Efficacy of fecal sampling as a gut proxy in the study of chicken gut microbiota

RUNNING TITLE: Efficacy of fecal sampling as a gut proxy

Wei Yan[a], Jiangxia Zheng[a], Chaoliang Wen[a], Congliang Ji[b], Dexiang Zhang[b], Yonghua Chen[b],

Congjiao Sun[a, #] and Ning Yang[a, #]

[a]National Engineering Laboratory for Animal Breeding and MOA Key Laboratory of Animal

Genetics and Breeding, College of Animal Science and Technology, China Agricultural University,

Beijing, 100193, China

[b]Wen's Nanfang Poultry Breeding Co. Ltd, Guangdong Province, Yunfu, 527400, China

[*]Correspondence: Ning Yang, Department of Animal Genetics and Breeding, College of Animal

Science and Technology, China Agricultural University, Beijing, 100193, China

E-mail: nyang@cau.edu.cn

or Congjiao Sun, Department of Animal Genetics and Breeding, College of Animal Science and

Technology, China Agricultural University, Beijing, 100193, China

E-mail: cjsun@cau.edu.cn

ABSTRACT Despite the convenience and noninvasiveness of fecal sampling, the fecal

microbiota does not fully represent that of the gastrointestinal (GI) tract, and the efficacy of fecal

sampling to accurately represent the gut microbiota in birds is poorly understood. Using chickens

as a model, we collected 1,026 samples from 206 animals, including duodenum, jejunum, ileum,

cecum and feces samples. Most taxa in the small intestine (94.10 – 94.82%) and ceca (99.57%)

could be identified in feces. Microbial community membership was reflected with a gut anatomic

feature, but community structure was not. Excluding shared microbes, the small intestine and ceca

contributed 26.69 and 2.36% of the total fecal members, respectively. The composition of

Firmicutes members in the small intestine and that of Actinobacteria, Bacteroidetes and

Proteobacteria members in ceca could mirrored that observed in fecal samples well (ρ = 0.68 –

0.79 and 0.66 – 0.79, respectively, $P < 0.05$). Enterotype-like clustering was performed in GI tract

and all sites were clustered into 2 or 3 enterotype-like clusters. Feces from different clusters

reflected the GI microbiota with different efficacies, giving a new insight into observing efficacy

of feces as a gut proxy. Our results provide evidences that the good potential of feces to identify

most taxa in chicken guts, but microbial structure analyses using feces as a proxy for gut should be

interpreted with caution.

IMPORTANCE Fecal sample is the important object used in gut microbial study, as the

collection of feces is convenient and noninvasive. It is well known that the microbial community

in fecal sample may not be fully representative of that in gastrointestinal tract (GI), but the extent

to which the fecal sample reflect the microbiota in GI is not fully clear. Instead of focus on the

variation of different sampling sites, this study demonstrated the spatial relationships of

microbiota among duodenum, jejunum, ileum, cecum and feces and partitioned the efficacy of

feces as a gut proxy to quantitatively identify the extent to which fecal samples are appropriately

used in gut study. Enterotype concept was used to confirm the efficacy of feces as a gut proxy.

This work gives insights and provide future directions regarding the usage of fecal samples in

studies of the gut microbiome.

KEYWORDS   feces, proxy, gut, spatial relationships, enterotype, chicken

 INTRODUCTION

Many studies have reported on the important roles of gut microbiota in affecting host

metabolism and health in humans (1), other mammals (2) and in birds (3). Because of the

convenience and noninvasiveness of fecal sampling, most studies use fecal samples as a proxy to

study the gut microbiota, despite the increasing recognition that fecal microbial populations may

not be fully representative of those in the gastrointestinal (GI) tract contents or mucosa (4, 5).

Therefore, a comprehensive understanding of the efficacy of using fecal samples as a proxy to

study the GI microbiota would help improve longitudinal analyses of microbiota.

Among birds, the chicken is frequently used as a research model in research, and its GI

microbiota has been studied previously(6-10). In several studies, the microbiota present in

different GI segments have been investigated using traditional sequencing methods (11) or

high-throughput sequencing techniques (12, 13). However, these studies had small sample sizes

(N = 3 − 8) and were primarily aimed at examining the spatial heterogeneity among different

segments and did not focus on the spatial microbiota relationships between feces and the GI tract.

Compared with most mammals, the cecum has a specific structure in birds and has been

reported to play important roles in bird metabolism, such as in the digestion of cellulose, starch

and other resistant polysaccharides (14, 15) and in the absorption of nutrients (16) and water (17).

Microbial compositions and functions in chicken ceca have been reported in many studies (18, 19).

In addition, Stanley et al. (20) examined the microbial relationships between the ceca and feces

and observed that 88.55% of all operational taxonomic units (OTUs) were shared. However, the

microbial relationships between the ceca and small intestine (including the duodenum, jejunum

and ileum) were not reported, data which could provide an integrated view of gut microbial

relationships.

The microbial communities in human feces were previously partitioned into 3 clusters, referred

to as enterotypes (21), which varied from 2 to 3 in subsequent studies in humans (22-24). The

concept was later extended to other mammalian hosts, such as chimpanzees (25) and mice (26).

