s with 1s before we did so.

PCRs were performed in triplicate using 10 μl reactions of 0.5 μM primers targeting the

16S rRNA gene (F: AGGATTAGATACCCTGGTAGTCC, R: YCGTACTCCCCAGGCGG) or the

A. mellifera Actin gene (F: TGCCAACACTGTCCTTTCTG, R: AGAATTGACCCACCAATCCA)

[46] in 1X *Power*SYBR Green PCR Master Mix (Applied Biosystems, Waltham, MA, USA) with 3

μl of DNA. Each primer set included a no template control. Reactions were performed in 384 well

plates using a QuantStudio 6 Flex (Applied Biosystems) with the following cycling conditions: 50°

2 min, 95° 10 min, 40x[95° 15 sec/60° 1 min], melt curve: 95° 15 sec, 60° 1 min, 95° 15 sec.

Standard curves using serial dilutions of plasmids (TOPO pCR2.1 (Invitrogen)) containing the

target sequence (10^8-10^3 copies per 3 μl) were calculated from the molecular weight of the

plasmid and the DNA concentration of the plasmid.

We found an effect of colony replicate on gut microbial community structure across all

experiments, indicating that bees from different colonies have different gut microbial communities

and that early environment and/or genetics influences the composition of the gut microbiome.

This was the case when comparing bees from different typical colonies (Supplementary Fig. 1A-

B), among SCC bees originating from different colony sources (Supplementary Fig. 1C-D), and

across bees from different big-back colonies (Supplementary Fig. 1E). Colony differences in

honey bee microbiome composition have been previously reported [52].

Single microbe inoculations

To identify effects of individual microbes on behavior we treated groups of newly eclosed

bees, who emerged under sterile lab conditions, with either an inoculum of a microbe of interest

or sterile food in order to produce bees whose microbiomes were composed of a single honey

bee-associated microbe (single microbe-inoculated) or no honey bee-associated microbes

("microbiota-depleted," historically referred to in this way since bees lacking a typical honey bee

microbiome are not completely microbe-free [46, 67]), respectively. Although single microbe

inoculations do not represent a natural microbiome, we chose this approach because it would

likely allow for the most control over microbiome composition across replicates. This is because

inoculations with multiple microbes may result in variation in microbiome structure between

individuals, groups, and/or replicates since honey bee microbiome structure is shaped by both

environmental and genetic factors [27, 48, 68]. These experiments occurred in the summer of

2021.

To obtain bees that lack the dominant honey bee bacterial species in the gut, modified

methods from [46, 67] were used. Specifically, tan-colored pupae with dark eyes were gently

removed from brood frames from three or four source colonies per trial using ethanol sterilized

forceps, placed dorsal side down in sterilized 3D-printed dental grade resin modular pupation

plates (Supplementary Fig. 2, design file on Mendeley Data), which were fitted into sterilized

Plexiglas cages, and kept in a 34°C incubator with 60-65% relative humidity for 2 days [69].

Eclosed bees were moved in groups of 30-40 to sterilized Plexiglas boxes (10 x 10 x 7 cm) [70]

in a 32°C incubator at 60-65% relative humidity (5 boxes per treatment for each replicate). Each

box was provisioned with a ~1 g sterilized pollen patty and 1.7 mL autoclave (15 minutes)

sterilized 25% sucrose with or without inoculum. An equal number of bees from each source

colony were added to each box, such that each treatment group (microbiota-depleted or single

microbe inoculated) had an equal number of bees of each colony background.

For single microbe inoculations, microbes were obtained from the DSMZ Leibniz-Institut

in Germany (*B. asteroides* DSM 20089, *B. mellis* DSM 26255, *L. melliventris* DSM 26256) and

were streaked on solid media and then cultured under anaerobic conditions at 35°C for 3 nights

in de Mann, Rogosa, Sharpe (MRS) (*B. asteroides*) or MRS + 20g/L fructose (*B. mellis and L.*

melliventris) broth, and glycerol stocks were made and kept at -80°C until future use. Ten days

prior to the onset of inoculation experiments, glycerol stocks were streaked on solid MRS (*B.*

asteroides) or MRS + 20g/L fructose (*B. mellis* and *L. melliventris*) media plates and were grown

under anaerobic conditions at 35°C for 3 nights (Oxoid Anaerojar with Anaerogen bags, Thermo

Scientific, Waltham, MA, USA). A single colony of bacteria from each plate was cultured under

anaerobic conditions (Anaerobic Hungate Culture Tubes with air displaced by CO_2, VWR, Radnor,

PA, USA), shaking at 35°C for 3 nights and was then retained at 4°C. Every day, starting 2 days

prior to the onset of inoculations, new cultures from this same original stock were prepared by

culturing 50 μl of the stock in 5 mL of fresh MRS/MRS + 20g/L fructose broth under anaerobic

conditions, shaking at 35°C for 2 nights. These cultures were spun down (3000 rpm for 5 min)

and resuspended in 1x PBS to an OD of 1 and 50 μl of a new, fresh culture solution was added

to 1.7 mL sterile 25% sucrose water in a new inverted microtube for each treatment box every

morning over the course of a 5-day inoculation period. For microbiota-depleted bees, 50 μl of

sterile 1x PBS was added to 1.7 mL 25% sucrose water in a new inverted microtube for each

treatment box every morning over the course of a 5-day inoculation period. Bees were then fed

sterile 50% sucrose water for 2 days following the 5-day inoculation period. Very few bees (1-2

per box) died during this inoculation period. Bees were kept in treatment boxes until 7 days old.

Unfortunately, we did not keep track of the treatment box each bee came from, rather bees for

each treatment were pooled and randomly chosen for 16S rRNA analysis or behavioral assays.

Microbe survival in 25% sucrose water was ensured by placing 50 μl of each microbial culture

(OD ~1) in 1.5 mL 25% sucrose water overnight and then plating this solution using species

specific culturing conditions.

Foraging assays and analysis using barcoded bees

To obtain comprehensive records of foraging activity, we used an automated behavioral tracking

system ("bCode") that uses a custom matrix barcode, enabling the unique identification of

individual bees [71]. After single microbe inoculations, 7-day old inoculated and microbiota-

depleted bees were cold-anaesthetized, a unique barcode was affixed to their thorax using super

glue, and a spot of colored paint associated with their treatment group was placed on their

abdomen. An equal number (~100) of bees from a single microbe inoculated group and a

corresponding microbiota-depleted group (same mix of colony backgrounds, same age) were

given barcodes from different sets, and were placed together in a DCC. Four colony replicates

per microbe treatment were performed. Two experimental days, corresponding with *B. mellis*

colony replicate 3 days 1-2 and *L. melliventris* colony replicate 2 days 1-2, experienced bad

weather, and thus no foraging occurred on those days.

