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Gut Microbiota and Subjective Memory Complaints in Older Women

Fen Wu¹,
Samuel Davey¹,
Tess V. Clendenen,
Karen L. Koenig,
Yelena Afanasyeva,
Boyan Zhou,
Sukhleen Bedi,
Huilin Li,
Anne Zeleniuch-Jacquotte,
Yu Chen*
Department of Population Health, New York University School of Medicine, New York, NY, USA

Abstract

Background: Epidemiological studies that investigate alterations in gut microbial composition associated with cognitive dysfunction are limited.

Objective: To examine the association between the gut microbiota and subjective memory complaints (SMCs), a self-reported, validated indicator of cognitive dysfunction.

Methods: In this cross-sectional study of 95 older women selected from the New York University Women's Health Study (NYUWHS), we characterized the gut microbial composition using 16S rRNA gene sequencing. We estimated odds ratio (OR) from beta regression which approximates the ratio of mean relative abundances of individual bacterial taxon from phylum to genus levels by binary (2+ versus < 2) and continuous SMCs.

Results: Women reporting 2 or more SMCs had higher relative abundances of genus *Holdemania* and family *Desulfovibrionaceae* compared with those reporting one or no complaint. Compared with women with < 2 SMCs, the relative abundances of *Holdemania* and family *Desulfovibrionaceae* were 2.09 times (OR: 2.09, 95% confidence interval [CI]: 1.38–3.17) and 2.10 times (OR: 2.10, 95% CI: 1.43–3.09) higher in women with 2+ SMCs, respectively (false discovery rate (FDR)-adjusted p = 0.038 and 0.010, respectively). A dose-response association was observed for genus *Sutterella* and family *Desulfovibrionaceae*. Every one-unit increase in

Correspondence to: Yu Chen, Departments of Population Health and Environmental Medicine, New York University School of Medicine, 180 Madison Ave, Room 514, New York, NY 10016, USA. Tel.: +1 646 501 3596; Fax: +1 212 263 8570; Yu.Chen@nyulangone.org.

¹These authors contributed equally to this work.

SUPPLEMENTARY MATERIAL

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SMCs was associated with 25% and 27% higher relative abundances of *Sutterella* (OR: 1.25; 95% CI: 1.11–1.40) and *Desulfovibrionaceae* (OR: 1.27; 95% CI: 1.13–1.42), respectively (FDR-adjusted p = 0.018 and 0.006, respectively).

Conclusion: Our findings support an association between alterations in the gut bacterial composition and cognitive dysfunction.

Keywords

16S RNA gene sequencing; cross-sectional; gut microbiota; NYUWHS; subjective memory complaints

INTRODUCTION

Alzheimer's disease (AD), the most common form of dementia, affects an estimated 10% of adults over 65 in the United States [1]. AD and AD-related dementias (AD/ADRD) constitute a considerable public health challenge because of the current lack of effective treatment [2]. The pathologic process of AD begins decades before the emergence of clinical symptoms [3, 4]. A continuum between early, self-perceived changes in memory, here termed subjective memory complaints (SMCs), clinical manifestations of mild cognitive impairment (MCI) and predementia, has been documented [5]. SMCs have been associated with the presence of various pathological changes of AD, including amyloid- β) (A β) deposition [6–8], increases in white matter lesions [9, 10], and gray matter atrophy [11–13]. Also, SMCs strongly predict future dementia and AD [14–16]. Thus, the investigation of risk factors for SMCs may provide an impactful opportunity for elucidating the potential mechanism of AD/ADRD and facilitating early intervention to reduce or delay the onset of the early stages of dementia [17].

Emerging studies have provided evidence that gut microbiota may be associated with AD/ ADRD [18] via bidirectional interactions within the microbiota-gut-brain axis [19, 20]. Gut microbes communicate to the central nervous system through neural, endocrine, immune, and microbial metabolite pathways [21, 22]. The brain can affect the community structure and function of the gut microbiota through the autonomic nervous system, by modulating regional gut motility, intestinal transit and secretion, and gut permeability, and potentially through the luminal secretion of hormones that directly modulate microbial gene expression [23]. Animal studies have shown alterations in the gut microbial composition in mouse models of AD, which could contribute to brain A β deposition and decrements in learning and memory [24–30]. Manipulation of gut microbiota in transgenic mouse models of AD can influence cerebral A β deposition [28]. However, epidemiologic studies investigating the role of gut microbiota in AD/ADRD are limited.

