## Metagenomic analysis reveals linkages between cecal microbiota and feed efficiency in Xiayan chickens

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ABSTRACT The cecal microbiota plays a critical role in energy harvest and nutrient digestion, influencing intestinal health and the performance of chickens. Feed efficiency (FE) is essential for improving economic efficiency and saving social resources in chicken production and may be affected by the cecal microbiota. Therefore, to investigate the composition and functional capacity of cecum microbes related to FE in Xiayan chicken, an indigenous breed in Guangxi province, metagenome sequencing was performed on chicken cecal contents. 173 male and 167 female chickens were divided into high and low FE groups according to the residual feed intake. The cecal microbial genome was extracted and sequenced. The results showed that the genera *Bacteroides*, *Prevotella*, and Alistipes were the 3 most abundant in each cecal microbiome. The linear discriminant analysis effect size revealed 6 potential biomarkers in male and 14 in

female chickens. Notably, the relative abundance of Lactobacillus in the high FE group was higher than that of the low FE group both in the male and female chickens, and the species *Limosilactobacillus oris* has a higher score in the high FE group of male chickens. In contrast, some potentially pathogenic microorganisms such as *Campylobacter avium* in females and *Heli*cobacter pullorum in males were enriched in the low FE group. Predictive functional analysis showed that the high FE group in male chickens had a greater ability of xenobiotics biodegradation and metabolism and signaling molecules and interaction. In addition, the host sex was found to exert effects on the cecal microbial composition and function associated with FE. These results increased our understanding of the cecal microbial composition and identified many potential biomarkers related to FE, which may be used to improve the FE of the chickens.

Key words: Xiayan chicken, feed efficiency, cecal microbiota, metagenome sequencing

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## INTRODUCTION

Owing to the expansion of the human population, the improvement of income level, and urbanization, the requirement for protein and meat is increasing. Low production costs, high feed conversion ratios, and low product prices have contributed to making poultry the meat of choice, both for producers and consumers. Over the next decade, poultry will continue to strengthen its dominant position within the meat complex, accounting for virtually half of all additional meat that will be produced. The feed cost in chicken meat production accounts for a high proportion of the total farming expense, being nearly 70% of the total cost of poultry production (Ong, 2010). Therefore, improvements in the feed efficiency (FE) of the chicken will decrease production costs and reduce the demand of land area than for feed production, while also reducing the environmental impact of broiler production. The residual feed intake (**RFI**) is defined as the difference between the actual feed intake and predicted requirements based on animal maintenance and growth (Koch et al., 1963). The RFI is superior sensitive and accurate in measuring the FE and is increasingly used in the genetic improvement of the FE in livestock. Besides, the heritability values for the RFI ranged from 0.21 to 0.49 in the previous studies (Do et al., 2014; Yuan et al., 2015; Zhang et al., 2017).

Exploring the microbial community composition has gained a growing interest in breeding animals because this has been allowed to predict the associated metabolites and compositional structure of such communities

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(Hiergeist et al., 2015). In recent years, due to metagenomic approaches based on high-throughput sequencing methods, the research of the gut microbiota has been made rapid progress (Milani et al., 2013; Mancabelli et al., 2017; Wu et al., 2020). The gastrointestinal tract (GIT) of chickens is the major site for nutrient absorption and food digestion and has a highly diverse and dynamic microbial ecosystem (Yan et al., 2017). Changes of microbial diversity in the GIT have been associated with the differences of breeds (Pandit et al., 2018), sex (Siegerstetter et al., 2017), growth stages (Yan et al., 2017), and intestinal segments (Al-Marzooqi et al., 2020) in chickens, as a result of various environmental and genetic factors. The cecum, which is the main functional section in the distal intestine, plays important roles not only in vitamin and amino acid production but also in the prevention of pathogen colonization (Stanley et al., 2014; Williams et al., 2015). The cecal microbiota was found to be highly related to the FE, which suggested an important role in chicken FE (Yan et al., 2017).

