



The Role of Intestinal Microbial Metabolites in the Immunity of Equine Animals Infected With Horse Botflies

Dini Hu¹, Yujun Tang², Chen Wang³, Yingjie Qi³, Make Ente², Xuefeng Li², Dong Zhang¹, Kai Li^{1*} and Hongjun Chu^{4*}

¹ Key Laboratory of Non-invasive Research Technology for Endangered Species, School of Ecology and Nature Conservation, Beijing Forestry University, Beijing, China, ² Xinjiang Research Centre for Breeding Przewalski's Horse, Ürümqi, China, ³ Altay Management Station of Mt. Kalamaili Ungulate Nature Reserve, Altay, China, ⁴ Institute of Forest Ecology, Xinjiang Academy of Forestry, Ürümqi, China

OPEN ACCESS

Edited by:

Jean-Christophe Bambou,
Institut National de Recherche Pour
l'Agriculture, l'Alimentation et
l'Environnement (INRAE), France

Reviewed by:

Tariq Jamil,
Friedrich Loeffler Institut, Germany
Gang Liu,
Chinese Academy of Forestry, China

*Correspondence:

Kai Li
likai_sino@sina.com
Hongjun Chu
hongjunchu@vip.163.com

Specialty section:

This article was submitted to
Parasitology,
a section of the journal
Frontiers in Veterinary Science

Received: 10 December 2021

Accepted: 23 May 2022

Published: 22 June 2022

Citation:

Hu D, Tang Y, Wang C, Qi Y, Ente M,
Li X, Zhang D, Li K and Chu H (2022)
The Role of Intestinal Microbial
Metabolites in the Immunity of Equine
Animals Infected With Horse Botflies.
Front. Vet. Sci. 9:832062.
doi: 10.3389/fvets.2022.832062

The microbiota and its metabolites play an important role in regulating the host metabolism and immunity. However, the underlying mechanism is still not well studied. Thus, we conducted the LC-MS/MS analysis and RNA-seq analysis on *Equus przewalskii* with and without horse botfly infestation to determine the metabolites produced by intestinal microbiota in feces and differentially expressed genes (DEGs) related to the immune response in blood and attempted to link them together. The results showed that parasite infection could change the composition of microbial metabolites. These identified metabolites could be divided into six categories, including compounds with biological roles, bioactive peptides, endocrine-disrupting compounds, pesticides, phytochemical compounds, and lipids. The three pathways involving most metabolites were lipid metabolism, amino acid metabolism, and biosynthesis of other secondary metabolites. The significant differences between the host with and without parasites were shown in 31 metabolites with known functions, which were related to physiological activities of the host. For the gene analysis, we found that parasite infection could alarm the host immune response. The gene of “cathepsin W” involved in innate and adaptive immune responses was upregulated. The two genes of the following functions were downregulated: “protein S100-A8” and “protein S100-A9-like isoform X2” involved in chemokine and cytokine production, the toll-like receptor signaling pathway, and immune and inflammatory responses. GO and KEGG analyses showed that immune-related functions of defense response and Th17 cell differentiation had significant differences between the host with and without parasites, respectively. Last, the relationship between metabolites and genes was determined in this study. The purine metabolism and pyrimidine metabolism contained the most altered metabolites and DEGs, which mainly influenced the conversion of ATP, ADP, AMP, GTP, GMP, GDP, UTP, UDP, UMP, dTTP, dTDP, dTMP, and RNA. Thus, it could be concluded that parasitic infection can change the intestinal microbial metabolic activity and enhance immune response of the host through the pathway of purine and pyrimidine metabolism. This results will be a valuable contribution to understanding the bidirectional association of the parasite, intestinal microbiota, and host.

Keywords: *Equus przewalskii*, horse botfly, RNA sequencing, immune response, differentially expressed genes

INTRODUCTION

Horse botfly species are members of the *Gasterophilus* genus. These parasites live primarily in the gastrointestinal tract of equine animals but sometimes infect pigs, dogs, birds, and humans (1, 2). It has been reported that the prevalence of the horse botfly is particularly high in the desert steppe of Xinjiang, China. It infects nearly 100% of the wild equine animals that live there, which include *Equus przewalskii*, *Equus caballus*, and *Equus hemionus* (3). Due to horse botfly infestation, manifested as dysphagia, gastric and intestinal ulceration, and gastric obstruction or volvulus, gastrointestinal illnesses in equine animals can lead to various complications such as anemia, diarrhea, gastric rupture, peritonitis, and perforated ulcers (2, 4). When the parasite infects its host, it produces specific antigens during different stages in its life cycle, which induce the host to produce specific immune responses (5, 6). Thus, horse botfly infestation affects the physiology and immunity of equine animals.

Intestinal microbiota as an immune modulator plays an important role in maintaining the health of the host, which has extraordinary effect on physiology, metabolism, and immune function (7, 8). The balanced intestinal microbiota requires high microbial taxa diversity, high gene richness, and stable microbiome function (9). However, external factors such as age, diet, gender, and parasite infestation can disrupt the balance of intestinal microbiota, which results in the development of dysbiosis (10–12). The precise causes of the interaction between the unbalanced intestinal microbial community and the host health are that the metabolites produced by these microbes regulate the function of immune cells (13). For instance, the short-chain fatty acids (SCFAs) of butyrate derived from microbiota can promote IL-22 production to maintain the homeostasis of intestines (14, 15). Another SCFAs of lactate produced by *Bifidobacterium* spp. and lactic acid bacteria inhibit pro-inflammatory signals in the GPR81-independent metabolic process (16, 17). The metabolites derived from the microbiota profoundly shape the immunity, which means that understanding the microbial metabolic activities can help us clarify the immune status of the host.

Previous studies have shown that cytokine- or leukocyte-related gene expression in the blood can effectively reflect the host immune response in horses (18–22). But most of the studies on associations between parasite infections and immune responses in equine animals have mainly studied parasite antigen-induced host antibodies (23–25), and immune cells (26), clinical signs (27–30), and pathophysiology (31) in the host. It is our knowledge that only one study has evaluated gene expression in the whole blood of domestic horses infected with cyathostomin parasites (32). In addition, the study of the interaction between parasite infestation and intestinal microbiota in equine animals is still in its infancy. Starting in 2018, Peachey et al. (33) provided a novel insight into the parasite–microbiota interaction in equids. Following that, two studies explored these interactions in 2019 Walshe et al. (34) and Peachey et al. (35). At present, only Walshe et al. (34) examined the blood samples in horses infected with cyathostomin, who observed that the

changes in the intestinal microbial community might be involved in increasing inflammatory markers like albumin and serum fibrinogen. Peachey et al. (35) identified different types of fecal metabolites when the parasite counts are high and low. However, so far, although Peachey et al. found that parasite infection can increase the metabolites of isobutyrate, trehalose, leucine, phenylalanine, glutamate, glucose, lysine, and propionate from microbiota and decrease metabolites of nicotinate, valerate, and butyrate, this study was conducted on equine youngstock that displayed a different microbial community pattern from adult animals (36). It is not known whether or not the immune genes are expressed in blood of infected horses. Therefore, there is still a lack of systematic study to clarify the role of intestinal microbiota in response to parasite infection and how the unbalanced microbiota affects the immunity of equine animals.

The previous study by our research group has determined that the horse botfly infestation on *E. przewalskii* can change some specific bacterial taxa (37) and also altered the archaea and eukaryotic and viral communities (38). In addition, the functional analysis showed that the intestinal microbiota mainly participated in the amino acid and carbohydrate metabolism in infected *E. przewalskii* (38). Our study will continuously investigate the associations between horse botfly infestation and intestinal microbiota of *E. przewalskii* through the identification of metabolites derived from microbiota in feces and immunity-related genes in blood, and link its effects on the metabolism and immunity more concretely. The results can provide further insights into the function of intestinal microbiota and host immune response of equine animals during parasite infection.

MATERIALS AND METHODS

Ethics Statement

This study was conducted in accordance with Chinese law and the regulations of the Beijing Forestry University and its guidelines on animal research. The experimental protocol was reviewed and approved by the Institution of Animal Care and the Ethics Committee of the Beijing Forestry University (EAWC_BJFU_2020012). The management authority of the Kalamaili Nature Reserve (KNR) in China's Xinjiang region approved the collection of feces, botfly larvae, and blood from wild *E. przewalskii*.

Fecal Metabolite Analysis

Fecal Sample Collection

The procedures for fecal sample collection in Hu et al. (37) were adopted in our study. This study was carried out at the KNR in 2019 (Figure 1). In total, seven adult Przewalski's horses with similar body weight were driven from the wild and kept individually in a temporary enclosure. The fecal samples were collected from the horses immediately prior to the anthelmintic treatment (M-PATPHs) and 7 days thereafter (M-FATPHs). The fecal samples were stored in liquid nitrogen immediately after the collection and then transported back to the laboratory and stored at -80°C .

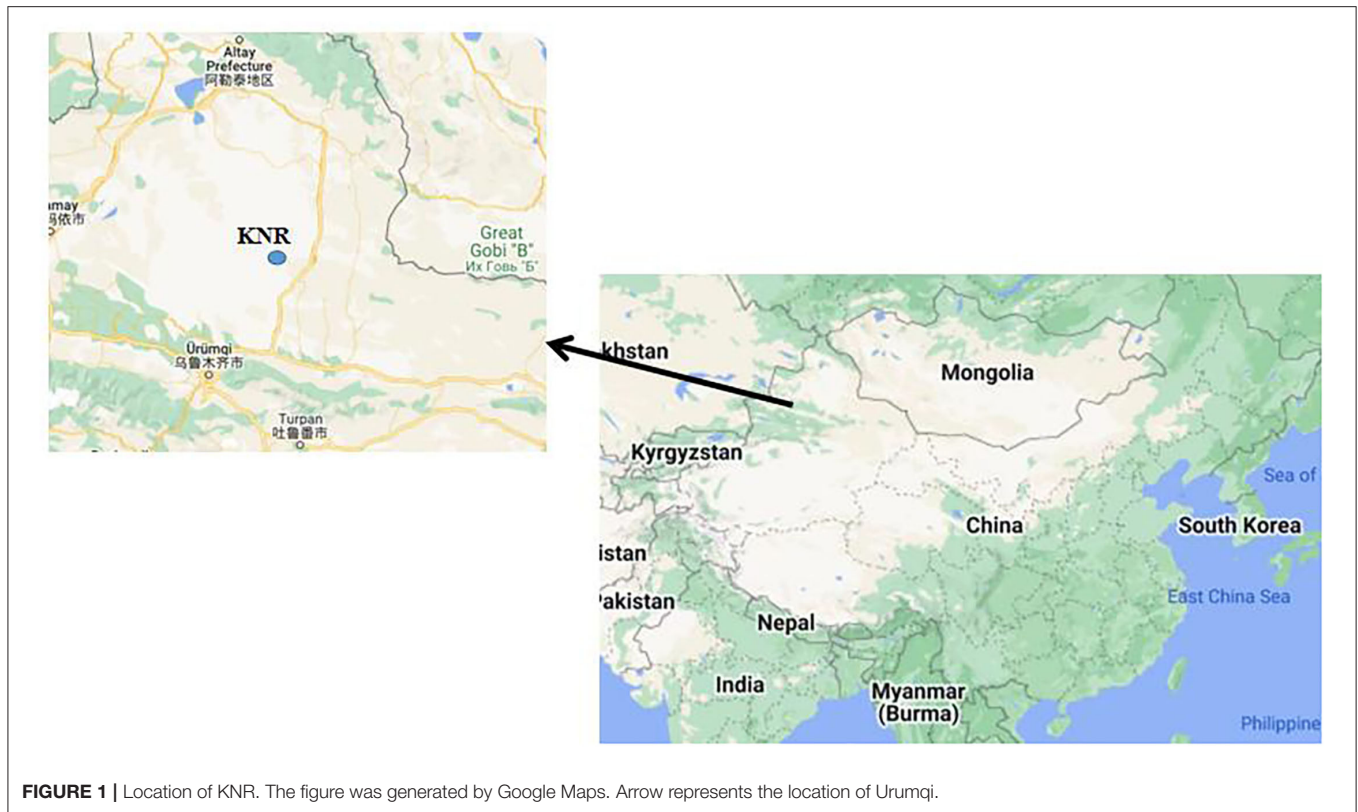


FIGURE 1 | Location of KNR. The figure was generated by Google Maps. Arrow represents the location of Urumqi.