A few cross-sectional studies have demonstrated differences in the gut microbial composition between individuals with MCI, AD, or dementia and healthy controls [31–41] (Supplementary Table 1). Three studies reported an enrichment of Proteobacteria in individuals with AD or MCI compared with healthy controls [33–35], while some other studies reported inconsistent findings on bacteria in Bacteroidetes [31, 32, 34, 36, 37, 40, 41]. However, these studies are limited in size (n < 50) [31, 34, 40], did not use next

generation sequencing to measure microbiome [36, 37], and/or included study subjects on medications or other interventions for AD/ADRD [32, 36, 37]. Only one hospital-based case-control study has assessed gut microbiota in relation to subjective cognitive decline [38]. We conducted a cross-sectional study to assess the association between gut microbiota and SMCs in 95 older women, recruited from a large prospective cohort study. We collected fecal samples and conducted 16S rRNA gene sequencing to quantify the gut microbial composition.

METHODS

Study population

The study population was derived from the New York University Women's Health Study (NYUWHS), a prospective cohort study of 14,274 women aged 35–65 years enrolled in 1985–1991 at a mammography screening center in New York City [42]. Women were ineligible for enrollment if they had used hormonal medications or had been pregnant or lactating in the previous 6 months. At baseline, data on demographics, anthropometric measures (height and weight), medical history, reproductive and lifestyle variables were collected through self-administered questionnaires. Active contact with cohort participants is achieved through questionnaires mailed every 3–5 years and, for non-respondents, telephone calls.

For the present study, we selected 250 women who responded to the NYUWHS questionnaire mailed in December 2018 (2018 follow-up). After excluding known cases of cancer and women > 90 years of age, we sent an invitation letter and a consent form to 227 women along with a screening questionnaire to collect information on exclusion criteria, including current dialysis; history of major gastrointestinal tract surgery; and use of systemic antibiotics, treatment with oral, nasal, or injected corticosteroids, treatment with immunosuppressive drugs other than corticosteroids, cleansing of the large intestine, colonoscopy, and substantial weight change (> 20 lbs) in the previous 6 months. A total of 118 (52%) eligible women consented to participate and 101 (44%) women returned a stool sample. Participants received a gift card of \$20 in the mail for their time and effort. The participation rate (44%), with the small amount of compensation, compares favorably with what has been reported in other cohorts (30–50%) that collected fecal samples [43–45].

At the time of fecal sample collection, participants completed a short questionnaire regarding current lifestyle (smoking, alcohol consumption, use of multivitamins and probiotics) and special diets (vegetarian, gluten-free, or other). All participants provided written informed consent to be involved in this study. All study procedures were approved by the institutional review board of NYU School of Medicine.

Assessment of SMCs

We assessed SMCs in the 2018 follow-up questionnaire based on 6 yes/no questions [7, 49], which have been validated against objective features of dementia, clinically established cognitive testing questionnaires for memory loss, and *APOE* e4 genotype [7, 50–52]. The 6 questions were based on reports of 1) a recent change in memory; 2) having more trouble

than usual remembering recent events; 3) remembering a short list of items; 4) following a group conversation or a plot in a TV program; 5) having difficulty understanding or following spoken instructions; and 6) having trouble finding one's way around familiar streets. We gave 1 point for each "yes" and computed a SMCs score as the total number of complaints reported by each woman [7]; thus, SMCs scores ranged from 0 to 6.

Collection of stool samples

We developed a protocol adapting procedures used in the Multiethnic Cohort (MEC) study [45] that allow participants to collect a stool sample in the privacy of their own homes. Briefly, once a signed consent was returned, we sent the participants a stool collection kit and easy-to-follow instructions with photos for stool collection. The stool collection kit consisted of a stool collection container (white tub with lid; Fisher Scientific, MA), a toilet adaptor tray (Fisher Scientific, MA), a triple-slide Beckman Coulter Hemoccult II SENSA[®] card (Beckman Coulter, CA), a pair of exam gloves, and a Ziploc bag to save the card. Participants collected stool onto all the three slides of the card. This method produces reproducible and accurate 16S rRNA gene-derived microbiota data [46, 47], and exhibits stability at room temperature up to 8 weeks [48]. Samples were mailed directly to NYU following at-home collection and stored immediately at -80° C until analysis.