This concept is rarely used for birds, except in a study on fecal samples from 31 chickens (27).

Thus, enterotype-like clustering was performed in the current study to identify the clusters present

in different sampling sites and help improve our understanding of the microbial relationships

among feces, ceca and small intestine.

Using chicken as a model, we performed large-scale sequencing surveys and focused on the

efficacy of using feces to represent the GI microbiota in birds. The efficacy was partitioned into

microbial community membership and structure to gain a comprehensive view concerning this

issue. Using enterotype clustering methods (21), this study reports the enterotype-like clusters in

the duodenum, jejunum, ileum and cecum of birds, improving our understanding of the efficacy of

the use of feces as a proxy to study the gut microbiota and their spatial relationships in the gut.

RESULTS

Landscape and quantification of microbial relationships among feces, ceca and small

intestine. To gain an overview of the microbial relationships among the chicken duodenum,

jejunum, ileum, ceca and feces, unweighted UniFrac distances (community membership;

presence/absence of taxa) and weighted UniFrac distances (community structure; taking the

relative abundances of taxa into account) were used to perform principal co-ordinates analysis

(PCoA; Fig. 1a, c). The variation in community memberships among different sites were primarily

explained by the sites origin (Fig. 1a), but the community structures showed both the sites origin

and inter-individual variation (Fig. 1c). In particular, the cecal microbial community exhibited a

distant relationship with the small intestine community, and the microbial community in feces

showed an intermediate relationship between those of the ceca and small intestines.

UniFrac distances between two samples from all assayed sites within each individual were

calculated to quantify the spatial relationships of the gut microbiota. When the community

membership was considered only, the UniFrac distance decreased along the gut anatomical

locations from the farthest to the nearest sites between fecal and duodenal, jejunal, ileal or cecal

samples (FD, FJ, FI or FC, respectively, in Fig. 1b), presenting clear anatomical differences.

However, when taking the community structure into account, the UniFrac distance increased in FI

and FC compared with that in FJ (Fig. 1d). This finding might be explained by the exchange of

contents between the ileum and ceca, suggesting that the specific cecal microbial structure

influences the microbial communities in the ileum and feces.

Among all pairs, the unweighted UniFrac distance between the cecal and duodenal samples was

highest ($P < 0.05$), and that between duodenal and jejunal samples was lowest ($P < 0.05$; Fig. 1b

and Table S1). Regarding the weighted UniFrac distances, cecal samples had similar distances to

the duodenal and jejunal samples and were higher than the other pairs ($P < 0.05$), whereas the

lowest distance was observed between duodenal and jejunal samples ($P < 0.05$; Fig. 1d and Table

S1). These results suggest that limited differences exist within small intestinal microbial

communities, while the microbial structure in the ceca is quite distinct from those in the small

intestine.

Analyses of shared and exclusive microbial members. Given that both the community

membership and structure influence the microbial relationships among the feces, ceca and small

intestine, we next evaluated the extent to which the spatial relationships were influenced by the

above two factors. The shared and exclusive OTUs were calculated to assess the influence of the

microbial community membership. We observed that 1604 OTUs, accounting for 50% of total

OTUs, were shared across all sites (Fig. 2a), and these shared OTUs can be referred to as the "core"

microbiota in the gut. These OTUs represented different proportions of sequences in different sites

and were especially high in fecal samples (98.48%; Fig. 2b), indicating that the most abundant

members detected in fecal samples belonged to these "core" microbiota. At the genus level, these

core taxa were primarily classified as unclassified *Clostridiaceae*, *Lactobacillus*, *Bacteroides*,

unclassified *Bacteroidales* and *Gallibacterium* (Fig. 2c). It is noteworthy that 5.66% of the "core"

microbiota sequences were not assigned and that most of these sequences (71.11%) were detected

in the cecal samples (small pie chart in Fig. 2c), suggesting that most of these unassigned taxa

tended to be anaerobic microbes.

Most OTUs in the small intestine (94.1 – 94.82%) and cecal (99.57%) samples could be

identified as fecal OTUs (Table 1), indicating that feces would be an excellent proxy for

identifying species in the gut microbiota. However, OTUs from the GI tract that were still present

(5.18 – 5.90% in small intestinal and 0.43% in cecal samples) remained undetected in fecal

samples (Table 1). Of these taxa, *Symbiobacterium*, members of the phylum Chloroflexi and α-, β-

and γ-Proteobacteria appeared to be particularly undetected feces (Supplementary Table 2).

Microbial communities in the small intestine and ceca did not contribute equally to the fecal

microbial members, as 27.01% of fecal OTUs were not identified in cecal samples, most of which

(26.69%) could be identified in small intestinal niches (Table 1 and Fig. 2d). These OTUs were

primarily from the orders *Alteromonadales*, *Bacillales*, *Lactobacillales*, *Actinomycetales*,

Sphingomonadales and *Enterobacteriales* (Fig. 2e) and were considered exclusive contributors of

the small intestinal microbiota to fecal microbial members. The ceca exclusively contributed 2.36%

of OTUs to the observed fecal members, representing 0.11% of the fecal sample sequences and

consisting of taxa primarily from the orders *Cenarchaeales*, *Bacteroidales*, *SB-45*, *Chlamydiales*

and *Thermococcales* (Fig. 2f).

