Check for updates

OPEN ACCESS

EDITED BY Decai Jin, Chinese Academy of Sciences (CAS), China

REVIEWED BY Anbu Poosakkannu, University of Turku, Finland Sunil Banskar, University of Arizona, United States

*CORRESPONDENCE Libiao Zhang ⊠ zhanglb@giz.gd.cn

RECEIVED 17 April 2023 ACCEPTED 10 July 2023 PUBLISHED 28 July 2023

CITATION

Guo M, Xie S, Wang J, Zhang Y, He X, Luo P, Deng J, Zhou C, Qin J, Huang C and Zhang L (2023) The difference in the composition of gut microbiota is greater among bats of different phylogenies than among those with different dietary habits. *Front. Microbiol.* 14:1207482. doi: 10.3389/fmicb.2023.1207482

COPYRIGHT

© 2023 Guo, Xie, Wang, Zhang, He, Luo, Deng, Zhou, Qin, Huang and Zhang. This is an openaccess article distributed under the terms of the Creative Commons Attribution License (CC BY). The use, distribution or reproduction in other forums is permitted, provided the original author(s) and the copyright owner(s) are credited and that the original publication in this journal is cited, in accordance with accepted academic practice. No use, distribution or reproduction is permitted which does not comply with these terms.

The difference in the composition of gut microbiota is greater among bats of different phylogenies than among those with different dietary habits

Min Guo¹, Siwei Xie², Junhua Wang¹, Yuzhi Zhang¹, Xiangyang He¹, Pengfei Luo¹, Jin Deng^{1,3}, Chunhui Zhou¹, Jiao Qin⁴, Chen Huang⁵ and Libiao Zhang¹*

¹Guangdong Key Laboratory of Animal Conservation and Resource Utilization, Guangdong Public Laboratory of Wild Animal Conservation and Utilization, Institute of Zoology, Guangdong Academy of Sciences, Guangzhou, China, ²College of Mathematics and Informatics, South China Agricultural University, Guangzhou, China, ³College of Life Sciences, South China Normal University, Guangzhou, China, ⁴Institute of Plant Protection, Guangdong Academy of Agricultural Sciences, Guangzhou, China, ⁵Dr. Neher's Biophysics Laboratory for Innovative Drug Discovery, State Key Laboratory of Quality Research in Chinese Medicine, Macau University of Science and Technology, Taipa, Macao SAR, China

Bats have a very long evolutionary history and are highly differentiated in their physiological functions. Results of recent studies suggest effects of some host factors (e.g., phylogeny and dietary habit) on their gut microbiota. In this study, we examined the gut microbial compositions of 18 different species of bats. Results showed that Firmicutes, Gammaproteobacteria, and Actinobacteria were dominant in all fecal samples of bats. However, the difference in the diversity of gut microbiota among bats of different phylogenies was notable (p = 0.06). Various species of Firmicutes, Actinobacteria, and Gammaproteobacteria were found to contribute to the majority of variations in gut microbiota of all bats examined, and Aeromonas species were much more abundant in bats that feed on both insects and fish than in those of insectivores. The abundance of various species of Clostridium, Euryarchaeota, and ancient bacterial phyla was found to vary among bats of different phylogenies, and various species of Vibrio varied significantly among bats with different dietary habits. No significant difference in the number of genes involved in various metabolic pathways was detected among bats of different phylogenies, but the abundance of genes involved in 5 metabolic pathways, including transcription; replication, recombination, and repair; amino acid transport and metabolism; and signal transduction mechanisms, was different among bats with different dietary habits. The abundance of genes in 3 metabolic pathways, including those involved in stilbenoid, diarylheptanoid, and gingerol biosynthesis, was found to be different between insectivorous bats and bats that feed on both insects and fish. Results of this study suggest a weak association between dietary habit and gut microbiota in most bats but a notable difference in gut microbiota among bats of different phylogenies.

KEYWORDS

bat, phylogeny, dietary habit, gut microbiota, dietary shift

Introduction

Gut microbiota is composed of many bacteria, fungi, viruses, and other organisms in the gastrointestinal system. The composition of gut microbiota in animals varies greatly in their early life but becomes stable in adulthood (Lim et al., 2015; Yin et al., 2020). Many largescale studies have revealed effects of gut microbiota on the well-being of the host (Wan et al., 2020; Kuang et al., 2022). Host factors such as phylogeny, physiology, behavior, and dietary habit have also been shown to affect the composition of gut microbiota (Benson et al., 2010; Kurilshikov et al., 2021).