Metabolite Extraction and Quality Control Sample (QC) Preparation

A measure of 50 mg feces from each sample were used for chemical analysis. The solution of 400 μ l methanol:water (4:1, v/v) was used for metabolite extraction. The mixture precipitated at -20°C was treated by using a high-throughput tissue crusher Wonbio-96c (Shanghai Wanbo Biotechnology Co., Ltd.) under the condition of 50 Hz for 6 min, followed by vortexing for 30 s, and ultrasound at 40 kHz and 5°C for 30 min. The prepared samples were placed at -20°C for 30 min for the precipitation of proteins, and the supernatant was centrifuged at 13,000 g at 4°C for 15 min and then carefully transferred to sample vials for liquid chromatography–mass spectrometry (LC-MS/MS) analysis. The QC was prepared by a mixture of equal volumes of all samples, followed by the same manner applied for analytic samples.

LC-MS/MS Analysis

A Thermo UHPLC system equipped with an ACQUITY BEH C18 column (100 mm \times 2.1 mm i.d., 1.7 μ m; Waters, Milford, USA) was used to perform the chromatographic separation of the metabolites. The mobile phases consisted of 0.1% formic acid in water (solvent A) and 0.1% formic acid in acetonitrile:isopropanol (1:1, v/v; solvent B). The solvent gradient was adjusted for equilibrating the systems according to the following conditions: 95% (A): 5% (B) to 80% (A): 20% (B), 0–3 min; 80% (A): 20% (B) to 5% (A): 95% (B), 3–9 min; 5% (A): 95% (B) to 5% (A): 95% (B), 9–13 min; 5% (A): 95% (B) to 95% (A): 5% (B), 13–13.1 min; 95% (A): 5% (B) to 95%

(A): 5% (B), 13.1–16 min. The 2 μ l samples were injected into the equipment with the flow rate at 0.4 ml/min and a column temperature at 40°C . All samples were stored at 4°C during the period of analysis.

The mass spectrometric data were collected based on a Thermo UHPLC-Q Exactive Mass Spectrometer equipped with an electrospray ionization (ESI) source operating in either positive or negative ion mode. The optimal conditions were set as follows: Aus gas heater temperature at 400°C , sheath gas flow rate at 40 psi, Aus gas flow rate at 30 psi, ion-spray voltage floating (ISVF) at $-2,800$ V in the negative mode and 3,500 V in the positive mode, and normalized collision energy at 20-40-60 V. The mode of data-dependent acquisition (DDA) was performed for data acquisition. The data were collected in the range of 70–1,050 m/z .

Data Preprocessing and Annotation

The Progenesis QI 2.3 (Nonlinear Dynamics, Waters, USA) was applied for detection of peak and alignment of raw data after LC-MS/MS analyses. A data matrix consisting of the retention time (RT), mass-to-charge ratio (m/z) values, and peak intensity can be obtained from the preprocessing analysis. The accurate mass, MS/MS fragments spectra, and isotope ratio difference were determined in comparison to the biochemical databases of Human Metabolome Database (HMDB; <http://www.hmdb.ca/>) and Metlin database (<https://metlin.scripps.edu/>), which were used for identification of metabolic features.

Multivariate Statistical and Differential Metabolite Analysis

The *ropls* (version 1.6.2, <http://bioconductor.org/packages/release/bioc/html/ropls.html>) R package from Bioconductor on Majorbio Cloud Platform (<https://cloud.majorbio.com>) was used for performing the multivariate statistical analysis. The differential metabolites between the comparable groups were determined through orthogonal partial least squares discriminate analysis (PLS-DA). Paired Student's *t*-test on single dimensional statistical analysis was adopted for estimating the *p*-values. The differential metabolites were selected based on the standard of *p*-value < 0.05 and VIP value > 1. The database of Kyoto Encyclopedia of Genes and Genomes (KEGG, <http://www.genome.jp/kegg/>) was applied for the clarification of the participated biochemical pathways and the involved functions of the differential metabolites.

Blood Transcriptome Analysis

Blood Sample Collections

Because no wild equine animals in the natural environment are free from parasites, the group of ivermectin-treated *E. przewalskii* was used as the control group (B-FATPHs). Lucja et al. (39) determined that immune cytokines were present at similar levels, and only four genes were differentially expressed in the blood when they compared children who were not infected with *Schistosoma haematobium* and children who had *S. haematobium* removed with praziquantel treatment 5 h after the treatment. Thus, in the present study, the groups of animals treated with the deworming drug are considered uninfected with *E. przewalskii* parasites.

With the veterinarian's assistance, blood samples from each horse immediately prior to the anthelmintic treatment (B-PATPHs) and 7 days thereafter (B-FATPHs) were collected through jugular vein puncture (37, 40). The whole blood drawn from each horse was placed in a blood collection tube (5 ml) and inverted several times. The blood samples were stored at -80°C until RNA was extracted.

RNA Extraction

Total RNA was extracted from the whole blood samples using TRIzol reagent in accordance with the manufacturer's instructions (Invitrogen, Carlsbad, CA, USA). DNase I (TaKara) was used to remove the genomic DNA. The 2100 Bioanalyzer (Agilent Technologies, Inc., Santa Clara CA, USA) and ND-2000 instrumentation (NanoDrop Thermo Scientific, Wilmington, DE, USA) were applied to determine the integrity and purity and quantify of the RNA. Only high-quality RNA samples (OD 260/280 = 1.8–2.2, OD 260/230 \geq 2.0, RIN \geq 8.0, 28S:18S \geq 1.0, > 1 μg) were used to construct sequencing libraries.

Library Construction and Transcriptome Sequencing

RNA purification, reverse transcription, library construction, and sequencing were performed at Shanghai Majorbio Bio-pharm Biotechnology Co., Ltd. (Shanghai, China) in accordance with the manufacturer's instructions (Illumina, San Diego, CA). The Illumina TruSeq™ RNA sample preparation kit (Illumina) was used to prepare the whole-blood RNA-seq transcriptome

libraries. The poly(A) mRNA purified from the total RNA using oligo-dT-attached magnetic beads was fragmented in a fragmentation buffer. Taking these short fragments as templates, double-stranded cDNA was synthesized using the SuperScript double-stranded cDNA synthesis kit (Invitrogen) with random hexamer primers (Illumina). The cDNA was synthesized after end-repair, and phosphorylation and "A" base addition were applied following the Illumina library construction protocol. Libraries were selected based on the size of cDNA fragments in the range of 200–300 bp on 2% Low Range Ultra Agarose, followed by PCR amplification using Phusion DNA polymerase (New England Biolabs, Boston, MA) for 15 PCR cycles. After fluorometric quantification on TBS380, two RNAseq libraries were sequenced in a single lane on the Illumina HiSeq xten/NovaSeq 6000 sequencer (Illumina) to obtain 2×150 -bp paired-end reads.

De novo Assembly and Annotation

The default parameters of SeqPrep (<https://github.com/jstjohn/SeqPrep>) and Sickle (<https://github.com/najoshi/sickle>) were applied to trim the raw paired-end reads for quality control. The *de novo* assembly was performed using Trinity (<http://trinityrnaseq.sourceforge.net/>) (41). All the assembled transcripts were compared with the following databases: National Center for Biotechnology Information (nonredundant proteins, NR, <https://www.ncbi.nlm.nih.gov/>), Swiss-prot, Pfam, COG, and KEGG using BLASTX to identify the proteins with the highest sequence similarities to the given transcripts to retrieve their functional annotations using a typical cut-off *E*-value of $\leq 1.0 \times 10^{-5}$. GO annotations of the unique assembled transcripts were obtained based on the BLAST2GO program to describe the biological processes, molecular functions, and cellular components (<http://www.blast2go.com/b2ghome>) (42). The KEGG database was employed to analyze the metabolic pathways (43).

Differential Expression Analysis and Functional Enrichment

To identify DEGs between two different samples, the expression level of each transcript was calculated in accordance with the transcripts per million reads (TPM) method. The gene abundances were quantified by RSEM (<http://deweylab.biostat.wisc.edu/rsem/>) (44). DESeq (45) and EdgeR (46) were used to perform the differential expression analysis. The *Q* values ≤ 0.05 and DEGs with $|\log_2\text{FC}| > 1$ were considered to be significant DEGs. Functional enrichment analyses, including GO and KEGG, were performed to identify which DEGs were significantly enriched in GO terms and metabolic pathways at a Benjamini–Hochberg (BH) *p*-adjusted value of ≤ 0.05 compared with the whole transcriptome background. GOATOOLS (<https://github.com/tanghaibao/Goatools>) and KOBAS (<http://bioinfo.org/kobas>) (47) were employed to analyze the GO functional enrichment and KEGG pathway. The R software package was used to conduct the statistical analyses.

RESULTS AND DISCUSSIONS

Detection of Metabolites in All Samples

The chemical analysis showed that a total of 5,913 peaks were observed in the positive ion mode, which were classified into 297 types of metabolites. Meanwhile, a total of 4,869 peaks and 215 metabolites were identified in the negative ion mode. The number of identified metabolites largely depend on the host species variation and the chemical analysis method (48). For the equine animal, a total of 28 metabolites in all fecal samples with high and low parasite burden were identified in a previous study (35), which might be resulted from the feces of youngstock sampled and proton nuclear magnetic resonance (1H-NMR) used. PLS-DA analysis revealed that the composition of the metabolites in *E. przewalskii* fecal samples prior to the treatment (M-PATPHs) and the metabolites in *E. przewalskii* fecal samples following the treatment (M-FATPHs) were different either in the positive mode or in the negative mode (Supplementary Figures S1, S2), which reflected that horse botfly infestation can impact the metabolic activities of intestinal microbiota in horses.

