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A genetic association study reveals the relationship between the oral microbiome and anxiety and depression symptoms

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Background: Growing evidence supports that alterations in the gut microbiota play an essential role in the etiology of anxiety, depression, and other psychiatric disorders. However, the potential effect of oral microbiota on mental health has received little attention.

Methods: Using the latest genome-wide association study (GWAS) summary data of the oral microbiome, polygenic risk scores (PRSs) of 285 salivary microbiomes and 309 tongue dorsum microbiomes were conducted. Logistic and linear regression models were applied to evaluate the relationship between salivary-tongue dorsum microbiome interactions with anxiety and depression. Two-sample Mendelian randomization (MR) was utilized to compute the causal effects between the oral microbiome, anxiety, and depression.

Results: We observed significant salivary-tongue dorsum microbiome interactions related to anxiety and depression traits. Significantly, one common interaction was observed to be associated with both anxiety score and depression score, Centipeda periodontii SGB 224 × Granulicatella uSGB 3289 (P depressionscore = 1.41×10^{-8} , P anxietyscore = 5.10×10^{-8}). Furthermore, we detected causal effects between the oral microbiome and anxiety and depression. Importantly, we identified one salivary microbiome associated with both anxiety and depression in both the UKB database and the Finngen public database, Eggerthia (P $_{IVW-majordepression-UKB} = 2.99 \times 10^{-6}$, 10^{-59} . Ρ 3.06 =× IVW-Self-reportedanxiety/panicattacks-UKB $P_{IVW-depression-Finngen} = 3.16 \times 10^{-16}$ P = 1.14 IVW-anxiety-Finngen $\times 10^{-115}$).

Conclusion: This study systematically explored the relationship between the oral microbiome and anxiety and depression, which could help improve our understanding of disease pathogenesis and propose new diagnostic targets and early intervention strategies.

KEYWORDS

anxiety, depression, oral microbiome, Mendelian Randomization (MR), polygenic risk scores (PRS)

Introduction

Anxiety disorders and depression are prevalent mental illnesses, with an estimated 264 million people suffering from anxiety disorders and 322 million people suffering from depression in 2015, according to the World Health Organization (WHO) global health estimates (1). As represented in the Global Burden of Disease Study 2013 (GBD 2013), the second most common cause of years lived with disability (YLDs) was depression, and the ninth was anxiety (2). Furthermore, according to the description of the GBD Compare, the disability-adjusted life years (DALYs) for depression were 1.84 and 1.13 for anxiety (https://vizhub.healthdata.org/gbdcompare/). In conclusion, the high prevalence of anxiety disorders and depression and the huge burden and harm caused by both diseases to society and individuals have made anxiety and depression highly concerning. Previous studies have found that human genetics regulates the pathogenesis of anxiety and depression (3, 4). Offspring of patients suffering from anxiety were 4-6 times more likely to develop anxiety than offspring of the general population (3). In twin studies, the heritability of anxiety disorders was approximately 30-50% (3), and the heritability of depression was approximately 35% (4).

The microbiota is a complex ecosystem of microorganisms containing viruses, fungi, bacteria, and protozoa that live in different parts of the human body, such as the mouth, the gastroenteric tube, the vagina, the respiratory system, and the skin (5). The microbiota plays an essential role in the host immune system's induction, cultivation, and function. Accordingly, the host immune system has developed multiple ways to maintain its symbiotic relationship with the microbiota (6). More than 70% of microbiota are found in the gastrointestinal tract and maintain a mutually beneficial relationship with the host (6). As a complex physiological ecosystem, the gut microbiota influences its host health. Growing evidence links the gut microbiota to various psychiatric and neurological disorders, such as schizophrenia, depression, bipolar disorder, autism spectrum disorder, Alzheimer's disease, and Parkinson's disease (7-12). Studies have shown that the links between the gut microbiota and these diseases may be due to a bidirectional communication system between the gut and the central nervous system (CNS) called the "microbiota-gut-brain axis" (MGBA) (13, 14). The signaling mechanism of the bidirectional communication system is that the brain regulates gut function via the hypothalamic-pituitary-adrenal (HPA) axis and the autonomic nervous system. For example, the brain releases norepinephrine during stress, and norepinephrine has been found to stimulate the proliferation of gut pathogens (15, 16).

Furthermore, the gut modulates CNS function through a variety of microbiota-derived metabolites and products. For instance, gut hormones and neuroactive substances are delivered to the brain *via* the circulatory system, enteric nervous system, immune system, and vagus nerve (14). In recent years, the

diversity of microbiota species in the mouth has made it the focus of research. Several mechanisms may exist for the link between poor oral health and mental health disorders, such as changes in the oral microbiota (17). Nevertheless, the relationship between the oral microbiome and anxiety and depression remains unclear.

Previously, researchers have revealed the impact of human genetics on the oral microbiome (18). A twin oral microbiome study showed that the oral microbiome was heritable, and the heritability of many microbiome phenotypes was more than 50% (19). Although the effect of host genetics on the composition and stability of the oral microbiome remains poorly understood, previous GWAS studies have identified several genetic loci associated with the oral microbiome. For example, an unbiased GWAS analysis by Demmitt et al. (19) suggested that the genes IMMPL2 and INHBA-AS1 could influence the oral microbiome. The study by Poole et al. (20) identified AMY1-CN as a genetic factor associated with microbial composition and function. However, these findings are still limited. Polygenic risk scores (PRSs) provide an overall estimate of the genetic predisposition for a trait at the individual level by calculating the sum of the effects of risk alleles, where each risk allele estimates the phenotype from an independent GWAS (21). Previously, PRS has also been used to assess the impact of d dietary habits and the gut microbiome on anxiety and depression (22).

This study conducted polygenic risk scores (PRSs) analysis of the salivary microbiome and tongue dorsum microbiome in the UK Biobank cohort. Logistic/linear regressions were applied to analyze the associations between salivary microbiome-PRSs, tongue dorsum microbiome-PRSs, and their interactions with anxiety and depression. Then, Mendelian Randomization (MR) was used to evaluate the causal relationship between each salivary microbiome, tongue dorsum microbiome, anxiety, and depression.