Some indigenous chicken breeds have a higher product quality, productivity, and pathogen resistance, which have been widely reported (Arora et al., 2011; Haunshi et al., 2011; Duah et al., 2020). For example, Xiayan chicken is a famous specialty in Guangxi province and one of the top 10 vellow-feather broilers in China. Xiayan chickens were characterized by the fast growth rate, strong survivability, large size, and tender meat and have been enjoying a high reputation in the broiler market in Guangxi, Guangdong, Hainan, Hong Kong, and Macau. At present, our understanding of the intestinal microbial community of indigenous chickens in China remains limited. Although research on the gut microbiota of poultry is increasing, most of the current information about the intestinal microbiota is still limited to humans (Marchesi et al., 2016; Wu et al., 2020). Moreover, the majority of previous research focused on the male chickens (Stanley et al., 2016) and the adult hens (Yan et al., 2017), whereas cecal microbiota associated with the FE were rarely explored simultaneously in chickens of both sexes. In this experiment, we have compared the microbial community composition in the cecum of Xiayan chickens with different FE by metagenomic sequencing and explored the interactions and relationships between the FE and cecal microbiota.

## MATERIALS AND METHODS

#### Animal Rearing and Management

A total of 340 indigenous Xiayan chickens (63 d old) including 173 males and 167 females from Guangxi Rongxian Zhouyi Breeding Co., Ltd., Rongxian Economic Development Zone, Guangxi were used in this study. The animal works were reviewed and approved by the Animal Care and Use Committee in Guangxi University (approval number GXU2018-058). All experimental chickens were raised in the scientific research base of the College of Animal Science and Technology of Guangxi University. During the whole experiment, all chickens were raised with the same commercial diet and management conditions, and water was freely available. Each chicken was raised in a different cage. One week after the start of feeding, we recorded the total feed consumption and total BW gain of each chicken from 70 d to 90 d of age. Average daily feed intake = total feed consumption/total days; average daily BW gain = total BW gain/total days.

## Phenotypic Data and Cecal Sample Collection

The RFI value was calculated using SAS linear simulation fitting function following the model (Koch et al., 1963; Luiting and Urff, 1991) of RFI = ADFI- $(b_0+b_1MBW^{0.75}+b_2ADG)$ , where the ADFI represents average daily feed intake; MBW refers to the mean BW; MBW<sup>0.75</sup> is the metabolic BW; ADG represents average daily BW gain;  $b_0$  is intercepted;  $b_1$  and  $b_2$  represent partial regression coefficients. The RFI value was used to estimate FE and was negatively associated with FE. We ranked the obtained RFI value, after which the 3 chickens with the highest RFI and the 3 chickens with the lowest RFI were selected from the male and female experimental chickens, respectively. These 12 chickens (3 replications  $\times$  2 genders  $\times$  2 groups) were slaughtered for collecting the cecal contents at the age of 90 d. All samples were immediately transferred into liquid nitrogen and then stored at  $-80^{\circ}$ C for subsequent metagenomic sequencing.

# DNA Extraction and Metagenomic Sequencing

We used the QIAamp Fast DNA Stool Mini Kit (QIAGEN, Germany) to extract cecal microbial genome DNA according to the manufacturer's instructions. The agarose gel electrophoresis was used to evaluate the DNA quality and integrity, and DNA concentration was measured using Thermo Scientific NanoDrop 2000 spectrophotometers.

