Gut Microbiota Differs Between Parkinson’s Disease Patients and Healthy Controls in Northeast China

Chunxiao Li1†‡, Li Cui1†, Yimin Yang2, Jing Miao1, Xiuzhen Zhao1, Jingdian Zhang1, Guohong Cui3* and Ying Zhang1*

1Department of Neurology and Neuroscience Center, The First Hospital of Jilin University, Changchun, China, 2Department of Intensive Care Unit, First Hospital of Jilin University, Changchun, China, 3Neurobiology Laboratory, National Institute of Environmental Health Sciences, National Institutes of Health, Durham, NC, United States

Background: There is accumulating evidence suggesting a connection between the gut and Parkinson’s disease (PD). Gut microbiota may play an important role in the intestinal lesions in PD patients.

Objective: This study aims to determine whether gut microbiota differs between PD patients and healthy controls in Northeast of China, and to identify the factors that influence the changes in the gut microbiota.

Methods: We enrolled 51 PD patients and 48 healthy controls in this study. Microbial species in stool samples were determined through 16S-rRNA gene sequencing. Dietary intakes were collected from a subset of 42 patients and 23 controls using a food frequency questionnaire (FFQ). Gut microbiota species richness, diversity, differential abundance of individual taxa between PD patients and controls, and the relationship between the gut microbiota abundance and the dietary and clinical factors were analyzed.

Results: PD patients showed decreased species richness, phylogenetic diversity, β-diversity, and altered relative abundance in several taxa compared to the controls. PD-associated clinical scores appeared to be the most influential factors that correlated with the abundance of a variety of taxa. The most consistent findings suggested by multiple analyses used in this study were the increase of Akkermansia and the decrease of Lactobacillus in PD patients in Northeast China.

Conclusion: Gut microbiota significantly differed between a group of PD patients and healthy controls in Northeast China, with decreased species richness, phylogenetic diversity, β-diversity, and altered relative abundance in several taxa compared to the controls.

Keywords: Parkinson’s disease, gut microbiota, dysbiosis, Akkermansia, Lactobacillus

Abbreviations: PD, Parkinson’s disease; OTUs, Operational taxonomic units; SD, standard deviation; BMI, body mass index; UPDRS-III, The third part of the Unified Parkinson’s Disease Rating Scale; SCOPA-AUT, Scale for Outcomes in PD for Autonomic Symptoms; NMSQ, Non-Motor Symptom Questionnaire; RBDQ-HK, REM Sleep Behavior Disorder Questionnaire-Hong Kong.
BACKGROUND

Parkinson's disease (PD) is a common neurodegenerative disorder characterized by motor and non-motor symptoms. The motor symptoms include bradykinesia, resting tremor, rigidity, and postural instability (Reichmann, 2010). These symptoms are mainly attributable to the degeneration and loss of dopaminergic neurons in the substantia nigra of the midbrain (Chung et al., 2001). In contrast, the non-motor symptoms of PD, such as constipation, hyposmia, depression, and cognitive impairment (Chen et al., 2013) cannot be explained by the changes in the substantia nigra. Braak et al. (2003) found that in PD, pathological changes were not limited to the substantia nigra, but rather involved multiple parts of the nervous system. In the pre-symptomatic stage, pathological changes are confined to the medulla oblongata/pontine tegmentum and olfactory bulb/anterior olfactory nucleus. As the disease progresses, the substantia nigra and other nuclei in the midbrain and forebrain become the main areas of pathological involvement. Eventually, the pathological changes spread to the neocortex (Braak et al., 2004) causing cognitive symptoms. Recently, the non-motor symptoms of PD have become a focus of PD studies. One such symptom is constipation, which is common and often precedes the motor symptoms (Cersosimo et al., 2013). Lewy bodies and α-synuclein, which are the neuropathological hallmarks of PD, may appear in the gut before they appear in the brain (Adler and Beach, 2016). In PD patients, constipation is often associated with α-synuclein accumulation and neurodegeneration in the enteric nervous system (Cersosimo and Benarroch, 2012), with local inflammation, oxidative stress, and increased intestinal permeability (Forsyth et al., 2011; Devos et al., 2013). These pathophysiological changes occur at the initial stages of PD, and in some cases, may predate the motor symptoms by years (Savica et al., 2009; Shannon et al., 2012). Therefore, it is speculated that lesion in the gut may be one of the triggering factors that contribute to the pathogenesis of PD.

