

# Association between post-stroke cognitive impairment and gut microbiota

### A PRISMA-compliant systematic review and meta-analysis

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#### Abstract

**Background:** Accumulating evidence has indicated a possible connection between post-stroke cognitive impairment (PSCI) and gut microbiota imbalance. To further investigate this association, the present work was designed to systematically assess the dissimilarity of gut microbiota between PSCI and healthy individuals or stroke patients.

**Methods:** A meta-analysis and systematic review was conducted by searching various databases including PubMed, Web of Science, Embase, VIP, CNKI, and Wangfang for relevant studies. The pooled outcomes were used to estimate the combined dissimilarity of gut microbiota composition between PSCI and healthy individuals or patients with stroke.

**Results:** Nine eligible studies were included in this meta-analysis. The results showed that there were no significant changes in observed richness indexes (Chao1 and ACE) and Shannon index. Notably, a significant decrease in Simpson index was observed in PSCI patients in comparison to the healthy individuals (-0.31, 95% CI: -0.62 to -0.01, P = 0.04). Moreover, the microbiota composition at the phylum level (increased abundance of *Proteobacteria*), family level (increased abundance of *Bacteroidaceae*; decreased abundance of *Enterobacteriaceae*), and genus level (increased abundance of *Bacteroides*; decreased abundance of *Prevotella* and *Ruminococcus*) was found to be significantly different between PSCI and controls.

**Conclusion:** This meta-analysis suggests a significant shift of observed species and microbiota composition in PSCI compared to healthy individuals or patients with stroke.

Abbreviations: GM = gut microbiota, PSCI = post-stroke cognitive impairment, SMD = standardized mean differences.

Keywords: diversity, gut microbiota, meta-analysis, post-stroke cognitive impairment

#### 1. Introduction

Stroke is one of the leading causes of death and disability worldwide<sup>[1]</sup> with Post-stroke cognitive impairment (PSCI) encompassing any type of cognitive impairment ranging from mild to severe, which may occur following a stroke event.<sup>[2]</sup> Recent studies suggest that PSCI is far more prevalent than previously thought; for example, 10-year stroke survivors have an incidence rate of up to 61%.<sup>[3]</sup> Besides, recurrent ischemic stroke in high-risk patients under adequate pharmacotherapy, such as antiplatelet therapy, could be associated with PSCI. The lesion location and related data are essential for predicting cognitive recovery, indicating the affected neural substrates.<sup>[4-6]</sup> Furthermore, common risk factors like hypertension and diabetes may increase the chances of PSCI ensuing from stroke.<sup>[7]</sup>

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However, due to the lack of scale application, early symptoms are often overlooked, caused by the absence of diagnosis and treatment. Therefore, it is of great importance to identify potential biomarkers for early detection and diagnosis of PSCI, as it usually takes 3 months or more before onset.<sup>[8]</sup>

Recent research has revealed that the gut microbiota (GM) structure of patients with neuropsychiatric disorders can be compromised.<sup>[9,10]</sup> Studies have demonstrated that the fecal microbial diversity and composition of Alzheimer Disease patients are significantly different from those of healthy controls.<sup>[11]</sup> Jiang et al discovered that the GM structure of patients with active severe depression was altered, with a notable elevation of *Bacteroidetes*, *Proteobacteria*, *Bacteroidetes*, and *Enterobacteriaceae*, and a marked decrease in *Firmicutes* and

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Faecalibacterium.<sup>[12]</sup> Moreover, following stroke, the signature of GM ecological disruption has been identified as an increased abundance of opportunistic pathogenic bacteria and decreased levels of butyrate-producing bacteria.<sup>[13]</sup> Accumulating evidence suggests that GM plays an important role in cognitive impairment in various disorders,<sup>[14–16]</sup> and GM has been confirmed as a noninvasive biomarker for disease diagnosis.

Recently, researchers have observed a reduction in the a diversity and abundance of certain microbial populations in patients with PSCI when compared to healthy control subjects or stroke patients.<sup>[17-20]</sup> However, due to conflicting reports, more investigation is necessary to explore the relationship between the GM and PSCI. To further understand the possible contribution of GM to the development of PSCI, a

meta-analysis was conducted to analyze the changes of microbial populations at different levels between PSCI patients and healthy controls or stroke patients. The aim of this study was to evaluate the degree of agreement of the changes in GM and to identify those microbial populations that may serve as potential biomarkers.

#### 2. Materials and Methods

#### 2.1. Search strategy

A rigorous search was conducted until February 2023 in PubMed, Web of Science, Embase, Cochrane databases, Wangfang, VIP and CNKI without language restrictions, using



Figure 1. Flow chart of the search strategy and study selection progress.