A recent large-scale study of the gut bacteria of bats, amphibians, and birds showed that bats and birds share similar profiles of gut bacteria (Song et al., 2020), suggesting that an innate ability, such as flying, may have similar effects on gut bacteria in animals of different phylogenies. In contrast, other studies disclosed a link between specific microorganisms and their hosts (Brucker and Bordenstein, 2012; Easson and Thacker, 2014; Sanders et al., 2014; Brooks et al., 2016). Such relationship is referred to as phylosymbiosis that predicts host clades harbor distinguishable microbial communities (Brucker and Bordenstein, 2012) and is common in some animals, such as sponges (Easson and Thacker, 2014), Nasonia (Brucker and Bordenstein, 2012), ants, and apes (Sanders et al., 2014). These findings suggest a relationship between the composition of gut microbiota and host evolutionary history, behaviors, and ecological factors (Zepeda Mendoza et al., 2018; Lutz et al., 2019).

Bat is the second largest order of mammals and have a long evolutionary history, distinct physiology, diverse dietary habits, echolocations, and flying ability (Adams and Pedersen, 2013). Similar to carnivores, the gut microbiota of bats are dominated by Firmicutes and Proteobacteria (Sun et al., 2020). It has been shown that the compositions of gut bacteria of bats vary with their diet, sex, age, mass, geographic location, physiology, reproduction, and symbiotic parasites (Phillips et al., 2012; Carrillo-Araujo et al., 2015; Lutz et al., 2019). Although the effects of host factors on gut bacteria have been described (Phillips et al., 2012; Carrillo-Araujo et al., 2015; Lutz et al., 2019), the mechanisms of such effects are unknown. We hypothesized that host phylogeny and dietary habits could affect the diversity of gut microbiota and therefore conducted this study to compare the compositions of gut microbiota in bats of different phylogenies and those of insect-eating (insectivores) bats, fruit-eating (frugivores) bat, and bats that feed on both insects and fish, aiming to uncover the impact of various host factors on the gut microbiota of bats.

Materials and methods

Samples

A total of 24 bats of 18 different species belonging to 8 genera and 3 families (Pteropodidae, Rhinolophidae, and Vespertilionidae) were captured from Yanyan Cave, Huizhou City, China. These bats included 6 *Monascus pilosus* bats, 2 *Monascus chinensis* bats, and one each of the following: *Monascus altarium, Monascus davidii, Monascus longipes, Rhinolophus pearsonii, Rhinolophus pusillus, Rhinolophus siamensis, Rhinolophus affinis, Rhinolophus macrotis, Hypsugo pulveratus, Eptesicus serotinus, Pipistrellus abramus, Murina huttoni, Murina aurata, Miniopterus pusillus, Macropus fuliginosus, Rousettus* *leschenaultii* (Wilson and Mittermeier, 2019). These bat species were selected because they have different dietary habits in addition to different phylogenies. Among the 18 bat species, *M. pilosus* bats feed on both insects and fish, *R. leschenaultii* bat is frugivores, and the remaining bats are insectivores (Wilson and Mittermeier, 2019).

To collect fecal samples, each bat was kept in a clean cloth bag for about 5 h. Fecal particles were then collected and placed in 2 mL tubes. For those with no fecal particles, anal swabs were used. A total of 24 fecal samples were collected (one sample from each bat). The samples were immediately snap frozen in liquid nitrogen and then stored in a -80° C freezer until used.

DNA extraction, metagenomic assembly, and taxonomical annotation

DNA was extracted from fecal samples using the QIAamp[®] Fast DNA Stool Mini kit following manufacturer's instructions without any modifications. DNA samples were then subjected metagenomic analysis by next generation sequencing (NGS) with a NovaSeq 6000 sequencer (read length: 150 bases; sequencing depth: 10 Gb; paired-end sequencing by Shanghai Origingene Corporation, China).

To filter out low-quality and adapter contaminated reads, the raw reads were subjected to quality control by SOAPnuke (1.5.6; Chen et al., 2018). The high-quality reads were assembled by Megahit¹ (Li et al., 2015). MetaGeneMark² (Zhu et al., 2010) was employed to search for coding sequences (CDSs) in assembled contigs longer than 100 bp. To construct a non-redundant gene set, genes of all samples were merged and clustered with 95% identity by Cd-hit (v4.6.4)³ (Li and Godzik, 2006). Taxonomic identification of genes in the non-redundant gene set was performed with Blastp (v2.2.28)⁴ (Camacho et al., 2009) by aligning the genes against those in the Nucleotide Sequence Database (Nr database; v20170519) at an e-value $\leq 1e-5$. When the gene length coverage was less than 50%, the gene was assigned an "Unclassified" status. Genes were annotated by aligning high quality reads to the non-redundant gene set using Bowtie2 (v2.2.5; Langmead and Salzberg, 2012), and gene abundance was calculated by PathoScope2 (v2.0.6; Hong et al., 2014). Species abundance was determined by the sum of assigned gene counts. The taxonomic tree of microbes was built according to the NCBI taxonomy. Divergence time of bat species in the phylogenetic tree was determined by TimeTree 5⁵ (Kumar et al., 2022).