It is known that more than 50% metabolites detected in the feces are from the intestinal microbiota ((49)). The metabolites were annotated based on the KEGG compound classification, which were compounds with biological roles, bioactive peptides, endocrine-disrupting compounds, pesticides, phytochemical compounds, and lipids. Of those functions, the metabolites with biological roles were classified into eight types in the KEGG database, of which the three most abundant metabolites were amino acids, phospholipids, and nucleosides and bases (Supplementary Figure S3). As reported, the fecal metabolites of amino acids, nucleosides, and nucleic acid bases are also abundant in donkeys (48, 50). Amino acids produced by microbes can enhance the immune response of hosts (51). In phytochemical compounds, the most abundant metabolites were flavonoids and isoflavonoids, followed by fatty acids, monolignols, sesquiterpenoids, and triterpenoids (Supplementary Figure S4). Flavonoids are a class of secondary metabolites from plants which can be degraded by intestinal microbes (52). Thus, the flavonoids identified in this study should be from the food of horses. It has been reported that fatty acids and isoprenoids can be involved in many physiological and immune processes (53, 54). Isoflavonoids possess a wide range of pharmacological properties, such as antibacterial function (55), which can alter the symbiosis relationship among microbes.

These identified metabolites can be involved in the seven KEGG pathways, which were organismal systems, human diseases, environmental information processing, cellular processes, metabolism, genetic information processing, and drug development (Supplementary Figure S5). The three pathways with the largest number of metabolites were lipid metabolism, amino acid metabolism, and biosynthesis of other secondary metabolites, all of which were from the pathway of metabolism. These results can confirm the previous function analysis for intestinal microbiota of *E. przewalskii*, which suggested that metabolism is the most important function of the microbiota in the intestine, and the intestinal microbiota of horses mainly

participate in the amino acid metabolism and carbohydrate metabolism (37). The identified metabolites in our study were also involved in the carbohydrate metabolism, which was not a dominant metabolism (Figure 6). In addition, the lipid metabolism is usually related to the synthesis of dietary lipids and the use of dietary nutrients (56, 57). The most identified metabolites produced by microbes through the lipid metabolism confirm our previous assumption that horse botfly infestation can influence the digestive ability of horses (37, 38).

Fecal Metabolites Altered by Horse Botfly Infestation

The different analyses showed that a total of 2,477 metabolites had significant differences ($p < 0.05$) between M-PATPHs and M-FATPHs. Of these metabolites, 143 had been named and 31 had known functions (Supplementary Table S1 and Table 1). Except for L-carnitine, all other functional metabolites were mainly involved in the pathway of metabolism (Table 1). The horse botfly increased the abundance of adenosine 3'-monophosphate, aldosterone, DTMP, guanosine, hyocholic acid, inosine, L-carnitine, LPC (18:1), LPC (18:3), LysoPC (15:0), LysoPC (18:0), LysoPC [18:3 (6Z,9Z,12Z)], phosphocholine, 5-dehydroavenasterol, 2-aminoethylphosphonic acid, 16-hydroxy hexadecanoic acid, and 21-deoxycortisol and decreased the abundance of the rest of the metabolites in Table 1. Among those increased metabolites, adenosine 3'-monophosphate has the function of inhibiting protein synthesis (58). Aldosterone and 21-deoxycortisol are the important physiological stress indicators (59–61). Guanosine and its derivatives have potent inhibitory property against viruses (62, 63). Hyocholic acid and its derivatives are known for its resistance to type 2 diabetes, and it is commonly found in pigs, and its trace amount can be found in humans (64). It is first found in horses. Inosine plays an important role in RNA modulation, gene translation, and purine biosynthesis (65). L-Carnitine is usually linked to the lipid metabolism in animals (66). Phosphocholine is a precursor and also a degradation product of choline (vitamin) (67). 2-Aminoethylphosphonic acid has antibacterial function (68). The aforementioned results suggest that the horse botfly can participate in the metabolism of hosts and exert some beneficial effects on the hosts. This can be the reason why even horse botfly infestation is harmful to horses but not fatal.

Comparison of Total Gene Expression Between Horses With and Without Horse Botfly Infestation

Only the RNA of the blood samples from three Przewalski's horses prior to and following the ivermectin treatment was of sufficient quality for library construction and transcriptome analysis. Altogether, 38.85 GB of clean data with Q30 percentages above 92.26% were obtained from the transcriptome analysis of the six samples (three horses, before and after treatment). Trinity was used to assemble the clean data, which identified 146,447 unigenes and 176,623 transcripts, and the average N50 length was 1,580 bp. The mapping rates ranged from 83.67 to 86.75% after comparing the clean reads from each sample with the

TABLE 1 | Functions of named 31 metabolites which showed significant differences between M-PATPHs and M-FATPHs.

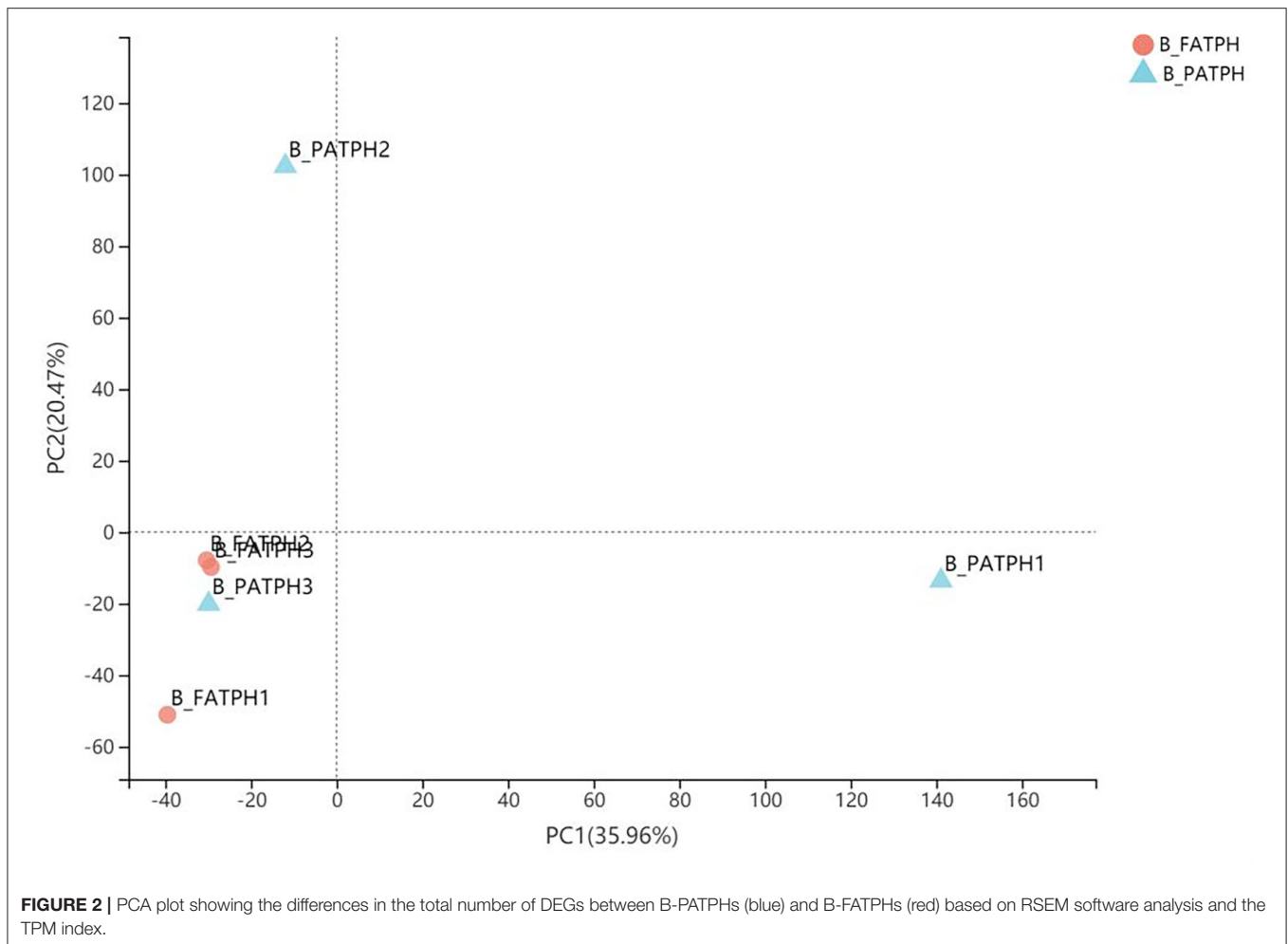
Compound name	Formula	KEGG pathway
16-Hydroxy hexadecanoic acid	C ₁₆ H ₃₂ O ₃	Metabolism
21-Deoxycortisol	C ₂₁ H ₃₀ O ₄	Metabolism
2-Aminoethylphosphonic acid	C ₂ H ₈ NO ₃ P	Metabolism; environmental information processing
2-Formaminobenzoylacetate	C ₁₀ H ₉ NO ₄	Metabolism
2-Indanone	C ₉ H ₈ O	Metabolism
4-(2-Aminophenyl)-2	C ₁₀ H ₉ NO ₄	Metabolism
4-Hydroxy-2-quinolone	C ₉ H ₇ NO ₂	Metabolism
5-Dehydroavenasterol	C ₂₉ H ₄₆ O	Metabolism
9(S)-HOTrE	C ₁₈ H ₃₀ O ₃	Metabolism
9-OxoODE	C ₁₈ H ₃₀ O ₃	Metabolism
Adenosine 3' -monophosphate	C ₁₀ H ₁₄ N ₅ O ₇ P	Metabolism
Aldosterone	C ₂₁ H ₂₈ O ₅	Metabolism; drug development; organismal systems
Atrolactic acid	C ₉ H ₁₀ O ₃	Metabolism
Cytosine	C ₄ H ₅ N ₃ O	Metabolism
Dopaquinone	C ₉ H ₉ NO ₄	Metabolism
DTMP	C ₁₀ H ₁₅ N ₂ O ₈ P	Metabolism; human diseases
Guanosine	C ₁₀ H ₁₃ N ₅ O ₅	Metabolism; environmental information processing
Hyocholic acid	C ₂₄ H ₄₀ O ₅	Metabolism
Inosine	C ₁₀ H ₁₂ N ₄ O ₅	Metabolism; environmental information processing
L-Carnitine	C ₇ H ₁₅ NO ₃	Organismal systems
LPC(18:1)	C ₂₆ H ₅₂ NO ₇ P	Metabolism; human diseases
LPC(18:3)	C ₂₆ H ₄₈ NO ₇ P	Metabolism; human diseases
L-Proline	C ₅ H ₉ NO ₂	Metabolism; organismal systems; environmental information processing; genetic information processing; human diseases
LysoPC(15:0)	C ₂₃ H ₄₈ NO ₇ P	Metabolism; human diseases
LysoPC(18:0)	C ₂₆ H ₅₄ NO ₇ P	Metabolism; human diseases
LysoPC(18:3(6Z,9Z,12Z))	C ₂₆ H ₄₈ NO ₇ P	Metabolism; human diseases
ParaXanthine	C ₇ H ₈ N ₄ O ₂	Metabolism
Phosphocholine	C ₅ H ₁₄ NO ₄ P	Metabolism; human diseases
Phytosphingosine	C ₁₈ H ₃₉ NO ₃	Metabolism
Stercobilin	C ₃₃ H ₄₆ N ₄ O ₆	Metabolism
Xanthurenic acid	C ₁₀ H ₇ NO ₄	Metabolism

reference sequences obtained from the Trinity assembly. In total, 1,564 DEGs were identified, of which 354 were upregulated and 1,210 were downregulated when comparing *E. przewalskii* blood samples prior to the treatment (B-PATPHs) with *E. przewalskii* blood samples following the treatment (B-FATPHs). The unigene annotation of six databases, including NR, Swiss-prot, Pfam, COG, GO, and KEGG, was used to determine functional information about the up- and downregulated genes. The analysis showed that most of the unigenes were annotated in NR (35,041, 24.90%), Swiss-prot (24,855, 17.67%), and COG (24,312, 17.28%). Among the unigenes, 37,954 successfully annotated in seven databases, accounting for 26.98%. The principal component analysis (PCA), which is used to visualize the expression differences and similarities between the B-PATPHs and B-FATPHs, showed that the whole-blood transcriptome profiles of the B-FATPHs clustered together, but the B-PATPHs were separated from each other, with 35.96 and 20.47% of the variation attributed to PC1 and PC2, respectively (Figure 2).