Table 1 Details of	the elig	lible studie	s in this	; meta-ana	ılysis.														
First author	۲	Province	Exp.	Ctr.	Total number	cExp.	cCtr.	Male	Female	Smoke	Alcohol	Diabetes	Hypertension	Age	BMI	MoCA	Country	16S region	Seq. Tech.
Yinting	2021	Guang-	PSCI	nPSCI	56	29	27	34	22	NA	NA	31.03	58.62	62.76	NA	5.38	China	V3V4	MiSeq Benchtop
Huang Yongqiang	2020	dong Shanghai	PSCI	nPSCI	65	30	35	49	16	NA	NA	NA	NA	64.90	23.41	17.46	China	V3V4	sequencer Illumina MiSeq
Liu Huidi Wang	2022	Guang-	PSCI	nPSCI	83	34	49	68	15	44.10	32.40	35.30	70.60	61.50	23.70	NA	China	V4	Illumina iSeq
Feng	2021	dong Sichuan	PSCI	HC	47	24	23	26	21	54.17	45.83	NA	NA	63.17	24.73	13.25	China	Whole 16S	1 UU PacBio Sequel
Rongjian Li Yamei	2022	Sichuan	PSCI	nPSCI(HC)	36	12	12 (12)	20	16	NA	NA	NA	NA	60.75	25.20	19.92	China	region Whole 16S	PacBio Sequel
Li Yonghua Liu Zhirong	2022 2021	Shandong Sichuan	PSCI PSCI	nPSCI(HC) HC	67 47	29 24	18 (20) 23	34 26	33 21	27.60 50.00	17.20 45.80	37.90 NA	44.80 NA	68.50 63.01	NA 25.34	13.30 13.25	China China	region V3-V4 Whole 16S	MiSeq PacBio Sequel
Song Xinna	2021	Guang-	PSCI	nPSCI	141	120	21	NA	NA	NA	NA	NA	NA	NA	NA	NA	China	region NA	MiSeq
Yi Ling	2020a	dong Zhejiang	PSCI	nPSCI	93	53	40	62	31	17.00	41.50	24.50	77.40	72.20	24.70	13.70	China	V3-V4	MiSeq Benchtop
Yi Ling	2020b	Zhejiang	PSCCID	nPSCCID	66	41	25	31	35	51.20	NA	29.30	58.50	69.63	25.14	13.17	China	V3-V4	Sequencer MiSeq Benchtop
Feng Dan	2022	Sichuan	PSCI	НС	39	20	19	22	17	NA	NA	NA	NA	61.25	24.96	12.15	China	Whole 16S	sequencer PacBio
Du Jun	2022	Jiangsu	VCIND	НС	60	40	20	29	31	35.00	42.50	NA	NA	61.73	NA	22.08	China	region V3-V4	NA
BMI = body ma impairment, Se	tss index, cl q. Tech = s	Ctr = cases in cv equencing techr	ontrol group iique.	), Ctr = control ç	group, Exp = ex	kperimental	group, HC =	healthy cc	introl, MoCA	= montreal c	cognitive asses	ssment, NA = ni	ot available, nExp = c	ases in expe	eriment gro	up, nPSCI =	= non-PSCI, PS	CI = post-stroke (	ognitive

the search terms "stroke," "cognitive impairment," and either "gut" or "intestinal," to identify studies eligible for inclusion in a meta-analysis. The Preferred Reporting Items for Systematic Reviews and Meta-analyses criteria was used as the guideline for this meta-analysis.

#### 2.2. Inclusion and exclusion criteria

Studies relevant to PSCI and GM were identified using the following criteria: comparison of gut microbial composition between PSC and healthy control subjects or stroke patients; availability of adequate data for consolidated comparison; and access to the full-text. Reviews, animal studies, and conference summaries that did not meet these criteria were excluded.

#### 2.3. Data extraction and quality assessment

The authors, date of publication, country, sample size, and detection methods of all studies included in the analysis are presented herein. The data from 16S rRNA-sequencing and the relative abundances of GM were abstracted independently by 2 reviewers, and any disagreements were resolved by consensus. The quality assessment of the studies included was conducted according to the criteria recommended by the Cochrane Non-Randomized Studies Methods Group, with special consideration given to selection, comparability, and outcome details.<sup>[21]</sup>

#### 2.4. Statistical analysis

All statistical analyses were conducted using STATA SE 15 software, with data extraction executed by WebPlotDigitizer tool. The standardized mean differences (SMD) were applied to compare the richness and diversity indexes, as well as GM between the PSCI and control groups. Heterogeneity was assessed by  $I^2$  statistic, and a fixed-effect model was employed when  $I^2 > 50\%$ , otherwise, the random-effect model was used.