The gut microbiota is an important component of the intestinal environment. It can affect immune function and development (Hooper et al., 2012), metabolism (Sonnenburg and Bäckhed, 2016), as well as the central nervous system (Mayer et al., 2014). Several psychiatric and neurological disorders have been found to be associated with changes in the gut microbiota, including schizophrenia (Severance et al., 2016), depression (Zheng et al., 2016), autism (Kang et al., 2013), multiple sclerosis (Berer et al., 2011; Lee et al., 2011) and Alzheimer’s disease (Minter et al., 2017). Recently, many studies have shown that the composition of gut microbiota significantly differs between PD patients and control subjects, but the results varied widely among these studies (Hasegawa et al., 2015; Keshavarzian et al., 2015; Mulak and Bonaz, 2015; Scherjans et al., 2015; Felice et al., 2016; Unger et al., 2016; Hill-Burns et al., 2017; Li et al., 2017). However, this is not altogether surprising, as the gut microbiota is affected by many factors, such as lifestyle, diet (Smits et al., 2017), age, geography (Yatsunenko et al., 2012), body mass index (BMI), and race (Chen et al., 2016). In this study, we aim to determine whether gut microbiota profiles differ between PD patients and control subjects living in the northeast region of China.

MATERIALS AND METHODS

Subjects

We enrolled 51 patients who had been diagnosed with PD according to the diagnostic criteria proposed by the International Parkinson Disease and Movement Disorder Society in 2015 (Postuma et al., 2015) in the First Hospital of Jilin University. To be included in the study, PD patients were required to meet the following criteria: no antibiotic use for at least 3 months prior to the study; no digestive system diseases, such as inflammatory bowel disease, no diseases affecting the liver, gall bladder, or pancreas, and no history of surgery on the digestive tract; non-smoking; no alcohol consumption for at least 2 years prior to the study; no autoimmune disease, such as diabetes; no family history of PD; and age at PD onset > 50 years. The patients were on PD medications including carbidopa/levodopa and dopamine agonists when the samples were collected. Subjects with a history of using medications that have been shown to affect gut microbiota, including COMT inhibitors, anticholinergics, anti-secretory drugs, or cardiovascular drugs within the 3 months before the start of the study were excluded.

We also recruited the spouses of the PD patients as healthy controls (n = 39) to minimize the influence of lifestyle factors on the gut microbiota. Recruiting the spouses of PD patients as controls could also help offsetting the potential impact on gut microbiota by the physical and socioeconomic stress that is present in both PD patients and their spouses (O’Reilly et al., 1996; Hemmerle et al., 2012; Anderson et al., 2016). To ensure a sufficient sample size, we further recruited nine age-matched, healthy controls from the local community. Thus, in total, we recruited 48 healthy controls who met the following criteria: no neurodegenerative diseases; no digestive system diseases, no diseases affecting the liver, gall bladder, or pancreas, and no history of surgery on the digestive tract; no hypertension, diabetes, or immune diseases; no antibiotic use, no proton pump inhibitors use, no cardiac drugs use for at least 3 months prior to the study; non-smoking; and no alcohol consumption for more than 2 years prior to the study (Table 1). All subjects were of Han Chinese ethnicity and resided in the northeast region of China. In addition, patients and controls were well-matched in terms of age and BMI.

Clinical Data Collection

Parkinsonian motor symptoms were measured using the Movement Disorder Society Unified Parkinson Disease Rating Scale (UPDRS)-III (Movement Disorder Society Task Force on Rating Scales for Parkinson’s Disease, 2003) and the modified Hoehn and Yahr scale (Goetz et al., 2004). Non-motor symptoms were assessed using the Non-Motor Symptom Questionnaire (NMSQ; Chaudhuri et al., 2006), Scales for Outcomes in Parkinson’s Disease—autonomic symptoms (SCOPA-AUT;
Visser et al., 2004), and Rapid Eye Movement Sleep Behavior Disorder questionnaire—Hong Kong (RBDQ-HK) scale (Shen et al., 2014) in the “on” state. PD duration was defined as the interval between the patient’s age at the time of the study and their age at PD onset (Table 1). Constipation was measured according to the Wexner scoring scale (Agachan et al., 1996). Medication data were extracted from the patients’ prescriptions at the time of this study.