Colonies were kept in a dark, temperature (32°C) and humidity (~50%) controlled building,

with access to the outside environment through a plastic tube. To monitor flight activity, an

entrance monitor with a video camera (Supplementary Methods) was attached to the hive

entrance, on the outside-facing end of this tube (Supplementary Fig. 3). The camera recorded

videos of bees entering and leaving the hive from 05:00 to 21:00 daily for a total of six days.

Barcodes in these videos were detected as in [71], and detections were filtered to remove tracking

errors and misidentifications. A mean of 97.82%, with a range between 96.24%-99.02%, of

detections were retained after these filtering steps. We then applied a flight activity detector [72,

73] (Supplementary Methods) to the remaining barcode detections to identify passes through the

entrance   monitor.   Computer   code   for   this   detector   can   be   found   at:

https://github.com/gernat/btools.

Passes through the entrance monitor determined by the flight activity detector were filtered

to remove unused barcodes, whose detections either occurred through detection error or due to

bees accidentally entering the wrong experimental colony. A mean of 99.19%, with a range of

95.4%-100%, of events were retained after this filtering. Passes were then classified as

"incoming," "outgoing," or "other." Since incoming passes had a lower error rate than outgoing

passes (Supplementary Table 8), incoming passes alone were used to denote a foraging trip.

Individual foraging trips that occurred within 5 min of each other were condensed into a single

event. A bee's first day of foraging was defined as the first day on which it performed at least 4

foraging trips, with at least 50% of these trips occurring during peak foraging hours (11:00-15:00

CST), as per previous studies [43, 74]. These criteria yield similar automated thresholds

corresponding to human observations of foraging behaviors [43]. All incoming passes from this

first day of foraging were counted as foraging trips, as were all subsequent incoming passes [43].

Likewise, an individual bee was considered a "forager" from this day forward. Due to the large

size of these data sets (owing to high dimensional automated behavioral monitoring) the number

of foraging trips per bee and treatment group and the number of foragers per treatment group on

each experimental day are available on Mendeley Data (DOI: 10.17632/f2s47y3nhn).

To quantify the degree of skew in foraging intensity among all workers, we calculated the

Gini coefficient [43, 75] for each experimental colony across all days and each experimental

colony on each day, since colony dynamics change with changes in weather and resource

availability, and thus likely changed throughout our studies. The values we obtained (Table 2) are

similar to those in previous studies [43, 75], indicating differences in foraging intensity between

individuals in each colony. To determine which bees performed the majority of the foraging for

each experimental colony on each day, we ranked individuals based on the proportion of total

foraging trips they performed for the colony on each day. A group of "elite foragers" was defined

as the subset of bees (variation, 5-40% of the foragers) performing $\geq$50% of the foraging trips for

each colony on each day [43, 75]. Number of elite foragers per treatment group on each day are

available on Mendeley Data (DOI: 10.17632/f2s47y3nhn).[74]

Statistical Analysis

All statistical analyses were performed in R (v 4.2.0) [76]. For all analyses, assumptions (e.g.,

normality, homogeneity of variances) were checked before statistical analysis. The following

functions were used in the base statistics package: "shapiro.test" (Shapiro-Wilk Normality Test),

"fligner.test" (Fligner-Killeen Test of Homogeneity of Variances), "t.test" (Student's t-test),

"wilcox.test" (Mann-Whitney Rank Sum Test), "aov" (ANOVA), "TukeyHSD" (Tukey's HSD post

hoc test), "Kruskal.test" (Kruskal-Wallis test). Kruskal-Wallis post hoc was performed using

"kwAllPairsDunnTest", PMCMRplus package [77]. Microbiome beta diversity was analyzed using

Permutation MANOVAs with 999 permutations ("adonis2", vegan package [78]) on clr-

transformed ("transform", microbiome package [79]) raw read counts, followed by Pairwise

Permutation MANOVAs with 999 permutations and FDR p-value adjustment ("pairwise.adonis",

pairwiseAdonis package [80]), and visualized using Principal Components Analysis ("ordinate",

phyloseq package [81]) with Aitchison distance ("distance", phyloseq package [81]). The relative

abundance of individual microbes was analyzed using raw read counts and ANCOM-BC with FDR

adjustment ("ancombc2", ANCOMBC package [62, 65]) with task as a fixed effect and colony as

random effect. Relative abundances were visualized using stacked barplots ("ggplot", ggplot2

package [82]). To compare the absolute abundance of individual microbes, Permutation ANOVAs

were used with 999 permutations ("perm.anova," RVAideMemoire package [83]), and p-values

were adjusted for multiple comparisons ("p.adjust" with FDR adjustment). Absolute abundance

data were visualized using dotplots ("ggplot," ggplot2 package [82]). Age at onset of foraging data

(using each individual bee's first day of foraging) were analyzed using a Cox Proportional Hazards

model ("coxph," survival package [84]), stratified by replicate [35, 74], and depicted as survival

plots using the cumulative proportion of bees from each treatment group that were identified as

foragers on each day. Proportions of foraging trips per individual and last day of foraging data

were checked for outliers ("identify_outliers," rstatix package [85]) and were analyzed as

generalized linear mixed-effects models ("glmer," lme4 package [86]) with log-normal distributions

and nAGQ=0, inoculation treatment and day as main factors, and replicate and individual as

random factors, followed by "Anova" (car package [87]) to determine main factor significance, and

"emmeans" (emmeans package [88]) with Tukey's p-value adjustment for pairwise comparisons.

Linear mixed-effects models ("lmer," lme4 package [86]) with inoculation treatment and day as

main factors, and replicate as random factors, followed by "Anova" (car package [87]) to

determine main factor significance, and "emmeans" (emmeans package [88]) with Tukey's p-

value adjustment for pairwise comparisons were used in the remaining behavioral analyses. Gini

coefficients were calculated using "Gini" (DescTools package [89]). All behavioral data, except

age at onset of foraging, were visualized using boxplots ("ggplot," ggplot2 package [82]).

