



Effect of Bining decoction on gouty nephropathy: a network pharmacology analysis and preliminary validation of gut microbiota in a mouse model

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Background: To use network pharmacology and gut microbiota sequencing to investigate the probable mechanism of Bining decoction (BN) in the treatment of gouty nephropathy (GN).

Methods: Firstly, the mechanism of therapeutic effects of BN on GN were collected by integrating network pharmacology. Secondly, the treatment effects of BN against GN in 30 Institute of Cancer Research (ICR) mice were evaluated by performing biochemical tests [uric acid, blood urea nitrogen, and creatinine (UA, BUN, and Cr)] and evaluating the renal weight index. Finally, 16S rRNA sequencing was utilized for elucidating the therapeutical effect of BN in GN.

Results: The results of gut microbiota sequencing analysis showed the abundance of *Faecalibaculum*, *Romboutsia*, *Bifidobacterium*, *Bacteroides*, *Odoribacter*, *Lachnospiraceae* NK4A136 group, unclassified_f__*Lachnospiraceae*, *Roseburia*, norank_f__*Lachnospiraceae*, *Lactobacillus*, *Dubosiella*, norank_f__*Muribaculaceae*, and *Turicibacter* in the BN group had a significant changed between-group comparisons. Using a network pharmacology-related database, 413 active components of BN were identified, as well as 1,085 GN-associated targets. The 118 targets of disease targets and component targets were mapped, of which the top 10 genes were selected. The Kyoto Encyclopedia of Genes and Genomes (KEGG) results showed that 157 pathways were enriched, which was partially consistent with the metabolic pathways of gut microbiota sequencing analysis.

Conclusions: Combining 16S rRNA gene sequencing and network pharmacology analysis, similar signaling pathways were followed: “Pathways in cancer” and “Adipocytokine signaling pathway”. The results reveal that BN increases the abundance of *Turicibacter*, regulates the expression of JAK2 in the JAK/STAT pathway, increases the beneficial bacteria *Turicibacter* associated with intestinal butyric acid, which could enhance the intestinal barrier, and exert anti-inflammatory effects.

Keywords: Gut microbiota; network pharmacology; gouty nephropathy (GN); 16S rRNA gene; traditional Chinese medicine formulation (TCM formulation)

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Introduction

Gouty nephropathy (GN) is a kind of kidney disorder that results from an elevation in uric acid (UA) level in the serum which has long been accumulated in the renal tubules and interstitium (1), causing a reaction in the surrounding macrophages (2). The representative histological features of UA nephropathy are characterized by urate deposits in the renal tubules and interstitium, which can be observed as birefringent, needle-like urate crystals (1). Disease progression may even lead to renal failure. Most patients experience lumbago, nocturia, or waist soreness, and some patients may develop hematuria, edema, hypertension, and other symptoms such as fever. Currently, GN is regarded as an essential public health issue. A recent study estimated that 24% of gout patients exhibit chronic kidney disease (3). According to the statistics, 12–18% of people in southern China experience UA stones (4). The development of modern society has contributed to GN gradually becoming one of the serious threats to human health.

Colchicine, non-steroidal anti-inflammatory drugs (NSAIDs), and non-gut/oral glucocorticoids are recommended as first-line treatments for gout based on the American College of Rheumatology (ACR) guidelines issued in 2020 (5). Nevertheless, no other treatment options are currently available for some clinical patients. NSAIDs are widely used for the treatment of gout attacks, but are not suitable for elderly patients with renal injury and many

comorbidities. Meanwhile, colchicine has been prohibited for use in the treatment of gout attacks in patients with renal damage (6).

Therefore, there is a growing need for safe and effective alternatives to substitute drug therapy, and investigational drugs with a wider application range, better efficacy, and lower toxicity, especially those derived from natural products. Herbal medicine is a valuable resource in the exploration of novel drugs to treat various diseases, including GN. Increasing numbers of traditional Chinese medicines (TCM) have been reported to have a protective effect on the kidneys; therefore, mounting attention has been paid to TCM due to their satisfactory and mild activity in treating nephropathy. A study has shown that TCM formulations can influence disease by mediating the homeostasis of intestinal flora (7).

The mechanism of GN is primarily associated with hyperuricemia (HUA) and the deposition of monosodium urate crystals in different parts of the body. Excessive UA deposits place a burden on the kidneys, eventually causing lesions. In the human body, approximately 80% of the total amount of UA is decomposed from nucleic acids, and the rest is generated by purine-rich foods. The human body produces and excretes 600–700 mg of UA per day, 70% of which is excreted by the kidneys, and the remaining 30% is eliminated through the intestines (8). An elevated UA level in blood circulation affects the intestinal environment, resulting in alternations of the gut microbiota. Additional UA will be eliminated by the intestine as compensation for the impairment of the kidney (9). Therefore, the gut microbiota has become a novel target to elucidate the GN pathogenesis and for the further treatment of GN.

Evidence has shown that over 1,000 genera of bacteria can colonize the human gut and form the gut microbiota (10). These bacteria are involved in host metabolism, immunoregulation, as well as maintenance of internal environment homeostasis (11). High-throughput sequencing has achieved tremendous advances in our understanding of the gut microbiome. As previously described, gut microbiota can affect UA levels. Current research has shown that the intestinal flora reduces UA levels due to its participation in purine metabolism and decomposition of UA. For example, a report suggested that the gut microbiota can secrete the oxidative metabolism-associated enzyme xanthine oxidase (12).

Intestinal flora-generated metabolites promote UA excretion by providing energy for intestinal epithelial

Highlight box

Key findings

- Combining *16S rRNA* gene sequencing and network pharmacology analysis, we speculated that Bining decoction (BN) increases the abundance of *Turicibacter*, regulates the expression of JAK2 in the JAK/STAT pathway, increases the beneficial bacteria associated with intestinal butyric acid, which could enhance the intestinal barrier, and exert anti-inflammatory effects.

What is known and what is new?

- Traditional Chinese Medicine (TCM) formulation formulations can influence disease by mediating the homeostasis of intestinal flora.
- After Intervention of Bining Decoction, BN could increase the richness and diversity of the intestinal flora. BN did play a certain role in gouty nephropathy (GN) structure.

What is the implication, and what should change now?

- After more in-depth research, BN could be recommended for GN treatment in the future.

cells (11). Meanwhile, the gut microbiota also produces short-chain fatty acids (SCFA) by decomposing indigestible carbohydrates ingested by humans (13), which play a part in maintaining homeostasis of the human gastrointestinal tract.

Bining decoction (BN) is a TCM formulation that is derived from the Er Miao San prescription, the action of which is described as dissipating heat and dampness, as well as eliminating edema in the State Pharmacopoeia of People's Republic of China, Danxi's Experiences in Medicine, and other TCM literature. Clinically, BN is used in the treatment of GN; it provides a breakthrough in the prevention and treatment of GN and improves the quality of life of patients. The ingredients of BN include *Dioscoreae Hypoglaucae Rhizoma*, *Plantaginis Semen*, *Atractylodes Lancea* (Thunb.) DC., *Coicis Semen*, *Pseudobulbus Cremastrae Seu Pleiones*, *Phellodendri Chinensis Cortex*, *Chuanxiong Rhizoma*, *Cyathulae Radix*, and *Lonicerae Japonicae Caulis* and *Atractylodes Lancea* (Thunb.) DC. is the principal drug in the decoction, and various active ingredients have been found to act on tumor necrosis factor- α (TNF- α), interleukin-6 (IL-6), IL-1 β , transcription factor p65 (RELA), prostaglandin-endoperoxide synthase 2 (PTGS2), as well as recombinant mitogen-activated protein kinase 14 (MAPK14), which has a vital effect on modulating inflammatory cytokines and apoptosis-associated pathways for the improvement of the internal imbalance associated with the disease (14). Meanwhile, it also has the function of removing turbidity and could remove the dampness and heat from urine, supporting the modern pharmacological discovery of reducing UA levels, and promoting UA excretion. As an adjuvant drug, *Phellodendri Chinensis Cortex* helps the principal drug to dispel dampness and heat, but also works to resolve phlegm, indicating its protective effect on the kidneys. *Chuanxiong Rhizoma* is considered an adjuvant to promote blood circulation and remove dampness; *Cyathulae Radix* is regarded as an agent to catalyze the efficacy of the drug. The above drugs are used together to remove pathogenic factors, balance the healthy qi, and simultaneously treat the symptoms and the root causes, thereby jointly removing turbidity and protecting the kidneys. This also demonstrates the treatment based on syndrome differentiation of BN, reflecting the advancement of ancient TCM theory. Nevertheless, how BN mediates GN has not been elucidated thoroughly.

Network pharmacology is a conduit for explaining the complex interactions among biological systems, disease, and drugs. It can identify synergistic functions in disease

treatment by analyzing huge datasets and explicating the possible complex bioactivity processes, which may accelerate clinical translation (15). The primary outcome of this study is the mechanism of therapeutic effects of BN on GN which were collected by integrating network pharmacology. The secondary outcomes constitute the treatment effects of BN against GN in mice, which were evaluated by performing biochemical tests [UA, blood urea nitrogen, and creatinine (BUN and Cr)] and evaluating the renal weight index. The last outcome of this study is the signaling pathway prediction of BN in GN by using 16S rRNA sequencing. We present the following article in accordance with the ARRIVE reporting checklist (available at <https://atm.amegroups.com/article/view/10.21037/atm-22-5523/rc>).

Methods

Collection of BN and GN targets

The Traditional Chinese Medicine Systems Pharmacology Database and Analysis Platform (TCMSP; <https://old.tcm-sp-e.com>) was used to seek the useful active ingredients of BN (16). The keywords were as follows: "*Dioscoreae Hypoglaucae Rhizoma*", "*Plantaginis Semen*", "*Atractylodes Lancea* (Thunb.) DC.", "*Coicis Semen*", "*Pseudobulbus Cremastrae Seu Pleiones*", "*Phellodendri Chinensis Cortex*", "*Chuanxiong Rhizoma*", "*Cyathulae Radix*", and "*Lonicerae Japonicae Caulis*". The screening conditions were "oral bioavailability (OB) $\geq 30\%$, and drug-like (DL) index ≥ 0.18 ". OB refers to the speed and extent at which the drug enters the systemic circulation after oral absorption (17). DL is applied to reflect the similarities of specific groups in compounds with known drug degrees. OB and DL play essential roles in the exploration of the TCM active ingredient. The human genes related to GN were screened from the databases of GeneCards (www.genecards.org/), Online Mendelian Inheritance in Man (www.omim.org/), Genbank (<http://www.ncbi.nlm.nih.gov/>), and DisGeNET (<https://www.disgenet.org/home/>) using the keywords of "gouty nephropathy", "uric acid nephropathy", and "gouty kidney" to search for disease targets. The study was conducted in accordance with the Declaration of Helsinki (as revised in 2013). The UniProt database (www.uniprot.org/) was implemented to convert all the aforesaid target proteins into standardized gene names.