Materials and methods

The UK Biobank cohort and the definition of anxiety and depression

The UK Biobank (UKB) (http://www.ukbiobank.ac.uk) recruited 502,656 participants aged 40–69 from 2006 to 2010 and recorded health information, hospital records, and genetic data of participants (23). UKB performed genotyping, imputation, and quality control (QC) for 487,409 individuals. For example, the UK Bileve axiom array and the UK Biobank axiom array were used for genotyping. Both arrays share over 95% of their marker content. IMPUTE4 was used for imputation in chunks of about 50,000 imputed markers with a 250 kb buffer region. QC, imputation, and post-imputation cleaning were performed centrally by UKB. Participants were excluded who were inconsistent between self-reported gender and genetic gender,

| | Depression | | Anxiety | | |
|----------------|----------------|---|----------------|--|--|
| | PHQ-9 | Self-reported depression | GAD-7 | Self-reported anxiety | |
| Participants | 154 360 | 157 459 (76 672 cases/ 80 787 controls) | 155 076 | 138 709 (27 898 cases/ 110 811 controls) | |
| Females, n (%) | 87 206 (56.50) | 89 746 (57.00) | 87 604 (56.49) | 77,088 (55.58) | |
| Mean age (SD) | 55.90 (7.74) | 56.11 (7.78) | 55.89 (7.74) | 56.16 (7.70) | |

TABLE 1 The basic characteristics of study samples.

PHQ, Patient Health Questionnaire; GAD, Generalized Anxiety Disorder.

without imputation data and ethical consent. Additionally, all genetically related individuals were removed by the KING software (http://people.virginia.edu/~wc9c/KING/) (24). More information about genotyping, imputation, QC, and physical measurements has been described previously (23). Our research has been approved by the UK Biobank (Application 46478). The UK Biobank has support from the North West Multicenter Research Ethics Committee (MREC) and the Human Tissue Authority (HTA). All participants agreed to use their anonymous data and samples for any health-related studies and to reconnect for further sub-studies (25).

The study obtained two common psychiatric disorders, including anxiety (self-reported anxiety and general anxiety disorder (GAD-7) scores) and depression (self-reported depression and the patient health questionnaire (PHQ-9) scores). Both anxiety and depression are derived from UKB. For anxiety, the present study included 155 076 participants with GAD-7 scores and 138,709 participants with self-reported anxiety status data (27,898 cases and 110,811 controls). For depression, 154,360 participants with PHQ-9 scores and 157,459 participants with self-reported depression status data (76,672 cases and 80,787 controls) were included in this study. The basic characteristics of the study subjects and detailed information are presented in Table 1.

GWAS summary datasets of the oral microbiome

The GWASs of 309 tongue dorsum microbiomes (N = 2,017) and 285 salivary microbiomes (N = 1,915) were obtained from a recently published metagenome-genomewide association study of the human oral microbiome (18). The conservative inclusion threshold of mean depth > 8×, Hardy Weinberg equilibrium (HWE) > 10⁻⁵, and genotype calling rate > 98% for variants were applied. Furthermore, the samples had to meet the following criteria: variant calling rate > 98%, mean sequencing depth > 20×, no population stratification in principal component analysis (PCA), and excluding related individuals by calculating pairwise identity by descent. After quality control, 2,984 individuals (2,017 tongue dorsum and 1,915 salivary) with about 10 million common and low-frequency (MAF \geq 0.5%) variants were contained. More detailed information on sample collection, sequencing, microbiome trait preparation, observational analysis, and genotyping analysis is described elsewhere (18).

Oral microbiome-related PRSs calculation

We calculated the oral microbiome-related PRSs of each subject based on the individual genotype data of UKB by using PLINK2.0. For oral microbiome PRS, PRS_n denotes the PRS value of the oral microbiome for the nth subject, defined as follows:

$$PRS_n = \sum_{i=1}^l E_i D_{in},$$

where l denotes the total number of gut microbiota-associated SNPs; E_i denotes the effect size of significant gut microbiota-associated SNP i; and D_{in} denotes the dosage of the risk allele of the ith SNP for the nth individual (0 is coded for homozygous protective genotype, one for heterozygous, and two for homozygous polymorphic genotypes).

Statistical analysis

The logistic and linear regression models were used to evaluate the associations of tongue dorsum microbiome-PRSs, salivary microbiome-PRSs, and their interactions with depression and anxiety. The logistic and linear regression models were established by R software (R-4.1.2). The tongue dorsum microbiome-PRSs, salivary microbiome-PRSs, and salivary microbiome-PRSs, salivary microbiome-PRSs, and salivary microbiome PRSs × tongue dorsum microbiome-PRSs were selected as independent variables; PHQ-9 score, self-reported depression, GAD-7 score, and self-reported anxiety were fitted as dependent variables; sex, age, 10PC, ever smoked, ever drank, and Townsend deprivation index (TDI) were set as covariates. Bonferroni corrected the *P* value of 1.42×10^{-7} [*P* = 0.05/ (285 × 309 × 4)] as the significance threshold for correcting the multiple testing.

Two-sample MR analysis was used to assess the causal relationship between the tongue dorsum microbiome, salivary

microbiome, anxiety, and depression. The tongue dorsum and salivary microbiome were selected as instrumental variables (IVs), and anxiety and depression were used as the outcome variables. The inverse variance weighted (IVW) was used as the primary causal effect estimate. The IVW method is an ideal estimation, and it is an effective analysis under the assumption that all genetic variants are effective instrumental variables, and it has a strong ability to detect causality (26). But the IVW specifically requires that genetic variants affect the target outcome only through exposure in the study. Although known confounding SNPs were excluded as much as possible in this study, there were still many unknown confounders that could lead to pleiotropy and biased effect size estimates. Therefore, we adopted two methods to check the reliability and stability of the results, namely the MR Egger regression (27) and the Weighted Median Estimator (WME) (28). Bonferroni corrected the *P*-value of 2.84×10^{-7} [*P* = 0.05/ (285 × 309 × 2)] as the significance threshold for correcting the multiple testing. The MR analysis was performed using the "Two-Sample MR" package for R 3.5.3 (29).

This study used the IVW and MR Egger regressions to test the heterogeneity. If the P > 0.05, it was considered that there was no heterogeneity in the included IVs, and the effect of heterogeneity on the estimation of causal effects could be ignored. Furthermore, MR Egger regression was used to evaluate the bias caused by horizontal pleiotropy, and the Egger intercept can evaluate the size of pleiotropy. In this study, the P-value of the pleiotropy test was used to measure whether there was pleiotropy in the analysis. If P > 0.05, the possibility of pleiotropy in the causal effect is considered to be weak, and its impact can be ignored. Furthermore, GWAS summary statistics of anxiety and depression in the Finngen public database (https://www.finngen.fi/en/access_results) were used to verify the causal effects between tongue dorsum microbiome, salivary microbiome, anxiety, and depression in UKB.