Gene function and metabolic pathway annotations

The non-redundant gene set was annotated by blasting translated proteins against the database of Clusters of Orthologous Groups of proteins (COG) (v20090331) and that of Kyoto Encyclopedia of Genes and Genomes (KEGG) (v81). COG-annotated gene functions were

4 http://blast.ncbi.nlm.nih.gov/Blast.cgi

¹ http://i.cs.hku.hk/~alse/hkubrg/projects/idba_ud/

² http://exon.gatech.edu/meta_gmhmmp.cgi

³ http://www.bioinformatics.org/cd-hit/

⁵ http://www.timetree.org/

visualized using the boxplot function of the Lattice package in the R software. Significantly changed gene functions and pathways between groups were detected by the Wilcoxon test of the R software (adjusted p < 0.05, by the method of Holm).

Evaluation of microbial diversity

To assess the compositional homogeneity of the microbial communities in samples, Shannon–Wiener index (Shannon, 2001) was calculated using vegan, ggplot2, RColorBrewer, and reshape2 packages of the R software. Weighted principal component analysis (PCA) was executed using ggbiplot, ggrepel, and ggplot2 packages. *Z*-score normalization (Shalabi et al., 2006) was performed with the prcomp package prior to calculation of the principal components. The ggbiplot package was used to visualize the distribution of each sample in PC1 and PC2. Significant difference between groups was determined using the adonis package of the R software (permutations=999, by the method of Euclidean; Anderson, 2001).

Identification of significant gut microorganisms and host metabolic pathways

To identify the significant gut microorganisms, 500 tree models were constructed by random forest (RF) modeling. Each tree model was trained by bootstrapping samples from all samples; those samples unemployed in the training (~36.8%) were used to test the accuracy of the tree model (Breiman, 2001). RF models were built by the package randomForest of the R software (with parameters: ntree=500, mtry = sqrt(p) where p is number of variables, replace = TRUE, maxnodes = NULL, oob.prox = proximity, nPerm = 1, maxnodes = NULL) to analyze data of bats with different dietary habits and those of different phylogenies by random permutation of species-level abundance profiles. The out-of-bag (OOB) error rate was used to assess the accuracy of modeling. Microbial influence value was determined by the mean decrease in the Gini index of each node of the constructed model, and a higher value indicated a more significant effect on the composition of gut microbiota. The number of microbial species was determined by repeating RF modeling 5 times with 10-fold cross validation.

The significantly changed (p < 0.01) abundance of bacterial species between insectivorous bats and bats that eat both insects and fish were determined by Metastats non-parametric multiple tests with p-values adjusted for false discovery rate (FDR) to aviod errors (White et al., 2009). The significant difference in the number of genes involved in certain functions or pathways between groups were detected by the Wilcoxon test (adjusted p < 0.05 by the method of Holm). The abundance profiles of various significant microbes were visualized in heat map by packages gplots, fields, and akima of the R software.

Results

Sequencing data

The NGS performed with the NovaSeq 6000 platform on the 24 bat fecal samples generated approximately 7.9 GB raw data per sample.

After cleaning, an average of 7.8 Gb clean data were obtained from each sample with 22,213–1,050,926 contigs and 25,969–688,291 genes. The average N50 of all genes was 570 bp (Figure 1), and 7,525,231 unique genes were obtained from all samples.

Variations in gut microbiota among bats of different phylogenies or with different diets

After taxonomic annotation, 394 archaea species, 18,271 bacterial species, 757 fungal species, and 1,912 viral species were detected. Except for *P. abramus*, all samples showed similar microbial profiles at the phylum level (Figure 2A) with Proteobacteria being most predominant (average relative abundance: 0.658), followed by Firmicutes (0.204), norank_d__Viruses (0.054), Chlamydiae (0.018), and Deinococcus-Thermus (0.010). The predominant bacterial genera were *Enterobacter* (0.054), *Escherichia* (0.047), norank_f__ Enterobacteriaceae (0.041), *Serratia* (0.040), *Klebsiella* (0.032), *Bacillus* (0.031), *Morganella* (0.030), *Lactococcus* (0.022), *Citrobacter* (0.021), *Aeromonas* (0.019), *Kluyvera* (0.017), and *Enterococcus* (0.015) (Supplementary Figure S1).