Results

Nurses and foragers differ in gut microbial community composition

To begin to investigate whether gut microbes influence honey bee division of labor, we compared

gut microbial communities in nurse and forager bees, which represent two of the canonical

behavioral task groups in honey bee division of labor, as well as two discrete time points in

behavioral maturation [28]. Nurses and foragers differ markedly in physiology [37, 90–95],

neuroanatomy [96], neurochemistry [42, 97, 98], gene expression [38, 99] and gene regulation

[74, 100–102]. Therefore, we reasoned that testing for consistent differences in gut microbial

community between these two groups would be a powerful way to test for associations between

gut microbiome variation and division of labor.

We performed 16S rRNA sequencing on gut samples from nurses and foragers from three

unrelated honey bee colonies and reanalyzed, using updated methods [56, 57], a similar dataset

from a previously published study that used a different 16S rRNA sequencing method [49]. We

found that nurses and foragers differed significantly in gut microbial community structure in both

the previously published dataset (Fig. 1A) and our new dataset (Fig. 1B), indicating robust

differences associated with division of labor and independent of sequencing method.

We next sought to identify individual microbes that differed in abundance between nurse

and forager bees to identify candidate microbes specifically associated with division of labor.

Since microbes can vary between individuals in relative and/or absolute abundance, we used

both of these measures of biological diversity. Both may contribute to ecosystem functioning

[103–106], but their relative effects on host behavior are currently unknown [30, 62]. Therefore,

we reasoned that using both of these abundance measures would give us a well-rounded sense

of individual microbes that are associated with honey bee division of labor. Specifically, we

performed a relative abundance analysis (ANCOM-BC [62, 64, 65]) that accounts for the

compositional nature of 16S rRNA sequencing data and uses relative abundance measures to

estimate the true microbial composition of each gut (hereafter referred to as "relative abundance").

We also combined qPCR with 16S rRNA sequencing data analyses [29, 30] to estimate the

absolute abundance of individual microbial taxa in each gut ("absolute abundance"). We found

that individual gut microbes significantly differed in both relative and absolute abundance between

nurses and foragers, while total microbe abundance did not differ between groups (Fig. 1C-D,

Supplementary Table 1, Table 1). These results indicate that nurse and forager bees differ

significantly in their gut microbial communities. They also identify candidate microbes that may

be associated with division of labor.

Differences in gut microbial community composition between nurses and foragers

are dependent upon differences in behavioral state

Under typical colony conditions, nurses and foragers differ in chronological age. However, honey

bee division of labor is flexible and responsive to changing colony needs. For example, if a colony

experiences a shortage of older bees, some younger individuals will begin to forage at an early

age, while a shortage of younger bees will cause bees to continue to act as nurse bees despite

advancing chronological age [35–39]. To determine whether the gut microbial differences

between nurses and foragers depend on differences in behavioral state and/or age, we exploited

this adaptive plasticity to create SCCs, a well-established experimental approach that separates

these two factors [36, 37, 51]. In two SCC replicates, we compared gut microbiomes between

age-matched nurse and forager bees when they were about one week of age (representing typical-age nurses and precocious foragers) and three weeks of age (representing over-age nurses and typical-age foragers).

Age-matched typical-age nurses and precocious foragers did not significantly differ in gut microbial community structure (Fig. 2A, Supplementary Fig. 4A), while age-matched over-age nurses and typical-age foragers did (Fig. 2B, Supplementary Fig. 4B). Similar results were obtained for the abundance of individual microbial taxa: no microbes consistently differed in abundance between typical-age nurses and precocious foragers across the two SCC replicates (Fig. 2C, Fig. 2E, Supplementary Fig. 4C, Supplementary Fig. 4E, Supplementary Table 2, Supplementary Table 3), while four microbes consistently differed in relative abundance and three microbes consistently differed in absolute abundance between over-age nurses and typical-age foragers across the two SCC replicates (Fig. 2D, Fig. 2F, Supplementary Fig. 4D, Supplementary Fig. 4F, Supplementary Table 2, Supplementary Table 3). In addition, the total absolute abundance of microbes differed between age-matched typical-age nurses and precocious foragers in one colony replicate (Fig. 2E, Supplementary Fig. 4E), and between age-matched over-age nurses and typical-age foragers (Fig. 2F, Supplementary Fig. 4F). Overall, these findings indicate that age-dependent differences in behavioral state are associated with differences in gut microbial communities between nurses and foragers.

Together, results from the typical colony and three-week old SCC indicate that of the honey bee-associated microbes, *Bombilactobacillus mellis* (previously *Lactobacillus mellis*) and *Lactobacillus melliventris* were consistently (i.e. identified in three or more analyses across the two abundance measures) more abundant in nurses relative to foragers (Table 1), while Acetobacteraceae and *Snodgrassella alvi* were consistently less abundant in nurses relative to foragers (Table 1). In SCCs alone, *Bifidobacterium asteroides* showed consistent lower relative abundance in over-age nurses compared to typical-age foragers (Table 1) and *Gilliamella apicola* (Table 1) showed consistent higher absolute abundance in over-age nurses compared to typical-

age foragers. Such robust associations highlight these six microbes as candidate microbes whose

abundances may have causal effects on division of labor in honey bees. In addition, *Klebsiella*

spp., a taxonomic group not typically considered a honey bee-associate, was more abundant in

over-age nurses compared to typical-age foragers (Fig. 1-2, Supplementary Fig. 4, Table 1). This

microbial species group has previously been found to be prevalent in nurse bees and, as an

environmentally derived potential pathogen to bees, was likely picked up outside of the hive by

foragers and disseminated to nurses within the hive [30]. As previously suggested, it may

accumulate in the guts of nurses, which are typically heavier and support higher bacterial loads

than forager guts [30]. This possibility is consistent with our finding that *Klebsiella* was absent in

typical-age nurses and precocious foragers but present in over-age nurses and typical-age

foragers, likely indicating that *Klebsiella* was inadvertently introduced to the colony after one

week, at which time it began accumulating in the guts of over-age nurses, who have higher

bacterial loads than typical-age foragers (Fig. 2, Supplementary Fig. 4).

The lack of gut microbial community differences between one-week old SCC nurses and

foragers suggests that differences in behavioral maturation rate are not associated with

differences in gut microbial community composition. Rather, the observed microbiome differences

between over-age nurses and typical-age foragers suggest that associations between gut

microbiome and behavioral state depend on worker experience, possibly, as others have

suggested, due to task-related differences in environment, diet, and/or metabolic needs [30–32].