Protein-protein interaction (PPI) network construction

The targets of BN for the GN therapy, identified by mapping

the active compound targets to the GN-associated targets, were input into the database of the Search Tool for the Retrieval of Interacting Genes/Proteins (STRING; <https://string-db.org>) to build a PPI network, which was subsequently analyzed with the organism species restricted to *Homo sapiens* with a high confidence score more than 0.9. Subsequently, the obtained PPI data were input into Cytoscape 3.7.2 (<https://cytoscape.org/>), and the top 10 genes were selected by cytoHubba plugin in the Cytoscape (18).

Analysis of signaling pathway enrichment using the Kyoto Encyclopedia of Genes and Genomes (KEGG) database

Next, a signaling pathway enrichment analysis was conducted with the Metascape database (<https://metascape.org/gp/index.html>) via the KEGG database. Significantly, species was restricted to *Homo sapiens*. The diagrams for the signaling pathway enrichment were plotted by R 3.5.2. (The R Project for Statistical Computing, Vienna, Austria).

Animals

In the study, n refers to number of animals and the individual mouse was considered the experimental unit within the studies. The sample size calculation was based on a similar experiment reported in the literature and a preliminary experiment conducted under identical conditions to the planned experiment (19). A total of 30 male Institute of Cancer Research (ICR) mice (18±2 g), were provided by the Heilongjiang University of Chinese Medicine (Harbin, China). Animals were kept in polyacrylic cages (temperature: 22±2 °C, humidity: 50%±5%, 12-h light/dark cycle, and sufficient water and food). Animal experiments were performed under the approval by the Animal Ethics Committee of Heilongjiang University of Chinese Medicine (No. 202009670), in compliance with the Heilongjiang University of Chinese Medicine guidelines for the care and use of animals. A protocol was prepared before the study without registration. For each animal, 3 different investigators were involved as follows: a first investigator (HH) was responsible for animal grouping work. This investigator was the only person aware of the group allocation. A second investigator (YY) was responsible for the intragastrical gavage, whereas a third investigator (YG) (also unaware of treatment) was responsible for data collection and assessment. Based on the 3R principle, the experiment was designed to minimize the use of animals, observe the physiological conditions

of experimental animals, and standardize the gavage by the same investigator, thus minimizing animal pain and unnecessary death. At the end of the experiment, the mice were sacrificed by decapitation, and then the animal carcasses were sent to the specified places according to the requirements, which were uniformly processed by the institution.

Chemical reagents

Dioscoreae Hypoglaucae Rhizoma, *Plantaginis Semen*, *Atractylodes Lancea* (Thunb.) DC., *Coicis Semen*, *Pseudobulbus Cremastrae Seu Pleiones*, *Phellodendri Chinensis Cortex*, *Chuanxiong Rhizoma*, *Cyathulae Radix*, and *Lonicerae Japonicae Caulis* were purchased from Heilongjiang Deshunchang Decoction Pieces Co., Ltd. (Harbin, China). The GN mouse model was prepared by combining yeast and adenine (Shanghai Aladdin Biochemical Technology Co., Ltd., Shanghai, China). Other reagents were from domestic reagent companies.

Preparation of BN extract

The decoction was prepared by soaking all botanical drugs [*Dioscoreae Hypoglaucae Rhizoma*, *Plantaginis Semen*, *Atractylodes Lancea* (Thunb.) DC., *Coicis Semen*, *Pseudobulbus Cremastrae Seu Pleiones*, *Phellodendri Chinensis Cortex*, *Chuanxiong Rhizoma*, *Cyathulae Radix*, and *Lonicerae Japonicae Caulis*] for 30 minutes in water and heating to 100 °C. Next, the formulation was decocted 2 times, and upon combination, the 2 decoctions obtained were subjected to condensing and refluxing; the compound extract (15.1 g) was prepared as a concentrated liquid by using a freeze dryer (VirTis BenchTop Pro, VirTis Co., Ltd., Gardiner, NY, USA).

Animal experiments

The intragastric gavage (IG) suspension was prepared with adenine and 0.5% sodium carboxymethyl cellulose (CMC-Na) solution. Meanwhile, dry yeast powder was added to the regular feed of mice, with the daily intake of yeast being controlled to 10 g/kg (16). A total of 30 mice were assigned into 3 groups in a random fashion (n=10): the blank group, the model group, and the BN group. Random numbers were generated using the random number generator in Microsoft Excel (Microsoft Corp., Redmond, WA, USA). Mice in the blank group were given 0.5% CMC-Na

solution at 100 mg/kg, and those in the other 2 groups were administered with a combined 0.5% CMC-Na (100 mg/kg) and adenine (100 mg/kg) by IG once a day at 8:00 a.m. for 28 consecutive days. The research team monitored animals twice daily. Mouse health was monitored by weight, food and water intake, and general assessment of animal activity, panting, and fur condition. When the modeling began, mice in the BN group were separately administered with BN decoction every afternoon. The IG order of each group was randomized. With the prolongation of the modelling time, the mice in the model group showed signs of weight loss and lethargy. The administration group mice were lighter than the model group, but not as healthy as the control group.

Sample collection

Significant increases of UA, BUN, and Cr indicated that the animal model had been established successfully. The animals were excluded if insertion of the IG needle resulted in perforation of the vessel wall, or if the animal died prematurely, preventing the collection of data. A total of 30 mice were therefore included in the analysis in this study. All of the 30 mice were fasted for 12 hours before the last administration. On the 28th day, the blood from all mice was collected for the analysis of UA, BUN, and Cr. Then, the renal tissues were removed, followed by decapitation and weighing. Next, serum was isolated by centrifugation for 10 minutes at 3,000/min at 4 °C, and the serum levels of BUN, Cr, and UA were tested using an automatic biochemical analyzer. In addition, fresh stool samples were immediately harvested in frozen pipes (20). All biological samples were kept at -80 °C until testing.

Renal weight index

On the 28th day, the 30 mice were euthanized quickly, and the kidney index was calculated as double kidney weight/mouse weight × 100%.

16S rRNA gene sequencing

A total of 3 mice were randomly selected from each group. The 16S rRNA gene sequencing was conducted to elucidate whether BN has an effect on GN in mice. Total bacterial DNA was harvested, amplified, and then sequenced on the MiSeq™ platform (Illumina, San Diego, CA, USA) by Majorbio (Shanghai, China). Bioinformatics analysis

was implemented based upon operational taxonomic units (OTUs), and these units were clustered based on the 100% sequence similarity according to Uparse11 (<http://www.drive5.com/uparse/>). For the α -diversity analysis, Mothur1.30.2 (https://www.mothur.org/wiki/Download_mothur) was utilized for the calculation of the Chao1 Index, Shannon Index, Observed Species Index, Good's Coverage Index, and Simpson's Diversity Index. The non-metric multidimensional scaling (NMDS) and principal coordinate analysis (PCoA) were employed for β -diversity analysis by the unweighted UniFrac method. All processes were performed on MajorbioCloud (<https://cloud.majorbio.com/>). PICRUSt2 (<http://picrust.github.io/>) was utilized for predicting the functionally-enriched signaling pathways in gut metabolism (21).

Statistical analysis

SPSS 21.0 was used for statistical analysis. The measurement data were expressed as mean ± standard deviation (SD). A One-way analysis of variance (ANOVA) was used for comparison between groups, and SNK method was further used for pairwise comparison, and P<0.05 was considered as statistically significant.

Results

Screened Targets and PPI Networks Associating with BN and GN

The databases yielded 413 pharmacological targets associated with BN upon the removal of the duplicate targets. A total of 1,085 GN-associated targets were available from all the databases, which were plotted to 118 active component targets. As shown in *Figure 1A*, the targets of BN for GN therapy were analyzed by the Venny 2.1 (<https://bioinfogp.cnb.csic.es/tools/venny/>), and the STRING database was employed for the visualization of the PPI networks of BN. In the PPI network, more significant interactions were observed by showing with 416 edges and 90 nodes (*Figure 1B*). Additionally, the core targets were screened out using the cytoHubba plugin Cytoscape, and the top 10 genes, including mitogen-activated protein kinase (MAPK)1, proto-oncogene tyrosine-protein kinase Src (SRC), heat shock protein HSP 90-alpha (HSP90AA1), tyrosine-protein phosphatase non-receptor type (PTPN)11, transforming protein RhoA (RHOA), MAPK14, estrogen receptor (ESR)1, tyrosine-