Results

Associations of the oral microbiome with depression

We found three and 10 significant salivary-tongue dorsum microbiome interactions for self-reported depression and PHQ-9 score, respectively (Table 2). Such as *Streptococcus unclassified SGB (uSGB) 891 × Rothia mucilaginosa SGB 3124* (P _{PHQ-9score} = 1.34×10^{-9}) and *Campylobacter A uSGB 1321 × Capnocytophaga uSGB 307* (P _{self-reporteddepression} = 1.03×10^{-8}). After integrating the two results, no common interactions were shared between the PHQ-9 score and self-reported depression. After gender stratification, we identified five significant interactions for the PHQ-9 score in females and 12 significant interactions for the PHQ-9 score in males

(Supplementary Tables S1, S2), such as *Gemella morbillorum* SGB 349 × Solobacterium uSGB 2587 (P _{PHQ-9-female} = 9.81 × 10^{-10}). The significant salivary-tongue dorsum microbiome interactions for depression are shown in Figure 1.

Associations of the oral microbiome with anxiety

One significant salivary-tongue dorsum microbiome interaction was associated with self-reported anxiety, and 14 interactions were related to the GAD-7 score (Table 3). Such as *Lancefieldella sp000564995 SGB 2467× Haemophilus D uSGB 3426* (P _{Self-reportedanxiety} = 5.50×10^{-22}) and *Lachnoanaerobaculum uSGB 2879× Lancefieldella uSGB 1356* (P _{GAD-7score} = 1.42×10^{-8}). No common interaction was shared by both the GAD-7 score and self-reported anxiety. After gender stratification, 105 significant interactions were found for the GAD-7 score in males (Supplementary Table S3). The significant salivary-tongue dorsum microbiome interactions for anxiety are shown in Figure 1.

The common oral microbiome is associated with depression and anxiety

We discovered one significant common salivary-tongue dorsum microbiome interaction shared by the PHQ-9 score and GAD-7 score, *Centipeda periodontii SGB 224* × *Granulicatella uSGB 3289* (P _{PHQ-9} = 1.41×10^{-8} , P _{GAD-7} = 5.10×10^{-8}) (Supplementary Table S4). After gender stratification, no significant common interactions were shared between anxiety and depression.

The causal relationship between the oral microbiome and depression

Figure 2 shows the results of a two-sample MR analysis identifying the relationship between salivary microbiomes, tongue microbiomes, and major depression. We observed that five salivary microbiomes and 10 tongue dorsum microbiomes were the causal risk factors for major depression in the UK Biobank. Furthermore, we discovered that four salivary microbiomes and six tongue dorsum microbiomes were causal risk factors for depression in the Finngen public database (Supplementary Table S5). Both the UK Biobank and Finngen public databases shared one common salivary microbiome and five tongue dorsum microbiomes as the risk factors for depression. Such as *Eggerthia* in salivary microbiomes (P $_{\rm UKB-IVW} = 2.99 \times 10^{-6}$, P $_{\rm UKB-WME}$

| Interaction (salivary microbiomes × tongue dorsum microbiomes) | Beta | SE | P value | |
|---|-----------------------|-----------------------|------------------------|--|
| Self-reported depression | | | | |
| Campylobacter A uSGB 1321 $	imes$ Capnocytophaga uSGB 307 | $6.50 	imes 10^{14}$ | $1.13 	imes 10^{14}$ | $1.03 	imes 10^{-8}$ | |
| Campylobacter A uSGB 1321 $	imes$ Streptococcus sanguinis SGB 1844 | $1.04 	imes 10^{15}$ | $1.85 	imes 10^{14}$ | $1.79 	imes 10^{-8}$ | |
| Neisseria meningitidis A SGB 552 \times Porphyromonas endodontalis SGB 3255 | 1.95×10^{12} | 9.95×10^{10} | 1.05×10^{-85} | |
| PHQ-9 score | | | | |
| Streptococcus uSGB 891 × Rothia mucilaginosa SGB 3124 | 0.064 | 0.011 | $1.34 	imes 10^{-9}$ | |
| Streptococcus uSGB 891 × Rothia sp001808955 SGB 2280 | 0.064 | 0.011 | $1.34 	imes 10^{-9}$ | |
| Granulicatella uSGB 3371 × CAG-917 uSGB 1705 | 0.042 | 0.007 | $4.21 	imes 10^{-9}$ | |
| Pauljensenia uSGB 2446 × f CAG-917 uSGB 1683 | -0.021 | 0.004 | 5.24×10^{-9} | |
| Centipeda periodontii SGB 224 $	imes$ Granulicatella uSGB 3289 | 0.021 | 0.004 | 1.41×10^{-8} | |
| Solobacterium extructa SGB 681 × Treponema A uSGB 3206 | -963.461 | 175.805 | $4.25 	imes 10^{-8}$ | |
| Solobacterium extructa SGB 681 \times Pauljensenia uSGB 3205 | 302.352 | 55.209 | $4.35 	imes 10^{-8}$ | |
| Solobacterium extructa SGB 681 \times Campylobacter A uSGB 1072 | -596.774 | 109.062 | $4.46 	imes 10^{-8}$ | |
| Neisseria uSGB 3379 × Granulicatella uSGB 3289 | 8.727 | 1.597 | $4.62 	imes 10^{-8}$ | |
| Saccharimonadaceae uSGB 1345 \times Neisseria uSGB 3010 | -8.72×10^{9} | $1.64 	imes 10^9$ | $1.05 	imes 10^{-7}$ | |
| | | | | |

TABLE 2 The significant salivary-tongue dorsum microbiomes interactions for self-reported depression and PHQ-9 score.

PHQ, Patient Health Questionnaire.



