

# Signature of Microbial Dysbiosis in Periodontitis

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1	Signature of microbial dysbiosis in periodontitis.
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5	
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10 11	Running Head: Dysbiosis in Periodontitis
12	
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15	

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#### 17 Abstract

with failed resolution of the chronic destructive inflammation. The objective of this study is to identify microbial signatures for health and chronic periodontitis at the genus level and to propose a model of dysbiosis including the calculation of bacterial ratios. Design and methods. Published sequencing data obtained from several different studies (196 sub-gingival samples from patients with chronic periodontitis and 422 sub-gingival samples from healthy subjects) were pooled and submitted to a new microbiota analysis using the same VAMPS (Visualization and Analysis of Microbial Population Structures) analysis pipeline to identify microbiota specific to health and disease. Microbiota were visualized using CoNet and Cytoscape. Dysbiosis ratio, defined as the percentage of genera associated with disease relative to the percentage of genera associated with health, were calculated to distinguish disease from health. Correlation between the proposed dysbiosis ratio and the periodontal pocket depth was tested on a different set of data obtained from a 33 recent study to confirm the relevance of the ratio as a potential indicator of dysbiosis. 34 Results. Beta diversity showed significant clustering of periodontitis-associated microbiota 35 according to clinical status at the genus level, independent of methods used. Specific genera 36 were highly prevalent (>95%) in health (Veillonella, Neisseria, Rothia, Corynebacterium and 37 Actinomyces) while other genera were associated with chronic periodontitis (Eubacterium, 38 Campylobacter, Treponema and Tannerella). The calculation of dysbiosis ratio based on the 39 relative abundance of these genera found in health versus periodontitis was tested. Non-40 periodontitis samples were significantly identifiable by low ratios as compared to chronic

Periodontitis is driven by a disproportionate host inflammatory immune response induced

by an imbalance in the composition of oral bacteria. It instigates microbial dysbiosis along

41 periodontitis samples. When applied to a sub-gingival set with well-defined clinical data, the 42 method showed a strong correlation between the dysbiosis ratio, as well as a simplified ratio 43 (Porphyromonas, Treponema and Tannerella to Rothia and Corynebacterium), and pocket 44 depth. 45 Conclusion. Microbial analysis of chronic periodontitis can be correlated with the pocket 46 depth through specific signatures of microbial dysbiosis. 47 48 Importance 49 Defining microbiota typical of oral health or chronic periodontitis is a difficult task. 50 Periodontal disease evaluation is currently based on probing of the periodontal pocket. However, the status of pockets "on the mend" or sulci at "risk of periodontitis" cannot solely 51 52 be addressed through pocket depth measurements, nor by current microbiological tests 53 available for practitioners. Thus, a more specific microbiological status of dysbiosis could 54 help in future diagnoses of periodontitis. In this work, data from different studies were 55 pooled to improve the accuracy of the results. However, analysis of multiple microbiota from 56 different studies intensified the bacterial network and complicated the search for 57 reproducible microbial signature. Despite different methods used in each study, the 58 microbiota investigated at the genus level showed that some genera are prevalent (up to 59 95% of the samples/microbiota) in health or disease, allowing the calculation of bacterial 60 ratios (*i.e.* dysbiosis ratios). The correlation between the proposed ratios and the 61 periodontal pocket depth was tested which confirmed the link between dysbiosis ratios and 62 the severity of the disease. The importance of this work is promising but longitudinal studies 63 will be required to improve the ratios and define microbial signatures of the disease, which

- 64 will allow the monitoring of the periodontal pocket recovery and conceivably, the potential
- 65 risk of periodontitis in healthy patients.

## 66 Introduction

67	Chronic periodontitis (CP) is a chronic inflammation characterized by alveolar bone loss with
68	intermittent periods of remission and relapse. CP is currently considered as an infection
69	mainly due to an increase of bacteria in the sulcus leading to the formation of a periodontal
70	pocket (for review: 1, 2). The major pathogen linked to CP is Porphyromonas gingivalis with
71	bacterial partners like Treponema denticola and Tannerella forsythia. These three bacteria
72	have been considered as the major pathogenic "red complex" since 1998 (3). However,
73	recent advances from metagenomics studies developed a new model of periodontal disease
74	pathogenesis. CP does not result from individual pathogens but rather from polymicrobial
75	synergy and dysbiosis (4) associated with a dysregulated immune response inducing
76	inflammation-mediated tissue damage (5). Host genetic components have also been
77	implicated in CP, where multiple genes contribute cumulatively to the hosts overall disease
78	risk (or protection) through effects on the host immune response and the microbiome (6).
79	Since the Human Microbiome Project (HMP) (7), microbiota were analyzed based on partial
80	sequencing of the 16S rRNA gene with different numbers of healthy and CP samples.
81	However, comparison between studies is difficult because of the differences in methods
82	used ( <i>i.e.</i> clinical examination and diagnosis of periodontitis and oral health, sample
83	collection protocols, DNA extraction protocols, hyper variable regions of the 16S rRNA gene
84	analyzed). As there is a growing interest in the human microbiome, despite the difficulties
85	mentioned earlier, the use of independent studies to look for "Signal in the Noise" should
86	proceed as previously suggested (8), by reanalyzing all data with the same protocol. The
87	difference between periodontal health- and disease-associated microbiota should be larger
88	than the technical variations of the different studies and enable the identification of
89	microbial signatures using NGS technologies. The first objective of this study was to explore