All 12 metagenomic DNA samples were sequenced on the Illumina platform by using 150 bp paired-end sequencing. Sequence reads were treated to remove low-quality reads, trim the read sequences, and remove the host genome sequences. Specifically, reads with low-quality bases (quality value  $\leq 38$ ) greater than 40 bp were removed. Reads with N bases greater than 10 bp were removed, and the reads which shared the overlap above 15 bp with adapter were removed. Host genomic sequences were removed (Karlsson et al., 2012, 2013b) by bowtie2 (v2.2.4). The remaining clean reads were assembled using SOAPdenovo (v2.04) (Luo et al., 2012) to gain Scaffigs. Bowtie2 was used to align the clean data of each sample to Scaftigs and collect unmapped reads for another SOAP denovo assembly with the same parameters. After combining the unused reads of each sample, SOAPdenovo undergo mixed

assembly with the same parameters as above. The mixed assembled Scaffolds were broken from the N junction to get new Scaftigs ( $\geq$ 500 bp). MetaGeneMark (v2.10) (Zhu et al., 2010) was performed for predicting the open reading frames according to Scaftigs ( $\geq$  500 bp) from single-sample assembly and mixed-sample assembly. The redundancy of the predicted open reading frames (>100 nt) was eliminated using CD-HIT (v4.5.8) (Li and Godzik, 2006) to obtain a nonredundant initial gene catalog (genes). The clean data from each sample were mapped to the initial gene catalog (genes) using bowtie2 and get the number of reads. The gene in which the number of reads  $\leq$  2 (Li et al., 2014) were filtered out and the final gene catalog (unigenes) used for subsequent analysis were obtained.

## Abundance Analysis and Taxonomy Annotation

The abundance information of each gene in each sample was counted based on the length of the gene and the number of mapped reads. The formula is as follows:  $G_K = \frac{r_K}{L_K} * \frac{1}{\sum_{i=1}^{n} \frac{r_i}{L_i}}$ , where r refers to the number of reads mapped to the genes and L is the gene's length (Qin et al., 2010; Karlsson et al., 2012). The abundance of each gene in each sample was used to analyze the number of differential genes between groups. The results were visualized by the Venn figure.

The sequences of bacteria, fungi, archaea, and viruses are all extracted from the RefSeq nonredundant proteins database (accessed 2 January 2018) of the NCBI. DIA-MOND software (Buchfink et al., 2015) (V0.9.9) was used to blast the unigenes to the sequences of the microbiome database. The aligned result (e value  $\leq$  the smallest e value \* 10) (Oh et al., 2014) was chosen and the LCA algorithm was performed to make sure the species annotation information of sequences. The table containing the number of genes and the abundance information of each sample in each taxonomy hierarchy (kingdom, phylum, class, order, family, genus, species) was obtained based on the LCA annotation result. The abundance of species in one sample equals the sum of the gene abundance annotated for the species (Qin et al., 2010; Karlsson et al., 2012, 2013b). To measure the differences of cecal microbiota composition between samples from different FE groups, the vegan package was used to calculate the Bray–Curtis distance matrixes. The results were visualized by principal coordinate analysis and ggplot2 using R (v3.6.1). The nonparametric Kruskal–Wallis sum-rank test was used to detect different species between different FE groups, and the linear discriminant analysis effect size (LEfSe) analysis was used to reduce dimensionality and evaluate the impact of different species to obtain the biomarkers with significant differences between groups (Segata et al., 2011).

## Functional Annotation Analysis

DIAMOND software (v0.9.9.110) was used to blast the unigenes against the functional databases such as the Kyoto Encyclopedia of Genes and Genomes (**KEGG**) database (v01012018) and the Carbohydrate-Active enZYmes (**CAZy**) database (v04072015) with parameters of blast -e 1e-5. The annotation information of KEGG Orthology (**KO**) from the KEGG database was acquired based on the relative abundance profile. The differential function pathway between different FE groups was demined by STAMP software (P < 0.05).

## Statistical Analysis

In our research, SAS (v9.2) was used to perform statistical analysis and significance analysis on the data.

## RESULTS

## Phenotypic Values of Chicken RFI

The feed conversion ratio, RFI, daily BW gain, and daily feed intake from 70 to 90 d of age were recorded separately in all 340 experimental chickens and ranked according to the RFI value. In this study, we selected 3 chickens with the highest RFI (low FE) and 3 chickens with the lowest RFI (high FE) among male and female chickens, respectively (Supplementary Table 1). The RFI value of the group with a high FE was significantly lower than that of the group with a lower FE (low FE) (P < 0.01), either in male or female experimental chickens. The daily feed intake and daily BW gain were significantly different between the high FE and low FE groups (P < 0.01) in female chickens; no differences were observed in male chickens (Table 1).