= 9.83 \times 10⁻⁸, P _{Finngen-IVW} = 3.16 \times 10⁻¹⁶, and P _{Finngen-WME} = 1.43 \times 10⁻¹⁶) and *Lancefieldella uSGB* 2940 in tongue dorsum microbiomes (P _{UKB-IVW} = 5.91

 \times 10⁻¹², P _{UKB-WME} = 1.81 \times 10⁻⁰⁷, P _{Finngen-IVW} = 1.06 \times 10⁻¹⁸, and P _{Finngen-WME} = 1.48 \times 10⁻¹⁰). The estimated cause-effect sizes of the SNPs on both the exposure

| Interaction (salivary microbiomes × tongue dorsum microbiomes) | Beta | SE | P value | |
|--|----------------------|----------------------|------------------------|--|
| Self-reported anxiety | | | | |
| Lancefieldella sp000564995 SGB 2467 \times Haemophilus D uSGB 3426 | -6.83×10^{15} | $7.08 	imes 10^{14}$ | 5.50×10^{-22} | |
| GAD-7 score | | | | |
| Lachnoanaerobaculum uSGB 2879 \times Lancefieldella uSGB 1356 | -0.074 | 0.013 | 1.42×10^{-8} | |
| Lachnoanaerobaculum uSGB 2879 \times Veillonella uSGB 1430 | -0.074 | 0.013 | $1.42 	imes 10^{-8}$ | |
| Lachnoanaerobaculum uSGB 2879 × Lancefieldella sp000564995 SGB 1669 | -0.074 | 0.013 | $1.42 	imes 10^{-8}$ | |
| Lachnoanaerobaculum uSGB 2879 \times Saccharimonadaceae uSGB 2252 | -0.074 | 0.013 | 1.42×10^{-8} | |
| Lachnoanaerobaculum uSGB 2879 \times Lancefieldella uSGB 2149 | -0.074 | 0.013 | 1.42×10^{-8} | |
| CAG-793 uSGB 1688 \times Bacteroidaceae uSGB 2747 | -166.356 | 29.462 | $1.64 	imes 10^{-8}$ | |
| CAG-793 uSGB 1688 × Streptococcus sp001553685 SGB 223 | 486.829 | 86.559 | $1.87 	imes 10^{-8}$ | |
| Centipeda periodontii SGB 224 \times Granulicatella uSGB 3289 | 0.019 | 0.003 | $5.10 	imes 10^{-8}$ | |
| Lancefieldella sp000564995 SGB 2677 \times Prevotella baroniae SGB 143 | 0.077 | 0.015 | $1.25 	imes 10^{-7}$ | |
| Lancefieldella sp 000564995 SGB 2677 \times Prevotella enoeca SGB 792 | 0.077 | 0.015 | $1.25 	imes 10^{-7}$ | |
| Lancefieldella sp 000564995 SGB 2677 \times Prevotella intermedia SGB 2942 | 0.077 | 0.015 | $1.25 	imes 10^{-7}$ | |
| Lancefieldella sp 000564995 SGB 2677 \times Prevotella lo escheii SGB 1756 | 0.077 | 0.015 | $1.25 	imes 10^{-7}$ | |
| Lancefieldella sp 000564995 SGB 2677 \times Prevotella sp 000467895 SGB 51 | 0.077 | 0.015 | $1.25 	imes 10^{-7}$ | |
| CAG-793 uSGB 1688 × Saccharimonadaceae uSGB 2893 | -222.061 | 42.009 | 1.25×10^{-7} | |

TABLE 3 The significant salivary-tongue dorsum microbiomes interactions for self-reported anxiety and GAD-7 score.

GAD is a general anxiety disorder.

(salivary microbiomes and tongue dorsal microbiomes) and outcome (depression) were displayed in scatter plots (Supplementary Figures S1, S2).

The causal relationship between the oral microbiome and anxiety

Figure 3 shows the results of a two-sample MR analysis identifying the relationship between salivary microbiomes, tongue dorsum microbiomes, and anxiety. We observed that seven salivary microbiomes and three tongue dorsum microbiomes were the causal risk factors for anxiety in UK Biobank. Furthermore, we discovered that six salivary microbiomes and four tongue dorsum microbiomes were causal risk factors for anxiety in the Finngen public database (Supplementary Table S6). Both the UK Biobank and Finngen public databases shared four common salivary microbiomes and one tongue dorsum microbiome as the risk factors for anxiety. Such as Actinomyces uSGB 2337 in salivary microbiomes (P $_{\rm UKB-IVW} = 4.11 \times 10^{-24}$, P $_{\rm UKB-WME} = 5.26 \times 10^{-14}$, P $F_{inngen-IVW} = 4.10 \times 10^{-30}, P_{Finngen-WME} = 1.82 \times 10^{-18})$ and Capnocytophaga sp002209445 SGB 3500 in tongue dorsum microbiomes (P_{UKB-IVW} = 9.35×10^{-10} , P_{UKB-WME} = 2.17 \times 10⁻⁰⁶, P _{Finngen-IVW} = 2.25 \times 10⁻⁸, and P _{Finngen-WME} = 6.26×10^{-6}). The estimated cause-effect sizes of the SNPs on both the exposure (salivary microbiomes and tongue dorsal microbiomes) and outcome (anxiety) were displayed in scatter plots (Supplementary Figures S3, S4).

The common causal relationship between anxiety and depression

Supplementary Table S7 displayed six shared risk factors between major depression and self-reported anxiety/panic attacks of UKB, including four salivary microbiomes and two dorsal tongue microbiomes. Supplementary Table S8 showed five shared risk factors between depression and anxiety in the Finngen database, including three salivary microbiomes and two dorsal tongue microbiomes. Concurrently, we detected a salivary microbiome to be a risk factor for all anxiety, and depression in the UKB and Finngen databases, *Eggerthia* (P _{IVW-majordepression-UKB} = 2.99×10^{-6} , P _{IVW-Self-reportedanxiety/panicattacks-UKB} = 3.06×10^{-59} , P _{IVW-depression-Finngen} = 3.16×10^{-16} and P IVW-anxiety-Finngen = 1.14×10^{-115}).

Discussion

In the present study, we used logistic and linear regression models to find that multiple significant salivadorsal tongue microbiome interactions were associated with depression/anxiety. Furthermore, we found causal associations between the salivary microbiomes, dorsal tongue microbiomes, anxiety, and depression with the MR analysis, and these associations were confirmed in the Finngen public database.