Identifying DEGs Between Horses With and Without Horse Botfly Infestation and Analyzing Functional Enrichment

Host responses to the botfly infestations were investigated by identifying the top DEGs between the B-PATPHs and B-FATPHs. The DEGs were considered to have a significant difference between the groups when the log₂FC values exceeded 2, and the BH *p*-adjusted values were below 0.05. The volcano plot shown in Figure 3 indicates that the top upregulated DEGs included “protein FAM19A3 isoform X2” (source: *E. caballus*, *p* = 2.47E-06), “RNA binding motif” (*p* = 5.00E-06), and “cathepsin W” (source: *E. caballus*, *p* = 9.89E-06), of which cathepsin W is involved in proteolysis and cellular protein catabolic process (GO:0051603), extracellular space (GO:0005615), lysosome pathway (GO:0005764), and cysteine-type endopeptidase activity (GO:0004197; GO:0008234). The protein product of the cathepsin W gene also participates in apoptosis and lysosome pathways (K08569). Previous studies have also shown that the cathepsin gene is involved in the development of innate and adaptive immune responses of mammalian hosts during infection with helminth parasites (69, 70). Apoptosis is an immune-related pathway involved in the genetic programming process used by multicellular organisms to regulate their cell numbers or to eliminate harmful cells (71).

Meanwhile, the top downregulated DEGs were also identified in this study. They were “protein S100-A12” (source: *Tupaia chinensis*, *p* = 4.93E-67), “protein S100-A8” (source: *E. przewalskii*, *p* = 1.89E-55), “PREDICTED: protein S100-A9-like isoform X2” (source: *E. przewalskii*, *p* = 1.24E-47), “neutrophil gelatinase-associated lipocalin isoform X1” (source: *E. caballus*, *p* = 6.14E-43), “NADPH: adrenodoxin oxidoreductase, mitochondrial isoform X4” (source: *E. caballus*, *p* = 2.34E-38), “histidine ammonia-lyase” (source: *Panthera tigris altaica*, *p* = 8.76E-36), and “cystatin-A” (source: *E. caballus*, *p* = 7.83E-33). Two of the downregulated genes are associated with immune functions (e.g., “protein S100-A8” and “protein S100-A9-like isoform X2”). Protein S100-A8 is

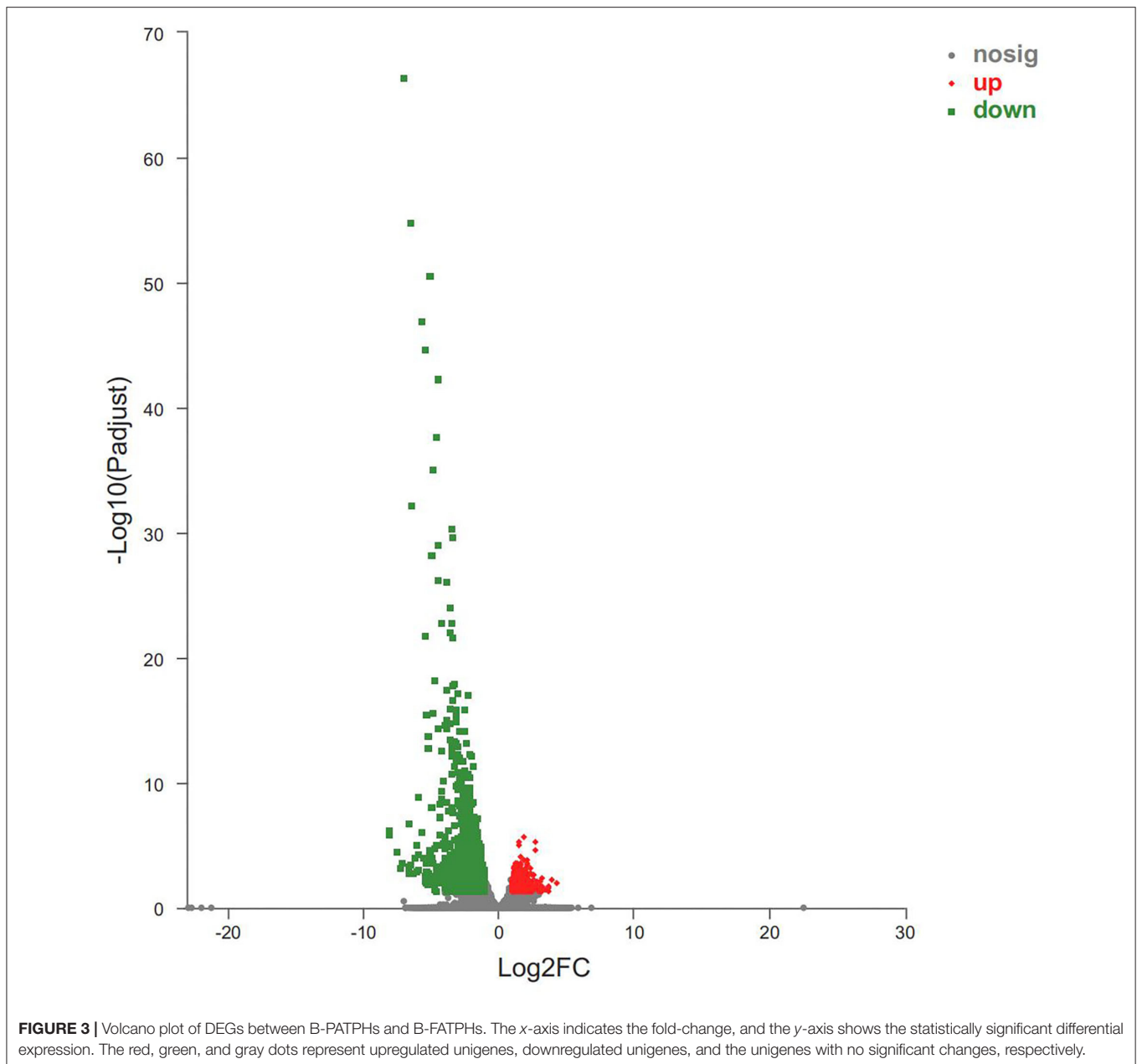


associated with chemokine production (GO:0032602), cytokine production (GO:0001816), toll-like receptor signaling pathway (GO:0002224; GO:0035662), and the immune and inflammatory response (GO:0002523; GO:0002526; GO:0002544; GO:0045087; GO:0050727; GO:0050729; GO:0006954). The gene encoding S100-A9-like isoform X2 functions in immune and inflammatory responses (GO:0045087 and GO:0050727) and toll-like receptor 4 binding (GO:0035662). As reported previously, cytokines are key mediators of immunity and participate in multiple biological processes in the host, such as defense against parasitic infections, tissue repair, allergic inflammation, and metabolic homeostasis (72, 73). Recent studies have suggested that mice infected with *Neospora caninum* parasites have higher levels of cytokines (mainly IFN- γ and IL-12) than uninfected controls (74), whereas mice infected with *Trypanosoma cruzi* have higher TNF- α and IL-1 β cytokine levels than uninfected controls (75). Chemokines, such as CXCL10, have also been shown to play important roles in the development of immunity, and their levels show a reduced trend in parasite-infected hosts (76). TLRs contribute a great deal to the immune response by activating the immune system through pathogen recognition (77), a finding applicable to equine animals (19, 21). One study concludes that TLR expression levels will decrease significantly after the host is infected with parasites,

to avoid an excessive immune response against the parasites (78). Our results are consistent with those of that study, that is, the toll-like receptor signaling pathway was downregulated in B-PATPHs. Some genes involved in immune-related inflammatory responses were found to be differentially expressed in the present study. Thus, the aforementioned results indicated that host immune response genes were activated in wild *E. przewalskii* horses infested with horse botflies, which can help understand the immune status of the host.

Finally, the functional enrichment analysis was performed to identify the functional differences in the DEGs between B-PATPHs and B-FATPHs. The GO annotation for unigenes suggested that these genes could be classified into three GO terms: biological process, cellular component, and molecular function (Figure 4). The GO functional enrichment analysis indicated that the most significantly different GO term was signaling pattern recognition receptor activity, while the defense response function contained the largest number of unigenes (64) (Supplementary Figure S6). This result is consistent with the findings of another study, where the defense response function is abundant in monkeys after immunization (79).

As for the participating KEGG functions, the first five annotated categories included human diseases, organismal