### Patient Groups and Subgroups

Our study population was divided into a PD group (N = 51) and a control group (N = 48). The PD group was further divided into subgroups according to the patients’ scores on various scales but resulted in too small sample sizes. Thus, the data were not further analyzed and were not presented in this study.

A subset of PD patients (N = 42) and controls (N = 23) completed a modified Chinese food frequency questionnaire (FFQ). The rest of the subjects only provided the stool samples but declined the request for further completing the questionnaires. We quantified their dietary fiber, carbohydrate, and fat intake according to Chinese food composition tables, and samples but declined the request for further completing the questionnaires. We quantified their dietary fiber, carbohydrate, and fat intake according to Chinese food composition tables, and found no difference between the PD and control groups (Table 2, Supplementary Figure S1).

### Stool Collection

Stool samples were collected at home in sterile, sealed tubes by the subjects themselves, according to our instructions. The subjects were told to pass stools in a clean basin, place the middle section of the stools into a sterilized tube, and store the samples at −80°C until analysis. 

Within 3 days, cooled them with liquid nitrogen, and stored them at −80°C until analysis.

### 16S rRNA Amplicon Analysis

DNA extraction from the stool samples and 16S rRNA amplicon sequencing were performed using the HiSeq2500 PE250 sequencing platform, according to the manufacturer’s instructions. Briefly, fecal DNA was extracted using QIAamp Fast DNA Stool Mini Kit (Qiagen) and diluted to 1 ng/µl before PCR amplification. Barcoded primers (515F and 806R) were used to target the V4 region of the 16S rRNA gene. Amplicons were purified and sequenced using HiSeq2500 PE250 sequencing platform. Raw tags were formed by linking the reads from each sample using FLASH (V1.2.7) and filtered to obtain high quality clean tags according to published protocols (Qiime V1.7.0). The final effective tags were obtained by removing the chimera sequence from clean tags (Caporaso et al., 2010; Edgar et al., 2011; Haas et al., 2011; Magoc and Salzberg, 2011; Bokulich et al., 2013). All samples were sequenced in the same laboratory. Operational taxonomic units (OTUs) were selected using Uparse v7.0.1001 (Edgar, 2013) at 97% similarity for all effective tags. The representative sequence of the OTUs was analyzed using the mothur program and SILVA SSU rRNA database (Wang et al., 2007; Quast et al., 2013).

### Species Richness and Alpha-Diversity

We analyzed within-community microbial diversity by using α-diversity analysis (Li et al., 2013). We calculated the Observed-species, Chao1, Shannon, Simpson, ACE, Goods-coverage, and PD_whole_tree indexes by using Qiime v1.7.0. We plotted the rarefaction curve, rank abundance curve (Lundberg et al., 2013), and species accumulation boxplot with R v2.15.3. We analyzed the differences in these indices between the PD and control groups by using t-test and Wilcoxon rank-sum test.

### Beta-Diversity

We calculated the dissimilarities (distance) in the microbiomes, including unweighted unique fraction metric (Unifrac; Lozupone and Knight, 2005; Lozupone et al., 2011) and weighted Unifrac (Lozupone et al., 2007), between the PD and control groups by using the weighted pair-group method with arithmetic mean (UPGMA) by Qiime software (Version 1.7.0). We performed principal co-ordinates analysis by using the WGCNA, stats, and ggplot2 packages of the R software. Analysis of similarities (Anosim), multi-response permutation procedures (MRPP) analysis, and Adonis (permutational multivariate analysis of variance; Zapala and Schork, 2006) were conducted using the R software. Differences in the weighted and unweighted UniFrac between the PD and control groups were analyzed using the t-test and Wilcoxon rank-sum test by the R software.

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### Table 1 | Demographic characteristics of all the participants.

| Demographic factor | PD | Control | P-value* |
|--------------------|----|---------|----------|
| Total subjects     | 51 | 48      | NA       |
| Male               | 32 | 19      | NA       |
| Female             | 19 | 29      | NA       |
| Age (years, mean ± SD) | 62.2 ± 8.3 | 62.2 ± 9.2 | 0.746  |
| BMI (kg/m², mean ± SD) | 24.9 ± 3.8 | 24.5 ± 2.6 | 0.711  |
| PD duration (years, mean ± SD) | 4.55 ± 3.59 | NA | NA |  

*Unpaired t-test; PD, Parkinson’s disease; SD, standard deviation; BMI, body mass index.