There was no significant difference in microbial diversity was detected among bats with insect (IB) or both insect and fish (IFB) diets by statistical (Wilcoxon test adjusted by the method of Holm, all p > 0.05) and PCA analyses (p = 0.69; Figure 2B). Meanwhile, in PCA analysis, there was no significant difference at the family level of microbes among bats of different phylogenies ($R^2 = 0.049$, p = 0.743) (Supplementary Figure S2), but a high variation in microbial diversity at the genus level was observed ($R^2 = 0.54$, p = 0.06; Figure 2B).

Identification of microbes responsible for major variations in gut microbiota and major metabolic pathways in bats

The random forest (RF) method was used to identify key microbes responsible for major variations in the composition of gut microbiota, and 53 different species of microbes were found among bats of different phylogenies (PG bats; OOB error rate: 41.67%). For bats with different dietary habits (DG bats), 38 different species of microbes were found (OOB error rate: 16.67%) (Figure 3A). By constructing a phylogenetic tree of all 91 microbes, species of Firmicutes, Actinobacteria, and Gammaproteobacteria were found to contribute the most variations in gut microbiota in all bats. For bats of different phylogenies, species of *Clostridium*, *Euryarchaeota*, and ancient bacterial phyla contributed the major variation. In contrast, species of *Vibrio* contributed almost solely the major variation in gut microbiota in bats with different dietary habits (Figure 3B).

For all bats, the predominant orthologous gene functions were Transcription, Replication, recombination, and repair, Amino acid transport and metabolism, and Signal transduction mechanisms (Supplementary Figure S3). The predominant metabolic pathways were Transporters for insectivorous bats (IB), Transporters and Carbon metabolism for bats that feed on both insects and fish (IFB), and Transporters and Biosynthesis of amino acids for frugivorous bat (FB) (Supplementary Figure S4). No significant difference in the





Diversity of gut microbiota in various species of bats. (A) Top 18 microbial phyla in gut microbiota of bats. The bar below the phylogenetic tree indicates divergence time of the bat species. (B) Diversity of gut microbiota in bats with different diet habits (the two upper panels) and those of different phylogeneis (at genus level, the two lower panels).



FIGURE 3

Major microbial species in gut microbiota of bats. (A) Gray lines show results of 5 repeats of RF modeling with 10-fold cross validations. Dark lines indicate average values of the 5 repeats. Dashed lines denote the lowest point of the dark line, where the model had the lowest out-of-bag (OOB) error. The bar length in MeanDecreaseGini shows the extent of microbes affecting node impurity in RF modeling, and a microbe with a longer bar has a higher effect on the composition of gut microbiota. (B) Phylogenetic tree of the 91 most significant microbes among bats. Red branches represent the 53 most significant microbes among bats of different phylogenies, and green branches denote the 38 most significant microbes among bats with different dietary habits. Scale bar length indicates the number of amino acid substitutions per site.

number of these orthologous genes or metabolic pathways was detected among all bats. However, the number of genes of metabolic pathways was found to vary among bats with different dietary habits as follows: Photosynthesis - antenna proteins between IB and IFB (p = 0.036); Flavonoid biosynthesis between IB and FB (p = 0.002) and between IFB and FB (p = 0.041); Isoflavonoid biosynthesis between IB and IFB (p = 0.035); Flavone and flavonol biosynthesis between IB and IFB (p = 0.032); and Stilbenoid, diarylheptanoid, and gingerol biosynthesis between IB and FB (p = 0.041).

Significantly changed microbes during dietary transition

By performing Metastats non-parametric multiple tests, the abundance of 62 bacterial species was found to vary (Figure 4A). Most of them were also observed in the RF modeling of bats with different dietary habits, e.g., members of *Vibrio, Aeromonas, Streptomyces, Cytophagaceae, Pseudoalteromonas, Alishewanella, Pseudoalteromonas, Marinobacter, Neisseria*, and *Gammaproteobacteria bacterium* TMED95. A large number of *Aeromonas* species were much more abundant in bats that feed on both insects and fish than in those with insect diet, and *Micromonospora* sp. RP3T was found to be enriched in bats that feed on both insects and fish (Figure 4B).

Discussion

The compositions of gut microbiota vary among different species of animals and may also vary under different physiological and behavioral conditions. As bats are highly differentiated animals with various dietary preferences and some bat species such as M. pilosus are in transition from insect-eating to fish-eating, they are ideal for investigation of the effects of diets on gut microbiota. In this study, we found that the gut microbiota profile of the one frugivorous bat (R. leschenaultii) examined was greatly different from that of insectivorous bats and bats that feed on both insects and fish, but those of the latter two dietary types of bats were similar. However, the abundance of 62 different species of bacteria was found to vary among different species of bats. These observations suggest that the difference in the profiles of gut microbiota in bats of different phylogenies was higher than those with different dietary habits as previously described (Lutz et al., 2019). We also found a much less difference in gut microbiota among bats of different phylogenies at the genus ($R^2 = 0.54$, p = 0.06) level than at the family level ($R^2 = 0.049$,



p = 0.743) by PCA analysis. A possible reason for such is that closely related bats have similar behaviors, physiologies, and environmental factors and therefore have similar gut microbial compositions (Archie, 2019).