Since SCCs are composed of bees of a single age cohort, three-week old SCCs are composed

of bees that have been performing their respective tasks for prolonged periods (1-2 weeks longer)

compared to one-week old SCCs. Therefore, we hypothesized that task experience plays an

important role in gut microbial community composition. In the next section, we tested this by

manipulating foraging experience since it is known to have strong effects on behavior, brain

chemistry, brain structure, and brain gene expression [75, 97, 107, 108].

Foraging experience influences gut microbial community structure, while behavioral state influences the abundance of most individual microbes

In order to determine the causal effects of behavioral state and foraging experience on honey bee gut microbiome composition, we controlled for behavioral state while manipulating foraging experience. To do this, we created "big-back colonies" in which some foragers were freely able to leave the colony ("active foragers"), while other bees, who appeared at the entrance and showed an inclination to forage, were prevented from ever leaving the hive due to the presence of a thick plastic tag on their backs ("inactive foragers") [51, 53]. We were thus able to compare bees of the same age and behavioral state that differed in foraging experience.

We found that nurses and inactive foragers did not differ in gut microbial community structure, while active foragers and inactive foragers did (Fig. 3A). This finding indicates that foraging experience plays an important role in defining differences in gut microbial community structure.

When we compared the abundance of individual taxa between age-matched active and inactive foragers, we found that unclassified *Lactobacillus spp.* were higher in relative and absolute abundance in inactive foragers compared to active foragers, *L. apis* was higher in relative abundance in inactive foragers compared to active foragers, and *B. mellis* was lower in relative abundance in inactive foragers compared to active foragers (Fig. 3B-C, Supplementary Table 4). In addition, inactive foragers had a higher absolute abundance of total microbes compared to active foragers (Fig. 3C). This effect seems to be driven by a low absolute total abundance in the active foragers, as the inactive foragers had a median absolute total abundance similar to foragers in typical colonies and three-week old SCCs (Supplementary Tables 1-4). Therefore, while inactive foragers are unable to leave the colony to defecate, differences in microbiome between inactive and active foragers are unlikely to be due to an unhealthy accumulation of bacterial cells in inactive forager guts.

The observed difference in *Lactobacillus spp.* abundance between inactive and active

foragers matches our earlier finding that this taxonomic group decreases in absolute abundance

between nurses and foragers in typical colonies (Fig. 1D), indicating that this microbe group may

be associated with task experience. Interestingly, in our earlier analyses *B. mellis* was higher in

relative and absolute abundance measures in nurses compared to foragers (Fig. 1, Fig. 2,

Supplementary Fig. 4, Table 1). While we did not find an effect on absolute abundance here, it is

surprising that *B. mellis* is higher in relative abundance in active foragers compared to inactive

foragers. While all of our analyses support a connection between *B. mellis* and division of labor,

this relationship may be more complex than originally considered from nurse/forager results

alone. The other division of labor associated microbes (Table 1) did not differ in abundance

between inactive and active foragers.

Overall, these results suggest that the abundance of some microbes is determined by the

experience of foraging outside of the hive, while the abundance of most individual microbes is

defined by behavioral state. Together with our SCC results, which indicate that differences in the

abundance of individual gut microbes may be dependent upon worker experience (Fig. 2D,F,

Supplementary Fig. 4D,F), we speculate that honey bee gut microbes and foraging behaviors

interact through a feedback model. For example, gut microbes may change in abundance due to

foraging experience, possibly incrementally (and thus indiscernibly) at first and building with more

experience, and this change in microbiome may then feed back to support host foraging behaviors

[32]. In the next section, we test the latter part of this model by assessing the effects of inoculation

with individual microbes on host foraging behavior.

Inoculation with individual microbes affects group- and individual-level foraging

intensity

To determine if gut microbes play a causal role in honey bee division of labor, we

performed single microbe inoculations with three out of the six individual honey bee microbes

robustly associated with division of labor as described above: *B. asteroides*, *B. mellis* and *L.*

melliventris (Supplementary Fig. 5A-C). In addition to the strong patterns of association reported

above, we chose these three microbes due to *B. asteroides*' known effects on host physiology

with links to behavior, the high metabolic output associated with taxa in Lactobacillaceae [46],

since gut microbial metabolites likely contribute to the microbiota-gut-brain axis [7], and due to

these three species' previously published associations with honey bee behavioral task [30–32,

109]. Therefore, we reasoned that these three species represented strong candidates to test for

causal effects on honey bee division of labor. Using an automated behavioral tracking system

[71], we determined the effect of inoculation with these three microbes on two measures of

foraging behavior important in division of labor in honey bees: rate of behavioral maturation (age

at onset of foraging) and foraging intensity (amount of foraging trips performed by a group or

individual) [44]. Our results indicate effects only during the first two days of behavioral monitoring,

for reasons that we speculate about in the Discussion.

B. asteroides inoculated bees did not differ from microbiota-depleted bees in behavioral

maturation rate (Fig. 4A) across experimental replicates, but did differ in foraging intensity.

Overall, *B. asteroides* inoculated bees performed the majority of the foraging trips for the colony

on the first and second days after colony formation (Fig. 4B, Supplementary Table 5). These

group-level effects were not due to a difference in number of foragers between inoculated and

microbiota-depleted bees (Fig. 4C, Supplementary Table 5). Rather they were at least partly due

to a difference in individual-level foraging intensity, as individual *B. asteroides* inoculated foragers

performed a majority of the foraging trips for the colony on the first day after colony formation (Fig.

4D, Supplementary Table 5). Likewise, *B. asteroides* experimental colonies displayed a degree

of skew in foraging intensity among individuals (Table 2) and *B. asteroides* inoculated bees

represented a higher proportion of "elite forager" bees, i.e., a small subset of foragers that

performed $\geq$50% of the colony's foraging trips [43, 75], relative to microbiota-depleted bees on

the first and second days after colony formation (Fig. 4E, Supplementary Table 5).

580   Similar to *B. asteroides* inoculated bees, *B. mellis* inoculated bees did not differ from

microbiota-depleted bees in behavioral maturation rate (Fig. 5A). By contrast, *B. mellis*

inoculation did not have an effect on group-level foraging intensity (Fig. 5B, Supplementary Table

5) or number of foragers between inoculated and microbiota-depleted bees (Fig. 5C,

Supplementary Table 5). Rather, *B. mellis* inoculation influenced individual-level foraging

intensity, as *B. mellis* inoculated bees performed a minority of foraging trips for the colony on the

first and second days after colony formation (Fig. 5D, Supplementary Table 5). Additionally, *B.*

mellis inoculation caused a skew in foraging intensity between individual foragers (Table 2), as *B.*

mellis inoculated bees represented a lower proportion of elite foragers than microbiota-depleted

bees on the second day after colony formation (Fig. 5E, Supplementary Table 5).