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### Table 2 | PD characteristics, constipation score and dietary fiber intake information collected from a subset of the participants.

| PD            | Control | P-value* |
|---------------|---------|----------|
| Number of subjects | 42 | 23 | NA |
| Modified Hoehn and Yahr scale | 1.70 ± 0.73 | NA | NA |
| UPDRS-III (mean ± SD) | 24.55 ± 12.38 | 0.78 ± 0.74 | < 0.0001 |
| NMSQ (mean ± SD) | 12.07 ± 3.85 | 1.043 ± 0.88 | < 0.0001 |
| SCOPA-AUT (mean ± SD) | 16.74 ± 8.23 | 1 ± 0.60 | < 0.0001 |
| RBDQ-HK (mean ± SD) | 16.93 ± 14.32 | 2.44 ± 1.81 | < 0.0001 |
| Constipation (Wexner score, mean ± SD) | 5.93 ± 4.05 | 0.85 ± 0.82 | < 0.0001 |
| Dietary fiber (g/day, mean ± SD) | 12.56 ± 5.47 | 12.64 ± 3.74 | < 0.0001 |

*Unpaired t-test; PD, Parkinson’s disease; SD, standard deviation; UPDRS-III, The third part of the Unified Parkinson’s Disease Rating Scale; SCOPA-AUT, Scale for Outcomes in PD for Autonomic Symptoms; NMSQ, non-movement syndrome questionnaire; RBDQ-HK, REM Sleep Behavior Disorder Questionnaire-Hong Kong.

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1. http://ccb.jhu.edu/software/FLASH/
2. http://qiime.org/scripts/split_libraries_fastq.html
Differential Abundance Analysis of Gut Microbiota Between PD Patients and Controls

Linear discriminant analysis (LDA) effect size (LEfSe) analysis (Segata et al., 2011) was conducted using the LEfSe software, and the threshold for LDA score was set at 4. Permutation tests were conducted between groups at various classification levels (phylum, class, order, family, genus, and species) by using the R software, and P-values were obtained. The P-values were modified to q-values (White et al., 2009) by using the Benjamini and Hochberg false discovery rate method. Differences between groups were also analyzed using the t-test and Wilcoxon rank-sum test in R software.

Correlations Between Microbiome Abundance and Clinical and Dietary Factors

To elucidate the potential relationship between the gut microbiota and clinical factors such as the UPDRS, NMSQ, and SCOPA scores and dietary factors, we conducted a Spearman correlation analysis to study the relationship between the clinical and dietary factors mentioned above and gut microbiota abundance, focusing on bacteria with significant differences between PD and control groups identified in the t-test analysis.

Quantification of the Effect Size

To quantify the effect size in the t-test analyses, we calculated the Hedge’s g. To quantify the effect size for the Wilcoxon rank-sum test, we calculated the rank biserial correlation coefficient r.

RESULTS

The Differences of Gut Microbiota Profiles Between PD and Control Groups

We first plotted the rarefaction curves to show that they approached a plateau, indicating that the sequencing depth was adequate in this study (Figure 1A). We also plotted the rank abundance curves to compare the richness and evenness between PD and control groups and found reduced richness (less spread on x-axis) and evenness (steeper slope) in PD patients (Figure 1B). We further plotted species cumulation boxplot to confirm that the sample size of our study is adequate for valid analysis (Figure 1C).