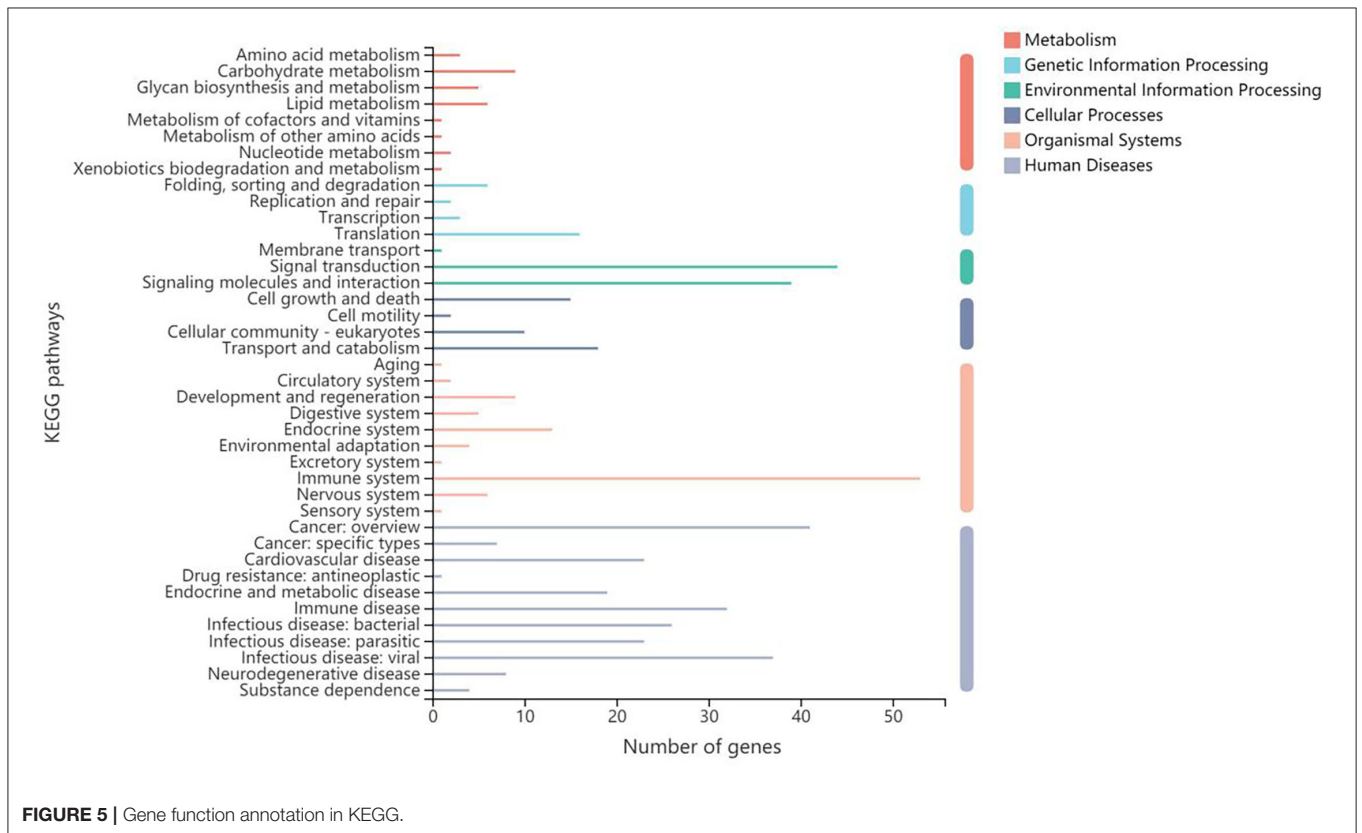
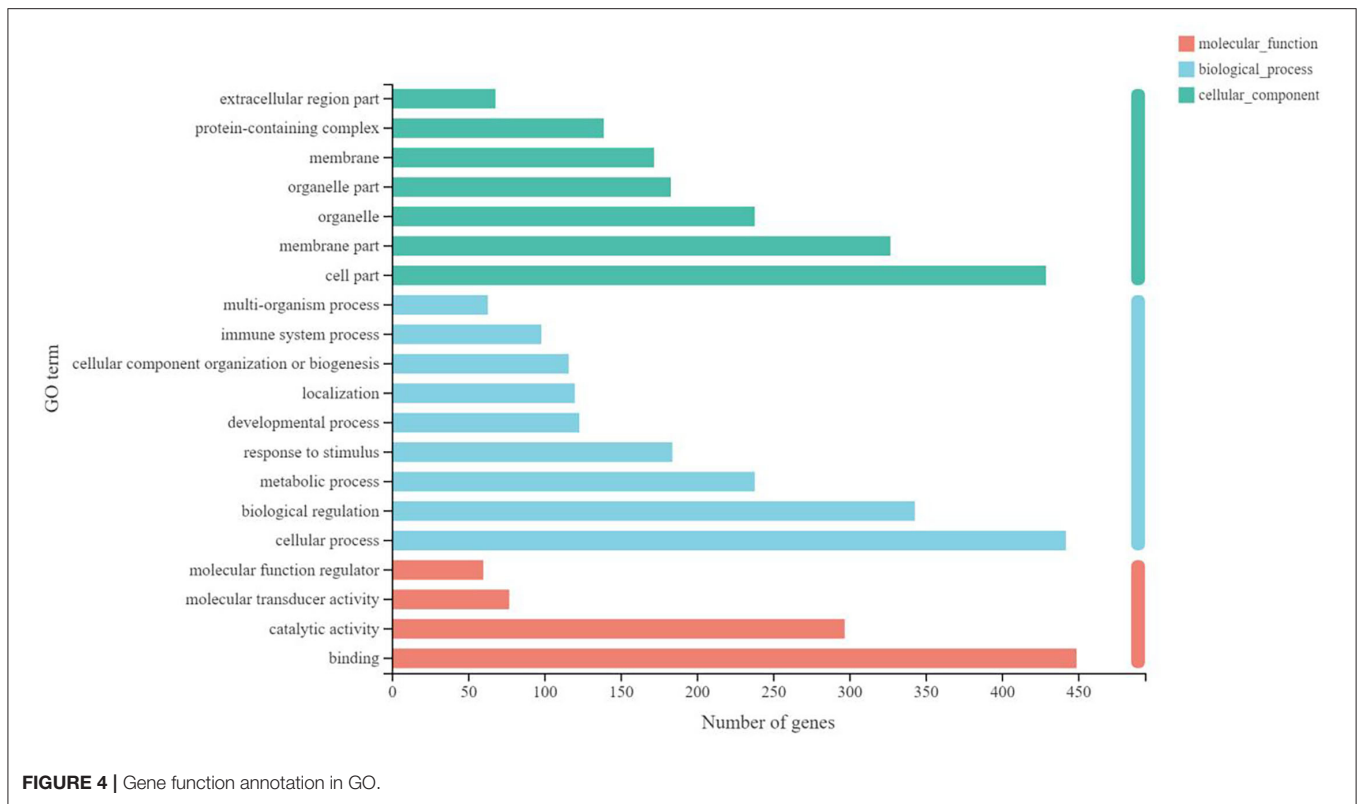


systems, metabolism, cellular processes, genetic information processing, and environmental information processing (**Figure 5**). The top three second categories included the immune system, signal transduction, and cancer: overview. The KEGG enrichment analysis determined that the second-category DEGs were mainly enriched in pathways relating to cancer (47 unigenes), Th17 cell differentiation (36 unigenes), and osteoclast differentiation (**Supplementary Figure S7**); the latter two of were significantly different between B-PATPHs and B-FATPHs. Th17 is found to be involved in the immune response against parasite infection. While Ortiz Wilczyński et al. (80) show that Th17 cell numbers are reduced in patients infected with helminths, in the mouse model, serum Th17 levels increase after infection with

helminths (81). Thus, we assume that horse botfly infestation would trigger an immune response in *E. przewalskii* that might affect the progression of parasite infections with botflies.

Associations Between Metabolites and Gene Expression

The altered metabolites and DEGs were commonly involved in 19 KEGG pathways, of which the top 10 were glycerophospholipid metabolism, choline metabolism in cancer, ABC transporters, purine metabolism, tryptophan metabolism, pyrimidine metabolism, steroid hormone biosynthesis, tyrosine metabolism, thermogenesis, and bile secretion (**Figure 6**). Among them, purine metabolism and pyrimidine metabolism contained



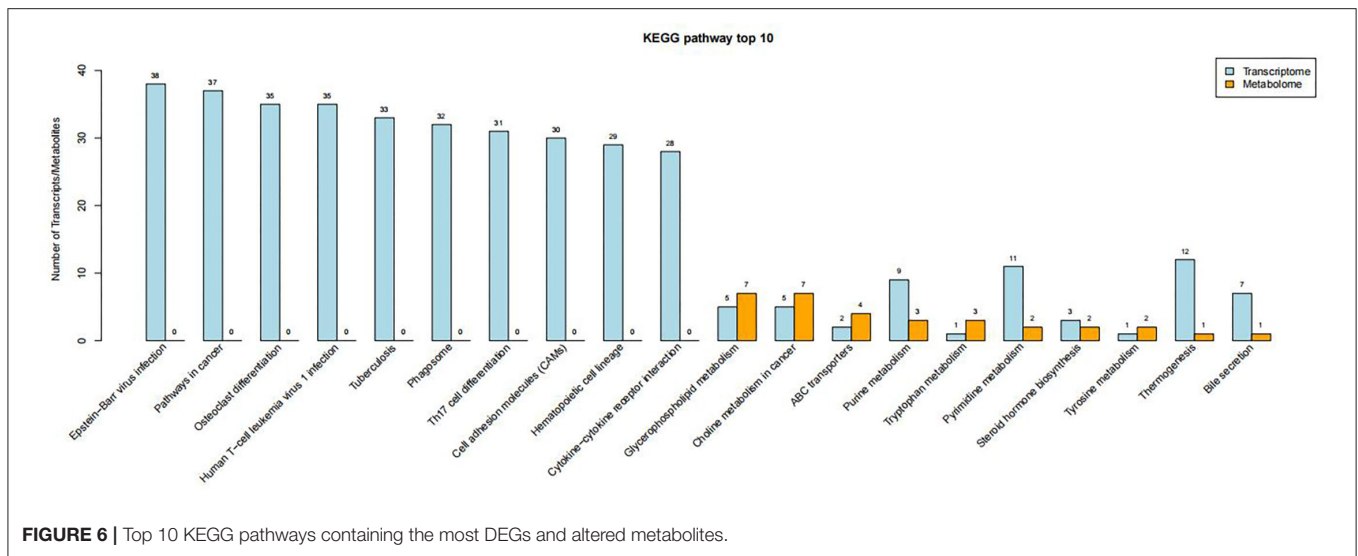


FIGURE 6 | Top 10 KEGG pathways containing the most DEGs and altered metabolites.

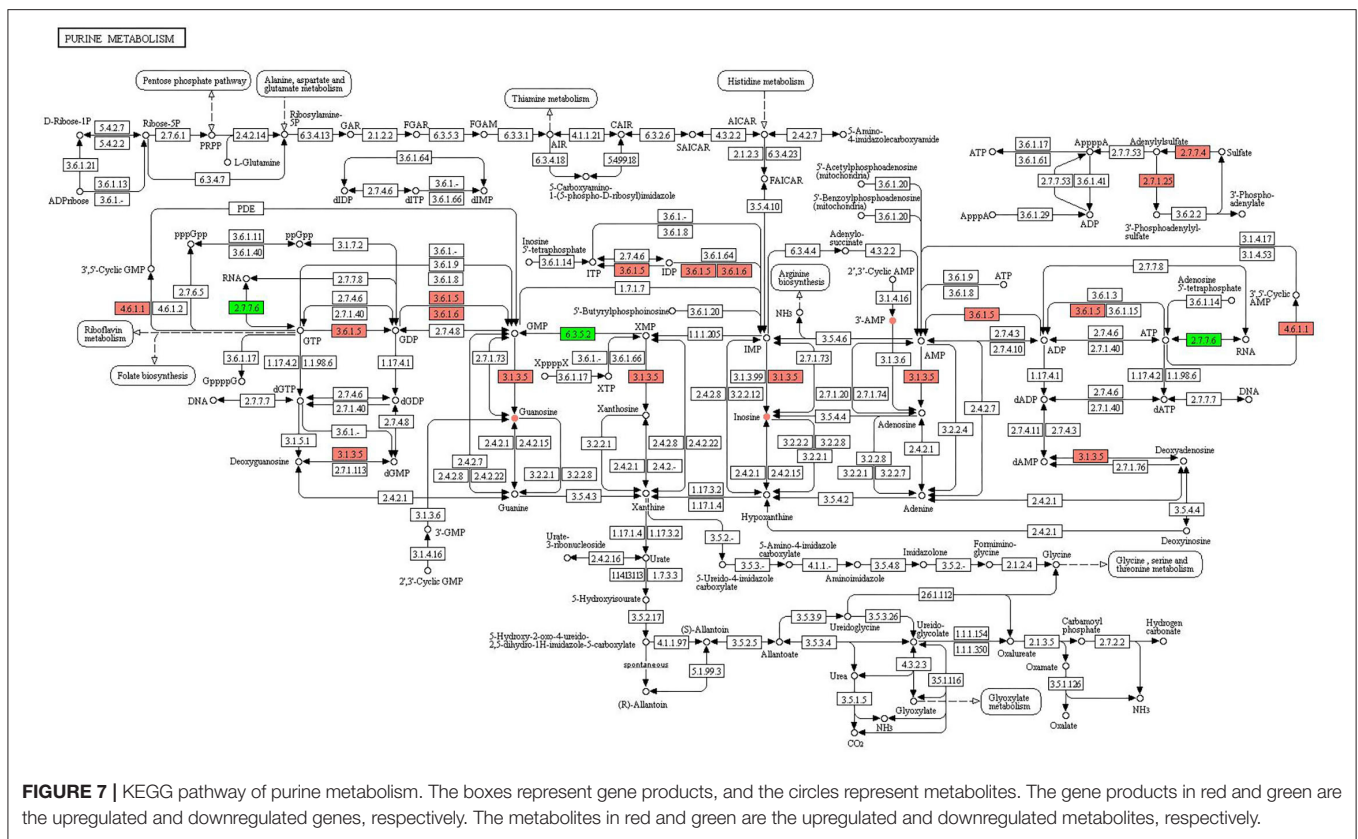


FIGURE 7 | KEGG pathway of purine metabolism. The boxes represent gene products, and the circles represent metabolites. The gene products in red and green are the upregulated and downregulated genes, respectively. The metabolites in red and green are the upregulated and downregulated metabolites, respectively.

the most altered metabolites and DEGs (Figures 7, 8, Supplementary Figures S8–S15).