Covariates

Time-invariant variables (race, education, height) were from the baseline questionnaire. We used data on age (continuous, years), alcohol consumption and current use of multivitamin and probiotics collected at the time of fecal sample collection. Other self-reported covariates, including weight (continuous), hypertension (defined as use of antihypertensive medicine or self-reported diagnosis, yes/no), use of cholesterol lowering medication (yes/ no), and use of antidepressants (yes/no) were from the 2018 follow-up questionnaire. Body mass index (BMI; kilogram per square meter) was calculated from height and weight. There were a limited number of women who reported current smoking (n = 1), diabetes (defined as use of insulin or diabetes pills or self-reported diagnosis, yes/no; n = 7), or special diets (vegetarian: n = 7, gluten-free: n = 5), and therefore these variables were not used.

Microbial assessment

From the 101 stool samples returned, one of the three sections from the cards containing the stool sample was cut and shipped on dry ice to Argonne National Laboratory for microbiome analysis. A second section from seven participants was included for quality control analyses. The samples underwent 16S rRNA gene sequencing at the Environmental Sample Preparation and Sequencing Facility of the Laboratory. Bacterial DNA was extracted using the Mo Bio PowerSoil DNA Isolation Kit (MO BIO Laboratories, Carlsbad, CA) according to the manufacturer's instructions. The V4 region of the 16S rRNA gene was PCR amplified with the 515F/806R primer pair [53, 54]. Six samples failed PCR amplification and were not further processed. PCR amplicons were purified using AMPure XP beads (Beckman Coulter, Brea, CA), pooled in equal amounts, and sequenced on the MiSeq platform (Illumina, San Diego, CA), as previously described [55].

Sequence read processing

Sequence reads were processed using QIIME 2 [56]. Briefly, sequence reads were demultiplexed and paired-end reads were joined, followed by standard quality filtering procedures [57]. We obtained an average of 28,148 reads per sample (range: 7,501–56,487; median: 27,453; interquartile range: 23,353–31,579). Next, the DADA2 pipeline [58] was applied, which uses sequence error profiles to obtain putative error-free sequences, referred to as amplicon sequence variants (ASVs). 0After building the ASV table and removing chimeras, taxonomy was assigned using the Ribosomal Database Project (RDP) Classifier (v2.13) [59] trained on the 99% OTUs in the Greengenes 13 8 database [60]. A phylogenetic tree was built using FastTree (v2.1.11) [61] from a multiple sequence alignment made with the PyNAST alignment tool [62]. The ASV tables was rarefied to 7,501 sequence reads per sample (the lowest sequencing depth among samples) prior to a- and β -diversity analysis. a-Diversity (within-subject diversity) was assessed using richness (the actual number of ASVs observed in a sample), Pielou's evenness, and Shannon diversity index, which measures diversity by accounting for evenness and richness [63]. β-Diversity (between-subject diversity) was assessed using the Bray-Curtis dissimilarity, and unweighted and weighted UniFrac distances [64]. Principal coordinate analysis (PCoA) was used for visualization. The ASVs were agglomerated to phylum, class, order, family, and genus levels and relative abundance of bacterial taxa (total sum scaling) was calculated at each taxonomic rank. After removing taxa that a) were present in < 10% of samples, b) had mean relative abundance < 0.01%, and c) were unclassified at the family or genus level, a total of 9 phyla, 17 classes, 20 orders, 34 families, and 52 genera were included in final analyses.