Correlation analyses of microbial abundances. Because community structure also affects the

spatial relationships of gut microbiota, we next performed Spearman correlation analyses between

the mean fecal and segmental genera abundance to evaluate the effects of community structure and

assess the extent to which the microbial community in GI tract was reflected in the fecal samples

(Fig. 3). If a high correlation was observed between two sites, the differences in abundance

between sites were considered highly consistent, so that the abundance at one site had the potential

to be a good proxy for the abundance at another. The microbial composition of feces was

moderately correlated with those in the small intestine (Spearman: $\rho = 0.43$; $P < 0.05$) and in the

combination of small intestine and ceca ($\rho = 0.50$; $P < 0.05$; Fig. 3a). We then performed similar

analyses to identify the correlation bias in predominant phyla (Actinobacteria, Bacteroidetes,

Firmicutes and Proteobacteria; Fig. S2). Genera of the Firmicutes and Proteobacteria phyla in

fecal samples showed moderate to high correlations with those in all of four GI sites (ρ = 0.43 –

0.79, P < 0.05). In particular, fecal samples were highly representative of Firmicutes members in

the small intestine (ρ = 0.68 – 0.79, P < 0.05) and of Actinobacteria, Bacteroidetes and

Proteobacteria members in ceca (ρ = 0.66 – 0.79, P < 0.05). However, Actinobacteria members in

the small intestine might not be well represented in fecal samples (ρ = 0.15 – 0.25, P > 0.05; Fig.

S2).

A follow-up question concerned the extent to which each microbe correlated between two sites.

To address this issue, Spearman correlation tests were performed for each genus between two sites.

The genera with abundances of > 0.1% at either compared site with a significant correlation (P <

0.05) are summarized in Fig. S3. Between the fecal and each of the 4 gut segmental samples, a

limited number of significant correlations (P < 0.05) were observed, and these correlations were

not high (ρ = -0.2 – 0.4, P < 0.05) for each genus. Most genera with significant correlations

belonged to the phyla Firmicutes and Proteobacteria. However, more significant and moderate

correlations were observed between two of the small intestinal segments, and most of the genera

with significant correlations were also from the phyla Firmicutes and Proteobacteria (Table S3).

The results suggest that the gut microbiota structures could be moderately reflected by fecal

samples when taking all genera into consideration simultaneously, but analyses of fluctuations in

abundance for a specific genus should be interpreted with caution.

Although microbes at one site were weakly correlated with the corresponding microbes at

another site, certain patterns were observed in some cases, as exemplified by the genus

Campylobacter (Table S3). The abundance of this genus in ceca exhibited consistent correlations

with that observed in the jejunum ($\rho = 0.21$, $P < 0.05$) and ileum ($\rho = 0.36$, $P < 0.05$). In ileal

samples, this genus was correlated with that measured in fecal samples ($\rho = 0.19$, $P < 0.05$), while

no correlation was observed between cecal and fecal samples. This finding indicates that

Campylobacter has great colonization ability in the distal gut of chickens, especially in ceca, and

most *Campylobacter* contributions to the fecal composition are probably from the ileum, but not

ceca.

Enterotype clustering analyses. The above analyses showed the spatial relationships of the gut

microbiota among chicken feces, ceca and small intestine by regarding each site as an entirety, and

the "entirety" could be clustered into different enterotype-like groups (21), which might improve

our observations. To address this issue, enterotype analyses of different sites were implemented

using the methods reported by Arumugam et al. (21).

The microbial communities in the duodenum, jejunum and ceca were clustered into two clusters,

while those in ileum and feces were clustered into three (Fig. 4a). Each of the clusters was

characterized by signature taxa that were overrepresented in one cluster compared to their

representation in the others within each site. The signature taxa of clusters in different sites

presented different but anatomy-linked features (Fig. 4b). In the small intestine, *Ochrobactrum*

and *Rhodococcus* were overrepresented in duodenal cluster 1, jejunal cluster 2 and ileal cluster 2.

An unassigned taxon was overrepresented in duodenal cluster 2, jejunal cluster 2 and ileal cluster

2, and this taxon was identified as New.ReferenceOTU2622 based on OTU level analyses. These

signature taxa showed conserved overlaps among clusters within the small intestine. Similarly, the

signature taxa in fecal clusters presented linkages with those in GI clusters. Fecal cluster 1 was

overrepresented by unclassified *Clostridiaceae*, which also occurred in jejunal cluster 1 and ileal

cluster 1. *Bacteroides* was overrepresented in fecal cluster 2, as well as in ileal cluster 4 and cecal

cluster 1. However, *Gallibacterium* and *Lactobacillus* were overrepresented in fecal cluster 3,

which did not occur any of the GI clusters (Fig. 4b).