In purine metabolism (Figure 7), the upregulated genes were involved in the processes of converting GTP to 3',5'-cyclic GMP (4.6.1.1), GTP to GDP (3.6.1.5), dGMP to deoxyguanosine (3.1.3.5), GDP to GMP (3.6.1.5, 3.6.1.6), GMP to guanosine (3.1.3.5), IDP to ITP (3.6.1.5), IDP to IMP (3.6.1.5, 3.6.1.6), IMP to inosine (3.1.3.5), AMP to adenosine (3.1.3.5), ADP to AMP (3.6.1.5), ATP to ADP (3.6.1.5), ATP to 3',5'-cyclic AMP (4.6.1.1),

sulfate to adenylylsulfate (2.7.7.4), and adenylylsulfate to 3'-phosphoadenylyl-sulfate (2.7.1.25). The downregulated genes were involved in the process of GTP to RNA (2.7.7.6), XMP to GMP (6.3.5.2), and ATP to RNA (2.7.7.6). The upregulated metabolites were guanosine, inosine, and 3' AMP, and no downregulated metabolites were identified.

Figure 8 shows the KEGG pathway of purine metabolism. The upregulated genes were associated with the pathways of UTP to UDP (3.6.1.5), UDP to UMP (3.6.1.5, 3.6.1.6), UMP to uridine

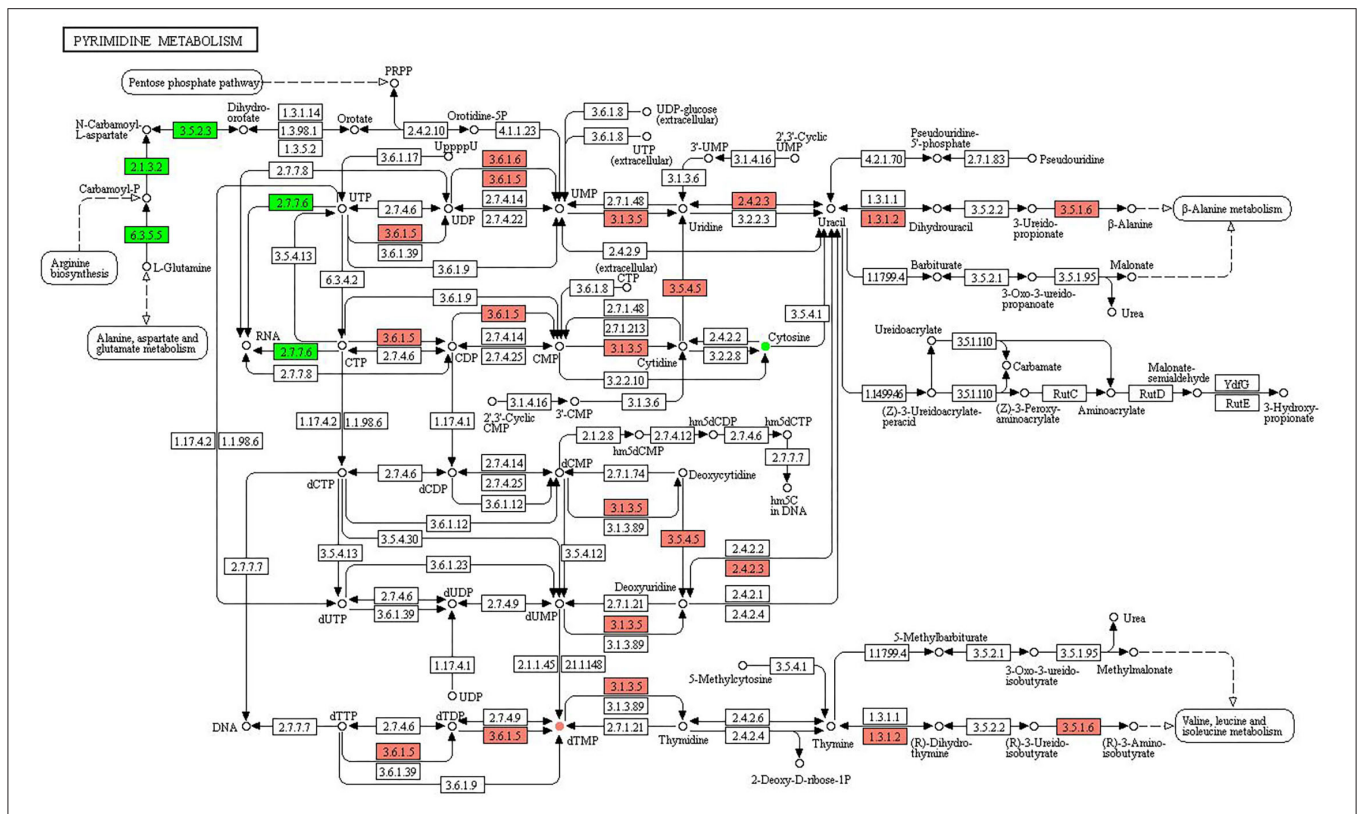


FIGURE 8 | KEGG pathway of pyrimidine metabolism. The boxes represent gene products, and the circles represent metabolites. The gene products in red and green are the upregulated and downregulated genes, respectively. The metabolites in red and green are the upregulated and downregulated metabolites, respectively.

(3.1.3.5), uridine to uracil (2.4.2.3), uracil to dihydrouracil (1.3.1.2), uracil to deoxyuridine (2.4.2.3), 3-ureido-propionate to β -alanine (3.5.1.6), CTP to CDP (3.6.1.5), CDP to CMP (3.6.1.5), CMP to cytidine (3.1.3.5), cytidine to uridine (3.5.4.5), dCMP to deoxycytidine (3.1.3.5), deoxycytidine to deoxyuridine (3.5.4.5), dUMP to deoxyuridine (3.1.3.5), dTTP to dTDP (3.6.1.5), dTDP to dTMP (3.6.1.5), dTMP to thymidine (3.1.3.5), thymine to (R)-dihydrothymine (1.3.1.2), and (R)-3-ureidoisobutyrate to (R)-3-aminoisobutyrate (3.5.1.6). The downregulated genes were involved in L-glutamine to carbamoyl-P (6.3.5.5), carbamoyl-P to N-carbamoyl-L-aspartate (2.1.3.2), N-carbamoyl-L-aspartate to dihydroorotate (3.5.2.3), CTP to RNA (2.7.7.6), and UTP to RNA (2.7.7.6). The upregulated and downregulated metabolites were dTMP and cytosine, respectively.

In mammals, when mitochondrial activity is unable to support energy demand, muscles produce ATP through the adenylate kinase reaction in the purine metabolism. The product of purine metabolism is uric acid (82). There are differences in purine metabolism in different species, which cause gout and renal stones in humans, but the effects of which on horses are not known (83). One study shows that uric acid levels increase when horses engage in strenuous activities (84). In contrast to purine metabolism, the pyrimidine metabolism shows little difference in different species. Pyrimidines are the critical nutrients and play an important role in maintaining the physiological homeostasis of the host, and the metabolic aberrations of pyrimidines disrupt nervous, mitochondrial, and

hematological systems (85). This led us to assume that horse botfly infestation may increase the energy consumption, affect the nutrient absorption, and damage normal physiological processes of the horses.

CONCLUSION

The parasite infection is well known to influence the intestinal microbial community and composition. This study aims to explain the whole process after the host was infected with parasites and uncover a major role of intestinal microbiota and its metabolites in the modulation of the host immune response. The results show that while changing intestinal microbiota, parasites also adjust the types and levels of metabolites microbiota produce. Then, the altered microbiota and their metabolites will stimulate the immune response of the host, which might occur through purine and pyrimidine metabolism. This study can provide some reference information for further studies, such as fecal microbiota transplant to restore the host immunity and develop probiotics and microbial-related drugs. Future studies are needed and valuable.

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 in the article/supplementary material.

ETHICS STATEMENT

The animal study was reviewed and approved by Ethics Committee of the Beijing Forestry University (EAWC_BJFU_2020012).

AUTHOR CONTRIBUTIONS

DH, CW, YQ, ME, XL, and KL conceived the experiments and undertook sampling work. DH, DZ, KL, and HC analyzed the results and wrote the manuscript. All authors read and approved the final manuscript.