It is no secret that dysbiosis of the oral microbiome can cause oral diseases like decayed teeth and periodontitis (30).

| nstrumental variable | Name | Method | | | P | OR (Odds 95% Cl |
|---------------------------|--------------------------------------|---------------------------|----------|-------|---------------|--------------------|
| Salivary microbiome | Lancefieldella sp000564995 SGB 1038 | MR Egger | | | 0.711 | 1.148 (0.555-2.372 |
| | | Weighted median | | | 2.33E-13 | 0.961 (0.950-0.97 |
| | | Inverse variance weighted | 1 | | 4.56E-23 | 0.958 (0.950-0.966 |
| | Lancefieldella sp000564995 SGB 1829 | MR Egger | | • | 0.278 | 2.147 (0.548-8.41 |
| | | Weighted median | | | 4.13E-06 | 0.961 (0.944-0.97 |
| | | Inverse variance weighted | | | 7.06E-09 | 0.961 (0.948-0.97 |
| | Lancefieldella uSGB 2805 | MR Egger | H | | 0.779 | 1.122 (0.504-2.49) |
| | | Weighted median | | | 2.66E-14 | 0.961 (0.951-0.97 |
| | | Inverse variance weighted | | | 1.92E-25 | 0.958 (0.951-0.96 |
| | Eggerthia | | H | 4 | 0.028 | 0.903 (0.825-0.98 |
| | 55 | Weighted median | | 4 | 9.83E-08 | 0.989 (0.985-0.99 |
| | | Inverse variance weighted | | ŧ. | 2.99E-06 | 0.993 (0.990-0.99 |
| | Gemella uSGB 2314 | MR Egger | | | 0.991 | 1.008 (0.252-4.03 |
| | | Weighted median | | м | 5.97E-05 | 1.058 (1.029-1.08 |
| | | Inverse variance weighted | | ы | 3.08E-07 | 1 058 (1 036-1 08 |
| onque dorsum microbiome | Lancefieldella uSGB 2940 | MR Egger | ⊢ | | 0.919 | 0.726 (0.002-320.5 |
| angue deletant merepiente | | Weighted median | | ы | 1.81E-07 | 1 117 (1 071-1 16 |
| | | Inverse variance weighted | | M | 5.91E-12 | 1 119 (1 084-1 15 |
| | Lancefieldella uSGB 3453 | MR Egger | | | 0.626 | 1 763 (0 189-16 47 |
| | | Weighted median | | M | 9.91E-07 | 1 117 (1 069-1 16 |
| | | Inverse variance weighted | | ы | 1.04E-10 | 1 121 (1 083-1 16 |
| | RE39 USGB 1766 | MR Egger | | | 0 292 | 0 506 (0 168-1 52 |
| | | Weighted median | | | 8 77E-10 | 0.890 (0.857-0.92 |
| | | Inverse variance weighted | | | 4 24E-15 | 0.895 (0.870-0.92 |
| | Lachnoanaerobaculum umeaense SGB 772 | MR Egger | | | 0.912 | 1 048 (0 459-2 39 |
| | Lacinoanaerobaculum unedense COD //2 | Weighted median | | | 3.63E-14 | 0.917 (0.896-0.93 |
| | | Inverse variance weighted | | | 1.03E-26 | 0.915 (0.900-0.93) |
| | Oribacterium uSGB 1215 | MR Egger | | | 0 129 | 1 071 (0 981-1 17 |
| | Chibacterian acces 1215 | Weighted median | | | 1.02E-82 | 0.929 (0.922-0.93 |
| | | Inverse variance weighted | | | 6 89E-125 | 0.935 (0.930-0.94 |
| | Oribacterium uSCB 1/11 | MP Egger | | | 0.002-120 | 0.637 (0.022-18.02 |
| | Onbacterium dogib 1411 | Weighted median | | | 1 705-05 | 0.007 (0.022-10.02 |
| | | Inverse verience weighted | | | 1 495 10 | 0.005 (0.979 0.03 |
| | Oribectorium uSCP 199 | MP Eggor | | L | 0.630 | 0.967 (0.942 1 11 |
| | Chibacterian acces 405 | Weighted median | | 1' | 1045 44 | 0.007 (0.042-1.11 |
| | | Inverse verience weighted | | | 1 465 61 | 0.022 (0.012-0.00 |
| | Lashnaspirasas uSCR 1003 | MD Egger | | 1 | 0.792 | 1 175 (0 270 2 64 |
| | Lacinospiraceae uSGB 1095 | Weighted median | | ы | 3 775 13 | 1.175 (0.379-3.64 |
| | | weighted median | | | 3.605 19 | 1.078 (1.050-1.10 |
| | TM7:: .:SCB 690 | MD France weighted | | | 3.60E-18 | 1.009 (1.053-1.08 |
| | TM7X USGB 689 | Wirk Egger | | | 1 525 08 | 1.095 (0.270-4.44 |
| | | weighted median | |] | 1.52E-00 | 0.951 (0.934-0.96 |
| | Saasharimanadaaaaa uSOB 1065 | Inverse variance weighted | | | 4.03E-13 | 0.949 (0.936-0.96 |
| | Saccharimonadaceae USGB 1065 | MR Egger | | | 0.802 | 1.255 (0.217-7.25 |
| | | vveighted median | | | 3.17E-07 | 1.194 (1.116-1.27 |
| | | inverse vanance weighted | 0.5 | 1 1.5 | 2.432-10 | 1.102 (1.120-1.20 |
| | | | Odds | Ratio | | |
| | | | | | | |
| | | | | | | |

However, a recent study showed that adolescent anxiety and depression symptoms were associated with the differential abundance of specific oral bacterial taxa, including Actinomyces, Spirochaetaceae, Fusobacterium, and Treponema (31). Moreover, Wingfield et al. (32) examined the structure and composition of the salivary microbiome in young adults with depression and a control group. They revealed that 21 bacterial taxa differed in abundance in the depressed cohort, including increased Neisseria spp. and Prevotella nigrescens, while the abundance of 19 taxa decreased. Although both studies have confirmed the relationship between the oral microbiome and depression, limited research has explained the specific mechanism. A previous study (33) has shown that periodontal bacteria can directly arrive in the brain through the bloodstream or areas with an incomplete or damaged blood-brain barrier (BBB). In addition, periodontitis can indirectly affect the central nervous system through pro-inflammatory cytokines (33). Proinflammatory cytokines activate endothelial cells to express TNF- α and interleukin-1 (IL-1) receptors, transmitting signals to perivascular macrophages that activate microglia, leading to neuroinflammation. Moreover, periodontitis can also lead to leaky periodontium and lipopolysaccharides (LPS) in the systemic circulation, activating the HPA axis, and increasing stress hormones or neurotransmitters (33).