590   Similar to *B. asteroides* and *B. mellis* inoculated bees, *L. melliventris* inoculated bees did

not differ from microbiota-depleted bees in behavioral maturation rate (Fig. 6A). However, *L.*

melliventris inoculated bees had an opposite effect on foraging intensity from *B. asteroides*

inoculated bees: *L. melliventris* inoculated bees performed the minority of foraging trips for the

colony on the first and second days after colony formation (Fig. 6B, Supplementary Table 5). This

group-level effect was not due to a difference in number of foragers between inoculated and

microbiota-depleted bees (Fig. 6C, Supplementary Table 5). Rather, it was at least partly due to

a difference in individual-level foraging intensity, as individual *L. melliventris* inoculated foragers

performed a minority of the foraging trips for the colony on the second day after colony formation

(Fig. 6D, Supplementary Table 5). Likewise, *L. melliventris* colonies displayed a degree of skew

in foraging intensity between individual foragers (Table 2), and *L. melliventris* bees represented

a lower proportion of elite foragers than microbiota-depleted bees on the second day after colony

formation (Fig. 6E, Supplementary Table 5).

Overall, these results indicate that inoculation with individual gut microbes is sufficient to

cause increases or decreases in foraging intensity between worker bees at both the group- and

individual- level.

Discussion

Here, we show that the gut microbiome plays an important role in modulating foraging

behavior in the honey bee. Specifically, our results indicate that differences in honey bee gut

microbial communities influence honey bee division of labor by affecting foraging intensity, but

not rate of behavioral maturation. This conclusion is supported by both microbe abundance and

behavioral results: *B. asteroides* is found in higher relative abundance in foragers compared to

nurses, and inoculation with this species caused bees to have increased foraging intensity

compared to microbiota-depleted controls. Similarly, *B. mellis* and *L. melliventris* are found lower

in abundance in foragers compared to nurses, and inoculation with either of these species caused

bees to have decreased foraging intensity compared to microbiota-depleted controls.

From a technical perspective, the observation that single microbe inoculations can cause

either an increase or a decrease in foraging intensity indicates that it is unlikely that the results

reported here are due to an artifact of the microbiological or behavioral methods we used. In

addition, the congruence of our microbe abundance and behavioral results indicate that single

microbe inoculations, although they do not represent natural microbiome variations, provide a

useful method for functional analysis.

Overall, our behavioral results show the strongest effects on the first and second days

after experimental colony formation, indicating that the most robust effects of microbiome on

behavior occur within the first few days after inoculation. We suspect that this is due to changing

social dynamics in the experimental honey bee colonies over the six day behavioral tracking

period. Although we did not measure gut microbiomes post colony introduction, it is likely that

changes in the gut microbiomes of the experimental bees occurred after colony formation due to

"homogenization" between individuals due to "trophallaxis," the sharing of gut fluids that contain

nutritive and signaling molecules, and/or colonization by microbes acquired from the colony

environment or non-experimental bees. Therefore, experimental treatment effects on behavior

may be weak or non-existent on later experimental days. Changes in colony dynamics, such as

the pool of available foragers, may also account for weak behavioral effects observed later in the

experiment. Foraging is an energetically costly task and is associated with high mortality

compared to in-hive behaviors [110–112]. Indeed, across all types of experimental colonies in our

studies, foragers that began foraging early after colony formation had an earlier last day of

foraging (which likely corresponds with day of death [44, 74]) (Supplementary Fig. 5D-F). This

indicates that our experimental colonies had a typical high forager turnover.

We observed additional interesting changes in colony foraging force due to foraging

activity. In the case of *B. mellis* experimental colonies, only microbiota-depleted foragers that

began foraging on the first day had a significantly earlier last day of foraging compared to later

foragers (Supplementary Fig. 5E). In addition, microbiota-depleted elite foragers had a

significantly earlier first day of foraging compared to *B. mellis* inoculated elite foragers

(Supplementary Fig. 5G). These results indicate that in *B. mellis* experimental colonies, forager

turnover was associated with treatment such that microbiota-depleted foragers had their foraging

career at younger ages compared to *B. mellis* inoculated foragers. A similar but non-significant

trend in elite forager onset was also observed in *L. melliventris* colonies (Supplementary Fig. 5G).

We suggest that this likely accounts for the (non-significant) switch in group-level and elite

foraging intensity seen on the later experimental days in *B. mellis* (Fig. 5B, Fig. 5E) and *L.*

melliventris (Fig. 6B, Fig. 6E), but not *B. asteroides* (Fig. 4) experimental colonies.

Previous studies reported correlational differences in gut microbiome associated with

honey bee behavioral task [30–32] and behavioral maturation [50]. These studies also reported

associations of *Bifidobacterium*, *B. mellis* and *L. melliventris* with behavioral task [30, 32] and

ontogeny [31], supporting a link between these microbes and division of labor in honey bees.

Here we found that *B. mellis* and *L. melliventris* were higher in abundance in nurses compared to

foragers, matching results from [30, 32]. However, while we found that *B. asteroides* was higher

in relative abundance in typical-age foragers relative to over-age nurses, results in [30, 32] found

that *Bifidobacterium* was higher in relative [32] and absolute abundance [30] in nurses compared

to foragers. This difference may be due to the difference in taxonomic level used in analysis

between these studies and ours, as we found that other *Bifidobacterium* species were higher in

abundance in nurses compared to foragers (trend in Fig. 1C-D and Fig. 2D, significant in Fig. 2F).

In addition, these studies found an increase in *S. alvi* with age [31], and a higher abundance of

Gilliamella and *Klebsiella* associated with nurses [30], which match our abundance results (Table

1). Here, we provide evidence that at least some of these gut microbes play a causal role in

defining behavioral differences between honey bees. Therefore, the congruence of our

abundance results with other studies indicate that our behavioral results are likely generalizable

across populations of honey bees.

Our findings are consistent with a previously proposed model in which positive feedback

interactions cause adaptive changes in both gut microbiome composition and host behavior [32].