REFERENCES

- Zumpt F. *Myiasis in Man and Animals in the Old World: Textbook For Physicians, Veterinarians and Zoologist*. London: Butterworth (1965).
- Soulsby E. *Helminths, Arthropods and Protozoa of Domesticated Animals*, 7th ed. London: Bailliere Tindall (1982).
- Li K, Wu Z, Hu DF, Cao J, Wang C. A report on new causative agent (*Gasterophilus* spp.) of the myiasis of Przewalski's horse occurred in China. *Chin J Anim Vet Sci.* (2007) 38:837–40. doi: 10.3321/j.issn:0366-6964.2007.08.015
- Sequeira, J., and Tostes, R. Oliveira-Sequeira, T. Prevalence and macro- and microscopic lesions produced by *Gasterophilus nasalis* (Diptera: *Oestridae*) in the Botucatu Region, SP, Brazil. *Vet Parasitol.* (2001) 102:261–6. doi: 10.1016/S0304-4017(01)00536-2
- Xing J, Li P, Yan W Y. The dynamic observation on the immunological function of rat infected with *Trichinella spiralis*. *Chin J Parasit Dis Control.* (2005) 18:26–7. doi: 10.3969/j.issn.1673-5234.2005.01.008
- Wang J, Cui J, Wang ZQ. Serum IgG levels in the mice experimentally infected with *Trichinella* spp. *J Pathogen Biol.* (2007) 2:266–7. doi: 10.3969/j.issn.1673-5234.2007.04.008
- Valdes AM, Walter J, Segal E, Spector T. Role of the gut microbiota in nutrition and health. *BMJ Clin Res.* (2018) 361:k2179. doi: 10.1136/bmj.k2179
- Tilg H, Zmora N, Adolph TE, Elinav E. The intestinal microbiota fuelling metabolic inflammation. *Nat Rev Immunol.* (2020) 20:40–54. doi: 10.1038/s41577-019-0198-4
- Fan Y, Pedersen O. Gut microbiota in human metabolic health and disease. *Nat Rev Microbiol.* (2020) 19:55–71. doi: 10.1038/s41579-020-0433-9
- Becker L, Spear ET, Sinha SR, Haileselassie Y, Habtezion A. Age-related changes in gut microbiota alter phenotype of muscularis macrophages and disrupt gastrointestinal motility. *Cell Mol Gastroenterol Hepatol.* (2019) 7:243–5.e2. doi: 10.1016/j.jcmgh.2018.09.001
- Hu D, Chao Y, Li Y, Peng X, Wang C, Wang Z, et al. Effect of gender bias on equine fecal microbiota. *J Equine Vet Sci.* (2020) 97:103355. doi: 10.1016/j.jevs.2020.103355
- Martín-Mateos R, Albillos A. The role of the gut-liver axis in metabolic dysfunction-associated fatty liver disease. *Front Immunol.* (2021) 12:660179. doi: 10.3389/fimmu.2021.660179
- Caffaratti C, Plazy C, Mery G, Tidjani AR, Fiorini F, Thiroux S, et al. What we know so far about the metabolite-mediated microbiota-intestinal immunity dialogue and how to hear the sound of this crosstalk. *Metabolites.* (2021) 11:406. doi: 10.3390/metabo11060406
- Louis P, Flint HJ. Formation of propionate and butyrate by the human colonic microbiota. *Environ Microbiol.* (2017) 19:29–41. doi: 10.1111/1462-2920.13589
- Yang W, Yu T, Huang X, Bilotta AJ, Xu L, Lu Y, et al. Intestinal microbiota-derived short-chain fatty acids regulation of immune cell IL-22 production and gut immunity. *Nat Commun.* (2020) 11:4457. doi: 10.1038/s41467-020-18262-6
- Belenguer A, Duncan SH, Holtrop G, Anderson SE, Lobley GE, Flint HJ. Impact of pH on lactate formation and utilization by human fecal microbial communities. *Appl Environ Microbiol.* (2007) 73:6526–33. doi: 10.1128/AEM.00508-07
- Errea A, Cayet D, Marchetti P, Tang C, Kluza J, Offermanns S, et al. Lactate inhibits the pro-inflammatory response and metabolic reprogramming in murine macrophages in a GPR81-independent manner. *PLoS ONE.* (2016) 11:e0163694. doi: 10.1371/journal.pone.0163694
- Castagnetti C, Mariella J, Pirrone A, Cinotti S, Mari G, Peli A. Expression of interleukin-1beta, interleukin-8, and interferon-gamma in blood samples obtained from healthy and sick neonatal foals. *Am J Vet Res.* (2012) 73:1418–27. doi: 10.2460/ajvr.73.9.1418
- Fossum C, Hjertner B, Olofsson K M, Lindberg R, Ahooghalandari P, Camargo M M, et al. Expression of tlr4, md2 and cd14 in equine blood leukocytes during endotoxin infusion and in intestinal tissues from healthy horses. *Vet Immunol Immunopathol.* (2012) 150:141–8. doi: 10.1016/j.vetimm.2012.09.005
- Vinther AM, Skovgaard K, Heegaard PM, Andersen PH. Dynamic expression of leukocyte innate immune genes in whole blood from horses with lipopolysaccharide-induced acute systemic inflammation. *BMC Vet Res.* (2015) 11:134. doi: 10.1186/s12917-015-0450-5
- Migdal A, Migdal Ł, Oczkiewicz M, Okólski A, Chełmońska-Soyta A. Influence of age and immunostimulation on the level of toll-like receptor gene (TLR3, 4, and 7) expression in foals. *Animals.* (2020) 10:1966. doi: 10.3390/ani10111966
- Zarski LM, Weber PSD, Lee Y, Soboll Hussey G. Transcriptomic profiling of equine and viral genes in peripheral blood mononuclear cells in horses during equine herpesvirus 1 infection. *Pathogens.* (2021) 10:43. doi: 10.3390/pathogens10010043
- Graciela LU, José B. Dominant IgM synthesis against the soluble form of the prevailing surface glycoprotein from TeAp-N/D1 Trypanosoma equiperdum throughout the experimental acute infections of horses with non-tsetse transmitted Trypanozoon parasites. *J Immunoassay Immunochem.* (2020) 41:745–60. doi: 10.1080/15321819.2020.1778029
- Jürgenschellert L, Krücken J, Austin CJ, Lightbody KL, Bousquet E, Samson-Himmelstjerna G. Investigations on the occurrence of tapeworm infections in German horse populations with comparison of different antibody detection methods based on saliva and serum samples. *Parasit Vectors.* (2020) 13:462. doi: 10.1186/s13071-020-04318-5
- Tzelos T, Geyer KK, Mitchell MC, Mcwilliam HEG, Matthews JB. Characterisation of serum IgG(T) responses to potential diagnostic antigens for equine cyathostomiasis. *Int J Parasitol.* (2020) 50:289–98. doi: 10.1016/j.ijpara.2020.01.004
- Mukhopadhyay D, Arranz-Solis D, Saeij JJP. Influence of the host and parasite strain on the immune response during *Toxoplasma* infection.

FUNDING

This research was funded by the China Postdoctoral Science Foundation with (No. 2020TQ0047); the National Science Foundation of China (No. 31670538); the Species Project (2018) of Department for Wildlife and Forest Plants Protection, NFGA of China; and the Forestry Fund of LiBin (02210823).

SUPPLEMENTARY MATERIAL

The Supplementary Material for this article can be found online at: <https://www.frontiersin.org/articles/10.3389/fvets.2022.832062/full#supplementary-material>

- Front Cell Infect Microbiol.* (2020) 10:580425. doi: 10.3389/fcimb.2020.580425
27. Vyas A. Mechanisms of host behavioral change in *Toxoplasma gondii* rodent association. *PLoS Pathogen.* (2015) 11:e1004935. doi: 10.1371/journal.ppat.1004935
 28. Tedford E, McConkey G. Neurophysiological changes induced by chronic *Toxoplasma gondii* infection. *Pathogens.* (2017) 6:19. doi: 10.3390/pathogens6020019
 29. Marra CM. Central nervous system infection with *Toxoplasma gondii*. *Handb Clin Neurol.* (2018) 152:117–22. doi: 10.1016/B978-0-444-63849-6.00009-8
 30. Schlüter D, Barragan A. Advances and challenges in understanding cerebral toxoplasmosis. *Front Immunol.* (2019) 10:242. doi: 10.3389/fimmu.2019.00242
 31. Brosnahan M. Eosinophils of the horse: part II: eosinophils in clinical diseases. *Equine Vet Educ.* (2020) 32:590–602. doi: 10.1111/eve.13262
 32. Steuer AE, Stewart JC, Barker VD, Adams AA, Nielsen MK. Cytokine and goblet cell gene expression in equine cyathostomin infection and larvicidal anthelmintic therapy. *Parasite Immunol.* (2020) 42:e12709. doi: 10.1111/pim.12709
 33. Peachey LE, Molena RA, Jenkins TP, Cesare AD, Traversa D, Hodgkinson JE, et al. The relationships between faecal egg counts and gut microbial composition in UK thoroughbreds infected by cyathostomins. *Int J Parasitol.* (2018) 48:403–12. doi: 10.1016/j.ijpara.2017.11.003
 34. Walshe N, Duggan V, Cabrera-Rubio R, Crispie F, Cotter P, Feehan O, et al. Removal of adult cyathostomins alters faecal microbiota and promotes an inflammatory phenotype in horses. *Int J Parasitol.* (2019) 49:489–500. doi: 10.1016/j.ijpara.2019.02.003
 35. Peachey LE, Castro C, Molena RA, Jenkins TP, Griffin JL, Cantacessi C. Dysbiosis associated with acute helminth infections in herbivorous youngstock – observations and implications. *Sci Rep.* (2019) 9:11121. doi: 10.1038/s41598-019-47204-6
 36. Costa MC, Stämpfli HR, Allen-Vercoe E, Weese JS. Development of the faecal microbiota in foals. *Equine Vet J.* (2016) 48:681–8. doi: 10.1111/evj.12532
 37. Hu D, Chao Y, Zhang B, Wang C, Qi Y, Ente M, et al. Effects of *Gasterophilus pecorum* infestation on the intestinal microbiota of the rewilded Przewalski's horses in China. *PLoS ONE.* (2021) 16:e0251512. doi: 10.1371/journal.pone.0251512
 38. Hu D, Yang J, Qi Y, Li B, Li K, Mok KM. Metagenomic analysis of fecal archaea, bacteria, eukaryota and virus in Przewalski's horses following anthelmintic treatment. *Front Vet Sci.* (2021) 8:708512. doi: 10.3389/fvets.2021.708512
 39. Lucja AL, Ayola AA, Bruce AR, John M, Ulysse A-N, Abena SA, et al. A praziquantel treatment study of immune and transcriptome profiles in Schistosoma haematobium-infected gabonese schoolchildren. *J Infect Dis.* (2020) 222:2103–13. doi: 10.1093/infdis/jiz641
 40. Jimenez J, Timsit E, Orsel K, van der Meer F, Guan LL, Plastow G. Whole-blood transcriptome analysis of feedlot cattle with and without bovine respiratory disease. *Front Genet.* (2021) 12:627623. doi: 10.3389/fgene.2021.627623
 41. Grabherr MG, Haas BJ, Yassour M, Levin JZ, Thompson DA, Amit I, et al. Full-length transcriptome assembly from RNA-Seq data without a reference genome. *Nat Biotechnol.* (2011) 29:644–52. doi: 10.1038/nbt.1883
 42. Conesa A, Gotz S, Garcia-Gomez JM, Terol J, Talon M, Robles M. Blast2GO: a universal tool for annotation, visualization and analysis in functional genomics research. *Bioinformatics.* (2005) 21:3674–6. doi: 10.1093/bioinformatics/bti610
 43. Goto M. KEGG: kyoto encyclopedia of genes and genomes. *Nucleic Acids Res.* (2000) 28:27–30. doi: 10.1093/nar/28.1.27
 44. Li B, Dewey CN. RSEM: accurate transcript quantification from RNA-Seq data with or without a reference genome. *BMC Bioinformatics.* (2011) 12:323. doi: 10.1186/1471-2105-12-323
 45. Love MI, Huber W, Anders S. Moderated estimation of fold change and dispersion for RNA-seq data with DESeq2. *Genome Biol.* (2014) 15:550. doi: 10.1186/s13059-014-0550-8
 46. Robinson MD, McCarthy DJ, Smyth GK. edgeR: a bioconductor package for differential expression analysis of digital gene expression data. *Bioinformatics.* (2010) 26:139–40. doi: 10.1093/bioinformatics/btp616
 47. Xie C, Mao X, Huang J, Ding Y, Wu J, Dong S, et al. KOBAS 2.0: a web server for annotation and identification of enriched pathways and diseases. *Nucleic Acids Res.* (2011) 39:W316–22. doi: 10.1093/nar/gkr483
 48. Tsuchida S, Hattori T, Sawada A, Ogata K, Watanabe J, Ushida K. Fecal metabolite analysis of Japanese macaques in Yakushima by LC-MS/MS and LC-QTOF-MS. *J Vet Med Sci.* (2021) 83:1012–5. doi: 10.1292/jvms.21-0076
 49. Zheng X, Xie G, Zhao A, Zhao L, Yao C, Chiu NHL, et al. The footprints of gut microbial-mammalian co-metabolism. *J Proteome Res.* (2011) 10:5512–22. doi: 10.1021/pr2007945
 50. Kobayashi A, Tsuchida S, Hattori T, Ogata K, Ueda A, Yamada T, et al. Metabolomic LC-MS/MS analyses and meta 16S rRNA gene analyses on cecal feces of Japanese rock ptarmigans reveal fundamental differences between semi-wild and captive raised individuals. *J Vet Med Sci.* (2020) 82:1165–72. doi: 10.1292/jvms.20-0003
 51. Du L, Sun Y, Wng Q, Wang L, Zhang Y, Li S, et al. Integrated metabolomics and 16S rDNA sequencing to investigate the mechanism of immune-enhancing effect of health Tonic oral liquid. *Food Res Int.* (2021) 144:110323. doi: 10.1016/j.foodres.2021.110323
 52. Vollmer M, Esders S, Farquharson FM, Neugart S, Duncan SH, Schreiner M, et al. Mutual interaction of phenolic compounds and microbiota: metabolism of complex phenolic Apigenin-C- and Kaempferol-O-Derivatives by human fecal samples. *J Agric Food Chem.* (2018) 66:485–97. doi: 10.1021/acs.jafc.7b04842
 53. Dono A, Patrizz A, McCormack RM, Putluri N, Ganesh BP, Kaur B, et al. Glioma induced alterations in fecal short-chain fatty acids and neurotransmitters. *CNS Oncol.* (2020) 9:CNS57. doi: 10.2217/cns-2020-0007
 54. Marques JG, Shokry E, Frivolt K, Werkstetter KJ, Brückner A, Schwerdt T, et al. Metabolomic signatures in pediatric Crohn's disease patients with mild or quiescent disease treated with partial enteral nutrition: a feasibility study. *SLAS Technol.* (2021) 26:165–77. doi: 10.1177/2472630320969147
 55. Zhang W, Jiang S, Qian D, Shang EX, Duan JA. Analysis of interaction property of calycosin-7-O-β-D-glucoside with human gut microbiota. *J Chromatogr B Analit Technol Biomed Life Sci.* (2014) 963:16–23. doi: 10.1016/j.jchromb.2014.05.015
 56. Martinez-Guryn K, Hubert N, Frazier K, Urlass S, Musch MW, Ojeda P, et al. Small intestine microbiota regulate host digestive and absorptive adaptive responses to dietary lipids. *Cell Host Microbe.* (2018) 23:458. doi: 10.1016/j.chom.2018.03.011
 57. Li S, Wang C, Wu Z. Dietary l-arginine supplementation of tilapia (*Oreochromis niloticus*) alters the microbial population and activates intestinal fatty acid oxidation. *Amino Acids.* (2021) 54:339–51. doi: 10.1007/s00726-021-03018-3
 58. Fuhr JE, Stidham JD. Inhibitory effect of cyclic adenosine 2,3'-monophosphate on leucine incorporation by L5178Y cells. *J Cell Physiol.* (1980) 103:71–5. doi: 10.1002/jcp.1041030111
 59. Fiet J, Gueux B, Raux-Demay MC, Kuttenn F, Vexiau P, Gourmelen M, et al. Le 21-désoxycortisol. Un nouveau marqueur de l'hyperandrogénie surrénalienne par déficit en 21-hydroxylase [21-deoxycortisol. A new marker of virilizing adrenal hyperplasia caused by 21-hydroxylase deficiency]. *Presse Med.* (1989) 18:1965–9.
 60. Burgess EA, Hunt KE, Kraus SD, Rolland RM. Adrenal responses of large whales: Integrating fecal aldosterone as a complementary biomarker to glucocorticoids. *Gen Comp Endocrinol.* (2017) 252:103–10. doi: 10.1016/j.ygcen.2017.07.026
 61. Miller LJ, Lauderdale LK, Walsh MT, Bryant JL, Mitchell KA, Granger DA, et al. Reference intervals and values for fecal cortisol, aldosterone, and the ratio of cortisol to dehydroepiandrosterone metabolites in four species of cetaceans. *PLoS ONE.* (2021) 16:e0250331. doi: 10.1371/journal.pone.0250331
 62. Elfiky AA. Novel guanosine derivatives as anti-HCV NS5b polymerase: a QSAR and molecular docking study. *Med Chem.* (2019) 15:130–7. doi: 10.2174/1573406414666181015152511
 63. Elfiky AA. Novel guanosine derivatives against Zika virus polymerase in silico. *J Med Virol.* (2020) 92:11–6. doi: 10.1002/jmv.25573
 64. Zheng X, Chen T, Jiang R, Zhao A, Wu Q, Kuang J, et al. Hypocholic acid species improve glucose homeostasis through a distinct TGR5 and FXR signaling mechanism. *Cell Metab.* (2021) 33:791–803.e7. doi: 10.1016/j.cmet.2020.11.017