Many of the physiological or behavioral characteristics of bats have a unique evolutionary history (Brualla et al., 2023). For example, the basic immune systems of bats (e.g., innate and adaptive immune systems) are more tolerant than those of other mammals (Randolph and Barreiro, 2018). A study of bat rabies virus found that bat antibodies failed to penetrate the blood-brain barrier and therefore were unable to defend the infection in the central nervous system (Roy et al., 2007). However, bats are known to carry many potentially lethal viruses, such as filoviruses, henipaviruses, lyssaviruses, paramyxoviruses, and coronaviruses (Karunarathna et al., 2020; Sun et al., 2020). Such symbiosis of various viruses may affect their physiologies, behaviors, and adaption to the virulent gut ecology. Bats also harbor many bacterial pathogens such as various species of Bacillus, Enterobacter, Enterococcus, Escherichia, Klebsiella, Pantoea, Pseudomonas, and Serratia (Daniel et al., 2013; Sun et al., 2020). We postulate that the symbiosis of various viruses and pathogenic bacteria is a survival strategy for bats to evade predators and that the host-microbe interactions may be a driving force for the evolution of bats to survive in various ecologic conditions. Dietary habits and other behaviors may also affect animal evolution. For instance, the bamboo diet promotes the adaptive development of strong teeth and paws in giant panda (Vallittu et al., 2021). Another example is that the flying ability of birds or bats is endowed with light bone, high degree of skull healing, shoulder strap, sternum with keel protuberance, and degradation of posterior tibia and fibula (Chang et al., 2019).

Different from most terrestrial mammals, the gut microbiota of bats with different dietary habits have few Bacteroidetes, but similar to carnivores, bat gut microbiota is enriched in Firmicutes and Proteobacteria (Sun et al., 2020). A previous study of bat gut bacteria showed that species of Gammaproteobacteria and Firmicutes were dominant in non-breeding adult bats; this observation is also consistent with ours in this study. We also did not observe a significant correlation between dietary habits and gut microbiota in bats that feed on insets or both insects and fish as that reported by other investigators (Phillips et al., 2012; Carrillo-Araujo et al., 2015; Lutz et al., 2019). We noticed that *Aeromonas* species were much more abundant in bats that feed on both insects and fish than those with insect diet. As *Aeromonas* is often detected in fish (Hanninen et al., 1997), it may alter the composition of gut microbiota in bats that feed on fish.

As observed in previous studies (Zepeda Mendoza et al., 2018; Ingala et al., 2021), we found that the predominant metabolic pathways were Transporters for insectivorous bats; Transporters and Carbon metabolism for bats that feed on both insects and fish; and Transporters and Biosynthesis of amino acids for frugivorous bat. A recent study showed that the numbers of genes involved in 37 metabolic pathways were different between herbivorous and animalivorous bats (Ingala et al., 2021). However, we found only 3 metabolic pathways involved in stilbenoid, diarylheptanoid and gingerol biosynthesis that were different in the number of genes between insectivorous bats and bats (*M. pilosus*) that feed on both insects and fish. A probable reason for this discrepancy is that we examined a very small dietary shift, i.e., from insect to both insect and fish diet. As *M. pilosus* bats have a high viral load, the enrichment of genes in these 3 metabolic pathways may be a symbiotic strategy with the viruses. The finding that the number of genes involved in photosynthesis of antenna proteins varied between IB and IFB (p=0.036) suggests that photosynthesis also plays a role in gut microbial variation among bats with different dietary habits as previously postulated (Jin et al., 2021).

By investigating the effects of host speciation and dietary habits on gut microbiota, this study disclosed the effect of host-microbe interactions on bat evolution. A major weakness in this study is in bat sampling. To avoid the influence of geographic variation on gut microbiota, bats used in this study were captured from the same cave. As a result, we only captured bats from 3 families (Pteropodidae, Rhinolophidae, Vespertilionidae), and the number of bats is very limited with only one bat in most species. Such problem may introduce deviations in the comparison of microbial diversity. Furthermore, a previous study surveyed the variations in gut microbiota among 9 bat families, and nearly half of the families showed significant differences in gut bacterial composition (Lutz et al., 2019). We investigated 3 bat families and detected no significant difference in gut microbiota among them, indicating that investigations with more families are warranted in future studies. Although we have investigated the effects of host speciation and diets on the composition of gut microbiota, there may be other host factors that also play a role in the formation of gut microbiota in bats. We also noticed that more than 50% of gut microbial genes in each sample were annotated as "function unknown." It has been postulated that these genes may be involved in echolocation, longevity, flying, and immune system (Zhang et al., 2013; Jebb et al., 2020).