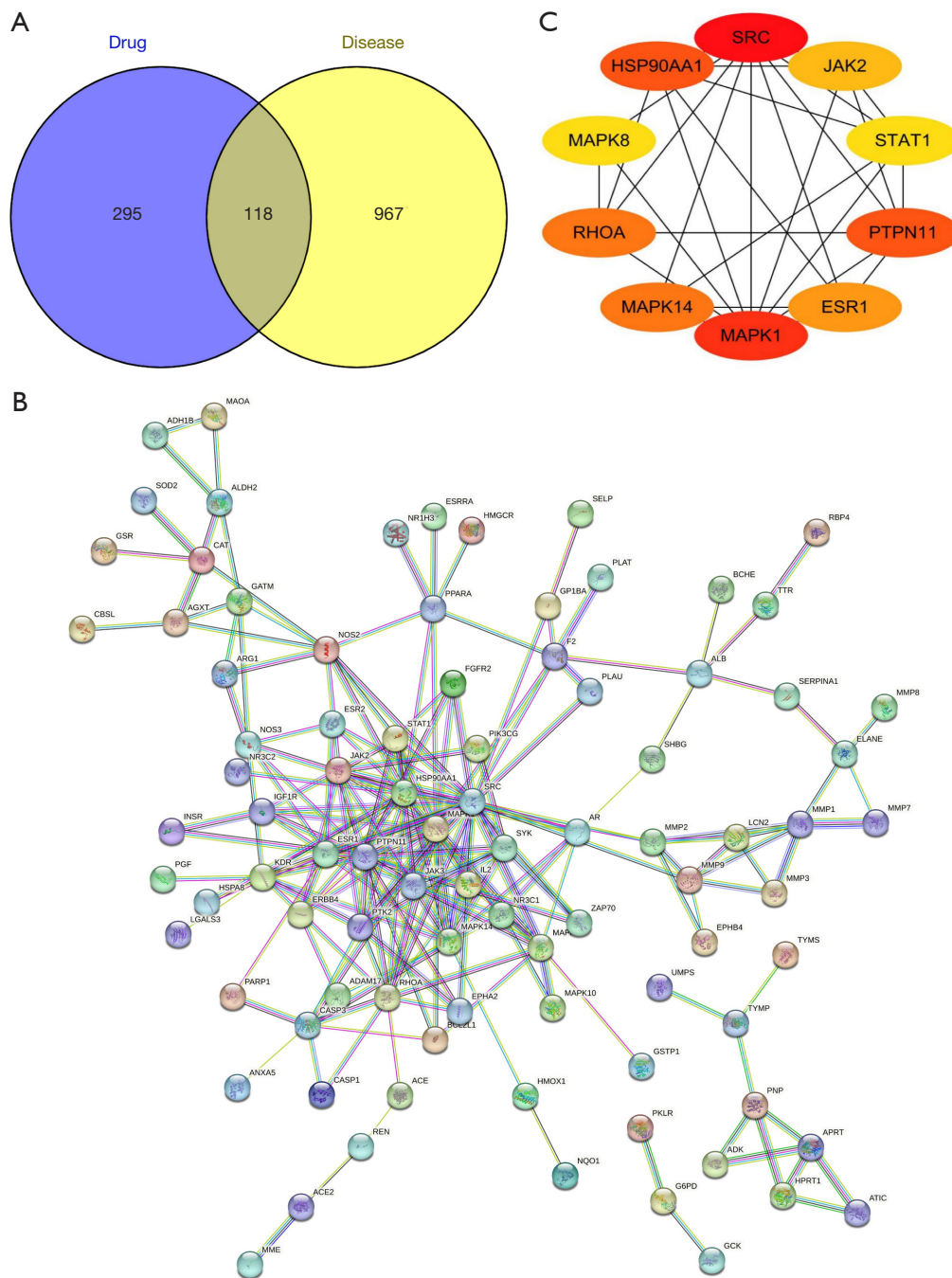


Figure 1 Features of the BN targets in GN therapy were ascertained using network pharmacology. (A) The common targets of BN in GN therapy are analyzed by Venn diagram. (B) The PPI network is employed according to the common targets of BN. The nodes refer to different proteins, the node size corresponds to the association strength, and the edges indicate the relation between proteins. (C) The screened top 10 core genes. BN, Bining decoction; GN, gouty nephropathy; PPI, protein-protein interaction.

protein kinase JAK (JAK)2, signal transducer and activator of transcription 1-alpha/beta (STAT)1, and MAPK8 (Figure 1C), were considered as core genes after combining

with the calculation method. These genes may be key targets through which BN plays a role in lowering the UA concentrations (22).

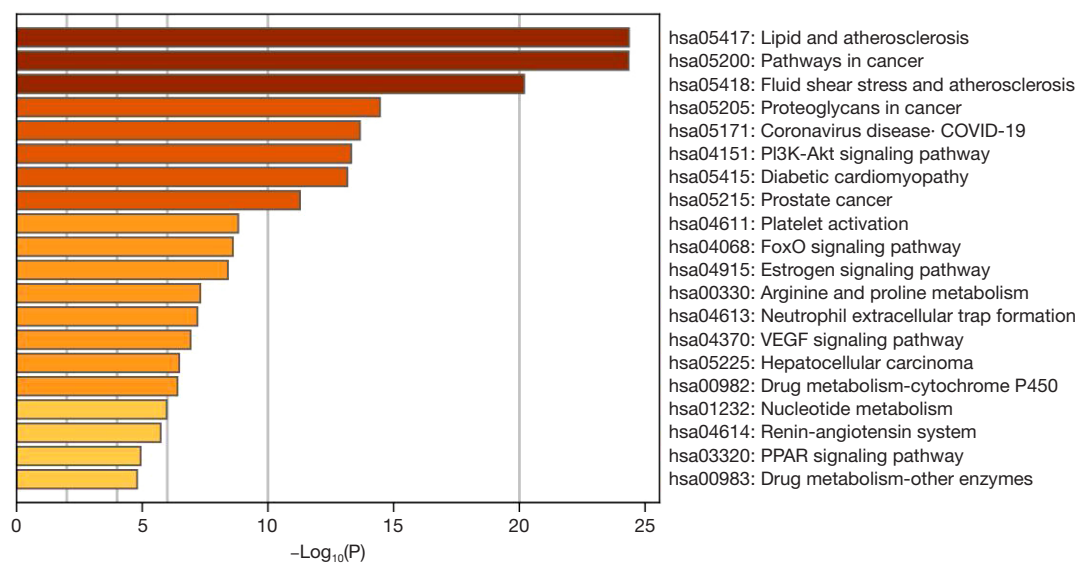


Figure 2 The pathway enrichment analysis is conducted using the KEGG database. KEGG, Kyoto Encyclopedia of Genes and Genomes.

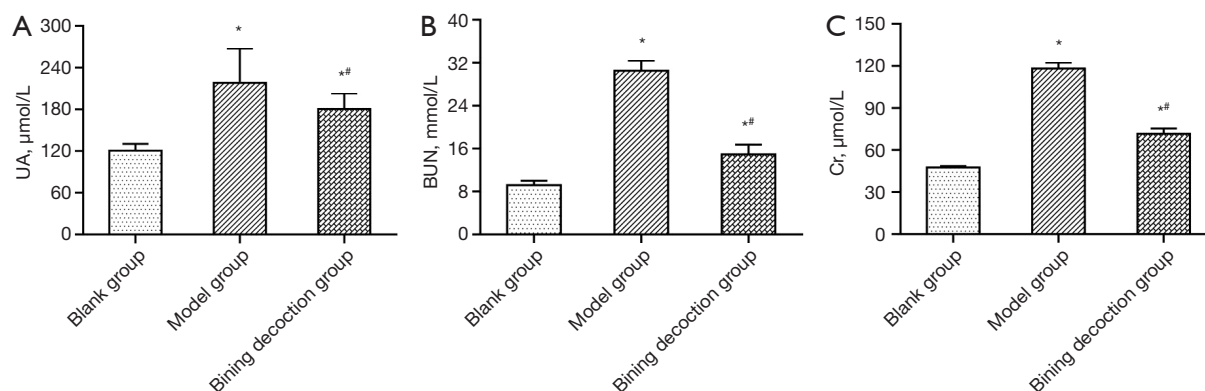


Figure 3 Comparisons of serum levels of UA (A), BUN (B), and Cr (C) in mice. n=10; *, P<0.05 vs. the blank group; #, P<0.05 vs. the model group. UA, uric acid; BUN, blood urea nitrogen; Cr, creatinine.

Signaling pathway enrichment using core targets

Subsequently, we performed the pathway enrichment analysis of the aforesaid 10 core genes using the KEGG database (Min Overlap: 3, P value cutoff: 0.01). Finally, 157 metabolic pathways were identified, and the top 20 pathways with the highest enrichment are exhibited in *Figure 2*.

Effect on serum biochemical indexes

As displayed in *Figure 3A-3C*, we observed that the plasma levels of UA, BUN, and Cr increased significantly. After

modeling, mice were euthanized to obtain the kidneys, followed by the calculation of the renal weight index (weight of kidney/ICR mouse weight, mg/g). The comparative findings of the serum levels of UA, BUN, and Cr in each group are displayed in *Table 1*. According to the results of the one-way ANOVA, we noticed that the serum levels of UA, BUN, and Cr in each group were statistically significant (all P<0.05). Further, the Student-Neuman-Keuls (SNK) method was used to compare the levels of UA, BUN, and Cr in each group, which indicated increased serum levels of UA, BUN, and Cr in the model and the BN groups in comparison to those in the blank group, and the levels of which were lower in the BN group than those in

Table 1 Biological index for model establishment

Group	n	UA ($\mu\text{mol/L}$)	BUN (mmol/L)	Cr ($\mu\text{mol/L}$)
Blank group	10	120.61 \pm 9.76	9.17 \pm 0.85	47.56 \pm 0.99
Model group	10	218.03 \pm 49.67*	30.50 \pm 1.92*	118.24 \pm 4.19*
Bining decoction group	10	180.56 \pm 22.63*#	14.91 \pm 1.85*#	71.64 \pm 3.79*#
F	–	23.564	466.359	1177.491
P	–	<0.001	<0.001	<0.001

*, compared with Blank group; #, compared with Model group, all $P < 0.05$. UA, uric acid; BUN, blood urea nitrogen; Cr, creatinine.

Table 2 The comparative results of the renal indexes of mice in each group

Group	n	Renal weight index (%)
Blank group	10	1.12 \pm 0.18
Model group	10	2.76 \pm 0.52*
Bining decoction group	10	1.65 \pm 0.26*#
F	–	23.564
P	–	<0.001

*, compared with Blank group; #, compared with Model group, all $P < 0.05$.

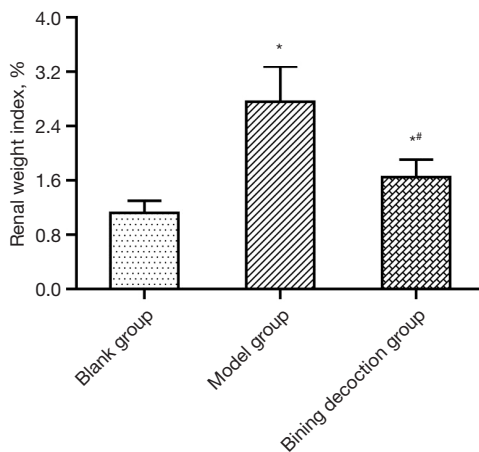


Figure 4 Comparison of the renal weight index of mice in each group. $n=10$; *, $P < 0.05$ vs. the blank group; #, $P < 0.05$ vs. the model group.

the model group (all $P < 0.05$).