#### 3. Results

#### 3.1. Characteristics of included studies

As presented in Figure 1, a total of 185 relevant articles were identified in 6 databases, with 54 of them being duplicates. After a rigorous assessment of titles and abstracts, 119 articles were eliminated due to not meeting the criteria, resulting in 12 eligible studies<sup>[17-20,22-29]</sup> included in the meta-analysis. Consequently, the analysis included 800 subjects, 456 of whom were suffering from PSCI and 344 in the control group. The pertinent

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Meta-analysis of	diversity of	aut microbiota	of patients	with PSCI.
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characteristics of the eligible articles are displayed in Table 1. All 12 selected researches were conducted in 6 provinces in China, with 6 of them being published in Chinese. The microbiota of the eligible studies was evaluated with high-throughput sequencing of the V4 region, V3 to V4 region, or the entire length of the 16S rRNA gene. The assessment of the methodological quality revealed that eleven studies had excellent quality<sup>[17–20,22–24,26–29]</sup> and one had decent quality.<sup>[25]</sup>

## 3.2. Differences in richness and microbial diversity between PSCI and controls

A meta-analysis of 10 studies was carried out for quantifying the pooled disparities in the general characteristics of high-throughput sequencing between PSCI and control groups. The indices of richness (Chao1, ACE) and alpha diversity (Shannon and Simpson) were then assessed. The summary of the meta-analysis results can be seen in Table 2. A total of 294 individuals from 7 studies were examined to compare the ACE index of 16S rRNA sequencing results between PSCI patients and control groups. A non-significant difference was observed between PSCI and the control group (SMD, 0.10, 95% CI: -0.55-0.74, P = .770) as per the random-effect model, due to high heterogeneity ( $I^2 = 85.8\%$ ). When compared to healthy individuals (SMD, -0.3, 95%) CI: -1.22 to 0.43, P = .350) or stroke patients (SMD, 0.75, 95% CI: -0.11 to 1.60, P = .08), no considerable variation was found in ACE index of PSCI patients. Data from 8 studies on Chao1 indicated an insignificant difference between PSCI and controls (SMD, -0.31, 95%CI: -0.77 to 0.15, P = .191). Compared to the healthy group, PSCI patients showed lower Chao1 (SMD, -0.57, 95%CI: -0.93 to -0.21, P = .002), though no important discrepancy was noticed when PSCI patients were compared to stroke patients (SMD, 0.04, 95% CI: -0.84 to 0.92,  $\hat{P} = .927$ ). Data from ten studies with 485 participants on Shannon demonstrated no considerable variation between PSCI and the control group (SMD, 0.08, 95% CI: -0.31 to 0.47, P = .680), while no remarkable difference was observed either between PSCI patients and healthy individuals (SMD, -0.12, 95%CI: -0.60 to 0.37, P = .639) or stroke patients (SMD, 0.29, 95%CI: -0.35 to 0.94, P = .375). Nine studies provided Simpson index data and the pooled effect was calculated using a fixed-effect model due to the low heterogeneity. The results revealed a significant discrepancy between PSCI and controls (SMD, -0.21, 95%CI: -0.40 to -0.02, P = .034). Compared to the healthy group, PSCI patients had significantly lower Simpson index (SMD, -0.31, 95%CI: -0.62 to -0.01, P = .042), while no considerable variation was seen between PSCI and stroke patients (SMD, -0.12, 95% CI: -0.38 to 0.11, P = .279).

meta-analy	sis of diversity of gut mid	crobiota of patients	with PSCI.				
Terns	Number of studies	Participants	l <sup>2</sup>	Р	Effect [95%CI]	Z	Р
ACE	7	294	85.8	<.001	0.10 [-0.55, 0.74]	0.293	.770
HC	4	167	84.9	<.001	-0.39 [-1.22, 0.43]	-0.934	.350
nPSCI	3	127	79.5	<.001	0.75 [-0.11, 1.60]	1.718	.08
Chao 1	8	464	77.9	<.001	-0.31 [-0.77, 0.15]	-1.307	.191
HC	4	178	25.3	.260	-0.57 [-0.93, -0.21]	-3.128	.002
nPSCI	4	286	88.3	<.001	0.04 [-0.84, 0.92]	0.091	.927
Shannon	10	485	76.9	<.001	0.08 [-0.31, 0.47]	0.412	.680
HC	5	258	68.1	.014	-0.12 [-0.60, 0.37]	-0.469	.639
nPSCI	5	227	83.9	<.001	0.29 [-0.35, 0.94]	0.888	.375
Simpion	9	530	0	.626	-0.21 [-0.40, -0.02]	-2.117	.034
HC	4	178	0	.542	-0.31 [-0.62, -0.01]	-2.030	.042
nPSCI	5	352	0	.520	-0.12 [-0.38, 0.11]	-1.082	.279

HC = healthy control, nPSCI = non-PSCI, PSCI = post-stroke cognitive impairment.