To understand whether enterotype-like clustering would affect the spatial relationships of gut

microbiota, individual animals were divided into 3 groups according to the fecal enterotype-like

clusters (Enterotype-like clusters 1, 2 and 3, F1, F2 and F3). Within each group, the UniFrac

distances between feces and each of four GI segments were calculated, and each distance was

compared among three groups. The results showed that both the community membership

(unweighted UniFrac distance) and structure (weighted UniFrac distance) were associated with the

enterotype-like clustering (Fig. 5). Although membership was not significantly affected by the

clustering between F1 and F3, fecal samples in F2 exhibited a higher membership similarity with

cecal samples but a lower similarity with small intestine samples than the other two cluster groups

(Fig. 5a). Regarding community structure, fecal samples in F2 also exhibited a higher similarity

with cecal samples than the other two groups. Nevertheless, similarities between fecal samples and

each of the GI segments were lower in F1 than in F2 and F3 (Fig. 5b). The results suggested that

the fecal microbial community in F2 might be influenced by cecal microbiota and be more

representative of the cecal microbial community. Furthermore, the fecal samples in F1 had

reduced abilities to reflect the microbial community structure of the GI tract.

Spearman correlation analyses at the genus level were performed to assess the extent to which

GI communities were reflected by fecal samples in different fecal clusters. All three enterotype

groups presented moderate correlations between fecal and GI samples (Fig. 3b-d), and the

214 correlations were consistent with the results in comparisons of the weighted UniFrac distance, as

215 shown in Fig. 5b, confirming the microbial relationships between feces and GI tract among three

216 enterotype groups. To assess the correlation bias in the abundant phyla (Actinobacteria,

217 Bacteroidetes, Firmicutes and Proteobacteria), we next performed similar correlation analyses

218 between feces and each of the GI segments within each fecal cluster and specific phylum, and the

219 distribution of correlations is shown in Fig. S4. Fecal samples in all three groups showed

220 correlations with corresponding cecal samples ($\rho = 0.36 - 0.84$, $P < 0.05$) in all four phyla. In

221 particular, F2 showed high correlations ($\rho = 0.69 - 0.84$, $P < 0.05$) between fecal and cecal

222 samples, confirming the close relationships between these two sites in F2. In the phylum

223 Actinobacteria, most correlations between fecal and small intestine samples in both F1 and F3

224 were low or not significant ($\rho = 0.34$, $P < 0.05$). In the phylum Firmicutes, fecal samples in F2

225 and F3 were highly correlated with small intestine samples ($\rho = 0.66 - 0.83$, $P < 0.05$). The results

226 provide additional evidence that the enterotype-like clustering influences the efficacy of fecal

227 sampling for studying gut microbial communities.

DISCUSSION

The efficacy of using fecal samples to reflect the gastrointestinal microbiota. To our

knowledge, this is the first large-scale sequencing assessment on the efficacy of using fecal

samples as a proxy for the gut microbiota in birds. In this study, we comprehensively examined

the membership, structure and enterotype-like clusters of the chicken gut microbiome at five

different biogeographic sites within 206 individual animals. We showed that fecal samples were

good proxies for detecting the presence/absence of GI microbial members because most GI tract

members could be detected within anatomic features in fecal samples (microbial communities in

feces   showed   increasing   similarities   to   those   in   the   GI   tract   along   the

duodenum-jejunum-ileum-ceca   axis).   However,   phyla   bias   and   inter-individual   and

enterotype-like clustering effects were observed to affect the efficacy of using fecal samples to

study GI microbial abundance.

Similar to the current study, a high proportion of shared OTUs has been previously observed

between fecal and cecal samples in chickens (20). Similarly, a study in house mice observed that

93.3% of OTUs were shared between fecal and lower GI samples (28). Another chicken study

indicated that the GI origin is a primary determinant for the chicken fecal microbiota composition

(29), supporting the high proportion of shared OTUs between feces and the four gut segments

observed in the current study. These results indicate that fecal samples have good potential for

identifying microbial members derived from the GI tract. However, another chicken study (12)

observed low percentages of shared OTUs between segments. A major reason for the differences

among studies might be the small sample size in Choi's study, which would increase the

sensitivity of the results with respect to individual variation. Moreover, the presence/absence of

microbial members in the GI tract was observed to be reflected by fecal samples in a given

anatomical feature, i.e., fecal samples had more similarities in community membership to those in

ileal and cecal samples than in duodenal and jejunal samples, consistent with previous reports in

birds (13) and mammals (28, 30).

As for microbial community structure, the efficacy of using fecal samples to represent the gut

microbiota structure did not work as well as for community membership. First, the weighted

UniFrac distances between feces and each of intestinal segments was significantly higher than the

corresponding unweighted UniFrac distances (Fig. S5), suggesting that taking the abundance into

account significantly increased the dissimilarity between feces and each of the GI segments.

Second, the abundances of most taxa were significantly different between fecal and GI samples

(Table S4), consistent with previous studies(11, 13, 31). Third, the correlations between the mean

fecal and segmental genera abundances were moderate, similar to the results obtained for rhesus

macaques (32). However, these correlations display bias among different phyla, i.e., different

phyla in the GI tract are differentially mirrored by fecal samples. Fourth, significant correlations

($P < 0.05$) of each microbe between fecal and segmental samples were low and rare, suggesting

that the efficacy of using fecal samples to represent microbial abundance was affected by the

inter-individual effect. A similar effect has also been observed in humans (4).