We next calculated the indices to compare the richness ('Observed species', Chao1, ACE, Figures 1D–F), α-diversity (Shannon, Simpson, Figures 1G,H), and phylogenetic diversity ('PD whole tree', Figure 1I) between PD and controls of all recruited subjects (Figures 1D–I), and between PD and controls with no or mild constipation (defined as Wexner score ≤ 3, Supplementary Figure S2). The obtained indices suggested decreased species richness and phylogenetic diversity, and increased α-diversity in PD patients when all subjects were included in the comparisons. However, the α-diversity indices (Shannon and Simpson) were not significantly different between PD patients and controls when only the subjects with Wexner score ≤ 3 were compared, suggesting that constipation could be a confounding factor that impacts the α-diversity (Zhu et al., 2014; Mancabelli et al., 2017) or due to reduced sample size in each group. To further rule out the potential impact of gastric distress on these results, we carried out correlation analyses to investigate the relationship between Wexner scores and the gut microbiota richness and diversity indices (Supplementary Figures S3A–F). The correlations between the measured indices and the Wexner constipation score are weak, suggesting that the severity of constipation does not have a strong effect on these indices. To investigate the potential impact of gender on these findings, we performed the two-way ANOVA and did not find the significant main effect of gender (Supplementary Figure S4).

We then calculated weighted and unweighted Unifrac, and compared the β-diversity between PD patients and controls. Principal Co-ordinates Analysis (PCoA) plots showed more clustered distribution of the samples from PD patients compared to the samples from the controls (Figures 1J,K, Supplementary Figures S3G,H). Anosim analysis also revealed reduced distances between PD samples than the distances between the control samples (Figure 1L). Both analyses suggested reduced β-diversity in PD patients, which was further supported by MRPP analysis (A = 0.013, observed Δ = 0.559, expected Δ = 0.566, p = 0.001) and Adonis analysis ($R^2 = 0.035$, $p = 0.001$).

To investigate the differential abundance of individual taxa, we first ran t-tests of the relative abundance of each taxon between PD and controls at different phylogenetic ranks, and found significant difference in one phylum, four classes, four orders, eight families, 20 genera, and seven species (Figure 2). To verify these findings, we also performed the LEfSe analysis, which couples statistical significance with biological consistency and effect relevance. With LDA threshold set at 4, we identified one class, four orders, three families, and two genera that showed significant different abundances between the PD and control groups (Figure 3). The taxa identified by LEfSe were among the taxa that showed the most significant difference in the t-tests, except for Prevotella_9, which was not significant in the t-test.

To examine if constipation was a confounding factor that caused the differences in abundance of the taxa identified by the LEfSe analysis, we compared the relative abundance of these taxa between a subset of PD patients and controls with Wexner scores smaller than 3 using Wilcoxon rank-sum test (Supplementary Figure S5). We found that Bacilli, Lactobacillales, Bacteroidales_S24_7_group, Lactobacillaceae and Lactobacillus remained significantly different between PD and controls, whereas Ruminococcaceae and Prevotella_9 lost significance. To further rule out the potential impact of gastric distress on these results, we carried out correlation analyses to investigate the relationship between Wexner scores and the individual taxa (Supplementary Figure S6). The correlations between the abundances and the Wexner constipation scores are weak, suggesting that the severity of constipation does not have a strong effect on these measurements. To examine the potential impact of gender on these findings, we again performed two-way ANOVA analysis to test the main effect of gender and PD on the
Altered gut microbiota richness and diversity indices in Parkinson’s disease (PD) patients. (A) Rarefaction curves to show the adequate depth of the sequencing. Data are presented as mean ± standard error of the mean (SEM). n = 51 for PD, n = 48 for control. * p < 0.05, unpaired t-test. (B) Rank abundance curve to show the relative species abundance of the samples from PD patients and controls. Data are presented as mean ± SEM. (C) Species accumulation boxplot to show that the number of samples is also adequate. (D–I) Comparison of species richness (D–F), α-diversity (G–H), and phylogenetic diversity (I) indices between PD patients and controls (n = 51 for PD, n = 48 for control). Data are presented as mean ± SEM with individual replicates also shown. * p < 0.05, ** p < 0.01, *** p < 0.001, unpaired t-test. (J–L) β-diversity analyses using Principal Co-ordinates Analysis (PCoA) plotted with weighted Unifrac (J) and unweighted Unifrac (K), and Analysis of similarities (ANOSIM; L).
FIGURE 2 | Gut microbiota differences between PD patients and controls detected by t-tests. Gut microbiota is compared between PD patients and healthy control subjects at phylum (A), class (B), order (C), family (D), genus (E), and species (F) levels. Only the taxa with statistically significant difference are plotted. The bars on the left side of each figure show the relative abundance. On the right side of each figure, the center of circles represents the difference between the means of the two groups. The error bars represent the 95% confidence interval. P-values of unpaired t-test are listed on the right (n = 51 for PD, n = 48 for control).
abundance of the taxa, and found that the main effect of PD, but not gender, was significant (Supplementary Figure S7).