Statistical analyses

We dichotomized the SMCs scores and compared women reporting two or more complaints with those reporting less than two complaints, as over 40% of the women reported at least one SMC. We also considered the SMCs as a continuous variable and compared participants with 2 and 3+ SMCs with those without SMCs. We calculated descriptive statistics for socio-demographic variables: a) mean and standard deviation for continuous variables and b) distribution (%) for categorical variables, by binary SMCs scores (2 + versus < 2). We used Fisher's exact test and the Wilcoxon rank-sum test to detect group differences in categorical and continuous variables, respectively. We used linear regression models to test whether *a*-diversity differed by SMCs scores (2+ versus < 2 or 1, 2, 3+ versus 0), adjusting for age, BMI, and race. Permutational multivariate analysis of variance (PERMANOVA; 'adonis' function, 'vegan' package, R) was used to examine differences in the overall microbial community composition between SMCs strata (2 + versus < 2 or 1, 2, 3 + versus 0), adjusting for age, BMI, and race. We hypothesized that certain alterations in gut microbial composition may lead to decline in cognitive function. However, given the cross-sectional nature of the study, we could only assess whether the relative abundance of certain bacterial taxa was higher or lower in women with SMCs. We fit a beta regression model, which is appropriate to model a percentage response [65], to evaluate differences in the mean relative abundances (percentage response) of gut bacterial taxa associated with SMCs, adjusting for age, BMI, and race, using the "betareg" R package, as previously described [66]. Briefly, bacterial taxa with zero abundance were imputed as 10^{-6} because the logarithm of zero

is undefined. For a given taxon with p mean relative abundance, the estimated odds ratio (OR) is a ratio of p/(1-p) in women with 2+ SMCs and p/(1-p) in women with < 2 SMCs. When p is small, as for most of the taxa under investigation (< 10% or < 5%), the OR approximates the ratio of the mean relative abundances by categories of SMCs. We thus interpreted the ORs as the ratios of mean relative abundances by the outcomes of interest. p values for these models were adjusted for multiple comparison at each taxonomic level separately by the false discovery rate (FDR) method [67]. Similar models were constructed using SMCs as a continuous independent variable to estimate OR associated with every one-unit increase in SMCs. For genera that were associated with SMCs after controlling for multiple comparisons, we further estimated ORs comparing women reporting any SMCs with women with no complaint. Sensitivity analyses were conducted with additional adjustment for education, and use of cholesterol lowering medication and antidepressants. We also conducted sensitivity analyses using a differential abundance analysis method based on read count data, namely, Analysis of Compositions of Microbiomes with Bias Correction (ANCOM-BC), which corrects for bias induced by differences in the unknown sampling fractions and identifies taxa that are differentially abundant according to the covariate of interest [68]. ANCOM-BC first estimates sampling fractions that are different across samples, and then models the log of read count data, in which zeros are replaced by pseudo-count 1, through a linear regression model including the estimated sampling fraction as an offset term.

RESULTS

Population characteristics

Among the invited participants, we compared women who provided a stool sample and were included in the present study with those who did not. Women who gave stool samples had a higher educational attainment (p < 0.01; Supplementary Table 2) and were also slightly younger (p = 0.07), compared with those who did not. The two groups did not differ significantly in other demographics, comorbidities, and SMCs scores. Participants in the present study were predominantly white (88.4%) and 78% were between the ages of 70-90 (mean: 76.0 ± 5.8 years old) at the time of fecal sample collection. Overall, 52.6% of the study participants reported no SMCs, 21.1% reported one, 14.7% reported two, and 11.7% reported three or more. Of the six SMCs included, a recent change in memory was the most common complaint, reported by 42.1% of the participants, followed by having more trouble than usual remembering recent events (20.2%) and a short list of items (18.5%), and having trouble finding one's way on familiar streets (3.2%). Compared with women reporting < 2SMCs, women reporting 2+ SMCs were more educated (Table 1, p = 0.09), less likely to use cholesterol lowering medication (p < 0.05), and more likely to use antidepressants (p= 0.07). No appreciable differences by SMCs were noted for the other demographic and lifestyle variables.

α - and β -diversity in relation to SMCs

Women with 2+ SMCs did not differ significantly from those reporting < 2 SMCs in any of the α -diversity and β -diversity indices (Fig. 1 and Supplementary Figure 1). Pairwise

comparisons between women with 1, 2, or 3+ SMCs and those without any complaint did not show any appreciable difference in α -diversity and β -diversity (data not shown).