#### 3.3. Differences in the microbial composition

**3.3.1.** *Phylum level.* Analysis of *Bacteroidetes* in 273 individuals revealed no significant difference in abundance between PSCI and control groups (SMD, -0.29, 95%CI: -1.58 to 0.99, P = .653, Fig. 2A). *Verrucomicrobiota* exhibited significantly increased relative abundance in PSCI compared to control groups (SMD, 0.01, 95%CI: -0.54 to 1.06, P = .934, Fig. 2B). Random-effect remodel was utilized to assess pooled differences in *Bacteroidetes* between PSCI and controls, with no significant disparity observed (SMD = -0.09, 95%CI: -0.39 to 0.21, P = .566, Fig. 2C). Similarly, no significant difference was determined in relative abundance of *Firmicutes* between PSCI and controls (Fig. 2D). Seven studies provided data on *Proteobacteria* abundance,

and the meta-analysis indicated a significantly higher relative abundance in PSCI than stroke patients (P = .006) and no significant difference from healthy individuals (Fig. 2E).

**3.3.2.** Family level. Table 3 presents the results of a metaanalysis on the levels of family in the GM of PSCI patients. Results indicated that, compared with the control group, there was a significant increase in the abundance of *Bacteroidaceae*, *Lachnospiraceae*, and *Veillonellaceae* in the GM, while the abundance of *Enterobacteriaceae* was significantly reduced, and the other families showed no significant difference. Further comparisons between the levels of family in the GM of PSCI patients and that of healthy people and stroke patients were made.

		%	8		
CTR and Author (Year)	Effect (95% CI)	Weight	CTR and Author (Year)	Effect (95% CI)	Wei
Yongqiang Liu (2020)	-0.90 (-1.41, -0.38)	17.14	нс		
ri Ling (2020b)	-3.93 (-4.77, -3.09)	16.38	Feng Rongjian (2021)	-0.19 (-0.76, 0.39)	3
₋i Yamei (2022)	<b>1.42 (0.52, 2.33)</b>	16.21	Liu Zhirong (2021)	0.26 (-0.32, 0.83)	3
PSCI 2			Li Yamei (2022)	-0.49 (-1.31, 0.32)	1
Subgroup, DL (1 <sup>2</sup> = 97.3%, p = 0.000)	-1.14 (-3.74, 1.46)	49.71	Subgroup, IV (I <sup>2</sup> = 18.1%, p = 0.295)	-0.07 (-0.43, 0.29)	8
Feng Rongjian (2021)	<ul> <li>0.24 (-0.33, 0.81)</li> </ul>	17.02			
iu Zhirong (2021)	- 0.07 (-0.51, 0.64)	17.02	nPSCI		
.i Yamei (2022)	1.36 (0.46, 2.25)	16.23	Li Yamei (2022)	0.26 (-0.54, 1.06)	1
HC	0 47 (-0 19 1 13)	50.28	Subgroup, IV (I <sup>2</sup> = 0.0%, p = .)	0.26 (-0.54, 1.06)	1
subgroup, DE (1 = 00.476, p = 0.001)	0.47 (-0.13, 1.10)	50.20	Hataraganaitu batusan araunau n 0.464		
Heterogeneity between groups: p = 0.240 Dverall, DL (I <sup>2</sup> = 95.4%, p = 0.000)	-0.29 (-1.58, 0.99)	100.00	Overall, IV ( $I^2 = 0.0\%$ , p = 0.395)	-0.01 (-0.34, 0.32)	10
-5 0	5		-1 (	1	
			D		
		%			
CTR and Author (Year)	Effect (95% CI)	Weight	CTR and Author (Year)	Effect (95% CI)	We
'ongqiang Liu (2020)	0.29 (-0.20, 0.78)	14.36	Yongqiang Liu (2020) 📃 🔳	-0.46 (-0.96, 0.03)	15
(i Ling (2020a)	-0.23 (-0.65, 0.18)	16.17	Yi Ling (2020a)	0.00 (-0.41, 0.41)	2
(i Ling (2020b)	0.46 (-0.05, 0.96)	14.06	Yi Ling (2020b)	-0.35 (-0.85, 0.16)	15
Song Xinna (2021)	0.12 (-0.41, 0.66)	13.30	Song Xinna (2021)		12
i Vamei (2022)	-0.11(-0.92,0.69)	8 78	Li Vamei (2022)	-0.19 (-0.99, 0.62)	5
PSCI	-0.11 (-0.52, 0.05)	0.70	nPSCI	-0.19 (-0.99, 0.02)	Ű
Subgroup, DL (I <sup>2</sup> = 24.4%, p = 0.259)	0.11 (-0.16, 0.37)	66.67	Subgroup, IV (I <sup>2</sup> = 56.0%, p = 0.059)	-0.09 (-0.32, 0.14)	71
eng Rongijan (2021)	-0.94 (-1.55, -0.34)	11.99	Feng Rongijan (2021)	-0.27 (-0.84, 0.31)	11
iu Zhirong (2021)	-0.21 (-0.78, 0.37)	12 59	Liu Zhirong (2021)	-0.18 (-0.76, 0.39	11
i Vernei (2022)	-0.21 (-0.78, 0.57)	0.74		-0.18 (-0.76, 0.39	, ,
	-0.20 (-1.07, 0.04)	0.74		-0.06 (-0.87, 0.74)	5
Subgroup, DL (I <sup>2</sup> = 40.9%,	-0.49 (-0.98, 0.00)	33.33	Subgroup, IV ( $I^2 =$	-0.19 (-0.55, 0.17)	28
p = 0.184)			0.0%, p = 0.920)		
Heterogeneity between groups: p = 0.037			Heterogeneity between groups: p = 0.629		
Overall, DL (1 <sup>2</sup> = 55.7%, p = 0.027)	-0.09 (-0.39, 0.21)	100.00	Overall, IV (I <sup>2</sup> = 26.2%, p = 0.220)	-0.12 (-0.31, 0.08)	10
	2		-1	0 1	
L	CTR and Author (Year)		% Effect (95% CI) Weight		
	Yongqiang Liu (2020)	· · ·	1.37 (0.82, 1.91) 15.01		
	Yi Ling (2020a)		0.70 (0.28, 1.13) 15.70		
	Yi Ling (2020b)		2 51 (1 85 3 17) 14 24		
	Li Vamoi (2022)				
	nPSCI		0.26 (-0.54, 1.07) 13.23		
	Subgroup, DL (I <sup>2</sup> = 88.4%, p = 0.0	000) <	1.21 (0.36, 2.07) 58.18		
	Feng Rongjian (2021)		<u> </u>		
	Liu Zhirong (2021)		0.39 (-0.19, 0.97) 14 79		
	Li Yamei (2022)		0.33 (-0.48 1.14) 13.21		
	Li rundi (2022)		0.00 (-0.40, 1.14) 10.21		
	HC Subgroup, DL ( $1^2 = 87.4\%$ , p = 0.0	000)	0.94 (-0.19, 2.08) 41.82		
	Heterogeneity between groups: p	= 0.710			
	Heterogeneity between groups: p = Overall, DL ( $I^2 = 86.0\%$ , p = 0.000	= 0.710	1.10 (0.48, 1.73) 100.00		