In addition, the efficacy of using fecal samples to represent the gut microbiota was affected by

enterotype-like clustering. In this study, we observed that fecal samples in cluster F2 were more

representative of the cecal microbiota than in F1 and F3. We infer that this phenomenon might be

explained by the greater susceptibility of the fecal microbial community in cluster F2 group to the

cecal microbiota, making the compositions of the microbial community in feces and ceca more

identical, e.g., the dominant genus in the cecum and the overrepresented genus in the feces in

cluster F2 was *Bacteroides*. The *Bacteroides* enterotype is broadly observed in fecal samples in

chickens (27), humans (21, 24) and other mammals (25, 26), but the relationships between fecal

and GI samples within this enterotype have been rarely reported. Therefore, more studies are

required to elucidate the enterotype-like clustering effects on the efficacy of using fecal samples to

reflect gut microbial profiles.

Previous studies in humans (4, 33) and other mammals (32, 34) have also addressed the issue of

whether fecal samples are good representatives for GI microbial analyses. Although the

conclusions may not be fully consistent, nearly all studies reached a consensus that microbial

communities in fecal samples do not represent the whole GI microbiota. Studies in humans

suggest that microbial communities in the duodenum and colon are not represented by those in

feces because of the large differences in microbial profiles (33), and these studies emphasized the

need to examine tissue biopsies in addition to fecal samples (5), proposing that standard forceps

mucosal biopsy samples can represent bacterial populations (4). Compared with human studies,

studies in other mammals are more comprehensive because a larger number of gut segments can

be involved in the analyses. Several studies in mice (28, 34) support the utility of fecal samples for

studying the gut microbiota, because microbial communities in fecal samples were observed to be

similar to those in the lower GI tract, which is supported by studies conducted in rhesus macaques

(32), pigs (35) and equines (36).

Compared with previous studies, the strength of the current study lies in the following: 1) it

involved the use of gut segments from the upper GI tract to the lower GI tract and feces, providing

a relatively comprehensive view of the spatial relationships of the gut microbiota; 2) the microbial

relationships were partitioned into two parts, i.e., microbial community membership and structure,

providing multi-angle observations to identify microbial relationships between feces and the GI

tract; 3) a massive number of individuals was sampled, which is significant for investigations of

gut spatial relationships, as the sizes of most of the above studies did not exceed twenty; and 4) the

use of enterotype-like clustering analyses provided new insights into exploring the utility of fecal

samples in studies of the gut microbiota.

Specific roles of the cecal microbial community in the gut. Because of the specific and

significant roles in nutrition and health (15, 37), ceca have been widely investigated in birds (38,

39), especially chickens (18, 19, 40). *Bacteroides* was observed as the dominant taxa in our sturdy

(Fig. S6) and in most other studies (41, 42), although some reports observed a predominance of

Clostridiales members in ceca (12, 43). Although the cecal microbial community may sometimes

be linked to diet (38), the nearly consistent results across studies suggests the cecal microbial

community is stable. This finding might be due to ceca having a special blind-ended structure and

are located in the lower GI tract, providing a stable and anaerobic environment for microbes and

longer storage periods of the contents, in contrast to the rapid transit environment in the small

intestine (44). In addition to the microbial composition, Stanley et al. (20) also compared

microbial differences and similarities between ceca and feces in chicken. They observed that 88.55%

of all OTUs, containing 99.25% of all sequences, were shared by the ceca and feces, similar to the

observations in the current study. These results indicate that except for some rare microbial

members, most microbes in the ceca can be detected in fecal samples.

The microbial relationships between the ceca and small intestine have been rarely reported in

birds. Choi et al. (12) compared the percentage of shared OTUs among ceca and three small

intestinal sections but observed low percentages between the segments (ranging from 1.2 to 2.9%,

representing from 38.7 to 65.5% of sequences). The percentages reported in another study (60.2%

for the duodenum, 50.5% for the jejunum and 43.5% for the ileum, which were calculated from

Fig. 3 in their article) were higher than those in Choi's study. In contrast, the results of Xiao's

study presented an opposite trend from our findings, i.e., the percentages of shared OTUs in

Xiao's study decreased from the duodenum to the jejunum and ileum, demonstrating a

reversed-anatomical feature compared with the current study. These inconsistent results might be

attributable to differences among species, diets or other environmental factors, but the small

sample size in Xiao's study may be an important reason for these inconsistencies.

Enterotype-like clustering in chicken gut microbiota. Enterotype-like clustering is of

increasing concern and has recently led to heated discussions (45, 46). To the best of our

knowledge, the current study describes the first attempt of enterotype-like clustering for the GI

tract of birds. Three enterotype-like clusters (*Ochrobactrum*, *Rhodococcus* and

New.ReferenceOTU2622) appeared to be conserved in the small intestine, demonstrating a close

microbial relationship within the small intestine. Additionally, the enterotype-like cluster

Bacteroides and unclassified *Clostridiaceae* appeared in the jejunum and ileum with the oxygen

concentration, pH and nutrient changes along the GI tract. Both New.ReferenceOTU2622 clusters

in the duodenum and ileum had a lower diversity than the corresponding clusters (Fig. S7),

suggesting a potential depression role of this taxon compared to many other microbes. Cecal

samples were clustered into two enterotype-like clusters, but the silhouette coefficient was low

(0.08, Fig. S1f). This issue might be attributable to similar dominant genera (*Bacteroides*,

unclassified *Bacteroidales* and unclassified *Clostridiales*) in both enterotype-like clusters, such

that the similarity between the two clusters was high (ADONIS: $R^2 = 0.021$, $P < 0.05$; Fig. S6).