## *Community Composition of Chicken Cecal Microbe*

To assess the cecal microbial composition at the species level, cecal contents of all 12 experimental chickens were collected, and metagenomic sequencing was performed. We obtained a total of 80,994.61 Mbp of clean reads after quality control. To obtain Scaftigs >500 bp, the sequence reads were *de novo* assembled. A total of 2.456 million Scaftigs with an average length of 1,496 bp and an average N50 length of 1,925 bp were produced after subsequent assembly. After removing redundancies, a total of 1,096,825 genes remained, which constituted the final gene catalog (unigenes) for subsequent analyses (Supplementary Table 2). The Venn diagrams indicated that 92,109 and 148,540 genes were unique in the high and low FE groups of male chickens, respectively. At the same time, 80,377 and 169,651 genes are unique in the high and low FE groups of female chickens, respectively (Supplementary Figure 1).

We determined the taxonomic composition of cecal microbiota by blasting the unigenes to the NCBI RefSeq nonredundant proteins database. The bacteroidetes, firmicutes, proteobacteria, and actinobacteria were the most prevalent phyla in all 12 samples, accounting for >77% of the cecal microbial populations (Table 2).

#### CECAL MICROBIOTA AND FEED EFFICIENCY

	Mal	es	Females		
Parameter	HFE	LFE	HFE	LFE	
Daily feed intake(g) Daily BW gain(g) FCR (g/g) RFI (g)	$\begin{array}{rrr} 75.30 \pm 10.58 \\ 27.50 \pm 4.58 \\ 2.75 \pm 0.21^{\rm A} \\ -13.02 \pm 0.92^{\rm A} \end{array}$	$\begin{array}{c} 95.85 \pm 6.73 \\ 22.00 \pm 2.05 \\ 4.37 \pm 0.20^{\rm B} \\ 13.56 \pm 1.47^{\rm B} \end{array}$	$\begin{array}{c} 65.63 \pm 4.82^{\rm A} \\ 21.83 \pm 0.95^{\rm A} \\ 3.01 \pm 0.15^{\rm A} \\ -11.79 \pm 2.55^{\rm A} \end{array}$	$82.95 \pm 1.69^{B}$ $18.17 \pm 0.29^{B}$ $4.57 \pm 0.09^{B}$ $11.89 \pm 1.62^{B}$	

**Table 1.** Feed efficiency and phenotype data for female and male chickens with high and low residual feed intake.

<sup>A,B</sup>Different capital letters indicate that the difference is statistically significant (P < 0.01). HFE and LFE denote high feed efficiency and low feed efficiency, respectively.

Abbreviations: FCR, feed conversion ratio; RFI, residual feed intake.

Nevertheless, at the genus level, the predominant genera in experimental chickens were different between the high and low FE group (Table 2 and Supplementary Figure 2). As can be seen from Table 1, the *Lactobacillus* of the low FE group was 1.41 and 2.31% lower than those of the high FE group in male and female chickens, respectively.

## *Similarities of the Cecal Microbial Communities*

To explore the differences in the cecal microbiota composition between high FE and low FE groups, we calculated the Bray-Curtis distance between high FE and low FE group of male chickens and high FE and low FE group of female chickens at the species level. The distance matrix was visualized by principal coordinate analysis (Figure 1). As can be seen from Figure 1, the samples could be clustered by FE, which was consistent with the grouping results.