Effect on the renal weight index

Table 2 shows the comparative results of the renal indexes of

mice in each group. A significant difference was witnessed in the renal weight index in each group based on the results of one-way ANOVA ($P < 0.05$). The results of the SNK method further demonstrated that the renal weight index of mice in the model group and the BN group was elevated when compared to that in the blank group, and the renal weight index was lower in the BN group than that in the model group ($P < 0.05$; Figure 4). It is suggested that BN could decrease the kidney weight index of GN mice.

16S rRNA gene sequencing

A growing number of articles have indicated that gut microbiota is regarded as a novel target for personalized drug therapy by impacting host metabolism and maintaining intestinal environment homeostasis (23).

Upon the completion of quality-filtering steps, denoising, and chimera removal, in the case of 100% of the sequence similarity threshold, 316,454 high-quality sequences were gained from 9 samples for further analyses of OTUs of gut microbiota. In the blank group, there were 109 OTUs and 0 unique OTUs; in the model group, 117 OTUs and 0 unique OTUs; in the BN group, 116 OTUs and 1 unique OTUs. These 3 groups possessed 105 OTUs (Figure 5A). The number of OTUs was increased in the model group in contrast to the blank group; the number of OTUs was downregulated in the BN group in comparison to the model group. The slopes of the rank abundance reduced smoothly and flattened eventually, and the wide span of the horizontal coordinates unveiled the rich and uniform species samples (Figure 5B).

Alpha diversity analysis of gut microbiota

The α -diversity analysis displayed in Table 3 shows an elevation in the Chao Index, Ace Index, Shannon Diversity

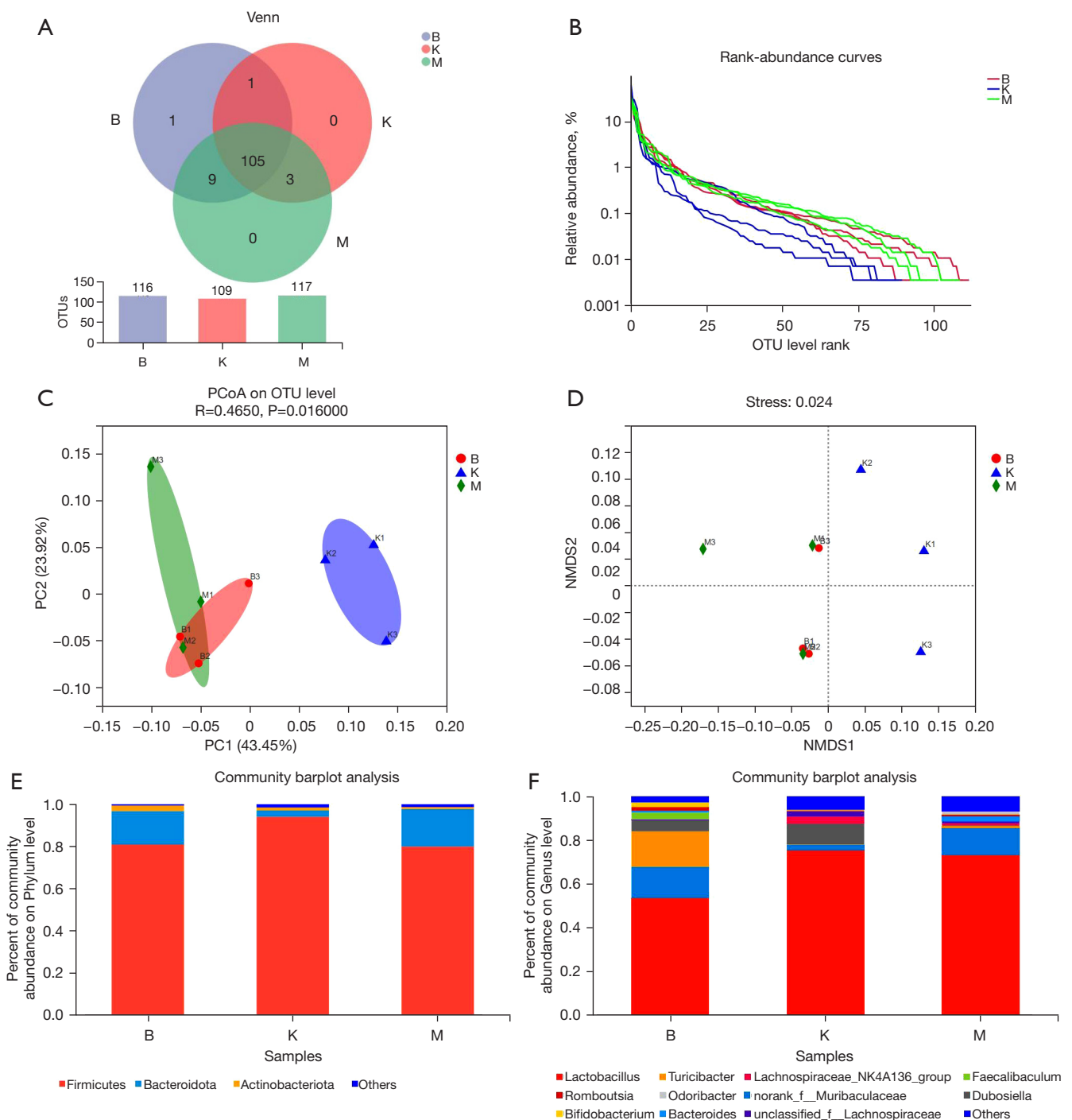


Figure 5 Effect of BN treatment on the intestinal microflora of mice suffering from gouty nephropathy. (A) The OTUs of gut microbiota in each group by Venn diagram analysis. (B) Rank-abundance curve; PCoA (C) and NMDS (D) are performed based on the ASVs for the β -diversity analysis. Bacteria with obvious differences in abundance at varying taxonomy levels. (E,F) The abundance of microbial species at the genus level and phylum level. B, the BN group; K, the blank group; M, the model group; PCoA, principal co-ordinates analysis; OTU, operational taxonomic unit; PC, principal co-ordinates; NMDS, non-metric multidimensional scaling; BN, Bining decoction; ASVs, actual sequence variants.

Table 3 The α -diversity analysis of gut microbiota in each group

Group	Ace	Chao	Shannon	Simpson	Good's coverage	Pd	Sobs
Blank group	97.080453	96.938492	1.952064667	0.286234667	0.999592667	10.28976	89
Model group	106.942599	109.75	2.694780667	0.149178667	0.999760333	11.10108	103
Bining decoction group	113.5857397	116.7333333	2.721098333	0.142312333	0.999688667	11.38382667	107.6666667

Pd, phylogenetic diversity; Sobs, the number of observed OTUs.

Index, Good's Coverage Index, Pd Index, and Sobs Index in the model group when compared with the blank group. This suggests that BN decoction could increase the richness and diversity of the gut microflora. In comparison to the model group, the Chao Index, Ace Index, Shannon Diversity Index, Pd Index, and Sobs Indexes in the BN group were almost increased, which demonstrated that BN could increase the richness and diversity of the intestinal flora. More than 98% of the Good's Coverage Index suggests high sample coverage and reliable sequencing findings.

Beta diversity analysis of gut microbiota

For the β -diversity analysis, PCoA and NMDS were performed based on the actual sequence variants (ASVs) (Figure 5C,5D). The findings revealed that the microflora of the mice was significantly separated in the model group versus the blank group, indicating a significant difference in the β -diversity between the 2 groups. The BN group exhibited the same tendency as that of the blank group, which disclosed that BN failed to reverse the GN structural alteration. However, BN did play a certain role in GN structure.

Taxonomic composition analysis of gut microbiota

Through the taxonomy analysis displayed in Figure 5E,5F, it can be observed that at the phylum level, *Firmicutes* and *Bacteroidota* were the most essential phyla, followed by *Actinobacteriota* and others. These 5 phyla belong to the dominant phyla and occupy a very high portion of the whole phyla. Gut microbiota sequencing analysis showed the abundance of *Faecalibaculum*, *Romboutsia*, *Bifidobacterium*, *Bacteroides*, *Odoribacter*, *Lachnospiraceae*_NK4A136_group, unclassified_f__*Lachnospiraceae*, *Roseburia*, norank_f__*Lachnospiraceae*, *Lactobacillus*, *Dubosiella*, norank_f__*Muribaculaceae*, and *Turicibacter* in the BN group had changed significantly.

As shown in Figure 6A-6C, *Firmicutes* (94.03–79.92%), *Bacteroidota* (3.01–17.79%), and *Actinobacteriota* (1.40–0.94%), The ratio of *Firmicutes/Bacteroidetes* in the blank group and the model group was 0.31 and 0.04, respectively, implying that the dominant flora structure of mice in the model and blank groups has changed significantly; the *Firmicutes/Bacteroidetes* ratio in the BN group was $P < 0.05$ compared with the model group, implying that BN could modulate the mouse' dominant flora.

From the subordinate level, by performing a cluster analysis of the absolutely-dominant bacterial genera in the top 20 positions, we discovered and analyzed the abundance of various bacterial genera, namely, *Faecalibaculum*, *Romboutsia*, *Bifidobacterium*, *Candidatus_Saccharimonas*, *Alloprevotella*, *Parabacteroides*, *Bacteroides*, *Odoribacter*, *Lachnospiraceae*_NK4A136_group, unclassified_f__*Lachnospiraceae*, *Eubacterium_xylanophilum_group*, *Helicobacter*, *Desulfovibrio*, *Roseburia*, norank_f__*Lachnospiraceae*, *Enterorhabdus*, *Lactobacillus*, *Dubosiella*, norank_f__*Muribaculaceae*, and *Turicibacter* (Figure 6D).

Metabolic pathway prediction in microbiota sample communities

In this study, to identify the metabolic pathway of BN for treating GN, the treatment group was analyzed by using picurst2, and 272 signaling pathways were identified by enrichment analysis (Table 4).

Integrating 16S rRNA gene sequencing and network pharmacology analysis

Combining 16S rRNA gene sequencing and network pharmacology analysis, similar signaling pathways which were followed included "Pathways in cancer" and "Adipocytokine signaling pathway".

According to the core genes from both the signaling pathways and network pharmacology, we plotted an integrated network map of BN decoction for GN therapy.