Similarly, the clustering in ceca did not significantly lead to a difference in microbial diversity

(Fig. S7). These results also provided evidence that the microbial community in ceca tended to be

more stable than other GI segments.

The chicken fecal samples were previously clustered into four potential enterotypes: enterotype

1, dominated by Firmicutes; enterotype 2, dominated by Firmicutes and Proteobacteria; enterotype

3, dominated by Firmicutes and Actinobacteria; and enterotype 4, dominated by Firmicutes and

Bacteroidetes (27). This result is not fully consistent with the enterotype-like clusters observed in

our study. Regarding cluster numbers, Kaakoush et al. (27) observed four potential enterotypes,

while we observed three. At the phylum level, the clusters in our study were overrepresented by

Firmicutes, Bacteroidetes and Proteobacteria, but not Actinobacteria. At the genus level, the

Lactobacillus and *Bacteroides* clusters were observed in both Kaakoush's and our studies, despite

differences in the species, environment and sample size. This finding indicates that *Lactobacillus*

and *Bacteroides* clusters have the potential to be conserved clusters in chickens. The *Lactobacillus*

cluster would account for the largest microbial proportion (53.0% in our study) in fecal samples,

because *Lactobacillus* has been widely reported as the dominant genus in the feces of domestic

poultry (29, 47) and pet birds (48). Because microbial diversity is linked to the stability of the

microbial community (49, 50), the community in the *Bacteroides* cluster has the potential to be

more stable with a higher diversity than in other fecal clusters ($P < 0.001$; Fig. S7). However, the

results are not fully consistent across studies in chickens (27), humans (23, 45) and mammals (26,

46), suggesting that the relationship between enterotype-like clustering and microbial diversity

might be influenced by differences in species, cohorts within species and the environment.

In humans, an enterotype classifier with a flow that includes two routes to obtain enterotype

assignments has been developed (45). We applied the classifier to our dataset, and none of the

duodenum, jejunum, ileum and feces samples were similar to stool samples from large-scale

projects, such as MetaHIT and HMP. In ceca, 71 samples matched the classifier sample criteria,

and most were clustered as enterotype *Bacteroides*, consistent with the current study (data not

shown). Differences in the physiological structure of the GI tract (e.g., the short colon with

numerous flat villi and relatively few goblet cells) (51), diet and the environment between humans

and birds make this enterotype classifier not fully suitable for data collected from birds, making it

necessary to develop an enterotype classifier for birds in the future that can be used to gain a better

understanding of their gut microbiota and to detect disorders.

In conclusion, we assessed the efficacy of using fecal samples to represent GI microbiota in

birds and analyzed potential factors affecting this efficacy. With highly shared microbial members,

fecal samples could be used to detect most microbial species in the small intestine and ceca with

gut anatomical features. However, analyses of microbial structures using fecal samples as the

proxy for the gut in longitudinal microbial studies should be interpreted with caution. Moreover,

we described the first attempt to perform enterotype-like clustering in GI segments and observed

that the clustering affected the efficacy of using fecal samples to represent the GI microbiota. This

study is one of the first attempts to identify the microbial relationships between feces and the

intestine in birds, which will help extend our understanding of the bird gut microbiota and provide

future directions regarding the usage of fecal samples in studies of the gut microbiome.

MATERIALS AND METHODS

Animal model. The complete procedure was performed according to the guidelines established by the Animal Care and Use Committee of China Agricultural University (permit number: SYXK 2013-0013).

The slow-growing yellow broiler was used as the animal model in this study, and the birds were obtained from Wen's Nanfang Poultry Breeding Co., Ltd. in Guangdong Province of China. Two hundred and six birds with similar body weights were selected and raised on the ground with *ad libitum* feeding and nipple drinkers. The birds were fed a common maize-soybean-based diet throughout the duration of the experiment. No antibiotics were applied during the thirty-five days before sample collection. Because chickens are the largest population of birds on earth, the chicken was selected as a bird model for this investigation. The slow-growing yellow broiler has not been highly selected for production, making this breed of chicken closer to the ancestral birds.

Sample collection. Fresh fecal samples were collected from each bird as soon as excreta was discharged through the cloaca at 77 days of age. Next, all the birds were humanely euthanized and dissected. The contents and mucosal surfaces of the duodenum, jejunum, ileum and cecum were collected immediately after dissection. To ensure the consistency of samples among individuals, a 10-cm-long fixed section of the duodenum and jejunum, and the whole ileum and a pair of ceca were selected for sampling from each bird. The contents and mucosa were mixed uniformly before collection. All samples were immediately placed in liquid nitrogen and then stored at -80°C. Both the intestinal contents and mucosa were sampled based on the consideration that the microbes from both sources may contribute to host interactions with respect to nutrient metabolism and immunity (52).