65. Srinivasan S, Torres AG, Ribas de Pouplana L. Inosine in biology and disease. *Genes*. (2021) 12:600. doi: 10.3390/genes12040600
66. Jacques F, Rippa S, Perrin Y. Physiology of L-carnitine in plants in light of the knowledge in animals and microorganisms. *Plant Sci*. (2018) 274:432–40. doi: 10.1016/j.plantsci.2018.06.020
67. Xu L, Berg L, Jamin KD, Townsend SD. An effective reagent to functionalize alcohols with phosphocholine. *Org Biomol Chem*. (2020) 8:767–70. doi: 10.1039/C9OB02582K
68. Stamper GF, Morollo AA, Ringe D. Reaction of alanine racemase with 1-aminoethylphosphonic acid forms a stable external aldimine. *Biochemistry*. (1998) 37:10438–45. doi: 10.1021/bi980692s
69. Alvarado R, To J, Lund ME, Donnelly S. The immune modulatory peptide FhHDM-1 secreted by the helminth *Fasciola hepatica* prevents NLRP3 inflammasome activation by inhibiting endolysosomal acidification in macrophages. *FASEB J*. (2016) 31:85. doi: 10.1096/fj.201500093r
70. Zhang X-X, Cwiklinski K, Hu R-S, Zheng W-B, Sheng Z-A, Zhang F-K, et al. Complex and dynamic transcriptional changes allow the helminth *Fasciola gigantica* to adjust to its intermediate snail and definitive mammalian hosts. *BMC Genomics*. (2019) 20:729. doi: 10.1186/s12864-019-6103-5
71. Nian YY, Chen BK, Wang JJ, Zhong WT, Yan DC. Transcriptome analysis of *Procambarus clarkii* infected with infectious hypodermal and haematopoietic necrosis virus. *Fish Shellfish Immunol*. (2019) 98:766–72. doi: 10.1016/j.fsi.2019.11.027
72. Chu C, Parkhurst CN, Zhang W, Zhou L, Artis D. The ChAT-acetylcholine pathway promotes group 2 innate lymphoid cell responses and anti-helminth immunity. *Sci Immunol*. (2021) 6:eabe3218. doi: 10.1126/sciimmunol.abe3218
73. O'Leary CE, Feng X, Cortez VS, Locksley RM, Schneider C. Interrogating the small intestine tuft cell–ILC2 circuit using *in vivo* manipulations. *Curr Protoc*. (2021) 1:e205. doi: 10.1002/cpz1.77
74. Li S, Zhang N, Liu S, Li J, Liu L, Wang X, et al. Protective immunity against *Neospora caninum* infection induced by 14-3-3 protein in mice. *Front Vet Sci*. (2021) 8:638173. doi: 10.3389/fvets.2021.638173
75. Borghi SM, Fattori V, Carvalho TT, Tatakihara VLH, Zaninelli TH, Pinho-Ribeiro FA, et al. Experimental *Trypanosoma cruzi* infection induces pain in mice dependent on early spinal cord glial cells and NFκB activation and cytokine production. *Front Immunol*. (2021) 11:539086. doi: 10.3389/fimmu.2020.539086
76. Wu X, Thylur RP, Dayanand KK, Punnath K, Norbury CC, Gowda DC. IL-4 Treatment mitigates experimental cerebral malaria by reducing parasitemia, dampening inflammation, and lessening the cytotoxicity of T cells. *J Immunol*. (2021) 206:118–31. doi: 10.4049/jimmunol.2000779
77. Han H-J, Kim J-H. Establishment of a TLR3 homozygous knockout human induced pluripotent stem cell line using CRISPR/Cas9. *Stem Cell Res*. (2021) 52:102187. doi: 10.1016/j.scr.2021.102187
78. Rivera JEM, Hecker YP, Burucua MM, Cirone KM, Cheuquepan FA, Fiorani F, et al. Innate and humoral immune parameters at delivery in colostrum and calves from heifers experimentally infected with *Neospora caninum*. *Mol Immunol*. (2021) 132:53–9. doi: 10.1016/j.molimm.2021.01.016
79. Sun S, Jiang H, Li Q, Liu Y, Gao Q, Liu W, et al. Safety and transcriptome analysis of live attenuated *brucella* vaccine strain S2 on non-pregnant cynomolgus monkeys without abortive effect on pregnant cynomolgus monkeys. *Front Vet Sci*. (2021) 8:641022. doi: 10.3389/fvets.2021.641022
80. Ortiz Wilczyński JM, Olexen CM, Errasti AE, Schattner M, Rothlin CV, Correale J, et al. GAS6 signaling tempers Th17 development in patients with multiple sclerosis and helminth infection. *PLoS Pathog*. (2020) 16:e1009176. doi: 10.1371/journal.ppat.1009176
81. Kong D, Li X, Zhang B, Yan C, Tang R, Zhang K. The characteristics of CDC+T-helper cell subset differentiation in experimental *Clonorchis sinensis*-infected FVB mice. *Iran J Basic Med Sci*. (2020) 23:12. doi: 10.22038/ijbms.2020.39436.9350
82. Alberghina D, Piccione G, Amorini AM, D'Urso S, Longo S, Picardi M, et al. Modulation of circulating purines and pyrimidines by physical exercise in the horse. *Eur J Appl Physiol Occup Physiol*. (2011) 111:549–56. doi: 10.1007/s00421-010-1673-6
83. Harkness RA. Purine metabolism in the horse—are evolutionary differences linked to muscular performance? *Equine Vet J*. (1986) 18:5–6. doi: 10.1111/j.2042-3306.1986.tb03525.x
84. Castejón F, Trigo P, Muñoz A, Riber C. Uric acid responses to endurance racing and relationships with performance, plasma biochemistry and metabolic alterations. *Equine Vet J*. (2006) 36:70–3. doi: 10.1111/j.2042-3306.2006.tb05516.x
85. Xu F, Pang Y, Nie Q, Zhang Z, Ye C, Jiang C, et al. Development and evaluation of a simultaneous strategy for pyrimidine metabolome quantification in multiple biological samples. *Food Chem*. (2021) 373(Pt A):131405. doi: 10.1016/j.foodchem.2021.131405

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.

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.

Copyright © 2022 Hu, Tang, Wang, Qi, Ente, Li, Zhang, Li and Chu. This is an open-access 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.