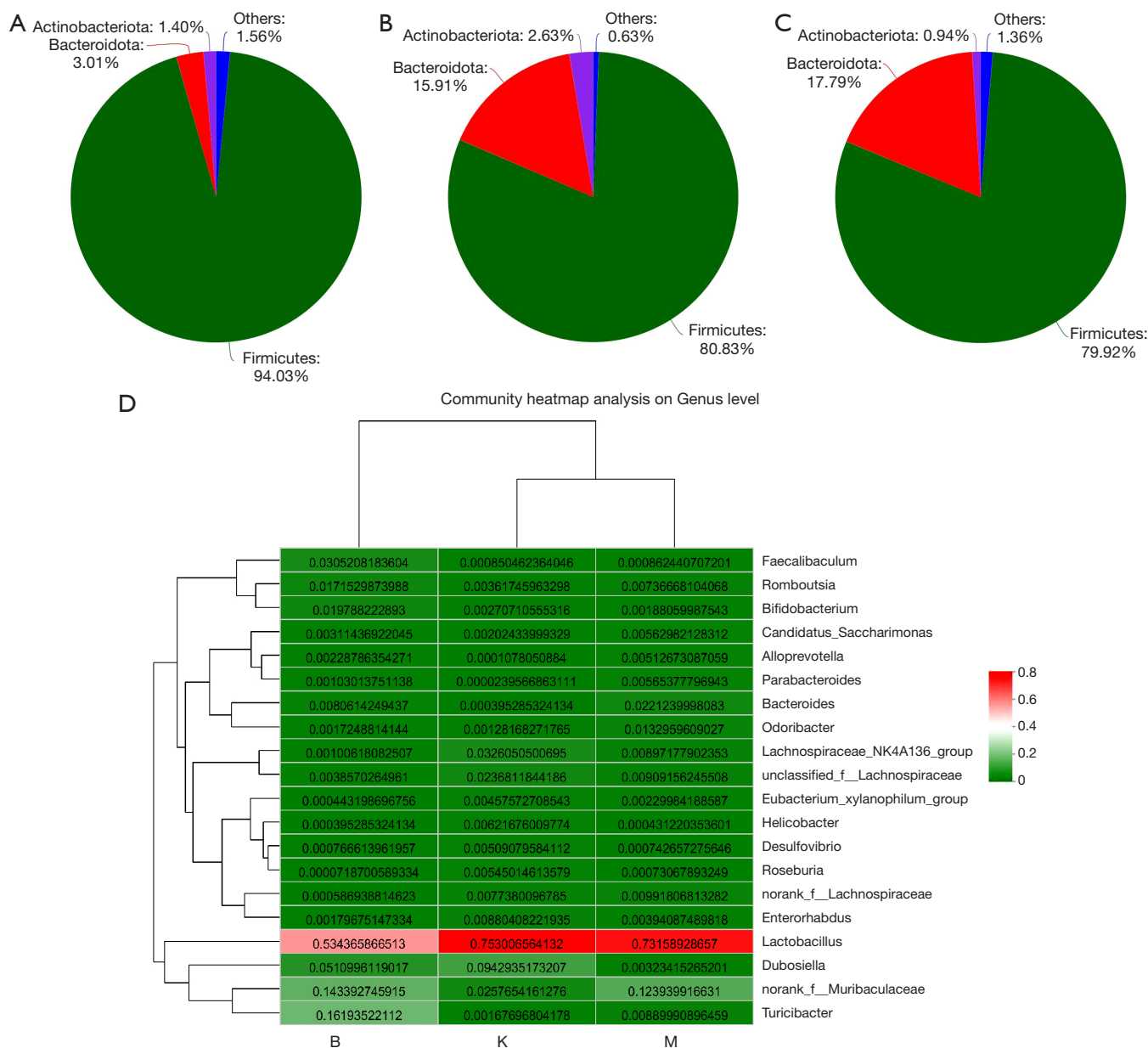


Figure 6 Community analysis pieplot on phylum level in the blank group (A), the BN group (B), and the model group (C). (D) The top 20 positions of the dominant bacterial genera. B, the BN group; K, the blank group; M, the model group; BN, Bining decoction.

As demonstrated in *Figure 7*, the “Pathways in cancer” had the most significant node, which exhibited the strongest association with JAK2.

Discussion

GN is a frequent complication of HUA that is caused by the deposition of urate crystals in the kidneys, and its

subsequent inflammatory reaction impairs renal function. In this study, ICR mice were given a larger number of urate precursors to accumulate UA in the kidney and damage the nephron for the model establishment. Similar to a previous study, this model was also induced by adenine and yeast powder (24). A higher BUN level is considered one of the most important biochemical indexes for high UA or injury to the kidney (25). Currently, the model is easy to operate

Table 4 Prediction of metabolic signaling pathways in microbiota sample communities

Pathway	Description
ko01100	Metabolic pathways
ko01110	Biosynthesis of secondary metabolites
ko01120	Microbial metabolism in diverse environments
ko01230	Biosynthesis of amino acids
ko03010	Ribosome
ko02010	ABC transporters
ko01200	Carbon metabolism
ko00230	Purine metabolism
ko00010	Glycolysis/gluconeogenesis
ko00500	Starch and sucrose metabolism
ko02020	Two-component system
ko00520	Amino sugar and nucleotide sugar metabolism
ko02024	Quorum sensing
ko00240	Pyrimidine metabolism
ko00970	Aminoacyl-tRNA biosynthesis
ko00620	Pyruvate metabolism
ko03440	Homologous recombination
ko02060	PTS
ko00052	Galactose metabolism
ko00550	Peptidoglycan biosynthesis
ko00270	Cysteine and methionine metabolism
ko03430	Mismatch repair
ko00250	Alanine, aspartate and glutamate metabolism
ko00260	Glycine, serine and threonine metabolism
ko00030	Pentose phosphate pathway
ko00680	Methane metabolism
ko00051	Fructose and mannose metabolism
ko03030	DNA replication
ko00190	Oxidative phosphorylation
ko00720	Carbon fixation pathways in prokaryotes
ko00300	Lysine biosynthesis
ko03060	Protein export
ko00640	Propanoate metabolism
ko03018	RNA degradation

Table 4 (continued)**Table 4** (continued)

Pathway	Description
ko00561	Glycerolipid metabolism
ko00900	Terpenoid backbone biosynthesis
ko00564	Glycerophospholipid metabolism
ko04112	Cell cycle-caulobacter
ko03070	Bacterial secretion system
ko00710	Carbon fixation in photosynthetic organisms
ko00650	Butanoate metabolism
ko03410	Base excision repair
ko01501	Beta-lactam resistance
ko00670	One carbon pool by folate
ko05230	Central carbon metabolism in cancer
ko01212	Fatty acid metabolism
ko00061	Fatty acid biosynthesis
ko03420	Nucleotide excision repair
ko00730	Thiamine metabolism
ko04922	Glucagon signaling pathway
ko00630	Glyoxylate and dicarboxylate metabolism
ko04066	HIF-1 signaling pathway
ko00195	Photosynthesis
ko00770	Pantothenate and CoA biosynthesis
ko00760	Nicotinate and nicotinamide metabolism
ko00983	Drug metabolism-other enzymes
ko00020	Citrate cycle (TCA cycle)
ko00220	Arginine biosynthesis
ko00450	Selenocompound metabolism
ko00521	Streptomycin biosynthesis
ko01210	2-oxocarboxylic acid metabolism
ko01502	Vancomycin resistance
ko01503	CAMP resistance
ko00480	Glutathione metabolism
ko00790	Folate biosynthesis
ko00780	Biotin metabolism
ko03020	RNA polymerase
ko05111	Biofilm formation-Vibrio cholerae
ko00040	Pentose and glucuronate interconversions

Table 4 (continued)

Table 4 (continued)

Pathway	Description
ko00330	Arginine and proline metabolism
ko04122	Sulfur relay system
ko00400	Phenylalanine, tyrosine and tryptophan biosynthesis
ko01220	Degradation of aromatic compounds
ko01523	Antifolate resistance
ko00473	D-alanine metabolism
ko05150	Staphylococcus aureus infection
ko00460	Cyanoamino acid metabolism
ko00430	Taurine and hypotaurine metabolism
ko00261	Monobactam biosynthesis
ko00350	Tyrosine metabolism
ko00471	D-glutamine and D-glutamate metabolism
ko00071	Fatty acid degradation
ko02026	Biofilm formation- <i>Escherichia coli</i>
ko00920	Sulfur metabolism
ko00600	Sphingolipid metabolism
ko04212	Longevity regulating pathway-worm
ko00280	Valine, leucine and isoleucine degradation
ko00511	Other glycan degradation
ko00910	Nitrogen metabolism
ko00523	Polyketide sugar unit biosynthesis
ko05132	Salmonella infection
ko05152	Tuberculosis
ko00625	Chloroalkane and chloroalkene degradation
ko00860	Porphyrin and chlorophyll metabolism
ko00290	Valine, leucine and isoleucine biosynthesis
ko00333	Prodigiosin biosynthesis
ko00130	Ubiquinone and other terpenoid-quinone biosynthesis
ko05418	Fluid shear stress and atherosclerosis
ko00750	Vitamin B6 metabolism
ko00626	Naphthalene degradation
ko00740	Riboflavin metabolism
ko00362	Benzoate degradation
ko04626	Plant-pathogen interaction

Table 4 (continued)

Table 4 (continued)

Pathway	Description
ko05134	Legionellosis
ko00660	C5-branched dibasic acid metabolism
ko00310	Lysine degradation
ko04727	GABAergic synapse
ko04621	NOD-like receptor signaling pathway
ko04217	Necroptosis
ko00340	Histidine metabolism
ko03013	RNA transport
ko00562	Inositol phosphate metabolism
ko03320	PPAR signaling pathway
ko00627	Aminobenzoate degradation
ko00525	Acarbose and validamycin biosynthesis
ko04724	Glutamatergic synapse
ko00603	Glycosphingolipid biosynthesis-globo and isoglobo series
ko05340	Primary immunodeficiency
ko04931	Insulin resistance
ko04070	Phosphatidylinositol signaling system
ko00380	Tryptophan metabolism
ko04016	MAPK signaling pathway-plant
ko00121	Secondary bile acid biosynthesis
ko00410	Beta-alanine metabolism
ko00072	Synthesis and degradation of ketone bodies
ko02040	Flagellar assembly
ko02030	Bacterial chemotaxis
ko04146	Peroxisome
ko00998	Biosynthesis of various secondary metabolites-part 2
ko00120	Primary bile acid biosynthesis
ko00311	Penicillin and cephalosporin biosynthesis
ko04940	Type I diabetes mellitus
ko00982	Drug metabolism-cytochrome P450
ko00980	Metabolism of xenobiotics by cytochrome P450
ko00830	Retinol metabolism
ko00540	Lipopolysaccharide biosynthesis
ko00524	Neomycin, kanamycin and gentamicin biosynthesis