Figure 2. Forest plot of meta-analysis of gut mcrobita in post-stroke cognitive impairment (PSCI) at the phylum levels. (A) Bacteroidetes, (B) Verrucomicrobiota, (C) Bacteroidetes, (D) Firmicutes, (E) Proteobacteria.

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Meta-analysis of gut microbiota of PSD at the family levels.

Gut microbiota	Number of studies	Participants	<b> </b> <sup>2</sup>	Р	Effect [95%CI]	Z	Р
Bacteroidaceae	4	289	88.0	<.001	-1.15 [-1.90, -0.39]	-2.985	.003
НС	1	71	0	NA	-0.13 [-0.70, 0.44]	-0.452	.651
nPSCI	3	218	75.0	.018	-1.47 [-2.06, -0.90]	-4.942	0
bifidobacteriaceae	4	289	85.0	<.001	-0.46 [-1.09, 0.17]	-1.445	.149
НС	1	71	0	NA	-0.53 [-1.36, 0.30]	-0.885	.376
nPSCI	3	218	89.6	<.001	-0.26 [-0.83, 0.31]	-1.245	.213
Enterobacteriaceae	4	289	97.0	<.001	2.91 [0.92, 4.90]	2.862	.004
HC	1	71	0	NA	0.23 [-0.34, 0.80]	0.788	.431
nPSCI	3	218	91.0	<.001	3.79 [2.34, 5.24]	5.131	0
Lachnospiraceae	4	289	38.3	.182	-0.57 [-0.81, -0.33]	-4.658	0
HC	1	71	0	NA	-0.29 [-0.87, 0.28]	-1.004	.315
nPSCI	3	218	47.3	.150	-0.63 [-0.89, -0.36]	-4.664	0
lactobacillaceae	3	242	97.8	<.001	0.63 [-1.22, 2.48]	0.672	.502
HC	1	71	0	NA	0.04 [-0.53, 0.61]	0.145	.884
nPSCI	2	171	98.5	<.001	0.84 [-1.79, 3.47]	0.627	.53
Prevotellaceae	4	289	97.5	<.001	1.47 [-0.47, 3.42]	1.483	.138
nPSCI	4	289	97.5	<.001	1.47 [-0.47, 3.42]	1.483	.138
Ruminococcaceae	4	289	68.7	.023	0.11 [-0.32, 0.54]	0.509	.611
HC	1	71	0	NA	-0.35 [-0.93, 0.23]	-1.195	.232
nPSCI	3	218	65.7	.054	0.24 [-0.20, 0.69]	1.074	.283
streptococcaceae	3	242	96.3	<.001	1.35 [-0.08, 2.78]	1.851	.064
HC	1	71	0	NA	0.25 [-0.33, 0.82]	0.85	.396
nPSCI	2	171	97.1	<.001	1.72 [-0.14, 3.58]	1.811	.07
Veillonellaceae	3	242	0	.470	-0.27 [-0.53, -0.01]	-2.047	.041
nPSCI	3	242	0	.470	-0.27 [-0.53, -0.01]	-2.047	.041

CI = confidence interval, HC = healthy control, NA = not available, nPSCI = non-PSCI, PSCI = post-stroke cognitive impairment.