Under this model, we speculate that gut microbes could change in abundance due to time spent

foraging, likely in association with environmental exposure and/or changes in diet and metabolic

needs. According to this speculation, the change in microbiome would then feed back to influence

host foraging behaviors. We suggest that changes in gut microbe abundance (e.g. increases in

B. asteroides with decreases in *B. mellis* and *L. melliventris*) and host behavior interact to increase

foraging intensity. Modeling interactions in this way can provide the basis for future studies that

probe the neural and metabolic mechanisms by which *B. asteroides*, *B. mellis* and *L. melliventris*

affect host behavior.

Previous studies have shown that gut microbes may influence host behavior through a

variety of mechanisms, including production of metabolites that cause changes in host brain gene

expression or host production of neurotransmitters and hormones [7]. Within honey bees,

Bifidobacterium, *Bombilactobacillus* Firm-4 and *Lactobacillus* Firm-5 are the major fermenters of

pollen-derived compounds and have large effects on the abundance of metabolites in the hindgut

[46]. In addition, *B. asteroides* inoculation has been associated with an increased production of

two juvenile hormone derivatives [46]. Changes in gene expression and the production of

neurotransmitters and hormones, most notably juvenile hormone, are associated with the

regulation of honey bee division of labor [42, 44, 74, 95–98, 101, 107, 113, 114], and therefore it

is possible that gut microbes influence honey bee division of labor through these pathways.

Likewise, previous studies indicate that some host factors associated with honey bee division of

labor, such as social interactions and diet, may also modulate gut microbiome composition,

including the abundance of these three microbes [7, 30, 48, 115, 116]. Future research will work

to identify specific mechanisms by which honey bees and their gut microbiomes interact to

influence worker behavior. Overall, studies using the honey bee as a model have the potential to

further elucidate the microbiota-gut-brain axis, and achieve a more comprehensive understanding

of interkingdom interactions.

695   Together, our results suggest that in naturally occurring bee populations, increases in *B.*

asteroides with simultaneous decreases in *B. mellis* and *L. melliventris* may act synergistically to

increase the intensity of foraging behavior in individual bees. Honey bees, with a complex eusocial

life history, live in large perennial colonies and therefore collect large amounts of nectar and pollen

to ensure colony survival during periods when floral resources are not available. Thus, the

influence of these three microbes on honey bee foraging intensity indicates that host-microbe

interactions likely help sustain the perennial lifestyle of this eusocial insect. Future studies may

address the interaction between these three microbes, other members of the honey bee

microbiome, and between the microbiome and the host in order to gain a more naturalistic and

comprehensive understanding of the role of the gut microbiome in social insect division of labor.

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Competing Interests

The authors declare no competing interests.

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Figures

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Figure 1. Nurses and foragers from typical honey bee colonies differ in gut microbial community. (A-B) Nurses and foragers differed in overall gut microbial community structure. (A) Reanalyzed data from Kapheim et al 2015. Two-way Permutation MANOVA using Aitchison Distance, Task: $F_{1,38}$ = 1.51, R^2 = 0.036, p = 0.016; Colony: $F_{4,38}$ = 1.72, R^2 = 0.165, p = 0.001,

Task*Colony: $F_{2,38}$ = 1.17, R^2 = 0.056, p = 0.100. n = 1-12 bees/colony, 5 colonies. (B) New data.

Two-way Permutation MANOVA using Aitchison Distance, Task: $F_{1,59}$ = 4.28, R^2 = 0.066, p =

0.001; Colony: $F_{2,59}$ = 2.31, R^2 = 0.071, p = 0.001, Task*Colony: $F_{2,59}$ = 1.20, R^2 = 0.037, p =

0.144. n = 10 bees/colony, 3 colonies. Depicted as Principal Components Analysis (PCA) plots.

Lowercase letters in legends denote statistically significant groups. **(C)** Nurses and foragers

differed in relative abundance of four individual microbial species (new data only). Depicted as

stacked bar plots, with each bar representing a single bee's microbiome. Asterisks in legend: *,

p$\leq$0.05, **, p$\leq$0.01, ANCOM-BC between nurses and foragers. See Supplementary Table 1 for all

p-values. **(D)** Nurses and foragers differed in absolute abundance of four individual microbial

species but not in the total normalized number of 16S rRNA gene copies (new data only). 10^x

number of 16S rRNA gene copies, calculated by multiplying the relative abundance each microbe

in each sample (determined through 16S rRNA sequencing) by the normalized number of 16S

rRNA gene copies in the sample (determined through qPCR). Depicted as dot plots with all data

points plotted, line represents median, n = 10 bees/colony, 3 colonies. *, p$\leq$0.05, **, p$\leq$0.01,

Permutation ANOVA Test between nurses and foragers. See Supplementary Table 1 for all p-

values.

Figure 2. Age-matched nurses and foragers differ in gut microbial community. (A-B) Age-matched typical-age nurses and precocious foragers from a single cohort colony did not differ in gut microbial community structure at about one week of age (A), but age-matched over-age nurses and typical-age foragers significantly differed in gut microbial community structure at three weeks of age (B). 1 week: Two-way Permutation MANOVA using Aitchison Distance, Task: $F_{1,39}$ = 0.92, R^2 = 0.023, p = 0.519; Source colony: $F_{1,39}$ = 2.60, R^2 = 0.064, p = 0.004, Task*Colony: $F_{1,39}$ = 1.15, R^2 = 0.028, p = 0.296. n = 10 bees/source colony, 2 source colonies. 3 weeks: Two-way Permutation MANOVA using Aitchison Distance, Task: $F_{1,79}$ = 6.49, R^2 = 0.074, p = 0.001; Source colony: $F_{3,79}$ = 1.82, R^2 = 0.062, p = 0.001, Task*Colony: $F_{3,79}$ = 1.23, R^2 = 0.042, p = 0.084. n = 10 bees/source colony, 4 source colonies. Depicted as PCA plots. Lowercase letters in legends denote statistically significant groups. **(C-D)** Age-matched typical-age nurses and precocious foragers from a single cohort colony did not differ in relative abundance of individual microbial species (C), while age-matched over-age nurses and typical-age foragers differed in relative abundance of five individual microbial species at three weeks of age (D). Depicted as stacked bar plots, with each bar representing a single bee's microbiome. Asterisks in legend: *, p≤0.05, **, p≤0.01, ANCOM-BC between nurses and foragers. See Supplementary Table 2 for all p-values. **(E-F)** Age-matched typical-age nurses and precocious foragers from a single cohort colony did not differ in absolute abundance of individual microbial species but did differ in the total normalized number of 16S rRNA gene copies (E), while age-matched over-age nurses and typical-age foragers differed in absolute abundance of six microbial species and the total number of 16S rRNA gene copies at three weeks of age (F). 10^x number of 16S rRNA gene copies, calculated by multiplying the relative abundance each microbe in each sample (determined through 16S rRNA sequencing) by the normalized number of 16S rRNA gene copies in the sample (determined through qPCR). Depicted as dot plots with all data points plotted, line represents median, n = 10 bees/source colony, 2 source colonies (1 week) or 4 source colonies (3 weeks).