Relationship of Gut Microbiota With Clinical and Dietary Factors
To reveal the relationships between clinical and dietary factors and the gut microbiota in control and PD patients, we performed Spearman correlation analysis to study the relationship between the clinical and dietary factors and gut microbiota abundance in 42 PD patients and 23 controls. We found that the PD-related clinical scores, such as UPDRS, NMSQ and SCOPA, were the most influential factors that either positively or negatively correlated with the abundance of the taxa at multiple phylogenetic ranks (Figure 4). The taxa that showed the strongest correlation with clinical factors included Bacillales, Lactobacillales, Acidaminococcaceae, Erysipelotrichaceae_1, Phascolarctobacterium, Akkermansia, Coprococcus_2, Pseudomonas_veronii and Ruminococcus_torques (Figure 5). We did not observe the strong correlations between clinical scores and the gut microbiota when the analysis was performed separately in PD patients (Supplementary Figure S8) or controls (Supplementary Figure S9), suggesting that the correlation we observed when all subjects were analyzed together more likely reflected the impact of PD as a disease condition, rather than being specific to individual clinical scores.

DISCUSSION
Our results revealed decreased species richness, phylogenetic diversity and β-diversity of gut microbiota in PD patients. More importantly, we identified several taxa that showed significant difference in abundance between PD patients and healthy controls using t-test and LEfSe analysis. Among these bacteria, some are particularly interesting because they have been linked to human health.

For example, Akkermansia, which was found higher in the gut of PD patients in our study (Figure 2E, consistent with previous findings; Keshavarzian et al., 2015; Unger et al., 2016; Hill-Burns et al., 2017), can degrade the colonic mucus barrier and increase pathogen susceptibility and intestinal permeability (Desai et al., 2016). Seregin et al. (2017b) have speculated that increased Akkermansia abundance may result in thinning of the intestinal mucus barrier, and allow for greater microbial access to the intestinal wall, which leads to the local inflammation driven by commensal organisms. In addition, Akkermansia abundance is negatively correlated with the innate immune receptor NLRP6 (Seregin et al., 2017b), the lack of which results in impaired interleukin-18 production and increased susceptibility to epithelial-induced injury (Chen et al., 2011; Elinav et al., 2011; Couturier-Maillard et al., 2013; Seregin et al., 2017a).

Another interesting finding of this study was the observed increase in the Prevotella genus and Prevotella copri species in PD patients. Increased abundance of Prevotella copri has been found in patients with rheumatoid arthritis (Scher et al., 2013) and HIV (Lozupone et al., 2014; Dillon et al., 2016), suggesting that it might be associated with inflammation. At the same time, Prevotella is involved in mucin synthesis in the gut mucosal layer, possibly contributing to the gut barrier function and the production of neuroactive short-chain fatty acids (SCFAs) through fiber fermentation (Arumugam et al., 2011). It has been recently reported that SCFAs can promote the α-synuclein-mediated neuroinflammation in a mouse model of PD (Sampson et al., 2016). Thus, the elevated gut Prevotella levels may play a role in worsening the PD in human patients. Our LEfSe analysis suggested that the abundance of Prevotella_9 was significantly higher in PD...
FIGURE 4 | Correlations between clinical and dietary factors and gut microbiota when PD patients and controls were analyzed together. Correlation heatmaps showing the relationship between clinical and dietary factors and gut microbiota at different taxonomic ranks. Each row in the heatmap represents a clinical or dietary factor. Each column represents a taxon. The color temperature encodes Spearman correlation coefficient \( r \). \( n = 65 \), * \( p < 0.05 \), ** \( p < 0.01 \).
patients than in controls (Figure 3). However, it was not significant in the Wilcoxon rank-sum test that compared its abundance between the PD patients and controls without severe constipation (Supplementary Figure S5). Furthermore, t-test comparisons (Figure 2E) and the two-way ANOVA (Supplementary Figure S3) assessing the effect of gender and PD on the abundance of *Prevotella_9* did not return statistical significance between PD and controls. *Prevotella copri*, a species within the *Prevotella* genus, showed significant increase in PD patients in t-test (Figure 2F). However, it was not identified as one of the significantly increased taxa in the LEfSe analysis when the threshold of LDA score was set at
4 (Figure 3). The possible increases in the Prevotella genus and Prevotella copri species in PD patients, though reported previously in a study with a large sample size (Hill-Burns et al., 2017), contrast with other studies (Hasegawa et al., 2015; Scheperjans et al., 2015; Unger et al., 2016; Bedarf et al., 2017; Petrov et al., 2017).