In the dorsal tongue microbiome, we found an interesting bacterial taxon, Porphyromonas. Porphyromonas gingivalis (Pg) is a gram-negative anaerobic pathogen of periodontitis that produces LPS, gingipain, and capsules that damage local periodontal tissue (34). In the etiology of periodontitis, Pg was considered to be the major pathogen. Numerous cross-sectional epidemiological studies have shown that the incidence of periodontitis is positively associated with the incidence of depression (35, 36). However, their causality and latent mechanisms are largely unknown. Nowadays, the neurotrophin deficiency hypothesis of depression has received extensive attention (37). The hypothesis revealed that reducing neurotrophic factors makes the brain unable to adapt to environmental stimulation, which contributes to the onset of depression (37). Brain-derived neurotrophic factor (BDNF) is an important member of the neurotrophic factor family and plays



a key role in the formation and plasticity of neuronal networks (37). According to research, infusion of BDNF into the midbrain or the dentate gyrus (DG) of the hippocampus increases antidepressant-like behavior in mice (38). A recent study has shown that Pg is an underlying risk factor for depression. They injected female mice with Pg every other day for 4 weeks and found that the Pg mice exhibited marked depression-like behavior (39). The possible mechanism is to increase activation of astrocytes in the hippocampus through the Pg-LPS/TLR4 signaling pathway, leading to downregulating astrocyte p75NTR and inhibiting BDNF maturation and, ultimately, depression. In addition, another study (40) also shown that Pg-LPS induces cognitive dysfunction mediated by neuronal inflammation *via* activation of the TLR4 signaling pathway in C57BL/6 mice.

Moreover, we also found some bacterial taxa, such as *Actinobacteria* and *Firmicutes*, live both in the oral cavity and the gut. Surprisingly, whether in the oral cavity or the gut, those bacterial taxa have been linked to depression (31, 41, 42). This may be because, although the composition of the microbiome is site-specific, there was evidence of some degree of overlap and crosstalk between the oral and gut microbiomes, and the oral bacteria may colonize the gut and cause chronic inflammation (43). In addition, microbes and their metabolites in the oral cavity were also likely to migrate or leak into the compromised BBB, leading to neuroinflammation, an important feature of the etiology of depression (44). In conclusion, it is known that periodontal pathogens are closely related to the

etiology and pathophysiology of neuropsychiatric disorders such as depression and schizophrenia, especially immune system dysregulation, which plays an integral role in the etiology and pathophysiology of these diseases (33). Therefore, we can focus on the oral microbiota associated with periodontal disease as a target for future therapeutic interventions to alleviate the symptoms of these debilitating psychiatric disorders. In addition, maintaining a healthy oral microbiome is expected to help improve general and mental health, so we can take some probiotics to prevent mental illness.

At the family level, Actinomycetaceae have also been linked to anxiety and depression (31). Existing research hypothesizes that salivary cortisol is elevated in mental health disorders due to dysregulated HPA axis activity, which may be associated with psychological symptoms and oral bacterial abundance (45). Simpson et al. (31) reported that cortisol regulated the relationship between anxiety and depression symptoms with a variety of microbial taxa. For example, they found that participants with high anxiety symptoms and above-average basal cortisol levels had significantly lower Actinomyces, but the relationship was not observed in participants with high anxiety levels but below average cortisol levels. Moreover, significant interactions were also observed at the order and family levels (Actinomycetaceae and Actinomycetales). Furthermore, at the Phylum level, Firmicutes were consistently proven to be associated with Alzheimer's disease (AD) (46).

After gender stratification, we found significant differences in the salivary-tongue dorsum microbiome interaction between males and females. For the PHQ-9 score, we found five interactions in females and 12 in males; for the GAD-7 score, we found 105 significant interactions in males. This suggested that the effects of the oral microbiome on anxiety and depression were different in males and females. This may be because of gender differences in the oral microbiome. Previous evidence suggested that oral and gut microbiota composition can be regulated by estrogen levels (47), and estrogen receptor- β has been identified in the oral mucosa and salivary glands (48). A previous study (49) analyzed the oral microbiota composition in fasted and fed states in 20 subjects (10 women/10 men). In addition, increased relative abundances of the family Pasteurellacae and the genus Haemophilus in the women were observed in the fed condition. The genus Capnocytophaga was significantly more abundant in the male subjects. In the fasted condition, the main difference between men and women was for the genus Eikenella, which is more abundant in male subjects. Their study showed that men have distinct oral microbiota compared to women in fed or fasted conditions, which may be related to the glycemic response after feeding.

There were several strengths and limitations in the present study. First, this study was based on a large number of study samples, reducing errors due to the small sample size. Second, we utilized the latest GWAS data on the oral microbiome (saliva microbiome and dorsal tongue microbiome), as well as the UK Biobank Cohort genotype data, the results of which improve the ability to detect significant interactions. Finally, we also used data from the Finngen database to verify the causal relationship between oral microbiota and anxiety and depression in the UK Biobank. However, this study still has several limitations that cannot be ignored. First, our results could have been affected by a possible confounding bias because of the impact of various factors on mental disorders, such as early adversity and comorbid illness. Second, changes in host behavior, such as nutrition, can alter the oral microbiome, and mental health conditions like anxiety and depression have also been linked to dietary changes. So, changes in anxiety, depression, and oral microbiome are complex and require further research. Most important, the GWASs data for the oral microbiota in this study were derived from Asian populations, while the genotype data and the GWASs data of anxiety and depression of UK Biobank were from European ancestry, so our findings may not be generalizable across ethnic groups.

In conclusion, despite these limitations, this study systematically explored the relationship between oral microbiota and anxiety and depression. This work highlights the need for more research on the potential role of the oral microbiome in mental health disorders to improve our understanding of disease pathogenesis, potentially leading to new diagnostic targets and early intervention strategies.

Data availability statement

The datasets analyzed during the current study are available in the UK Biobank [https://www.ukbiobank.ac.uk] and the CNGB [https://db.cngb.org/search/project/CNP0001664].

Ethics statement

Ethical review and approval was not required for the study on human participants in accordance with the local legislation and institutional requirements.

Author contributions

All authors listed have made a substantial, direct, and intellectual contribution to the work and approved it for publication.