## *Comparison of Chicken Cecal Microbiota in High and Low FE Groups*

An LEfSe analysis was performed to identify any significant differences in the relative abundances of microbial taxa between chickens with different FE group, which could be used as biomarkers. In total, 22 biomarkers were identified with linear discriminant analysis scores >4 (Supplementary Table 3). In the microbial population of male chickens, *Barnesiella sp An22* and Limosilactobacillus oris in the high FE group and Bacteroides sp An322, Subdoligranulum variabile, and Helicobacter pullorum in the low FE group were characteristic of the respective FE groups (Figure 2A). The Enterobacteriaceae and Proteobacteria in the high FE group and Blautia, Campylobacter avium, and Faecalibacterium in the low FE group could be considered as a potential biomarker in the microbial population of female chickens (Figure 2B).

## *Comparison of the Functionality of Chicken Cecal Microbiome*

To investigate the functional capacity of cecal microbiota related to chicken FE, unigenes were annotated based on the KEGG and CAZy databases. First, the unigenes were aligned to the KEGG gene database, and a total of 5.407 KOs were obtained. Based on the functional annotations and abundance information of KO in the KEGG gene database, we selected the KOs of the top 35 and hierarchically clustered from the functional difference (Figure 3 and Supplementary Table 4). These KOs enriched in male chickens with high FE were associated with genetic information processing (KO3169, KO3111, K18220, KO3496). These KOs associated with amino acid metabolism (KO1915), fatty acid metabolism (KO1897), amino sugar and nucleotide sugar metabolism (K12373), and genetic information processing (KO1153, KO3797) were more abundant in male chickens with low FE. In the female chickens, these KOs abundant in high FE were associated with genetic

**Table 2.** Relative abundance of the dominant phyla and genera in the cecum of the high and low FEgroups of male and female chickens.

	Male		Female			Male		Female	
Phylum (%)	HFE	LFE	HFE	LFE	Genus (%)	HFE	LFE	HFE	LFE
Bacteroidetes	55.08	58.03	48.23	49.86	Bacteroides	22.10	24.41	18.39	21.54
Firmicutes	20.89	19.32	22.93	25.98	Prevotella	7.52	7.02	4.30	4.74
Proteobacteria	4.48	3.43	4.74	4.08	Alistipes	4.70	4.41	8.01	6.18
Actinobacteria	1.65	1.06	1.71	2.18	Muribaculum	2.29	3.42	1.42	1.46
Verrucomicrobia	0.32	0.35	0.99	0.64	Lactobacillus	2.35	0.94	3.43	1.12
Spirochaetes	0.29	0.47	0.94	0.25	Megamonas	1.14	0.81	0.83	1.94
Euryarchaeota	0.38	0.87	1.08	0.81	$\check{B}lautia$	1.42	1.68	1.37	2.54
Fusobacteria	0.05	0.15	0.64	0.11	Lachnoclostridium	1.55	1.34	1.80	2.15
Synergistetes	0.83	0.99	0.91	1.18	Parabacteroides	2.14	1.96	1.21	1.75
Elusimicrobia	0.3	0.08	0.24	0.25	Mediterranea	2.04	2.42	1.82	1.55
Others	15.72	15.25	17.57	14.67	Others	52.74	51.60	57.42	55.03

HFE and LFE denote high feed efficiency and low feed efficiency, respectively.



Figure 1. Principal coordinate analysis (PCoA) of the cecal microbiota based on the Bray-Curtis distance at the species level. (A) PCoA between high FE and low FE of female chickens. (B) PCoA between high FE and low FE of female chickens. Abbreviation: FE, feed efficiency.

information processing (KO0558, KO3655, KO3701). Notably, those KOs associated with energy transportation (KO6147, KO2003, KO1990, and KO2004) were abundant in low FE. The presence of these KOs may reduce the FE of female chickens, as Virkel et al. pointed out in 2019 that ATP-binding cassette transporting protein may influence the bioavailability in domestic animals (Virkel et al., 2019).