One of the most consistent findings in our study supported by both t-tests, LEfSe and the phylogenetic connection was the decreased abundance of Lactobacillales/Lactobacillaceae/Lactobacillus in PD patients. This observation is in accordance with another recent study in Chinese PD patients (Qian et al., 2018) and a study in Germany (Scheperjans et al., 2015), but contrasted with the findings in other studies (Hasegawa et al., 2015; Scheperjans et al., 2015; Hill-Burns et al., 2017; Hopfner et al., 2017; Petrov et al., 2017). These differences may be attributed to racial and regional differences in the PD patients and controls being studied. Local cuisines vary between countries and areas, and different diets lead to significantly different compositions of the gut microbiota (Faith et al., 2011; Rey et al., 2013; David et al., 2014). For example, a staple food regularly consumed in Northeast China, locally called Suan Cai (Sour Cabbage, very similar to sauerkraut) is enriched with Lactobacillus. Regular consumption of this type of food may elevate the gut Lactobacillus abundance in healthy controls, and manifest its decrease in PD patients in Northeast China.

Gut microbiota alterations have recently been reported in Chinese PD patients (Lin et al., 2018; Qian et al., 2018). Some of the findings in these two studies were consistent with our results (e.g., reduced abundance of Lactobacillus in PD patients; Qian et al., 2018), and some were not. This is not too surprising because these two reports were studying the patients from the Southwest (Lin et al., 2018) and Southeast (Qian et al., 2018) of China. Geographic factors, such as ethnicity and local cuisines vary widely across China.

Finally, our correlation analysis revealed strong correlations between clinical factors and the gut microbiota abundances when all subjects were included (Figures 4, 5), but not when the patients and controls were analyzed separately (Supplementary Figures S8, S9). This suggests that the observed correlations between clinical factors and gut taxa are more likely reflecting the impact of PD as a general disease condition, rather than the correlation between specific clinical scores and specific gut microorganisms.

The direction of change in the relative abundance of a few species reported in this study appears to be contradictory to some previous studies in PD patients from other regions. This could be attributed to region-specific factors affecting the gut microbiota of the recruited subjects, or the relatively limited sample size of this study. For example, we only had 23 controls from whom the clinical scores and dietary information were collected. Further assessments of PD patients in Northeast China with larger cohorts are needed to resolve any discrepancies. Another limitation of this study is that the PD patients included in this study were on levodopa/carbidopa treatment, which may have some impact on the gut microbiota (Hill-Burns et al., 2017).

CONCLUSION

Gut microbiota significantly differed between a group of PD patients and healthy controls in Northeast China. PD patients showed decreased species richness, phylogenetic diversity and β-diversity. PD patients also showed altered relative abundance in several microbes compared to the controls. The most consistent finding revealed by both t-tests and LEfSe analysis was the reduction of the lactic acid-producing, potentially beneficial bacteria in PD patients at multiple phylogenetic ranks. A healthy diet or dietary supplements that promote the growth of Lactobacillus could potentially be beneficial in mitigating PD-related pathological changes in the gut.

ETHICS STATEMENT

This study was carried out in accordance with the recommendations of the medical ethics committee of the First Hospital of Jilin University with written informed consent from all subjects. All subjects gave written informed consent in accordance with the Declaration of Helsinki. The protocol was approved by the the medical ethics committee of the First Hospital of Jilin University.

AUTHOR CONTRIBUTIONS

LC, GC and YZ conceived the study. CL and LC collected and analyzed the data. YY, JM, XZ, JZ and GC helped with data analysis. CL, GC and YZ wrote the manuscript with the input from YZ.

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SUPPLEMENTARY MATERIAL

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fnmol.2019.00171/full#supplementary-material

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**Conflict of Interest Statement:** 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.

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