Results suggested that there was no significant difference between PSCI patients and healthy people in terms of the abundance of the family, while compared with stroke patients, a significant increase in *Bacteroidaceae*, *Lachnospiraceae*, and *Veillonellaceae*, and a significant decrease in *Enterobacteriaceae* were observed.

**3.3.3. Genus level.** Table 4 presents the results of the metaanalysis at the genus level. Compared to the control group, significantly increased abundances of *Bacteroides*, *Clostridium* XIVa, and *Parabacteroides* were observed in PSCI patients, while significantly decreased abundances of Blautia, *Prevotella*, and *Ruminococcus* were detected. In contrast, compared to the healthy group, no significant differences in the abundances of all genera were found in PSCI patients. Additionally, compared to the stroke patients, significantly decreased abundances of *Bacteroides*, *Clostridium* XIVa, and *Parabacteroides*, as well as significantly increased abundances of *Prevotella* and *Ruminococcus* were observed in PSCI patients.

#### 4. Discussion

Recently, reports have indicated changes in the composition of the GM and disruptions to the intestinal metabolic process in those suffering from PSCI, potentially influencing brain activity through the microbiota-gut-brain axis.<sup>[30]</sup> Nonetheless, the results of various studies have been discordant. In order to include as much information as possible, we determined the abundance of microbial communities instead of the raw datasets.<sup>[31]</sup> In this systematic review, we used 12 case-control studies to compare the gut microbial communities of PSCI and healthy individuals, or of stroke patients. We observed a significantly decreased Simpson index, suggesting reduced diversity in the GM of PSCI patients, although the other indices showed no significant variations. At the phylum level, Proteobacteria was found to be significantly more abundant in PSCI patients. At the family level, Bacteroidaceae, Lachnospiraceae, and Veillonellaceae were significantly increased, while Enterobacteriaceae was significantly decreased. At the genus level, Bacteroides, Clostridium XIVa,

and *Parabacteroides* were significantly more abundant in PSCI patients, while Blautia, *Prevotella*, and *Ruminococcus* were significantly decreased.

Dynamic alterations in the composition and metabolic byproducts of the gut microbiome are influenced by both internal and external factors. Recent studies have found that it may not only have an indirect impact on cerebral infarcts,<sup>[32]</sup> but may also directly influence their occurrence, progression and prognosis.[33] This gut microbiome-gut-brain axis, the bi-directional interaction mechanism between the gut microbiome and the brain, encompasses the vagus nerve, endocrine, metabolic and immunologic pathways, and facilitates mutual communication.<sup>[34]</sup> In 2020, initial reports indicated the changes in the gut microbiome of patients with PSCI, demonstrating a notably modified diversity and relative abundance in comparison to those individuals without cognitive deficits after stroke, thus indicating a possible relationship between gut microbiome dysfunction and PSCI.<sup>[26]</sup> Nonetheless, the characteristics of the gut microbiome disruptions associated with PSCI remain unknown, prompting further investigation into the changes of the gut microbiome in patients exhibiting early cognitive decline after stroke, and their potential as markers for early recognition, intervention and prognosis of PSCI.

*Proteobacteria* are present in different parts of the human body, such as the oral cavity, skin, gastrointestinal tract and vagina, with their diverse shapes and physiological functions. Its unique oxygen requirement keeps it in the gastrointestinal environment, and the imbalance of intestinal microecology can lead to an increase in the number of *Proteobacteria*, which can be observed in post-neonatal gastroenteric diversion, metabolic disorders or obese patients. In addition, *Proteobacteria* have also been shown to induce insulin resistance and obesity in mouse models, and may constitute a risk factor for cognitive impairment after stroke, and an increase in *Proteobacteria* can be observed in the intestinal environment of elderly and autistic children. However, different results were obtained in the studies of Alzheimer mouse models regarding the changes of *Proteobacteria*.

Table 4

Meta-analysis of gut microbiota of PSD at the genus levels.