We investigated pathways of significant differences between the high FE and low FE group with P < 0.05. We identified 2 kinds of xenobiotics biodegradation and metabolism such as chlorocyclohexane and chlorobenzene



Figure 2. Linear discriminant analysis (LDA) effect size (LEfSe) analysis of cecal microbiota. The same analysis was performed in the (A) male chickens, and (B) female chickens to compare the microbial communities in the cecum between high FE and low FE groups. The LDA plots indicate species that can be used as biomarkers. Abbreviation: FE, feed efficiency.

degradation and drug metabolism—cytochrome P450 (**CYP**), and signaling molecules and interaction, and agarose gel electrophoresis–receptor for advanced glycation end products signaling pathway in diabetic complications (Figure 4A). These function terms were enriched in the high FE group of male chickens. In the female chickens, amoebiasis and cell motility in high FE group and carbohydrate metabolism in low FE group were abundant (Figure 4B).

Subsequently, we further aligned the sequences of unigenes to protein sequences in the CAZy database and classified the sequences into 6 enzyme classes. Glycoside hydrolases and glycosyltransferases were the 2 classes enriched the most in all samples (Supplementary Figure 3). The function pathways of significant differences between the high FE and low FE groups were identified. At the enzyme family level, we observed the differential pathway associated with the carbohydrate metabolism in the high FE group was less than the low FE group in male chickens. Whereas in the female chickens, 6 differential pathways associated with the carbohydrate metabolism in the high FE group and 2 pathways in the low FE group were identified (Supplementary Figure 4).

## DISCUSSION

The gut microbiota in chickens and mammals (McCormack et al., 2017) play important roles in the nutrient digestion, harvesting of ingested energy, and regulation of intestinal function, and its variations affected the metabolism (Stanley et al., 2013) and immune functions (Derrien et al., 2011) of the host. The finding that differentially abundant microbes were detected in the cecum might have implied a crucial role for cecal microbiota in FE (Yan et al., 2017). In this experiment, we explored the cecal microbial composition of Xiayan chickens and systematically estimated the interactions and relationships between



Figure 3. Heat map of the top 35 predicted functions by KEGG. The X-axis shows the group IDs; M-HFE, M-LFE, F-HFE, and F-LFE denote the high feed efficiency group in male chickens, the low feed efficiency group in male chickens, the high feed efficiency group in female chickens, and the low feed efficiency group in female chickens, respectively. The Y-axis refers to the KEGG orthologues. Abbreviation: KEGG, Kyoto Encyclopedia of Genes and Genomes.

cecal microbiota and FE with metagenomic sequencing data.

To minimize the influence of genetics, diet, and external factors, all experimental chickens were selected with similar genetic backgrounds and the same management, environmental, and dietary conditions. However, FE evaluated by the RFI still had great differences in experimental chickens, which indicated that the difference of gut microbiota might result in this phenomenon. At the phylum level, bacteroidetes, firmicutes, and proteobacteria were the most abundant phyla in cecal microbiota of chickens, which were consistent with previous reports (Lee et al., 2017; Dong et al., 2019; Shi et al., 2019). A recent study has shown that the abundance of bacteroidetes in the intestinal microbiota of the high FE pigs was lower than that of low FE pigs (Quan et al., 2020). In this experiment, we found the same result in chickens. Turnbaugh et al. (2006) point out that firmicutes play an important role in energy extraction in mice (Turnbaugh et al., 2006) and Yang et al. (2016) argued that the fatness might be improved when *firmicutes* increases in pigs' intestine (Yang et al., 2016). The results of these studies may suggest that the

predominant phyla keep balance, and the stability of gut function can be ensured. However, compared with previous results, the composition and relative abundance of cecal-dominant microbes were different at the genus level (Pandit et al., 2018; Shi et al., 2019). Because the breed, sex, or feed of experimental animals between each study were different, the highly alterable microbial community composition may be affected by a variety of factors at a lower taxonomic level. For example, in the study by Shi et al. (2019), Bacteroides, Rikenellaceae RC9, and Faecalibacterium were the 3 most abundant genera in laying hens (Shi et al., 2019). In Tamil Nadu, Bacteroides, unclassified Bacteria, and Alistipes were the 3 dominant genera in Ross 308 chicken breeds/lines, and unclassified Clostridiales, Clostridium, and *Faecalibacterium* presented higher proportions in commercial Cobb 400 in the study by Pandit et al. (2018) (Pandit et al., 2018).