Taxa associated with SMCs

The relative abundances of 11 taxa were associated with SMCs at the nominal level when comparing women reporting 2+ SMCs with those reporting < 2 SMCs (Table 2), including the Betaproteobacteria-Burkholderiales-*Alcaligenaceae-Sutterella* and Deltaproteobacteria-Desulfovibrionales-*Desulfovibrionaceae-Bilophila* lineage within phylum Proteobacteria. After correction for multiple comparisons, genus *Holdemania* (family *Erysipelotrichaceae*) within phylum Firmicutes, class Deltaproteobacteria, order Desulfovibrionales, and family *Desulfovibrionaceae* were significantly enriched among women reporting 2+ SMCs compared with those reporting < 2 SMCs. The mean relative abundance of genus *Holdemania* was 2.09 times (OR: 2.09, 95% confidence interval [CI]: 1.38–3.17, FDR-adjusted *p* = 0.038) higher in women with 2+ SMCs, compared with women with < 2 SMCs. The mean relative abundance of the Deltaproteobacteria-Desulfovibrionales-*Desulfovibrionaceae* lineage was 2.10 times (OR: 2.10, 95% CI: 1.43–3.09, FDR-adjusted *p* = 0.005, 0.006, and 0.010, respectively) higher comparing women with 2+ SMCs with women who reported < 2 SMCs.

The associations between SMCs as a continuous score and the relative abundance of bacterial taxa were consistent with those with binary SMCs (Table 2). Specifically, after correction for multiple comparisons, the ORs for Proteobacteria and its downstream Betaproteobacteria-Burkholderiales-*Alcaligenaceae-Sutterella* and Deltaproteobacteria-Desulfovibrionales-*Desulfovibrionaceae* lineage associated with one-unit increase in SMCs were 1.19–1.27 (FDR-adjusted p = 0.003-0.033). Sensitivity analysis of additional adjustment for education, and use of cholesterol lowering medication and antidepressants in the model generated similar results. For instance, after correction for multiple comparisons, the mean relative abundances of genus *Holdemania* and family *Desulfovibrionaceae* were 2.57 times (OR: 2.57, 95% CI: 1.57–4.18, FDR-adjusted p = 0.016) and 2.05 times (OR: 2.05, 95% CI: 1.32–3.18, FDR-adjusted p = 0.062) higher in women reporting 2+ SMCs compared with women with < 2 SMCs.

Figure 2 depicts the enrichment of genus *Holdemania* and *Sutterella* in women reporting 1, 2, or 3+ SMCs compared with women without any complaint. For instance, the mean relative abundance of *Holdemania* was 2.36 times (OR: 2.36, 95% CI: 1.41–3.96) and 1.93 times (OR: 1.93, 95% CI: 1.08–3.45) higher in women reporting 2 and 3+ SMCs compared with women reporting no complaint, respectively. In addition, the mean relative abundance of *Sutterella* was 2.59 times (OR: 2.59, 95% CI: 1.56–4.30) higher in women reporting 3+ SMCs compared with women with no complaint.

Overall, the results from ANCOM-BC are consistent with those from beta regression at the nominal level, although none of the associations retained statistical significance after adjusting for multiple comparisons, possibly due to the conservative nature of the method. Specifically, ANCOM-BC also identified the genus *Holdemania* and the Deltaproteobacteria-Desulfovibrionales-*Desulfovibrionaceae* lineage as significantly enriched in women reporting 2+ SMCs compared with those reporting < 2 SMCs,

with a fold-change ranging from 1.95-2.41 (p = 0.009-0.022). When SMCs were treated as a continuous variable, per-unit increase in the SMCs scores was associated with a fold-change of 1.23-1.35 in the read counts of the genus *Holdemania*, the Betaproteobacteria-Burkholderiales-*Alcaligenaceae-Sutterella* lineage, and the Deltaproteobacteria-Desulfovibrionales-*Desulfovibrionaceae-Bilophila* lineage (p = 0.004-0.039). Also consistent with beta regression, ANCOM-BC demonstrated that the read counts of *Holdemania* and *Sutterella* were higher among women with 2 or 3+ SMCs relative to those without any complaint (data not shown). For instance, the read counts of *Sutterella* were 3.13-fold (95% CI: 1.46-6.72; p = 0.003) higher in women reporting 3+ SMCs than those with no complaint, with a p for trend of 0.006.