Table 4 (continued)

Table 4 (continued)

Pathway	Description
ko01524	Platinum drug resistance
ko04141	Protein processing in endoplasmic reticulum
ko01054	Nonribosomal peptide structures
ko00053	Ascorbate and aldarate metabolism
ko00908	Zeatin biosynthesis
ko00332	Carbapenem biosynthesis
ko05010	Alzheimer disease
ko04152	AMPK signaling pathway
ko01055	Biosynthesis of vancomycin group antibiotics
ko04930	Type II diabetes mellitus
ko05130	Pathogenic <i>Escherichia coli</i> infection
ko05165	Human papillomavirus infection
ko05203	Viral carcinogenesis
ko03008	Ribosome biogenesis in eukaryotes
ko05205	Proteoglycans in cancer
ko05146	Amoebiasis
ko05206	MicroRNAs in cancer
ko00360	Phenylalanine metabolism
ko04917	Prolactin signaling pathway
ko05120	Epithelial cell signaling in <i>Helicobacter pylori</i> infection
ko04142	Lysosome
ko05200	Pathways in cancer
ko00622	Xylene degradation
ko00621	Dioxin degradation
ko01051	Biosynthesis of ansamycins
ko00440	Phosphonate and phosphinate metabolism
ko00966	Glucosinolate biosynthesis
ko04910	Insulin signaling pathway
ko00940	Phenylpropanoid biosynthesis
ko00572	Arabinogalactan biosynthesis-mycobacterium
ko04213	Longevity regulating pathway-multiple species
ko00791	Atrazine degradation
ko04934	Cushing syndrome
ko05211	Renal cell carcinoma

Table 4 (continued)

Table 4 (continued)

Pathway	Description
ko00531	Glycosaminoglycan degradation
ko00401	Novobiocin biosynthesis
ko04964	Proximal tubule bicarbonate reclamation
ko04214	Apoptosis-fly
ko04216	Ferroptosis
ko00785	Lipoic acid metabolism
ko02025	Biofilm formation-pseudomonas aeruginosa
ko00361	Chlorocyclohexane and chlorobenzene degradation
ko00960	Tropane, piperidine and pyridine alkaloid biosynthesis
ko04918	Thyroid hormone synthesis
ko04138	Autophagy-yeast
ko05016	Huntington disease
ko04973	Carbohydrate digestion and absorption
ko04972	Pancreatic secretion
ko00405	Phenazine biosynthesis
ko04714	Thermogenesis
ko00513	Various types of N-glycan biosynthesis
ko04920	Adipocytokine signaling pathway
ko01055	Glycosphingolipid biosynthesis-ganglio series
ko04930	Isoquinoline alkaloid biosynthesis
ko05130	PI3K-Akt signaling pathway
ko04622	RIG-I-like receptor signaling pathway
ko04612	Antigen processing and presentation
ko04915	Estrogen signaling pathway
ko04657	IL-17 signaling pathway
ko04914	Progesterone-mediated oocyte maturation
ko05215	Prostate cancer
ko04659	Th17 cell differentiation
ko04068	FoxO signaling pathway
ko04211	Longevity regulating pathway
ko05014	ALS
ko00633	Nitrotoluene degradation
ko05231	Choline metabolism in cancer
ko00590	Arachidonic acid metabolism

Table 4 (continued)

Table 4 (continued)

Pathway	Description
ko05145	Toxoplasmosis
ko04013	MAPK signaling pathway-fly
ko00903	Limonene and pinene degradation
ko04361	Axon regeneration
ko04072	Phospholipase D signaling pathway
ko00981	Insect hormone biosynthesis
ko00643	Styrene degradation
ko00510	N-glycan biosynthesis
ko05133	Pertussis
ko04210	Apoptosis
ko05131	Shigellosis
ko04974	Protein digestion and absorption
ko00944	Flavone and flavonol biosynthesis
ko00140	Steroid hormone biosynthesis
ko01053	Biosynthesis of siderophore group nonribosomal peptides
ko00565	Ether lipid metabolism
ko00592	Alpha-Linolenic acid metabolism
ko04011	MAPK signaling pathway-yeast
ko00591	Linoleic acid metabolism
ko05020	Prion diseases
ko04932	NAFLD
ko05012	Parkinson disease
ko00623	Toluene degradation
ko00472	D-arginine and D-ornithine metabolism
ko04260	Cardiac muscle contraction
ko04614	Renin-angiotensin system
ko00281	Geraniol degradation
ko01040	Biosynthesis of unsaturated fatty acids
ko00930	Caprolactam degradation
ko04113	Meiosis-yeast
ko05225	Hepatocellular carcinoma
ko05017	Spinocerebellar ataxia
ko00571	LAM biosynthesis
ko05204	Chemical carcinogenesis

Table 4 (continued)

Table 4 (continued)

Pathway	Description
ko04978	Mineral absorption
ko04215	Apoptosis-multiple species
ko05210	Colorectal cancer
ko05169	Epstein-Barr virus infection
ko05161	Hepatitis B
ko05160	Hepatitis C
ko05168	Herpes simplex virus 1 infection
ko05163	Human cytomegalovirus infection
ko05170	Human immunodeficiency virus 1 infection
ko05164	Influenza A
ko05167	Kaposi sarcoma-associated herpesvirus infection
ko05162	Measles
ko05222	Small cell lung cancer
ko05416	Viral myocarditis
ko04115	p53 signaling pathway
ko05143	African trypanosomiasis
ko05142	Chagas disease (American trypanosomiasis)
ko03050	Proteasome
ko03450	Non-homologous end-joining
ko00906	Carotenoid biosynthesis
ko04919	Thyroid hormone signaling pathway
ko00642	Ethylbenzene degradation
ko05034	Alcoholism
ko05031	Amphetamine addiction
ko00965	Betalain biosynthesis
ko05030	Cocaine addiction
ko04728	Dopaminergic synapse
ko04928	Parathyroid hormone synthesis, secretion and action
ko00624	Polycyclic aromatic hydrocarbon degradation
ko04726	Serotonergic synapse
ko04071	Sphingolipid signaling pathway
ko00984	Steroid degradation
ko04723	Retrograde endocannabinoid signaling

ABC, ATP-binding cassette; PTS, phosphotransferase system; CoA, co-ordinate analysis; TCA, tricarboxylic acid; CAMP, cationic antimicrobial peptide; NOD, nucleotide-binding and oligomerization domain; ALS, amyotrophic lateral sclerosis; MAPK, mitogen-activated protein kinase; NAFLD, non-alcoholic fatty liver disease; LAM, lipoarabinomannan; BN, Bining decoction; GN, gouty nephropathy.

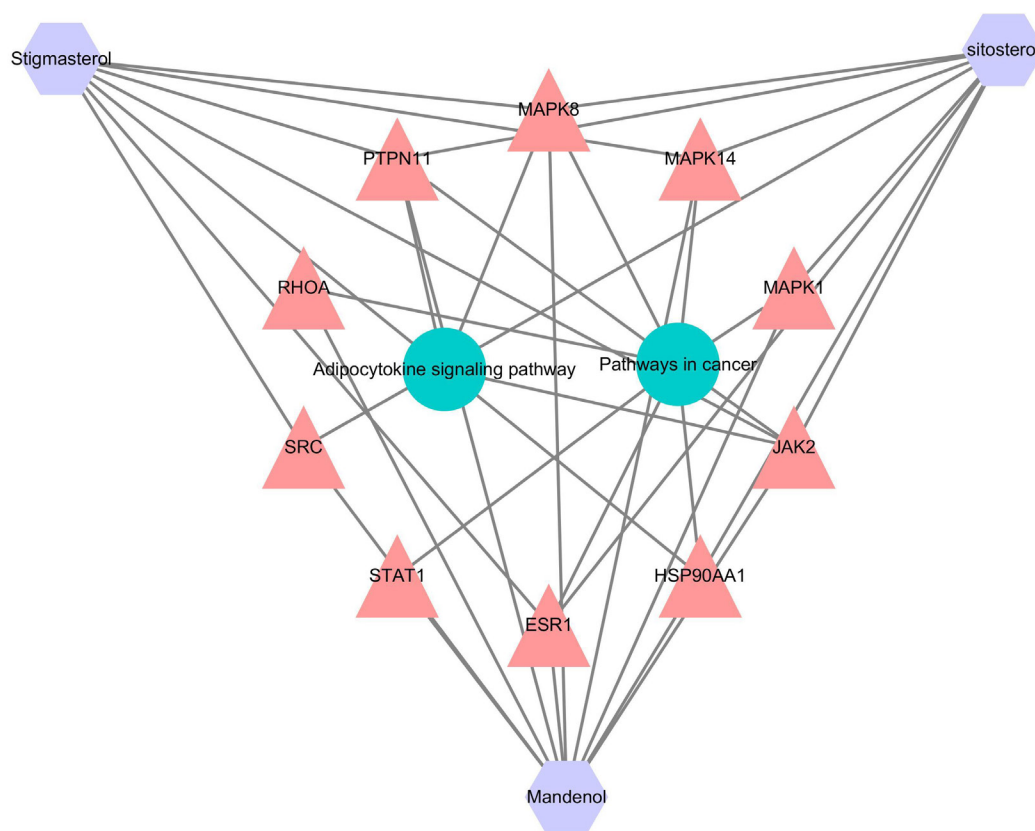


Figure 7 The component-target-disease interaction network of BN in the treatment of GN. BN, Bining decoction; GN, gouty nephropathy.

and repeat operations.

Revealed by the network pharmacology analysis, the “Pathways in cancer” had the most essential node, which had the strongest correlation with JAK2. These findings suggested that the possible molecular mechanism of BN on GN may be the inhibition of the JAK/STAT pathway.