Notably, the relative abundance of *Lactobacillus* in high FE groups was higher than the low FE groups, both in the male and female chickens (Table 2). We speculated that *Lactobacillus* may have a potential link to the FE. The LEfSe confirmed our hypothesis. That analysis indicated that the species *Lactobacillus oris* in the high FE group of male chickens could be considered as a potential biomarker. Yan et al. (2017) point out that Lactobacillus was one of the differentially abundant taxa and was highly related to the host FE (Yan et al., 2017). Studies have shown that *Lactobacilli* generally undergo 2 metabolic pathways: 1) converting glucose to lactic acid, which is called homofermentation, and 2) transforming glucose to lactic acid, acetic acid, ethanol, and  $CO_2$ , which is called heterofermentation (Forte et al., 2018). These metabolites decrease the pH of the intestinal lumen to put potential pathogenic bacteria in an unfavorable environment (Menconi et al., 2011). The presence of *Lactobacillus* can effectively increase some beneficial bacteria and decrease potentially harmful bacteria to maintain the stability of microbial microbiota in the GIT (Forte et al., 2016). Wang et al. (2017) point out that Lactobacillus johnsonii BS15 reduces fat deposition and promotes the growth performance of chickens (Wang et al., 2017). Besides, Lactobacilli are often used as the formulation of prebiotics and probiotics that enhance the intestinal health for improved the colonization resistance to gut pathogens such as *Campylobacter* and *Salmonella* and the host performance (Kobierecka et al., 2017; Muyyarikkandy and Amalaradjou, 2017; Saint-Cyr et al., 2017; Khan and Chousalkar, 2020; Khan et al., 2020).

Campylobacter avium that could be considered as a potential biomarker was observed in the low FE group of the female chickens. Campylobacteriosis is a primary food-borne zoonosis all over the world (Carron et al., 2018). Some investigators have proven that poultry can serve as a natural host for Campylobacter species and a reservoir during dissemination (Torralbo et al., 2014). Besides, broilers are often colonized by Campylobacter (Garcia-Sanchez et al., 2018), and previous studies have reported that the colonization of Campylobacter jejuni in broiler chickens has a medium impact on



Figure 4. Differential KEGG function pathways between high feed efficiency group and low feed efficiency group. (A) The KEGG pathways detected in male chickens. (B) The KEGG pathways detected in female chickens. Abbreviation: KEGG, Kyoto Encyclopedia of Genes and Genomes.

the composition of intestinal microbiota (Kaakoush et al., 2014; Sofka et al., 2015). Furthermore, in the low FE group of the male chickens, H. pullorum has a higher score with LEfSe analysis. Helicobacter pullorum is a putative enterohepatic pathogen that has been associated with hepatobiliary and gastrointestinal diseases in chickens and humans (Sirianni et al., 2013). Another report showed that the chicken was infected with Helicobacter with no symptoms during infection, but the cecum of sacrificed chickens had mild lesions (Ceelen et al., 2007). The analysis by Pineda-Quiroga (2018) showed that lower abundance H. pullorum in the cecum was beneficial to the growth of broiler weight (Pineda-Quiroga et al., 2018). And the reduction of this bacterium also minimizes the risk of being transmitted to humans by chicken product consumption (Borges et al., 2015). Both C. avium and H. pullorum were associated with the inflammatory response. Therefore, we hypothesize that the colonization of these bacteria in cecum may reduce the FE of chickens.