1056 *, p≤0.05, **, p≤0.01, Permutation ANOVA Test between nurses and foragers. See

1057 Supplementary Table 2 for all p-values.

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Figure 3. Age-matched inactive and active foragers differ in gut microbial community structure, but not the abundance of individual microbes. (A) Age-matched nurses and inactive foragers were similar in gut microbial community structure, while age-matched active and inactive foragers differed in gut microbial community structure. Two-way Permutation MANOVA using Aitchison Distance, Task: $F_{2,86}$ = 1.84, R^2 = 0.038, p = 0.011; Colony: $F_{2,86}$ = 6.02, R^2 =

0.124, p = 0.001, Task*Colony: $F_{4,86}$ = 0.798, R^2 = 0.033, p = 0.883. n = 10 bees/colony, 3

colonies. Depicted as PCA plot. Lowercase letters in legends denote statistically significant

groups as determined by Pairwise Permutation MANOVA. **(B)** Age-matched inactive and active

foragers did not differ in the relative abundance of individual microbial species (B). Depicted as

stacked bar plots, with each bar representing a single bee's microbiome. Asterisks in legend: *,

p≤0.05, **, p≤0.01, ANCOM-BC between inactive and active foragers. See Supplementary Table

4 for all p-values. **(C)** Age-matched inactive and active foragers did not differ in absolute

abundance of individual microbial species but did differ in the total normalized number of 16S

rRNA gene copies. 10^x number of 16S rRNA gene copies, calculated by multiplying the relative

abundance each microbe in each sample (determined through 16S rRNA sequencing) by the

normalized number of 16S rRNA gene copies in the sample (determined through qPCR). Depicted

as dot plots with all data points plotted, line represents median, n = 10 bees/colony, 3 colonies. *,

p≤0.05, **, p≤0.01, Permutation ANOVA Test between inactive and active foragers. See

Supplementary Table 4 for all p-values.

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Figure 4. Bees inoculated with *B. asteroides* do not differ from microbiota-depleted bees in behavioral maturation (age at onset of foraging), but do differ in foraging intensity. (A) *B. asteroides* inoculated bees did not differ from microbiota-depleted bees in age at onset of foraging. Cox Proportional Hazards, $z = -0.331$, $p = 0.741$. **(B)** *B. asteroides* inoculated foragers, as a group, performed the majority of foraging trips for the colony on the first and second days after experimental colony formation. Linear Mixed Effects Model, Treatment: $F_{1,33} = 21.16$, $p < 0.001$, Day: $F_{5,33} = 0$, $p = 1$, Treatment*Day: $F_{5,33} = 1.987$, $p = 0.107$. See Supplementary Table 5 for Pairwise comparisons. **(C)** *B. asteroides* inoculated bees represented a similar number of foragers as microbiota-depleted bees on all days after experimental colony formation. Linear Mixed Effects Model, Treatment: $F_{1,33} = 0.409$, $p = 0.527$, Day: $F_{5,33} = 12.095$, $p < 0.001$, Treatment*Day: $F_{5,33} = 0.140$, $p = 0.982$. See Supplementary Table 5 for Pairwise comparisons. **(D)** Individual *B. asteroides* inoculated foragers performed a majority of foraging trips for the colony the first day after experimental colony formation. Generalized Linear Mixed Effects Model

with log-normal distribution, Treatment: χ^2 = 4.985, p = 0.026, Day: χ^2 = 382.933, p < 0.001,

Treatment*Day: χ^2 = 26.315, p < 0.001. See Supplementary Table 5 for Pairwise comparisons.

(E) *B. asteroides* inoculated bees represented a higher proportion of elite foragers than

microbiota-depleted bees on the first and second days after experimental colony formation. Linear

Mixed Effects Model, Treatment: $F_{1,33}$ = 15.799, p < 0.001, Day: $F_{5,33}$ = 0, p = 1, Treatment*Day:

$F_{5,33}$ = 3.151, p = 0.020. See Supplementary Table 5 for Pairwise comparisons. (A) depicted as

survival plot. All other data depicted as box plots with data points plotted, thick horizonal line

represents median, x represents mean, whiskers represent the minimum and maximum values,

thin lines connect paired points for each colony, n = 4 colonies. Asterisks used to denote

comparisons between treatment groups on each day only: *, p≤0.05, **, p≤0.001.

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Figure 5. Bees inoculated with *B. mellis* do not differ from microbiota-depleted bees in behavioral maturation (age at onset of foraging), but do differ in foraging intensity. (A) *B. mellis* inoculated bees did not differ from microbiota-depleted bees in age at onset of foraging. Cox Proportional Hazards, z = 1.525, p = 0.127. **(B)** *B. mellis* inoculated foragers, as a group, performed a similar proportion of foraging trips for the colony as microbiota-depleted bees all days after experimental colony formation, with a marginal effect of performing a minority of foraging trips for the colony on the second day after colony formation. Linear Mixed Effects Model, Treatment: $F_{1,29.091}$ = 0.036, p = 0.851, Day: $F_{5,29.719}$ = 0, p = 1, Treatment*Day: $F_{5,29.091}$ = 1.981, p = 0.111. See Supplementary Table 5 for Pairwise comparisons. **(C)** *B. mellis* inoculated bees represented a similar number of foragers as microbiota-depleted bees on all days after experimental colony formation. Linear Mixed Effects Model, Treatment: $F_{1,33}$ = 3.183, p = 0.084, Day: $F_{5,33}$ = 14.918, p < 0.001, Treatment*Day: $F_{5,33}$ = 0.765, p = 0.582. See Supplementary Table

5 for Pairwise comparisons. **(D)** Individual *B. mellis* inoculated foragers performed a minority of

foraging trips for the colony on the first and second days after experimental colony formation.