Conclusion

In this study, we found a very weak correlation between dietary habits and gut microbiota in bats that feed on insects or both insects and fish. However, the compositions of gut microbiota were found to vary among different species of bats, suggesting that host speciation plays a major role in the formation of gut microbiota. We also found differences in the number of genes related to certain metabolic pathways among bats with different dietary habits, suggesting the effect of diet on physiological functions of bats.

Data availability statement

The datasets presented in this study can be found in online repositories. The names of the repository/repositories and accession number(s) can be found at: https://www.ncbi.nlm.nih.gov/, PRJNA954561.

Ethics statement

The animal study was reviewed and approved by the Animal Ethics Committee of Institute of Zoology, Guangdong Academy of Sciences (No. GIABR20200810).

Author contributions

MG and LZ designed the research. MG performed the research. JW, YZ, and XH collected samples and extracted fecal DNA. MG, SX, JQ, PL, CZ, and CH provided analytic tools. MG, SX, JD, and JQ analyzed data. MG and LZ wrote the paper. All authors contributed to the article and approved the submitted version.

Funding

This work was supported by grants from the GDAS Special Project of Science and Technology Development (2021GDASYL-20210103052), the Guangdong Provincial Science and Technology Program (2021B1212110003 and 2021B1212050021), and the National Natural Science Foundation of China (32172435).

Acknowledgments

The authors thank Feifan Chen (Zhongkai University of Agriculture and Engineering) and Liwei Liu (Guangzhou University) for their advice in the study.

References

Adams, R. A., and Pedersen, S. C. (2013). Bat evolution, ecology and conservation. New York: Springer.

Anderson, M. J. (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

Archie, E. A. (2019). Bat microbiomes are socially synchronized. Nat. Ecol. Evol. 3, 18–19. doi: 10.1038/s41559-018-0749-2

Benson, A. K., Kelly, S. A., Legge, R., Ma, F., Low, S. J., Kim, J., et al. (2010). Individuality in gut microbiota composition is a complex polygenic trait shaped by multiple environmental and host genetic factors. *Proc. Natl. Acad. Sci. U. S. A.* 107, 18933–18938. doi: 10.1073/pnas.1007028107

Breiman, L. (2001). Random forests. Mach. Learn. 2001, 45, 5-32. doi: 10.1023/A:1010933404324

Brooks, A. W., Kohl, K. D., Brucker, R. M., van Opstal, E. J., and Bordenstein, S. R. (2016). Phylosymbiosis: relationships and functional effects of microbial communities across host evolutionary history. *PLoS Biol.* 14:e2000225. doi: 10.1371/journal. pbio.2000225

Brualla, N. L. M., Wilson, L. A. B., Doube, M., Carter, R. T., McElligott, A. G., and Koyabu, D. (2023). The vocal apparatus: an understudied tool to reconstruct the evolutionary history of echolocation in bats? *J. Mamm. Evol.* 30, 79–94. doi: 10.1007/s10914-022-09647-z

Brucker, R. M., and Bordenstein, S. R. (2012). The roles of host evolutionary relationships (genus: *Nasonia*) and development in structuring microbial communities. *Evolution* 66, 349–362. doi: 10.1111/j.1558-5646.2011.01454.x

Camacho, C., Coulouris, G., Avagyan, V., Ma, N., Papadopoulos, J., Bealer, K., et al. (2009). BLAST+: architecture and applications. *BMC Bioinformatics* 10:421. doi: 10.1186/1471-2105-10-421

Carrillo-Araujo, M., Taş, N., Alcántara-Hernández, R. J., Gaona, O., Schondube, J. E., Medellín, R. A., et al. (2015). Phyllostomid bat microbiome composition is associated to host phylogeny and feeding strategies. *Front. Microbiol.* 6:447. doi: 10.3389/ fmicb.2015.00447

Chang, Y., Song, S., Li, A., Zhang, Y., Li, Z., Xiao, Y., et al. (2019). The roles of morphological traits, resource variation and resource partitioning associated with the dietary niche expansion in the fish-eating bat *Myotis pilosus. Mol. Ecol.* 28, 2944–2954. doi: 10.1111/mec.15127

Chen, Y., Chen, Y., Shi, C., Huang, Z., Zhang, Y., Li, Y., et al. (2018). SOAPnuke: a MapReduce acceleration-supported software for integrated quality control and preprocessing of high-throughput sequencing data. *Gigascience* 7, 1–6. doi: 10.1093/gigascience/gix120

Daniel, D. S., Ng, Y. K., Chua, E. L., Arumugam, Y., Wong, W. L., and Kumaran, J. V. (2013). Isolation and identification of gastrointestinal microbiota from the short-nosed

Conflict of interest

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

The handling editor DJ declared a shared parent affiliation with some of the authors MG, JW, YZ, XH, PL, JD, CZ, and LZ at the time of review.