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Conflict of interest

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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Supplementary material

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/ fpsyt.2022.960756/full#supplementary-material

References

1. World Health Organization. *Depression and Other Common Mental Disorders: Global Health Estimates World Health Organization*. World Health Organization. (2017). Available online at: https://apps.who.int/iris/bitstream/handle/10665/ 254610/WHO-MSD-MER-2017.2-eng.pdf (accessed May 7, 2021).

2. Risal A, Manandhar K, Linde M, Steiner TJ, Holen A. Anxiety and depression in Nepal: prevalence, comorbidity and associations. *Bmc Psychiatry*. (2016) 16:102. doi: 10.1186/s12888-016-0810-0

3. Meier SM, Deckert J. Genetics of anxiety disorders. *Curr Psychiat Rep.* (2019) 21:16. doi: 10.1007/s11920-019-1002-7

4. Otte C, Gold SM, Penninx BW, Pariante CM, Etkin A, Fava M, et al. Major depressive disorder. *Nat Rev Dis Primers.* (2016) 2:16065. doi: 10.1038/nrdp.2016.65

5. Sender R, Fuchs S, Milo R. Revised estimates for the number of human and bacteria cells in the body. *Plos Biol.* (2016) 14:e1002533. doi: 10.1371/journal.pbio.1002533

6. Belkaid Y, Harrison OJ. Homeostatic immunity and the microbiota. *Immunity.* (2017) 46:562–76. doi: 10.1016/j.immuni.2017.04.008

7. Li SJ, Song J, Ke PF, Kong LY, Lei BY, Zhou J, et al. The gut microbiome is associated with brain structure and function in schizophrenia. *Sci Rep-Uk*. (2021) 11:9743. doi: 10.1038/s41598-021-89166-8

8. Cepeda MS, Katz EG, Blacketer C. Microbiome-gut-brain axis: probiotics and their association with depression. J Neuropsych Clin N. (2017) 29:39-44. doi: 10.1176/appi.neuropsych.15120410

9. Hu SH, Li A, Huang TT, Lai JB, Li JJ, Sublette ME, et al. Gut microbiota changes in patients with bipolar depression. *Adv Sci.* (2019) 6:1900752. doi: 10.1002/advs.201900752

10. Tomova A, Husarova V, Lakatosova S, Bakos J, Vlkova B, Babinska K, et al. Gastrointestinal microbiota in children with autism in Slovakia. *Physiol Behav.* (2015) 138:179–87. doi: 10.1016/j.physbeh.2014.10.033

11. Caputi V, Giron MC. Microbiome-gut-brain axis and toll-like receptors in Parkinson's disease. *Int J Mol Sci.* (2018) 19:1689. doi: 10.3390/ijms19061689

12. Konjevod M, Perkovic MN, Saiz J, Strac DS, Barbas C, Rojo D. Metabolomics analysis of microbiota-gut-brain axis in neurodegenerative and psychiatric diseases. J Pharmaceut Biomed. (2021) 194:113681. doi: 10.1016/j.jpba.2020.113681

13. Rhee SH, Pothoulakis C, Mayer EA. Principles and clinical implications of the brain-gut-enteric microbiota axis. *Nat Rev Gastro Hepat.* (2009) 6:306–14. doi: 10.1038/nrgastro.2009.35

14. Louwies T, Johnson AC, Orock A, Yuan T, Greenwood-Van Meerveld B. The microbiota-gut-brain axis: an emerging role for the epigenome. *Exp Biol Med.* (2020) 245:138–45. doi: 10.1177/1535370219891690

15. Mart'yanov SV, Botchkova EA, Plakunov VK, Gannesen AV. The impact of norepinephrine on mono-species and dual-species staphylococcal biofilms. *Microorganisms*. (2021) 9:820. doi: 10.3390/microorganisms9040820

16. Liu L, Huh JR, Shah K. Microbiota and the gut-brain-axis: implications for new therapeutic design in the CNS. *EBioMedicine*. (2022) 77:103908. doi: 10.1016/j.ebiom.2022.103908

17. Maitre Y, Micheneau P, Delpierre A, Mahalli R, Guerin M, Amador G, et al. Did the brain and oral microbiota talk to each other? A review of the literature. *J Clin Med.* (2020) 9:3876. doi: 10.3390/jcm9123876

18. Liu XM, Tong X, Zhu J, Tian L, Jie ZY, Zou YQ, et al. Metagenome-genomewide association studies reveal human genetic impact on the oral microbiome. *Cell Discov*. (2021) 7:117. doi: 10.1038/s41421-021-00356-0

19. Demmitt BA, Corley RP, Huibregtse BM, Keller MC, Hewitt JK, McQueen MB, et al. Genetic influences on the human oral microbiome. *Bmc Genomics*. (2017) 18:659. doi: 10.1186/s12864-017-4008-8

20. Poole AC, Goodrich JK, Youngblut ND, Luque GG, Ruaud A, Sutter JL, et al. Human salivary amylase gene copy number impacts oral and gut microbiomes. *Cell Host Microbe*. (2019) 25:553. doi: 10.1016/j.chom.2019.03.001

21. Crouch DJM, Bodmer WF. Polygenic inheritance, GWAS, polygenic risk scores, and the search for functional variants. *P Natl Acad Sci USA*. (2020) 117:18924–33. doi: 10.1073/pnas.2005634117

22. Yao Y, Qi X, Jia YM, Ye J, Chu XM, Wen Y, et al. Evaluating the interactive effects of dietary habits and human gut microbiome on the risks of depression and anxiety. *Psychol Med.* (2022). doi: 10.1017/S0033291721005092

23. Bycroft C, Freeman C, Petkova D, Band G, Elliott LT, Sharp K, et al. The UK Biobank resource with deep phenotyping and genomic data. *Nature*. (2018) 562:203. doi: 10.1038/s41586-018-0579-z

24. Manichaikul A, Mychaleckyj JC, Rich SS, Daly K, Sale M, Chen WM. Robust relationship inference in genome-wide association studies. *Bioinformatics*. (2010) 26:2867–73. doi: 10.1093/bioinformatics/btq559

25. Sudlow C, Gallacher J, Allen N, Beral V, Burton P, Danesh J, et al. UK Biobank: an open access resource for identifying the causes of a wide range of complex diseases of middle and old age. *Plos Med.* (2015) 12:e1001779. doi: 10.1371/journal.pmed.1001779