Recent works have suggested that JAK/STAT signaling, NLRP3 inflammasome signaling, NF- κ B signaling, Toll-like receptor signaling, and cGAS-STING signaling control inflammation in cells (26). The JAK/STAT signaling pathway is part of the “pathways in cancer”. Among them, all members of the JAK/STAT signaling are observed in the injured kidney (27,28). This pathway consists of 3 parts: signal factor receptor, JAK, and STAT protein. JAKs are comprised of 4 members, namely JAK1, JAK2, JAK3, and TYK2. Among these, the JAK2 linkage to inflammatory processes is most evident, as it serves as a signaling transmitter downstream of the main cytokine receptors (29,30). The STAT protein, the downstream target of the JAKs family, is responsible for the signal transduction to the nucleus. Most cytokine-initiated immune responses

depend on STAT proteins (31). There are 7 numbers for the STAT protein, including STAT1-4, STAT5a, STAT5b, and STAT6. Moreover, multiple cytokines associated with the JAK/STAT information pathway activate specific STAT proteins. The NLRP3 inflammasome is an intracellular supramolecular complex consisting of sensor molecules, the adaptor apoptosis-associated speck-like protein containing a CARD (ASC), as well as the effector protease caspase 1. NLRP3 inflammasome and the successive caspase 1 are activated by reactive oxygen species (ROS), K^+ efflux, oxidized mitochondrial DNA release, and lysosomal disruption-induced ion fluxes, thereby executing inflammatory factors production and cell death (32). The type I interferon (IFN) receptors have been found to regulate Caspase-11 via the JAK/STAT signaling pathway and the caspase-1 activation is dependent on the expression of Caspase-11. Activated caspase 1 cleaves both pro-IL-1 and pro-IL-18 for the production of mature IL-1 and IL-18. Meanwhile, it also cleaves and activates the cytoplasmic gasdermin proteins, contributing to cell swelling, cell membrane rupture, cytoplasm efflux,

and eventually pyroptosis (33). The inflammatory factors outflow from the gasdermin-contained holes (34), inducing inflammation and injury. Furthermore, recent articles have disclosed that the JAK/STAT signaling pathway could be regarded as a potential target for rheumatoid arthritis (RA) therapy. Recently, novel JAK/STAT inhibitors have been approved by the Food and Drug Administration (FDA) for RA therapy. However, their associated risk and high cost limit their applications for broader use. These restrictions provide a basis for the exploration of novel JAK/STAT inhibitors of natural origin with improved safety, tolerability, and cost-effectiveness (35). One of the aims of investigating the BN is to confirm that the JAK/STAT signaling pathway is closely implicated in the pathological process of GN.

Based on the gut-kidney axis theory, the human gut microbiome is in equilibrium under normal circumstances, and *Firmicutes* and *Bacteroidetes* are the dominating bacterial phyla (36). These 2 dominating bacterial phyla were found to occupy more than 95% of the total gut microbiota, which is similar to the findings of other experiments in China and internationally (37). Additionally, *Bacteroidetes* and *Firmicutes* differed significantly at the phylum level; both of them were the dominant bacterial at the genus level in the GN model group.

Following previous research findings, the abundance of *Firmicutes* phylum was decreased and the *Bacteroides* phylum abundance was elevated in the model group when compared with the blank group (38). *Bacteroides* was the dominant species that distinguished the intestinal flora between the model and the blank groups. Also, the serum uremic toxins levels were positively correlated with the relative abundance of the *Bacteroides* genus (39), suggesting that these bacteria play a part in the GN progression. In the model group, the *Lactobacillus* and *Bacteroides* abundances were the dominant bacteria. The correlation analysis revealed that *Lachnospiraceae* was reduced significantly in the model group, which was in line with the previous articles, implying the effectiveness of SCFAs in kidney function (40).

In a chronic kidney disease study, the normal gut microbiome showed an enrichment in *Blautia*, *Bacteroides*, *Turicibacter*, *Escherichia-Shigella*, and *Lachnospiraceae* (41), which was similar to our study to some extent. These biota are correlated to systemic inflammation, intestinal barrier integrity, and renal fibrosis (42,43).

Some active ingredients in TCM are capable of impeding the activity of XOD and diminishing the production of UA (44-47). *Lactobacillus*, *Bacteroides*, and *Turicibacter* became the dominant bacteria after the BN intervention.

Intestinal bacteria have been shown to decompose UA. Many symbiotic bacteria in the gut, such as *Lactobacillus*, can express allantoicase and uricase, which participate in UA breakdown (48). Besides, *Lactobacillus* can also reduce the purine absorption in the gut, preventing the increase of serum uric acid (SUA) and further causing HUA (49). Furthermore, *Turicibacter* has been found to correlate to intestinal butyric acid, thereby exerting significant anti-inflammatory effects (50).

The combination of network pharmacology and the holistic view of TCM syndrome differentiation and treatment is a methodology with dynamic analysis and holistic view. Through the TCM database website and computer software, the diagnosis and treatment methods of TCM diseases are combined with the compatibility of TCM. The network pharmacology method is used to analyze various active ingredients of TCM or compound prescriptions, and to explore the mechanism of action of compound prescriptions in the treatment of diseases. It solves the difficulties in modern Chinese medicine research and improves the scientific and technological level of modern Chinese medicine research. In this paper, the empirical prescription BN is used as the research object of network pharmacology to predict the key targets and pathways of BN in the treatment of GN. However, it should be noted that network pharmacology technology is only a qualitative prediction of drug components and targets, to determine its clear pharmacological effects still need to be verified by animal experiments or even clinical trials. The in-depth and reasonable application of network pharmacology in the study of the utility mechanism and material basis of TCM compound is worth looking forward. Up to now, we found no research on GN combining network pharmacology and gut microbiota.

To date, further results with a larger sample size are warranted to validate the findings from the limited sample size in this study. Additionally, a 24-hour urine composition analysis for ICR mice was not performed in this study. Finally, the SCFAs were not conducted with the collected fecal samples due to financial constraints, and a metabolomic analysis was also lacking. Our group will launch in-depth research to address these issues in future and hope to offer a novel avenue for UA-induced nephropathy.

Conclusions

In this study, network pharmacology analysis combined with a 16S rRNA sequencing approach was performed to reveal

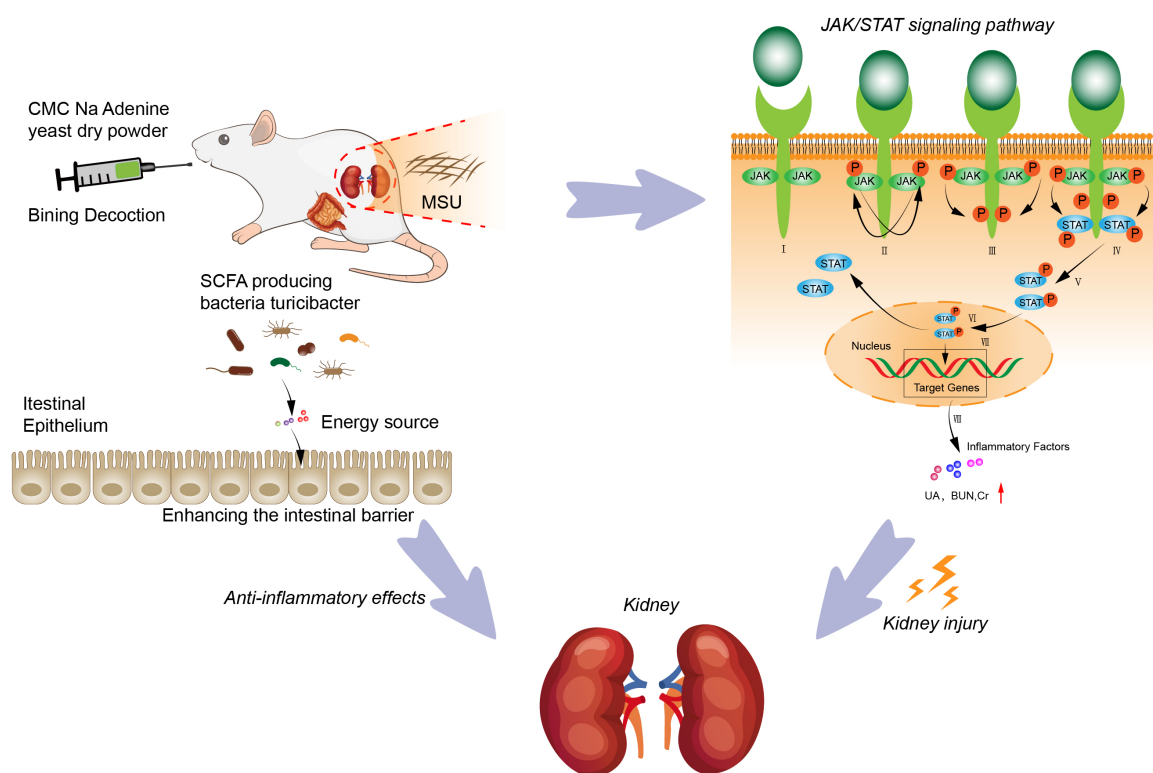


Figure 8 BN decoction function in GN. CMC, carboxymethyl cellulose; MSU, monosodium urate; SCFA, short-chain fatty acids; UA, uric acid; BUN, blood urea nitrogen; Cr, creatinine; BN, Bining decoction; GN, gouty nephropathy; black arrows, activated; red arrow, increased.

the gut microbiota composition in GN, and the effect of BN on GN (Figure 8).

We speculated that BN functioned in GN therapy by inhibiting the JAK/STAT signaling pathway, increasing the beneficial bacteria *Turicibacter* associated with intestinal butyric acid, which could enhance the intestinal barrier, and exert an anti-inflammatory effect. There have been few *Turicibacter*-related articles in recent studies. Our findings highlight that BN should be recommended for GN treatment in the development of novel chemopreventive or chemotherapeutic agents.

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Footnote

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Ethical Statement: The authors are accountable for all aspects of the work in ensuring that questions related to the accuracy or integrity of any part of the work are appropriately investigated and resolved. The study was conducted in accordance with the Declaration of Helsinki (as revised in 2013). Animal experiments were performed under the approval by the Animal Ethics Committee

of Heilongjiang University of Chinese Medicine (No. 202009670), in compliance with the Heilongjiang University of Chinese Medicine guidelines for the care and use of animals.