Gut microbiota	Number of studies	Participants	l <sup>2</sup>	Р	Effect [95%CI]	z	Р
Akkermansia	4	152	93.2	<.001	-0.28 [-1.66, 1.09]	-0.404	.686
НС	2	63	91.5	.001	0.00 [-1.84, 1.85]	0.003	.997
nPSCI	2	89	96.3	<.001	-0.56 [-3.23, 2.12]	-0.408	.683
Bacteroides	9	463	96.3	<.001	1.43 [0.25, 2.61]	2.369	.018
НС	3	112	95.9	<.001	0.48 [-1.64, 2.60]	0.444	.657
nPSCI	6	351	96.8	<.001	1.91 [0.38, 3.43]	2.453	.014
Bifidobacterium	8	407	96.6	<.001	-0.88 [-2.24, 0.48]	-1.266	.206
HC	3	112	93.4	<.001	-1.16 [-2.85, 0.53]	-1.345	.179
nPSCI	5	295	97.7	<.001	-0.70 [-2.80, 1.39]	-0.658	.511
Blautia	5	183	94.1	<.001	-1.98 [-3.54, -0.42]	-2.482	.013
НС	3	112	94.6	<.001	-1.75 -3.81, 0.31	-1.665	.096
nPSCI	2	71	96.4	<.001	-2.33 [-5.80, 1.14]	-1.317	.188
Clostridium XIVa	3	224	92.9	<.001	1.12 [0.01, 2.24]	1.973	.048
nPSCI	3	224	92.9	<.001	1.12 [0.01, 2.24]	1.973	.048
Enterococcus	3	135	98.8	<.001	-4.20 [-10.80, 2.40]	-1.247	.212
HC	2	88	99.0	<.001	-2.31 [-10.09, 5.47]	-0.582	.561
nPSCI	1	47	0	NA	-8.03 [-9.78, -6.27]	-8.941	0
Escherichia/Shigella	8	416	94.9	< .001	-0.34 [-1.33, 0.66]	-0.666	.505
HC	3	112	97.7	<.001	-2.36 [-5.83, 1.12]	-1.329	.184
nPSCI	5	304	90.5	<.001	0.58 [-0.23, 1.38]	1.406	.16
Fubacterium	5	183	96.9	< 001	1.77 [-0.43, 3.98]	1.573	.116
HC	3	112	97.0	< 001	0.80 [-1.87, 3.48]	0.588	.556
nPSCI	2	71	97.8	< .001	3.30 [-2.19, 8.79]	1.177	239
Faecalibacterium	8	424	97.8	< .001	1.27 [-0.59, 3.12]	1.34	.18
HC	2	73	97.6	< .001	1.27 [-2.61, 5.15]	0.641	.521
nPSCI	6	351	98.2	< .001	1.27 [-1.07, 3.61]	1.063	.288
Klebsiella	5	320	98.3	< .001	-1.56 [-4.11, 0.99]	-1.2	.23
HC	1	49	0	NA	-6.94 [-8.45, -5.42]	-8.982	0
nPSCI	4	271	97.5	< 001	-6.94 [-8.45, -5.42]	-0.173	863
Lachnoclostridium	4	225	94.4	< 001	_0.28 [_1.42_0.86]	-0.483	629
HC	1	49	0	NA	-0.93 [-1.53 -0.33]	-3.029	0020
nPSCI	3	176	95.0	< 001	-0.07 [-1.56, 1.43]	-0.087	93
Parabacteroides	3	187	4 9	350	2.30 [1.92.2.69]	11 71	0
nPSCI	3	187	4 9	.000	2.30 [1.92, 2.69]	11 71	0
Prevotella	4	280	98.4	< 001	-2.60[-5.19] -0.01]	-1.965	049
nPSCI	4	280	98.4	< 001	-2 60 [-5 19 -0 01]	-1.965	049
Roseburia	3	200	93.4	< 001	$0.63 [-0.47 \ 1.73]$	1 117	264
nPSCI	3	224	93.4	< 001	0.63 [-0.47, 1.73]	1 117	264
Ruminococcus	6	250	98.2	< 001	-2.52 [-4.95 -0.09]	-2 03/	.204
	2	73	00.2	< 001	1 07 [ 11 61 0 47]	0 100	.042
nDSCI	2	286	07.7	< 001	3 21 [ 5 54 0 87]	2 602	.042
Strentococcus	7	200	96.5	< 001	-0.51 [-1.88 0.85]	-0.737	.007
	2	110	07.4	< 001		0.737	104.
nPSCI	Л	11Z 220	97.4 Q6 2	<.001	-0.00 [-0.00, 2.04] _0 /2 [_2 11 1 27]	-0.431	000. aca
Subdoligranulum	<del>ч</del> Б	192	0.00	< 001		-0.407 0 520	.020
	2	100	90.0 06.3	<.001	0.02 [-1.44, 2.00]	1 220	.000
nDCCI	0	71	90.3 07.2	<.001		0.659	.210
11501	Z	/ 1	91.3	<.001	2.03 [-1.00, 7.30]	000.0-	IIC.