DNA extraction and 16S rRNA gene sequencing. DNA was extracted from intestinal and

fecal samples using a QIAamp DNA stool mini kit (QIAGEN, cat#51504) (53) following the

manufacturer's instructions. PCR amplification of the V4–V5 region of the bacterial 16S rRNA

gene was performed using the forward primer 515F (5'-GTGCCAGCMGCCGCGGTAA-3') and

the reverse primer 907R (5'-CCGTCAATTCMTTTRAGTTT-3'). Sample-specific 7-bp barcodes

were incorporated into the primers for multiplex sequencing. The PCR reactions contained 5 μl of

Q5 reaction buffer (5×), 5 μl of Q5 High-Fidelity GC buffer (5×), 0.25 μl of Q5 High-Fidelity

DNA Polymerase (5 U/μl), 2 μl of dNTPs (2.5 mM), 1 μl (10 μM) of each forward and reverse

primer, 2 μl of DNA template, and 8.75 μl of ddH$_2$O. Thermal cycling consisted of initial

denaturation at 98°C for 2 min, followed by 25 cycles of denaturation at 98°C for 15 s, annealing

at 55°C for 30 s, and extension at 72°C for 30 s, with a final extension at 72°C for 5 min. PCR

amplicons were purified using Agencourt AMPure Beads (Beckman Coulter, Indianapolis, IN) and

quantified using a PicoGreen dsDNA Assay kit (Invitrogen, Carlsbad, CA, USA). After the

quantification step, amplicons were pooled in equal amounts, and 2 × 300 bp paired-end

sequencing was performed using an Illumina MiSeq platform with the MiSeq Reagent kit v3 at

Shanghai Personal Biotechnology Co., Ltd. (Shanghai, China). The raw data on which the

conclusions of the manuscript rely has been deposited in the National Center for Biotechnology

Information (NCBI) database (accession number SRP139192, SRP139193 and SRP139195).

Analysis of sequencing data. Data analysis was performed using the Quantitative Insights Into

Microbial Ecology (QIIME, v1.8.0) pipeline (54). Briefly, raw sequencing reads with exact

matches to the barcodes were assigned to respective samples and identified as valid sequences.

The low-quality sequences were filtered based on the following criteria (55, 56): length < 150 bp,

average Phred score < 20, ambiguous bases, and mononucleotide repeats > 8 bp. Paired-end reads

were assembled using FLASH (57), and chimera detection was performed with QIIME. After

quality control, four fecal samples were excluded due to low sequence quality that was potentially

caused by a technical artifact. The remaining high-quality sequences were clustered into

operational taxonomic units (OTUs) at a 97% sequence identity using an open-reference OTU

picking protocol against the Greengenes database (58).

We focused on open-reference OTU picking for these analyses because this method yields

substantially more taxonomic identifications with sequences that failed to hit the reference

database than do closed-reference methods. The open-reference method can provide more

information for enterotype-like clustering and comparisons among intestinal segments or feces.

The singleton OTUs were discarded because such OTUs can occur due to sequencing errors. Only

OTUs representing more than 0.001% of the total filtered OTUs were retained to improve the

efficiency of the analysis. Because the sequencing and sampling quantity varied among

individuals, we rarefied the data to the lowest sequences per sample to control for sampling effort

in diversity analyses. Alpha and beta diversity of individual OTUs were calculated with

post-rarefaction data and the phylogenetic tree. Principal coordinate analysis (PCoA) was

performed using the unweighted or weighted UniFrac distance (59) for different intestinal

segments and feces.

Enterotype-like clustering was performed as previously described (21). In brief, the Calinski–

Harabasz (CH) index was calculated with PAM clustering to determine the optimal number of

clusters (Fig. S1a-e). The silhouette scores were calculated for each cluster to assess the robustness

of the clustering (Fig. S1f). Principal coordinate analysis (PCoA) was implemented with the

dudi.pco function using the *ade4* package, and the cluster plots were performed using the *rgl*

packages in **R**. The correlations between the mean fecal and segmental genera abundance were

calculated using the method described in a study of rhesus macaques (32).

Statistical analysis. We used Mann-Whitney tests to identify overrepresented genera in each

cluster within the same intestinal segment or fecal sample. Because abundant unassigned taxa

were detected that were significantly different in the duodenal, jejunal and ileal clusters,

identification of overrepresented taxa was also implemented at the OTU level in these samples.

ADONIS analyses were performed with 999 permutations for analysis of similarities and

dissimilarities using the *vegan* package in **R**. Venn plots were generated for intestinal segment or

feces samples at the OTU level using the VennDiagram package in **R**. The package *psych* in **R** was

used to calculated Spearman correlations.

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WY, JXZ and NY designed the study. WY, JXZ, CLW, CLJ, DXZ, YHC and CJS collected the samples. WY analyzed the data and wrote the manuscript. CLW assisted in construction of the figures. CJS and NY assisted in data analyzing and contributed to the revisions. All authors read and approved the final manuscript.

Conflict of Interest

The authors declare no conflicts of interest.