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References

- Mei Y, Dong B, Geng Z, et al. Excess Uric Acid Induces Gouty Nephropathy Through Crystal Formation: A Review of Recent Insights. *Front Endocrinol (Lausanne)* 2022;13:911968.
- Towiwat P, Chhana A, Dalbeth N. The anatomical pathology of gout: a systematic literature review. *BMC Musculoskelet Disord* 2019;20:140.
- Roughley MJ, Belcher J, Mallen CD, et al. Gout and risk of chronic kidney disease and nephrolithiasis: meta-analysis of observational studies. *Arthritis Res Ther* 2015;17:90.
- Ma Q, Fang L, Su R, et al. Uric acid stones, clinical manifestations and therapeutic considerations. *Postgrad Med J* 2018;94:458-62.
- FitzGerald JD, Dalbeth N, Mikuls T, et al. 2020 American College of Rheumatology Guideline for the Management of Gout. *Arthritis Care Res (Hoboken)* 2020;72:744-60.
- Bardin T, Richette P. Impact of comorbidities on gout and hyperuricaemia: an update on prevalence and treatment options. *BMC Med* 2017;15:123.
- Li Y, Xu W, Zhang F, et al. The Gut Microbiota-Produced Indole-3-Propionic Acid Confers the Antihyperlipidemic Effect of Mulberry-Derived 1-Deoxynojirimycin. *mSystems* 2020;5:00313-20.
- Yanai H, Adachi H, Hakoshima M, et al. Molecular Biological and Clinical Understanding of the Pathophysiology and Treatments of Hyperuricemia and Its Association with Metabolic Syndrome, Cardiovascular Diseases and Chronic Kidney Disease. *Int J Mol Sci* 2021;22:9221.
- Dalbeth N, Merriman TR, Stamp LK. Gout. *Lancet* 2016;388:2039-52.
- Gilbert JA, Blaser MJ, Caporaso JG, et al. Current understanding of the human microbiome. *Nat Med* 2018;24:392-400.
- Yin H, Liu N, Chen J. The Role of the Intestine in the Development of Hyperuricemia. *Front Immunol* 2022;13:845684.
- Crane JK, Naeher TM, Broome JE, et al. Role of host xanthine oxidase in infection due to enteropathogenic and Shiga-toxicogenic *Escherichia coli*. *Infect Immun* 2013;81:1129-39.
- Topping DL, Clifton PM. Short-chain fatty acids and human colonic function: roles of resistant starch and nonstarch polysaccharides. *Physiol Rev* 2001;81:1031-64.
- Li C, Wang C, Guo Y, et al. Research on the effect and underlying molecular mechanism of Cangzhu in the treatment of gouty arthritis. *Eur J Pharmacol* 2022;927:175044.
- Yang M, Lao L. Emerging Applications of Metabolomics in Traditional Chinese Medicine Treating Hypertension: Biomarkers, Pathways and More. *Front Pharmacol* 2019;10:158.
- Ru J, Li P, Wang J, et al. TCMSP: a database of systems pharmacology for drug discovery from herbal medicines. *J Cheminform* 2014;6:13.
- Liu H, Wang J, Zhou W, et al. Systems approaches and polypharmacology for drug discovery from herbal medicines: an example using licorice. *J Ethnopharmacol* 2013;146:773-93.
- Zheng Y, Zeng X, Chen P, et al. Integrating Pharmacology and Gut Microbiota Analysis to Explore the Mechanism of *Citri Reticulatae Pericarpium* Against Reserpine-Induced Spleen Deficiency in Rats. *Front Pharmacol* 2020;11:586350.
- Zhao H, Gao Q, Kong LZ, et al. Study on Network Pharmacological Analysis and Preliminary Validation to Understand the Mechanisms of Plantaginis Semen in Treatment of Gouty Nephropathy. *Evid Based Complement Alternat Med* 2020;2020:8861110.
- Zhen Z, Xia L, You H, et al. An Integrated Gut Microbiota and Network Pharmacology Study on Fuzi-Lizhong Pill for Treating Diarrhea-Predominant Irritable Bowel Syndrome. *Front Pharmacol* 2021;12:746923.
- Yu X, Jiang W, Kosik RO, et al. Gut microbiota changes and its potential relations with thyroid carcinoma. *J Adv Res* 2022;35:61-70.
- Alam MS, Rahaman MM, Sultana A, et al. Statistics and network-based approaches to identify molecular mechanisms that drive the progression of breast cancer.

- Comput Biol Med 2022;145:105508.
23. Wang Y, Tong Q, Shou JW, et al. Gut Microbiota-Mediated Personalized Treatment of Hyperlipidemia Using Berberine. *Theranostics* 2017;7:2443-51.
 24. Yang L, Chang B, Guo Y, et al. The role of oxidative stress-mediated apoptosis in the pathogenesis of uric acid nephropathy. *Ren Fail* 2019;41:616-22.
 25. Wang N, Li P, Pan J, et al. *Bacillus velezensis* A2 fermentation exerts a protective effect on renal injury induced by Zearalenone in mice. *Sci Rep* 2018;8:13646.
 26. Yuan Q, Tang B, Zhang C. Signaling pathways of chronic kidney diseases, implications for therapeutics. *Signal Transduct Target Ther* 2022;7:182.
 27. Brosius FC 3rd, He JC. JAK inhibition and progressive kidney disease. *Curr Opin Nephrol Hypertens* 2015;24:88-95.
 28. Fragiadaki M, Lannoy M, Themanns M, et al. STAT5 drives abnormal proliferation in autosomal dominant polycystic kidney disease. *Kidney Int* 2017;91:575-86.
 29. Koschmieder S, Mughal TI, Hasselbalch HC, et al. Myeloproliferative neoplasms and inflammation: whether to target the malignant clone or the inflammatory process or both. *Leukemia* 2016;30:1018-24.
 30. Hasselbalch HC. Perspectives on chronic inflammation in essential thrombocythemia, polycythemia vera, and myelofibrosis: is chronic inflammation a trigger and driver of clonal evolution and development of accelerated atherosclerosis and second cancer? *Blood* 2012;119:3219-25.
 31. Zhang S, Gan X, Qiu J, et al. IL-10 derived from Hepatocarcinoma cells improves human induced regulatory T cells function via JAK1/STAT5 pathway in tumor microenvironment. *Mol Immunol* 2021;133:163-72.
 32. Mangan MSJ, Olhava EJ, Roush WR, et al. Targeting the NLRP3 inflammasome in inflammatory diseases. *Nat Rev Drug Discov* 2018;17:588-606.
 33. Cuevas S, Pelegrín P. Pyroptosis and Redox Balance in Kidney Diseases. *Antioxid Redox Signal* 2021;35:40-60.
 34. Shi J, Gao W, Shao F. Pyroptosis: Gasdermin-Mediated Programmed Necrotic Cell Death. *Trends Biochem Sci* 2017;42:245-54.
 35. Kour G, Choudhary R, Anjum S, et al. Phytochemicals targeting JAK/STAT pathway in the treatment of rheumatoid arthritis: Is there a future? *Biochem Pharmacol* 2022;197:114929.
 36. Wexler AG, Goodman AL. An insider's perspective: *Bacteroides* as a window into the microbiome. *Nat Microbiol* 2017;2:17026.
 37. Ruan D, Fouad AM, Fan QL, et al. Dietary L-arginine supplementation enhances growth performance, intestinal antioxidative capacity, immunity and modulates gut microbiota in yellow-feathered chickens. *Poult Sci* 2020;99:6935-45.
 38. Ye G, Zhou M, Yu L, et al. Gut microbiota in renal transplant recipients, patients with chronic kidney disease and healthy subjects. *Nan Fang Yi Ke Da Xue Xue Bao* 2018;38:1401-8.
 39. Kikuchi M, Ueno M, Itoh Y, et al. Uremic Toxin-Producing Gut Microbiota in Rats with Chronic Kidney Disease. *Nephron* 2017;135:51-60.
 40. Pan L, Han P, Ma S, et al. Abnormal metabolism of gut microbiota reveals the possible molecular mechanism of nephropathy induced by hyperuricemia. *Acta Pharm Sin B* 2020;10:249-61.
 41. Xu X, Wang H, Guo D, et al. Curcumin modulates gut microbiota and improves renal function in rats with uric acid nephropathy. *Ren Fail* 2021;43:1063-75.
 42. Chen L, Chen DQ, Liu JR, et al. Unilateral ureteral obstruction causes gut microbial dysbiosis and metabolome disorders contributing to tubulointerstitial fibrosis. *Exp Mol Med* 2019;51:1-18.
 43. Feng YL, Cao G, Chen DQ, et al. Microbiome-metabolomics reveals gut microbiota associated with glycine-conjugated metabolites and polyamine metabolism in chronic kidney disease. *Cell Mol Life Sci* 2019;76:4961-78.
 44. Yang Q, Wang Q, Deng W, et al. Anti-hyperuricemic and anti-gouty arthritis activities of polysaccharide purified from *Lonicera japonica* in model rats. *Int J Biol Macromol* 2019;123:801-9.
 45. Adachi SI, Kondo S, Sato Y, et al. Anti-hyperuricemic effect of isorhamnetin in cultured hepatocytes and model mice: structure-activity relationships of methylquercetins as inhibitors of uric acid production. *Cytotechnology* 2019;71:181-92.
 46. Cheng LC, Murugaiyah V, Chan KL. Flavonoids and phenylethanoid glycosides from *Lippia nodiflora* as promising antihyperuricemic agents and elucidation of their mechanism of action. *J Ethnopharmacol* 2015;176:485-93.
 47. Lü JM, Yao Q, Chen C. 3,4-Dihydroxy-5-nitrobenzaldehyde (DHNB) is a potent inhibitor of xanthine oxidase: a potential therapeutic agent for treatment of hyperuricemia and gout. *Biochem Pharmacol* 2013;86:1328-37.
 48. Crane JK. Role of host xanthine oxidase in infection due

- to enteropathogenic and Shiga-toxigenic *Escherichia coli*. *Gut Microbes* 2013;4:388-91.
49. Yamada N, Iwamoto C, Kano H, et al. Evaluation of purine utilization by *Lactobacillus gasseri* strains with potential to decrease the absorption of food-derived purines in the human intestine. *Nucleosides Nucleotides Nucleic Acids* 2016;35:670-6.
50. Wu M, Yang S, Wang S, et al. Effect of Berberine on Atherosclerosis and Gut Microbiota Modulation and Their Correlation in High-Fat Diet-Fed ApoE^{-/-} Mice. *Front Pharmacol* 2020;11:223.

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