DISCUSSION

In this study of older women from a large prospective cohort, we observed significant overrepresentation of the Deltaproteobacteria-Desulfovibrionales-*Desulfovibrionaceae* lineage within phylum Proteobacteria as well as genus *Holdemania* in women reporting 2 or more SMCs, compared to women reporting one or no complaint. Furthermore, increasing relative abundances of Proteobacteria and its downstream Betaproteobacteria-to-*Sutterella* and Deltaproteobacteria-Desulfovibrionales-*Desulfovibrionaceae* lineage were associated with increasing SMCs scores in a dose-dependent manner. The relative abundances of *Holdemania* and *Sutterella* were higher among women with 2 or 3+ SMCs relative to those without any complaint. These findings suggest that specific gut microbial taxa may have altered in women with SMCs.

One mechanism by which the gut bacteria may affect cognitive functions and promote neurodegeneration involves lipopolysaccharide (LPS), a pro-inflammatory endotoxin derived from the cell wall of gram-negative bacteria. Changes in gut microbiota composition can lead to increased intestinal barrier permeability, which allows for translocation of LPS to the circulation [69], impairing the blood-brain barrier and eliciting systemic inflammation [70] that promotes neuroinflammation, neuronal loss, and ultimately AD [71]. Moreover, LPS was found in amyloid plaques and around vessels in AD brain [72]. Our findings that increases in the phylum Proteobacteria and its downstream taxa were associated with SMCs may lend support to the neuroinflammation hypothesis, as Proteobacteria are LPS containing, gram-negative commensal bacteria. An increased abundance of Proteobacteria in the gut reflects an unstable structure of the gut microbial community and may serve as a biomarker for dysbiosis [73, 74]. Transgenic AD mice were characterized by an increase in Proteobacteria during aging [75]. In conventional mice re-colonized with microbiota from high-fat fed mice, proportions of Proteobacteria were negatively associated with learning and memory performance [24]. Enrichment of Proteobacteria has been reported in cognitively impaired, defined based on cognitive test performance, but neurologically healthy, community-dwelling older adults [34].

Within Proteobacteria, we found enrichment of the Deltaproteobacteria-to-*Bilophila* lineage in women reporting 2 or more SMCs. In mice, the abundance of family *Desulfovibrionaceae* was significantly higher in transgenic AD mice than in wild-type mice [76]. *Bilophila* was more abundant in AD individuals relative to age- and sex-matched control individuals [40].

We also found positive associations between the Betaproteobacteria-to-*Sutterella* lineage and SMCs. Interestingly, a cohort study of 1,551 largely female (90%) members of the TwinsUK British twin cohort (mean age 63, ranging 40–89) reported a significant association of class Betaproteobacteria and order Burkholderiales with poor cognitive function [39]. In addition, higher abundance of the family *Alcaligeneceae* was associated with poor performance across multiple cognitive tests in individuals with cirrhosis and hepatic encephalitis [77]. *Alicaligenaceae* are typically associated with opportunistic infections and known to degrade urea to ammonia, which may explain part of this association [78]. Previous studies have concluded that *Sutterella* species are possible pro-inflammatory agents [79]. Transgenic AD mice were characterized by an increase in *Sutterella* during aging [75]. Taken together, our results support the hypothesis that over-representation of the pro-inflammatory bacteria in the gut may occur in the early stage of preclinical AD.

In addition to the taxa in Proteobacteria, we also observed positive associations of genera Holdemania (phylum Firmicutes), Odoribacter (phylum Bacteroidetes), and Faecalibacterium (phylum Firmicutes) with SMCs, and an inverse association between genus Phascolarctobacterium and SMCs; however, only Holdemania retained statistical significance after adjustment for multiple comparisons. One study has reported that Holdemania was enriched in Parkinson's disease patients [80]. In mice, the relative abundance of Holdemania was positively correlated with motor deficits [81], suggesting effects of Holdemania on neuropsychological disorders. The associations between other species and SMCs were supported by some but in contrary to other studies. For instance, genes of an *Odoribacter* species have been associated with the AD pathway [82], and increased relative abundance of this species was observed in transgenic AD mice [76] and in elderly AD subjects (mean > 80 years) [32]. *Faecalibacterium* and its metabolite butyrate, one of the gut microbiota-produced short-chain fatty acids (SCFAs), have well-documented anti-inflammatory properties [83]. A significant reduction in *Faecalibacterium* was reported in MCI [84] and AD subjects [32], and in subjects with subjective cognitive decline [38]. However, we observed that the relative abundance of *Faecalibacterium* was higher in women with SMCs. Phascolarctobacterium produces propionate, a less-studied SCFA that was reported to have equipotent anti-inflammatory effects as butyrate in vitro [85, 86]. It has been hypothesized that SCFAs may provide an alternative energy source to counteract neuronal damage [87] that contributes to neuronal dysfunction in AD [88]. SCFAs may also protect against blood-brain barrier permeability and modulate maturation and function of microglia in the brain [89]. However, a significant increase in Phascolarctobacterium was reported in MCI [35] and AD subjects [40] in other studies. Future larger studies are warranted to elucidate the role of these bacterial taxa across the stages of AD.