CI = confidence interval, HC = healthy control, NA = not available, nPSCI = non-PSCI, PSCI = post-stroke cognitive impairment.

Our study of GM demonstrated a marked transition from a high prevalence of Enterobacteriaceae to a cohort of organisms that is enriched with Bacteroidaceae, which is typically associated with a healthy microbiome.[35] Gramnegative members of the Enterobacteriaceae family are characterized by the presence of lipopolysaccharide in the outer membrane, which can trigger an immune response through Toll-like receptor transduction pathways.<sup>[36,37]</sup> Inflammatory conditions of the gut can be beneficial for the proliferation of Enterobacteriaceae pathobionts, as they can use host-derived factors for anaerobic respiration and outcompete strictly anaerobic microbes that inhabit the gut.[38] Increase abundance of Enterobacteriaceae can further fuel a dysbiosis, resulting in an inflammatory status of the gut epithelium.<sup>[39]</sup> Lachnospiraceae family are known for producing beneficial butyrate salts, which can be beneficial for the integrity of intestinal epithelial cells.<sup>[40]</sup> Deficiencies of Lachnospiraceae may lead to exacerbated intestinal inflammation, increased

production of toxins, and impaired intestinal epithelial barrier.<sup>[41]</sup> It has also been reported that a decrease in *Lachnospiraceae* abundance is associated with longer duration of Parkinson disease and decreased cognitive ability.<sup>[41]</sup> On the other hand, a higher abundance of *Veillonellaceae* has been linked with greater severity of schizophrenia<sup>[42]</sup> as well as a worse prognosis of cancer immunotherapy.<sup>[43]</sup>

By analyzing the abundance at the genus level, we can achieve a better understanding of the composition differences of the GM in patients with PSCI, as these genera and species have been widely accepted as having a close correlation with human health. Among them, *Bacteroides* plays a critical role in the health of the host, with impairment of the normal microecological balance of the host potentially triggering endogenous infections or colitis.<sup>[44]</sup> Previously, it had been hypothesized that *Clostridium* XIVa could produce butyrate,<sup>[45]</sup> which may help suppress systemic inflammatory responses. *Parabacteroides* merdae has been found to protect against cardiovascular damage through an increase in branched-chain amino acid catabolism.<sup>[46]</sup> Moreover, Parabacteroides distasonis has been observed to alleviate obesity and metabolic dysfunctions via the production of succinate and secondary bile acids.<sup>[47]</sup> The lipopolysaccharide derived from Parabacteroides goldsteinii has anti-inflammatory properties and has been reported to significantly ameliorate chronic obstructive pulmonary disease through acting as an antagonist of toll-like receptor 4 signaling pathway.<sup>[48]</sup> The oral ingestion of Blautia wexlerae has been observed to effect metabolic alterations and anti-inflammatory activities in mice, which has resulted in a reduction of both high-fat diet-induced obesity and diabetes.<sup>[49]</sup> Additionally, studies have indicated that a lack of gut Ruminococcus can lead to the development of antibiotic-associated diarrhea.<sup>[50]</sup> Not withstanding, the abundance and functions of different genera or species may vary across different diseases, and although these species' functions have been partially elucidated, their roles in the development of PSCI remain to be further explored.

Despite the numerous advantages of the present meta-analysis, it has certain limitations. Firstly, there is a marked statistical heterogeneity among the included studies, which is hard to be explained by differences in sample size, geographic area, study methods, etc, due to the small number of studies. Secondly, it is challenging to acquire the primary data from all the included studies, and the application of a digital extraction method could potentially lead to another bias in the findings. Moreover, our study merely addressed the structure and composition of intestinal microbiota, without delving into transcriptome and proteome studies, which could potentially provide a deeper understanding of the functionalities of intestinal microbiota. All these issues should be tackled in future research.

#### 5. Conclusion

In this study, we discovered distinct microbiota distributions in the gut of patients with PSCI compared to healthy controls or stroke patients at the phylum, family and genus levels. The Simpson index was markedly lower in PSCI patients than the control group, without any significant deviation from the norm with regards to richness and diversity indices. Nonetheless, due to the limited sample size and number of studies, as well as considerable disparity among the sampled population, we cannot yet generalize these results to a greater population. To reinforce these findings, further high-quality studies are needed.

#### **Author contributions**

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- Data curation: Xiaozhen Hu, Yajun Mao, Fang Luo. Formal analysis: Xiaozhen Hu, Fang Luo. Funding acquisition: Xiaozhen Hu, Xijun Wang. Investigation: Xiaozhen Hu. Methodology: Xiaozhen Hu. Project administration: Xiaozhen Hu. Resources: Xiaozhen Hu. Software: Xiaozhen Hu, Yajun Mao.
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