Generalized Linear Mixed Effects Model with log-normal distribution, Treatment: χ^2 = 3.626, p =

0.057, Day: χ^2 = 255.497, p < 0.001, Treatment*Day: χ^2 = 6.552, p = 0.256, See Supplementary

Table 5 for Pairwise comparisons. No statistical outliers were detected in these data. **(E)** *B. mellis*

inoculated bees represented a lower proportion of elite foragers than microbiota-depleted bees

on the second day after experimental colony formation. Linear Mixed Effects Model, Treatment:

$F_{1,29.091}$ = 1.896, p = 0.179, Day: $F_{5,29.719}$ = 0, p = 1, Treatment*Day: $F_{5,29.091}$ = 2.183, p = 0.083.

See Supplementary Table 5 for Pairwise comparisons. (A) depicted as survival plot. All other data

depicted as box plots with data points plotted, thick horizonal line represents median, x represents

mean, whiskers represent the minimum and maximum values, thin lines connect paired points for

each colony, n = 4 colonies. Asterisks used to denote comparisons between treatment groups on

each day only: *, $p \leq 0.05$, **, $p \leq 0.001$.

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Figure 6. Bees inoculated with *L. melliventris* do not differ from microbiota-depleted bees in behavioral maturation (age at onset of foraging), but do differ in foraging intensity. (A) *L. melliventris* inoculated bees did not differ from microbiota-depleted bees in age at onset of foraging. Cox Proportional Hazards, z = -0.435, p = 0.664. (**B**) *L. melliventris* inoculated foragers, as a group, performed the minority of foraging trips for the colony on the first and second days after experimental colony formation. Linear Mixed Effects Model, Treatment: $F_{1,31.021}$ = 0.351, p = 0.558, Day: $F_{5,31.443}$ = 0, p = 1, Treatment*Day: $F_{5,31.021}$ = 4.169, p = 0.005. See Supplementary Table 5 for Pairwise comparisons. **(C)** *L. melliventris* inoculated bees represented a similar number of foragers as microbiota-depleted bees on all days after experimental colony formation. Linear Mixed Effects Model, Treatment: $F_{1,33}$ = 0.079, p = 0.781, Day: $F_{5,33}$ = 18.923, p < 0.001, Treatment*Day: $F_{5,33}$ = 0.132, p = 0.984. See Supplementary Table 5 for Pairwise comparisons. **(D)** Individual *L. melliventris* inoculated foragers performed a minority of foraging trips for the colony the second day after experimental colony formation. Generalized Linear Mixed Effects

Model with log-normal distribution, Treatment: $\chi^2 = 0.141$, p = 0.707, Day: $\chi^2 = 334.831$, p < 0.001,

Treatment*Day: $\chi^2 = 18.336$, p = 0.003, See Supplementary Table 5 for Pairwise comparisons.

No statistical outliers were detected in these data. **(E)** *L. melliventris* inoculated bees represented

a lower proportion of elite foragers than microbiota-depleted bees on the first and second days

after experimental colony formation. Linear Mixed Effects Model, Treatment: $F_{1,31.021} = 0.0006$, p

= 0.981, Day: $F_{5,31.443} = 0$, p = 1, Treatment*Day: $F_{5,31.021} = 3.961$, p = 0.007. See Supplementary

Table 5 for Pairwise comparisons. (A) depicted as survival plot. All other data depicted as box

plots with data points plotted, thick horizonal line represents median, x represents mean, whiskers

represent the minimum and maximum values, thin lines connect paired points for each colony, n

= 4 colonies. Asterisks used to denote comparisons between treatment groups on each day only:

*, $p \leq 0.05$, **, $p \leq 0.001$.

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Microbe	Relative abundance			Absolute abundance		
	Typical colonies	SCC Replicate 1 (week 3)	SCC Replicate 2 (week 3)	Typical colonies	SCC Replicate 1 (week 3)	SCC Replicate 2 (week 3)
Acetobacteraceae		lower abundance	lower abundance	lower abundance		lower abundance
Bartonella spp.		lower abundance				
B. asteroides		lower abundance	lower abundance			
B. coryneforme					higher abundance	
Bom. apis						
F. perrara		lower abundance				
G. apicola					higher abundance	higher abundance
Lactobacillus spp.				higher abundance		
L. apis						higher abundance
L. helsingborgensis	higher abundance					
L. kullabergensis				higher abundance		higher abundance
A. kunkeei						
B. mellifer						
B. mellis	higher abundance				higher abundance	higher abundance
L. melliventris	higher abundance	higher abundance	higher abundance	higher abundance	higher abundance	
S. alvi		lower abundance		lower abundance	lower abundance	
Arsenophonus spp.						
Klebsiella spp.		higher abundance	higher abundance	higher abundance	higher abundance	higher abundance
Limosilactobacillus spp.						
Mixta spp.					higher abundance	

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1160 **Table 1: Summary table of individual microbe abundance results.** Microbes that significantly

1161 differed in relative (left) and absolute (right) abundance measures in nurses relative to foragers in

1162 typical colonies and individual single cohort colony replicates. Higher abundance, higher in

1163 abundance in nurses relative to foragers (pink). Lower abundance, lower in abundance in nurses

1164 relative to foragers (blue).

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Test Microbe	Colony Replicate	Overall Colony Gini Coefficient	Day 1 Gini Coefficient	Day 2 Gini Coefficient	Day 3 Gini Coefficient
B. asteroides	1	0.376	0.056	0.116	0.425
B. asteroides	2	0.362	0.352	0.32	0.278
B. asteroides	3	0.415	0.133	0.445	0.368
B. asteroides	4	0.305	0.199	0.303	0.242
B. mellis	1	0.339	0.13	0.32	0.304
B. mellis	2	0.459	0.304	0.453	0.355
B. mellis	3	0.31	NA	NA	0.31
B. mellis	4	0.302	0.126	0.23	0.363
L. melliventris	1	0.398	0.186	0.271	0.365
L. melliventris	2	0.283	NA	NA	0.281
L. melliventris	3	0.419	0.167	0.386	0.415
L. melliventris	4	0.391	0.174	0.453	0.267

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Table 2: Gini coefficients for each experimental colony depicted in Figures 4-6. Gini

coefficients, representing the degree of skew in foraging intensity between individual foraging

bees, where 0 represents complete equality in foraging intensity between individuals and 1

represents complete inequality in foraging intensity between individuals, measured for each

experimental colony across all days ("Overall Colony Gini Coefficient"), and for each experimental

colony on each individual day ("Day X Gini Coefficient"). NAs represent days with bad weather in

which foraging did not occur.

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