Publisher's note

All claims expressed in this article are solely those of the authors and do not necessarily represent those of their affiliated organizations, or those of the publisher, the editors and the reviewers. Any product that may be evaluated in this article, or claim that may be made by its manufacturer, is not guaranteed or endorsed by the publisher.

Supplementary material

The Supplementary material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fmicb.2023.1207482/ full#supplementary-material

fruit bat Cynopterus brachyotis. Microbiol. Res. 168, 485-496. doi: 10.1016/j. micres.2013.04.001

Easson, C. G., and Thacker, R. W. (2014). Phylogenetic signal in the community structure of host-specific microbiomes of tropical marine sponges. *Front. Microbiol.* 5:532. doi: 10.3389/fmicb.2014.00532

Hanninen, M. L., Oivanen, P., and Hirvela-Koski, V. (1997). Aeromonas species in fish, fish-eggs, shrimp and freshwater. *Int. J. Food Microbiol.* 34, 17–26. doi: 10.1016/s0168-1605(96)01163-4

Hong, C., Manimaran, S., Shen, Y., Perez-Rogers, J. F., Byrd, A. L., Castro-Nallar, E., et al. (2014). PathoScope 2.0: a complete computational framework for strain identification in environmental or clinical sequencing samples. *Microbiome* 2:33. doi: 10.1186/2049-2618-2-33

Ingala, M. R., Simmons, N. B., Dunbar, M., Wultsch, C., Krampis, K., and Perkins, S. L. (2021). You are more than what you eat: potentially adaptive enrichment of microbiome functions across bat dietary niches. *Anim. Microbiome* 3:82. doi: 10.1186/s42523-021-00139-8

Jebb, D., Huang, Z., Pippel, M., Hughes, G. M., Lavrichenko, K., Devanna, P., et al. (2020). Six reference-quality genomes reveal evolution of bat adaptations. *Nature* 583, 578–584. doi: 10.1038/s41586-020-2486-3

Jin, H., Wang, Y., Zhao, P., Wang, L., Zhang, S., Meng, D., et al. (2021). Potential of producing flavonoids using cyanobacteria as a sustainable chassis. *J. Agric. Food Chem.* 69, 12385–12401. doi: 10.1021/acs.jafc.1c04632

Karunarathna, S. C., Dong, Y., Karasaki, S., Tibpromma, S., Hyde, K. D., Lumyong, S., et al. (2020). Discovery of novel fungal species and pathogens on bat carcasses in a cave in Yunnan Province, China. *Emerg. Microbes. Infect.* 9, 1554–1566. doi: 10.1080/22221751.2020.1785333

Kuang, Z., Li, F., Duan, Q., Tian, C., Nevo, E., and Li, K. (2022). Host diet shapes functionally differentiated gut microbiomes in sympatric speciation of blind mole rats in upper Galilee, Israel. *Front. Microbiol.* 13:1062763. doi: 10.3389/fmicb.2022.1062763

Kumar, S., Suleski, M., Craig, J. M., Kasprowicz, A. E., Sanderford, M., Li, M., et al. (2022). TimeTree 5: an expanded resource for species divergence times. *Mol. Biol. Evol.* 39:msac174. doi: 10.1093/molbev/msac174

Kurilshikov, A., Medina-Gomez, C., Bacigalupe, R., Radjabzadeh, D., Wang, J., Demirkan, A., et al. (2021). Large-scale association analyses identify host factors influencing human gut microbiome composition. *Nat. Genet.* 53, 156–165. doi: 10.1038/s41588-020-00763-1

Langmead, B., and Salzberg, S. L. (2012). Fast gapped-read alignment with Bowtie 2. *Nat. Methods* 9, 357–359. doi: 10.1038/nmeth.1923

Li, W., and Godzik, A. (2006). Cd-hit: a fast program for clustering and comparing large sets of protein or nucleotide sequences. *Bioinformatics* 22, 1658–1659. doi: 10.1093/ bioinformatics/btl158

Li, D., Liu, C. M., Luo, R., Sadakane, K., and Lam, T. W. (2015). MEGAHIT: an ultrafast single-node solution for large and complex metagenomics assembly via succinct de Bruijn graph. *Bioinformatics* 31, 1674–1676. doi: 10.1093/bioinformatics/btv033

Lim, E. S., Zhou, Y., Zhao, G., Bauer, I. K., Droit, L., Ndao, I. M., et al. (2015). Early life dynamics of the human gut virome and bacterial microbiome in infants. *Nat. Med.* 21, 1228–1234. doi: 10.1038/nm.3950