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Table 1 Shared and exclusive OTUs between two of sampling sites

Site1	Site2	Shared OTUs		Exclusive OTUs	
		In site1, %	In site2, %	In site1, %	In site2, %
D[1]	J	95.03[2]	96.00	4.97	4.00
		99.95[3]	99.96	0.05	0.04
D	I	98.57	89.77	1.43	10.23
		99.90	99.80	0.10	0.20
J	I	98.78	89.05	1.22	10.95
		99.93	99.81	0.07	0.19
C	D	82.71	65.79	17.29	34.21
		97.16	43.13	2.84	56.87
C	J	81.38	65.40	18.62	34.60
		96.27	48.30	3.73	51.70
C	I	95.31	69.05	4.69	30.95
		99.58	59.14	0.42	40.86
F	D	86.81	94.20	13.19	5.80
		99.51	98.78	0.49	1.22
F	J	85.83	94.10	14.17	5.90
		99.49	99.05	0.51	0.95

F	I	95.94	94.82	4.06	5.18
		99.65	98.84	0.35	1.16
F	C	72.99	99.57	27.01	0.43
		99.12	>99.99	0.88	<0.01

630 [1]D, J, I, C and F denoted the microbial community of duodenum, jejunum, ileum,

631 cecum and feces, respectively. [2]The percentage of shared or exclusive OTUs; [3]The

632 percentage of sequence reads shared or exclusive OTUs represented.

FIG 1   Site origin and inter-individual effects on the shape of microbial community membership

and structure. **(a)** Principal coordinates analysis (PCoA) with unweighted UniFrac distance. Each

dot represents a sample from a duodenum (D), jejunum (J), ileum (I), cecum (C) or feces (F). PC1

and PC2 represent the top two principal coordinates that captured the most variation, with the

fraction of variation captured by that coordinate shown as a percent. **(b)** Unweighted UniFrac

distance (mean ± SEM) between two sampling sites. DJ represents the UniFrac distance between

the duodenal and jejunal microbial community, and it was the same as DI, JI, CD, CJ, CI, FD, FJ,

FI and FC. Asterisks indicate the significance of the paired *t*-test: [***]$P < 0.001$, [**]$P < 0.01$, and [*]P

< 0.05. **(c)** PCoA plot with weighted UniFrac distance, similar to **(a)**. **(d)** Weighted UniFrac

distance between two sampling sites, similar to **(b)**.

FIG 2 OTUs shared across different sampling sites. **(a)** Venn diagram demonstrating that the

taxa overlap among different sampling sites. D, J, I, C and F denote the microbial communities of

the duodenum, jejunum, ileum, cecum and feces, respectively. **(b)** The percentage of core OTUs

and sequences represented by these OTUs in the duodenal, jejunal, ileal, cecal and fecal samples.

(c), related to **(b)**. The dominant taxa of the core microbiota. The sequences of unassigned taxa are

partitioned to different sampling sites (duodenum, jejunum, ileum, ceca and feces) shown in the

small pie chart. **(d)** The percentage of OTUs in feces exclusively contributed by the small intestine,

and the percentage of OTUs in feces was below the limit of detection in the gastrointestinal tract.

SI = the microbial community in small intestine. Taxa exclusively shared between feces and small

intestine **(e)** or ceca **(f)** are shown at the order level. Each section in the pies **(c, e** and **f)** indicate

the percentage of sequences represented by the corresponding taxon in the core taxa **(c)** or

exclusively shared taxa **(e** and **f)**.

FIG 3 Microbial compositions in feces mirror those in the gastrointestinal tract. Each dot

represents the average relative abundance of a genus across all individuals **(a)**, individuals in fecal

enterotype-like cluster 1 **(b)**, individuals in fecal enterotype-like cluster 2 **(c)** and individuals in

fecal enterotype-like cluster 3 **(d)** for each region (feces at the x-axis; SI: small intestine or SI + C:

intestine including small intestine and ceca at the y-axis). Spearman's rho was calculated between

fecal and SI (or SI + C) negative logarithm-transferred relative abundances. All Spearman

correlations shown are significant ($P < 0.05$).

FIG 4 Enterotype-like clusters in different sites. **(a)** Analyses of enterotype-like clustering with

the partitioning around medoid (PAM) clustering algorithm, which visualizes results from PCoA

and clustering at the OTU level using a 97% similarity threshold. **(b)** Relative abundances of

overrepresented taxa of each enterotype. D, J, I, C and F denote the microbial communities of the

duodenum, jejunum, ileum, cecum and feces, respectively. The numbers following D, J, I, C and F

represent enterotype-like cluster 1, 2 or 3. Unclassified genera under a higher rank are marked by

asterisks.

FIG 5   Enterotype-like clustering affects microbial relationships between fecal and

gastrointestinal samples. Unweighted **(a)** and weighted **(b)** UniFrac distance (mean ± SEM)

between fecal and each of the four segmental samples in different fecal enterotype-like clusters.

F1, F2 and F3 denote fecal enterotype-like clusters 1, 2 and 3. D', J', I', C' and F' represent the

duodenal, jejunal, ileal, cecal and fecal samples, respectively, in corresponding fecal

enterotype-like clusters. Asterisks indicate the significance of the paired *t*-test: [***]$P < 0.001$, [**]$P <$

0.01, and [*]$P < 0.05$.