26. Burgess S, Butterworth A, Thompson SG. Mendelian randomization analysis with multiple genetic variants using summarized data. *Genet Epidemiol.* (2013) 37:658–65. doi: 10.1002/gepi.21758

27. Bowden J, Smith GD, Burgess S. Mendelian randomization with invalid instruments: effect estimation and bias detection through Egger regression. *Int J Epidemiol.* (2015) 44:512–25. doi: 10.1093/ije/dyv080

28. Bowden J, Smith GD, Haycock PC, Burgess S. consistent estimation in mendelian randomization with some invalid instruments using a weighted median estimator. *Genet Epidemiol.* (2016) 40:304–14. doi: 10.1002/gepi.21965

29. Yavorska OO, Burgess S. Mendelian randomization: an R package for performing Mendelian randomization analyses using summarized data. *Int J Epidemiol.* (2017) 46:1734–9. doi: 10.1093/ije/dyx034

30. Yamashita Y, Takeshita T. The oral microbiome and human health. J Oral Sci. (2017) 59:201–6. doi: 10.2334/josnusd.16-0856

31. Simpson CA, Adler C, du Plessis MR, Landau ER, Dashper SG, Reynolds EC, et al. Oral microbiome composition, but not diversity, is associated with adolescent anxiety and depression symptoms. *Physiol Behav.* (2020) 226:113126. doi: 10.1016/j.physbeh.2020.113126

32. Wingfield B, Lapsley C, McDowell A, Miliotis G, McLafferty M, O'Neill SM, et al. Variations in the oral microbiome are associated with depression in young adults. *Sci Rep-Uk*. (2021) 11:15009. doi: 10.1038/s41598-021-94498-6

33. Martinez M, Postolache TT, Garcia-Bueno B, Leza JC, Figuero E, Lowry CA, et al. The role of the oral microbiota related to periodontal diseases in anxiety, mood and trauma- and stress-related disorders. *Front Psychiatry*. (2021) 12:814177. doi: 10.3389/fpsyt.2021.814177

34. How KY, Song KP, Chan KG. Porphyromonas gingivalis: an overview of periodontopathic pathogen below the gum line. *Front Microbiol.* (2016) 7:53. doi: 10.3389/fmicb.2016.00053

35. Ng SKS, Leung WK, A. community study on the relationship between stress, coping, affective dispositions and periodontal attachment loss. *Community Dent Oral.* (2006) 34:252–66. doi: 10.1111/j.1600-0528.2006.00282.x

36. Hsu CC, Hsu YC, Chen HJ, Lin CC, Chang KH, Lee CY, et al. Association of periodontitis and subsequent depression a nationwide population-based study. *Medicine*. (2015) 94:e2357. doi: 10.1097/MD.00000000002347

37. Castren E, Rantamaki T. The Role of BDNF and its receptors in depression and antidepressant drug action: reactivation of developmental plasticity. *Dev Neurobiol.* (2010) 70:289–97. doi: 10.1002/dneu.20758

38. Hu YH, Russek SJ, BDNF. and the diseased nervous system: a delicate balance between adaptive and pathological processes of gene regulation. *J Neurochem.* (2008) 105:1–17. doi: 10.1111/j.1471-4159.2008.05237.x

39. Wang YX, Kang XN, Cao Y, Zheng DX, Lu YM, Pang CF, et al. Porphyromonas gingivalis induces depression via downregulating p75NTRmediated BDNF maturation in astrocytes. *Brain Behav Immun.* (2019) 81:523– 34. doi: 10.1016/j.bbi.2019.07.012

40. Zhang J, Yu CB, Zhang X, Chen HW, Dong JC, Lu WL, et al. Porphyromonas gingivalis lipopolysaccharide induces cognitive dysfunction, mediated by neuronal inflammation via activation of the TLR4 signaling pathway in C57BL/6 mice. *J Neuroinflamm*. (2018) 15:37. doi: 10.1186/s12974-017-1052-x

41. Jiang HY, Ling ZX, Zhang YH, Mao HJ, Ma ZP, Yin Y, et al. Altered fecal microbiota composition in patients with major depressive disorder. *Brain Behav Immun.* (2015) 48:186–94. doi: 10.1016/j.bbi.2015.03.016

42. Chung YCE, Chen HC, Chou HCL, Chen IM, Lee MS, Chuang LC, et al. Exploration of microbiota targets for major depressive disorder and mood related traits. *J Psychiatr Res.* (2019) 111:74–82. doi: 10.1016/j.jpsychires. 2019.01.016

43. Lira-Junior R, Bostrom EA. Oral-gut connection: one step closer to an integrated view of the gastrointestinal tract? *Mucosal Immunol.* (2018) 11:316-8. doi: 10.1038/mi.2017.116

44. Morris G, Fernandes BS, Puri BK, Walker AJ, Carvalho AF, Berk M. Leaky brain in neurological and psychiatric disorders: Drivers and consequences. *Aust Nz J Psychiat*. (2018) 52:924–48. doi: 10.1177/0004867418796955

45. Lamb AL, Hess DE, Edenborn S, Ubinger E, Carrillo AE, Appasamy PM. Elevated salivary IgA, decreased anxiety, and an altered oral microbiota are associated with active participation on an undergraduate athletic team. *Physiol Behav.* (2017) 169:169–77. doi: 10.1016/j.physbeh.2016.12.004

46. Vogt NM, Kerby RL, Dill-McFarland KA, Harding SJ, Merluzzi AP, Johnson SC, et al. Gut microbiome alterations in Alzheimer's disease. *Sci Rep-Uk.* (2017) 7:13537. doi: 10.1038/s41598-017-13601-y

47. Vieira AT, Castelo PM, Ribeiro DA, Ferreira CM. Influence of oral and gut microbiota in the health of menopausal women. *Front Microbiol.* (2017) 8:1884. doi: 10.3389/fmicb.2017.01884 48. Sawczuk B, Golebiewska M, Mazurek A, Chyczewski L. Evaluation of estrogen receptor-beta expression in the epithelium of the oral mucosa in menopausal women under hormone replacement therapy. *Adv Med Sci.* (2014) 59:85–9. doi: 10.1016/j.advms. 2013.09.003

49. Minty M, Loubieres P, Canceill T, Azalbert V, Burcelin R, Terce F, et al. Gender-associated differences in oral microbiota and salivary biochemical parameters in response to feeding. *J Physiol Biochem.* (2021) 77:155–66. doi: 10.1007/s13105-020-00757-x