Lutz, H. L., Jackson, E. W., Webala, P. W., Babyesiza, W. S., Kerbis Peterhans, J. C., Demos, T. C., et al. (2019). Ecology and host identity outweigh evolutionary history in shaping the bat microbiome. *mSystems* 4:e00511. doi: 10.1128/mSystems.00511-19

Phillips, C. D., Phelan, G., Dowd, S. E., McDonough, M. M., Ferguson, A. W., Delton hanson, J., et al. (2012). Microbiome analysis among bats describes influences of host phylogeny, life history, physiology and geography. *Mol. Ecol.* 21, 2617–2627. doi: 10.1111/j.1365-294X.2012.05568.x

Randolph, H. E., and Barreiro, L. B. (2018). Holy immune tolerance, batman! *Immunity* 48, 1074–1076. doi: 10.1016/j.immuni.2018.05.016

Roy, A., Phares, T. W., Koprowski, H., and Hooper, D. C. (2007). Failure to open the blood-brain barrier and deliver immune effectors to central nervous system tissues leads to the lethal outcome of silver-haired bat rabies virus infection. *J. Virol.* 81, 1110–1118. doi: 10.1128/JVI.01964-06

Sanders, J. G., Powell, S., Kronauer, D. J. C., Vasconcelos, H. L., Frederickson, M. E., and Pierce, N. E. (2014). Stability and phylogenetic correlation in gut microbiota: lessons from ants and apes. *Mol. Ecol.* 23, 1268–1283. doi: 10.1111/mec.12611

Shalabi, L. A., Shaaban, Z., and Kasasbeh, B. (2006). Data mining: a preprocessing engine. J. Comput. Sci. 2, 735–739. doi: 10.3844/jcssp.2006.735.739

Shannon, C. E. (2001). A mathematical theory of communication. ACM SIGMOBILE Mob. Comput. Commun. Rev. 5, 3–55. doi: 10.1145/584091.584093

Song, S. J., Sanders, J. G., Delsuc, F., Metcalf, J., Amato, K., Taylor, M. W., et al. (2020). Comparative analyses of vertebrate gut microbiomes reveal convergence between birds and bats. *mBio* 11:e02901. doi: 10.1128/mBio.02901-19 Sun, D. L., Gao, Y. Z., Ge, X. Y., Shi, Z. L., and Zhou, N. Y. (2020). Special features of bat microbiota differ from those of terrestrial mammals. *Front. Microbiol.* 11:1040. doi: 10.3389/fmicb.2020.01040

Vallittu, P. K., Varrela, J., Salo, J., Rengui, L., Shanshan, L., Shan, H., et al. (2021). Temporomandibular joint and Giant Panda's (*Ailuropoda melanoleuca*) adaptation to bamboo diet. *Sci. Rep.* 11:14252. doi: 10.1038/s41598-021-93808-2

Wan, Y., Tang, J., Li, J., Yuan, J., Wang, F., and Li, D. (2020). Contribution of diet to gut microbiota and related host cardiometabolic health: diet-gut interaction in human health. *Gut Microbes* 11, 603–609. doi: 10.1080/19490976.2019. 1697149

White, J. R., Nagarajan, N., and Pop, M. (2009). Statistical methods for detecting differentially abundant features in clinical metagenomic samples. *PLoS Comput. Biol.* 5:e1000352. doi: 10.1371/journal.pcbi.1000352

Wilson, D. E., and Mittermeier, R. A. (2019). Handbook of the mammals of the world volume 9: bats. Barcelona: Lynx Edicions.

Yin, Z., Sun, K., Li, A., Sun, D., Li, Z., Xiao, G., et al. (2020). Changes in the gut microbiota during Asian particolored bat (*Vespertilio sinensis*) development. *PeerJ* 8:e9003. doi: 10.7717/peerj.9003

Zepeda Mendoza, M. L., Xiong, Z., Escalera-Zamudio, M., Runge, A. K., Thézé, J., Streicker, D., et al. (2018). Hologenomic adaptations underlying the evolution of sanguivory in the common vampire bat. *Nat. Ecol. Evol.* 2, 659–668. doi: 10.1038/ s41559-018-0476-8

Zhang, G., Cowled, C., Shi, Z., Huang, Z., Bishop-Lilly, K. A., Fang, X., et al. (2013). Comparative analysis of bat genomes provides insight into the evolution of flight and immunity. *Science* 339, 456–460. doi: 10.1126/science.1230835

Zhu, W., Lomsadze, A., and Borodovsky, M. (2010). *Ab initio* gene identification in metagenomic sequences. *Nucleic Acids Res.* 38:e132. doi: 10.1093/nar/gkq275