The present study has strengths including the use of 16S rRNA gene sequencing, the inclusion of participants from a prospective cohort study, and the use of validated questions for SMCs. Compared with previous studies, which often involved patients in the hospital, the participants in the present study did not have serious health conditions or existing AD, minimizing the impact of systemic behavioral or diet changes, or medication use due to AD/ ADRD on the gut microbial composition. Also, the focus on a homogeneous population with respect to risk factors of cognitive impairment enhanced the internal validity of the study. The relatively small sample size and the lack of detailed information on diet and physical

activity are limitations of the study. Stool samples were collected on average 7.6 months (93% within 12 months) after completion of the 2018 follow-up questionnaire. Cognitive impairment and SMCs are long-term conditions. Though stool samples were collected after the questionnaire, we consider the status and prevalence of SMCs remained similar during the period. Potential selection bias cannot be excluded as women who gave stool samples had a higher educational attainment than those who did not (Supplementary Table 2). To the extent that educational attainment is associated with both gut microbial composition and SMCs, it can bias the results. However, we did sensitivity analysis of additional adjustment for education in the model which generated similar results. Nevertheless, we cannot account for unmeasured factors that could potentially result in selection bias. In addition, the cross-sectional design does not allow us to assess whether the gut microbiota are associated with longitudinal changes in SMCs. Finally, lack of shotgun-sequenced metagenome data did not allow us to characterize metagenomic functions.

In summary, our findings provide supporting evidence of alterations in the gut microbial composition in association with self-perceived memory decline, characterized by an enrichment of pro-inflammatory taxa. The present study indicated feasibility of future fecal sample collection on a large scale in this aging population. Our observation that some of the SMCs-associated taxa were consistent with those found for AD/ADRD in the literature suggests a critical role of the gut microbiota in early stages of cognitive dysfunction. Future large or prospective studies with longitudinal data, objective measures of cognitive function, and shotgun metagenomic sequencing are needed to confirm the associations and evaluate the generalizability of the findings.

Supplementary Material

Refer to Web version on PubMed Central for supplementary material.

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Authors' disclosures available online (https://www.j-alz.com/manuscript-disclosures/22-0011r1).

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Fig. 1.

Alpha diversity indices of gut microbiota by subjective memory complaints scores (2+ versus < 2). Violin plots based on (A) richness (number of ASVs), (B) Evenness, and (C) Shannon diversity index. Linear regression was used to test the group difference, adjusting for age, BMI and race.

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Fig. 2.

Odds ratios of mean relative abundance of *Holdemania* and *Sutterella* comparing women reporting 1, 2, or 3+ subjective memory complaints with women reporting no complaint. Beta regression was used to estimate the odds ratios, adjusting for age, BMI, and race.

Table 1

Distribution of lifestyle and demographic variables by subjective memory complaints scores (2+ versus < 2)

Characteristics	Subjective memory complaints scores < 2	Subjective memory complaints scores 2+	p a
Total N	70	25	
$Age^{b}(y)$, mean (SD)	76.1 (5.9)	75.7 (5.8)	0.75
Weight ^C (lbs), mean (SD)	147.2 (37.3)	150.9 (41.3)	0.55
BMI ^C (kg/m ²), mean (SD)	24.6 (4.8)	24.5 (6.1)	0.62
$\operatorname{Race}^{d}, n(\%)$			0.28
Caucasian	60 (85.7%)	24 (96.0%)	
Other	10 (14.3%)	1 (4.0%)	
Education $\overset{d}{d}$, n (%)			0.0
Did not complete college	20 (28.6%)	2 (8.3%)	
Completed college	13 (18.6%)	4 (16.7%)	
Some schooling after college	37 (52.8%)	18 (75.0%)	
Any alcoholic drink $^{\mathcal{C}}$ use per week b , n (%)	36 (54.5%)	11 (44.0%)	0.48
Current multivitamin use b , n (%)	36 (52.9%)	9 (39.1%)	0.34
Current probiotic $f_{ m use}^{}$, n (%)	11 (16.7%)	4 (17.4%)	1.00
Current use of cholesterol lowering medication $^{\mathcal{C}}$, $n\left(\% \right)$	35 (50.7%)	4(16.0%)	0.004
Hypertension $^{\mathcal{C}}$, n (%)	38 (54.3%)	12 (48.0%)	0.65
Current antidepressant use c' , n (%)	6 (8.6%)	6 (24.0%)	0.07