Notably, a number of species under the Bacteroides genus such as Bacteroides sp An322, Bacteroides sp Marseille P3166, and Bacteroides sp CAG 598 were observed. Bacteroides were gram-negative, spore-free, obligate anaerobic small bacilli. *Bacteroides* can cause endogenous infections when the organism's immune function is disordered or the microbiota is imbalanced. Previous studies have reported that the *bacteroidales* S24 7 group was more abundant in the low FE group of pigs (He et al., 2019). Ivarsson et al. (2014) pointed out that the abundance of Bacteroides-Prevotella-Porphyromonas in pigs was positively associated with the capacity of fermenting nonstarch polysaccharides to short-chain fatty acids (Ivarsson et al., 2014), and the short-chain fatty acids were linked to promoting human obesity (Cho et al., 2012). As is known to all, fat deposition decreases the FE of pigs (Martinsen et al., 2015).

Taken together, we speculate that these bacteria may lead to a decrease in FE by promoting the development of host fatness. The results of functional annotation that was used to predict the function pathway associated with FE confirmed our hypothesis. At the enzyme family level, we observed the differential pathway associated with the carbohydrate metabolism in the low FE group was higher than that of the high FE group in male chickens. Besides, the KEGG pathway of the metabolism of monosaccharides was enriched in low FE group of female chickens; the same result was observed in pigs (Yang et al., 2017). A previous study showed that gut microbiota related to obesity had a higher capacity for energy extraction (Turnbaugh et al., 2006).

Consist with the previous study, the pathway of the flagellar assembly was enriched in the high FE group of female chickens (Tan et al., 2017), which suggests that the growth environment for microorganisms in the high FE group was better than that of the low FE group. In addition, the pathway associated with xenobiotics biodegradation and metabolism was abundant in the high FE group of male chickens. The previous report showed the toxic effects of hexabromocyclododecane on mammals, and chlorocyclohexane and chlorobenzene degradation and drug metabolism-CYP were associated with the degradation of hexabromocyclododecane (Wang et al., 2019). The CYP in the intestinal mucosa serves as a main metabolic barrier against orally ingested xenobiotics (Obach et al., 2001). In addition, CYP1A, CYP3A, and CYP2H subfamilies play a vital role in hepatic drug metabolism in chickens (Ourlin et al., 2000). Interestingly, consistent with the previous finding (Lee et al., 2017), the FEassociated gut microbes and pathway were diverse in male and female chickens, suggesting that the host gender had a significant effect on gut microbiota in chickens.

Currently, research on the intestinal microbiota of indigenous breed chickens in China, especially local breeds with special genetic traits, is scarce. The number of indigenous breed chickens in China is huge, and almost every indigenous breed chicken has its outstanding traits such as low abdominal fat rate, lowtemperature resistance, and low-hypoxia resistance. These traits may be related to gut microbial composition and function. The Tibetan chicken lives in a high altitude and hypoxic environment and has the characteristics of strong resistance to hypoxia and roughage. Comparing the cecal microbiota of free-range Tibetan chickens with large-scale chickens, the researchers found that the cecal microbial composition and abundance were different (Zhou et al., 2016). This difference may be related to the excellent characteristics of Tibetan chickens. Therefore, research on the intestinal microbiota of Chinese indigenous breed chickens should be strengthened to enrich and improve information on the gut microbial composition and function of indigenous chicken.

## CONCLUSION

In conclusion, we have revealed the compositional differences within the cecal microbiota associated with FE in Xiayan chicken, suggesting a potential association between cecal microbiota and FE. Meanwhile, we identified a total of 22 potential biomarkers associated with FE, beneficial bacteria including *Lactobacillus* and *Limosilactobacillus oris*, and harmful bacteria such as *C. avium* and *H. pullorum* in female and male chickens, respectively. The present study increased our understanding of the cecal microbial composition and identified many potential biomarkers related to FE, which could help improve the FE of Xiayan chickens.

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#### SUPPLEMENTARY DATA

Supplementary data associated with this article can be found in the online version at https://doi.org/10.1 016/j.psj.2020.09.076.

## DISCLOSURES

The authors declare no conflict of interest.

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