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fPill, powder, or liquid.

 e^{0} One drink equals to 12 oz of beer, 5 oz of wine, or 1½ ounces of liquor.

b Data collected at the time of fecal sample collection.

 $^{\mathcal{C}}$ Data from the 2018 follow-up questionnaire.

 $d_{\text{Data collected at baseline.}}$

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Table 2

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Taxa (class; order; family; genus)		Binar	y (2+ ver	sus < 2))	Continuo	SI
	Mean relative abundance, %	OR ^b (95% CI)	d	FDR adjusted p^c	OR ^b (95% CI)	d	FDR adjusted p^c
Bacteroidetes							
Bacteroidia; Bacteroidales; [Odoribacteraceae], Odoribacter (genus)	0.28	1.58 (1.03–2.43)	0.038	0.399	1.18 (1.02–1.36)	0.025	0.262
Firmicutes							
Clostridia; Clostridiales; Ruminococcaceae; Faecalibacterium (genus)	7.07	1.50 (1.06–2.11)	0.023	0.399	1.12 (1.00–1.25)	0.053	0.393
Clostridia; Clostridiales; Lachnospiraceae; Anaerostipes (genus)	0.18	1.47 (0.95–2.25)	0.085	0.628	1.15 (1.00–1.32)	0.049	0.393
Clostridia; Clostridiales; Veillonellaceae; Phascolarctobacterium (genus)	0.26	0.69 (0.44–1.10)	0.094	0.712	0.80 (0.68–0.95)	0.014	0.180
Erysipelotrichi; Erysipelotrichales; Erysipelotrichaceae; Holdemania (genus)	0.04	2.09 (1.38–3.17)	0.0007	0.038	1.24 (1.08–1.41)	0.002	0.055
Proteobacteria	4.56	1.27 (0.89–1.83)	0.193	0.998	1.19 (1.06–1.34)	0.004	0.033
Betaproteobacteria (class)	1.43	1.61 (1.10–2.35)	0.016	0.135	1.20 (1.07–1.35)	0.002	0.021
Betaproteobacteria; Burkholderiales (order)	1.43	1.61 (1.10–2.35)	0.016	0.158	1.20 (1.07–1.35)	0.002	0.025
Betaproteobacteria; Burkholderiales; Alcaligenaceae (family)	1.38	1.76(1.20 - 2.59)	0.005	0.082	1.25 (1.11–1.40)	0.0003	0.006
Betaproteobacteria; Burkholderiales; Alcaligenaceae; Sutterella (genus)	1.38	1.76 (1.20–2.59)	0.005	0.131	1.25 (1.11–1.40)	0.0003	0.018
Deltaproteobacteria (class)	0.35	2.10 (1.43–3.09)	0.0003	0.005	1.27 (1.13–1.42)	0.0002	0.003
Deltaproteobacteria; Desulfovibrionales (order)	0.35	2.10 (1.43–3.09)	0.0003	0.006	1.27 (1.13–1.42)	0.0002	0.003
Deltaproteobacteria; Desulfovibrionales; Desulfovibrionaceae (family)	0.35	2.10(1.43 - 3.09)	0.0003	0.010	1.27 (1.13–1.42)	0.0002	0.006
Deltaproteobacteria; Desulfovibrionales; Desulfovibrionaceae; Bilophila (genus)	0.22	1.59 (1.03–2.43)	0.038	0.399	1.22 (1.07–1.40)	0.005	060.0

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 $b_{\mbox{Adjusting for age, BMI, and race (White versus other).}$

 $c_{\text{FDR}} = 5\%.$