90	the disease-associated changes in the sub-gingival microbiota using a unique VAMPS
91	(Visualization and analysis of microbial population structure) pipeline (9) at the genus level
92	for beta-diversity (Bray-Curtis dissimilarity) on a large number of samples (from 6 different
93	studies) and to confirm that the microbiota identified did not cluster according to the
94	methods used (primer or study type). Sub-gingival microbiota from patients with diagnosed
95	chronic periodontitis (196) and from healthy subjects (422) were included as well as external
96	control samples (from dentine caries, supra-gingival plaque and the mid-vagina). The second
97	objective was to determine a dysbiosis ratio of bacteria that could predict health or severity
98	of the disease from the sub-gingival samples and finally to test it on an independent cohort
99	of patients with well-described periodontal pocket measurements.

## 100 MATERIAL and METHODS

10	)1	Microbiome datasets for comparison. Read sequences from healthy and chronic
10	)2	periodontitis sub-gingival samples of five different studies <i>i.e.</i> Abusleme <i>et al.</i> (10), Kirst <i>et</i>
10	)3	al. (11), Griffen et al. (12) (shallow site samples also included), Zhou et al. (13), and Camelo-
10	)4	Castillo et al. (14) were retrieved from either the NCBI Sequence Read Archive (SRA), or
10	)5	the MG-RAST server (Table 1). Twenty-four samples analyzed using V3V4 primer from
10	)6	patients with chronic periodontitis recruited between June 2010 and September 2011 at the
10	)7	University Hospital (Rennes, France) were added (publication in progress). Each data set
10	)8	was manually imported into VAMPS, while numerous healthy sub-gingival samples were
10	)9	added from the HMP (2 different sub-gingival datasets using V1V3 and V3V5 primers
11	10	available in VAMPS (9), <a href="https://vamps.mbl.edu/">https://vamps.mbl.edu/</a> ). Three mouth control microbiota datasets
11	1	were used from the HMP (saliva and supra-gingiva, both V1V3 and V3V5 regions, available in
11	12	VAMPS) and dentine caries from Kianoush <i>et al.</i> (V3V4 regions, PRJEB5178) (15). One mid
11	13	vagina microbiome dataset from the HMP (V1V3 region, available in VAMPS) was used as an
11	14	external mouth control.
11	15	Finally, the dataset from Bizzarro S. et al. (16), containing well-described sample
11	16	pocket depth (from 2 to 8 mm) was used to independently challenge the relevance of the
11	17	dysbiosis ratio of bacteria involved in periodontitis.
11	18	
11	19	Ecology diversity and taxonomy identifications. Reads from the different datasets were
12	20	analyzed using VAMPS using default parameters for taxonomy assignment to the genus level
12	21	through the GAST process using the RDP classification to produce the best taxonomic
12	22	assignment for each read. Reads identified as Archaea, Eukarya, Organelle and unknown
12	23	reads were excluded for further analysis. The frequency of each taxonomic assignment

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124	in the dataset was reported as a percentage (number of reads assigned to a taxonomy
125	over total number of reads in the dataset). Alpha diversity as observed richness and
126	Shannon-Weaver index were determined from the raw data sets. Differences between
127	microbiota structures (beta-diversity) were assessed using principal coordinate analysis
128	(PCoA) 2D tree on the Bray-Curtis distance through VAMPS. Samples were divided into five
129	clusters (1 to 5), where visualizations were realized using Figtree software (v1.4.2) and 3D
130	PCoA plots were generated using Emperor software. Relative abundances were studied
131	when the average abundance was above 1% in at least one sample. Assessments of
132	significant patterns of microbial co-occurrence or mutual exclusion at the genus level were
133	performed using Cytoscape v3.2.1 (17) and the CoNet plugin (18). Only genera found in the
134	high majority, in at least 95%, of the healthy sub-gingival samples or the CP samples (from
135	the fifth cluster) are represented.
136	
137	Dysbiosis calculations – ratios of bacteria: To measure the dysbiosis, a first ratio based on
138	the relative abundance of genera highly prevalent (>95%) in CP samples (Eubacterium,
139	Campylobacter, Treponema and Tannerella) to genera highly prevalent (>95%) in healthy
140	microbiota (Veillonella, Neisseria, Rothia, Corynebacterium and Actinomyces) was calculated.
141	The ratios were normalized between samples using GraphPad Prism V6 software before
142	comparison. A second simplified ratio of Porphyromonas, Treponema and Tannerella to
143	Rothia and Corynebacterium was also tested. When no specific genus was detected and as

144 "no detection" does not mean "absence", a value of 0.1% was attributed.

145

146 Statistical analysis. Normality tests for data distribution were realized. Data were studied by

147 Spearman correlation test between biological origins, primers used, publication of sample

148	origins and microbiota clusters. Observed richness (S, number of taxa per sample), Shanon-
149	Weaver index and dysbiosis ratio of the genera found in disease to the genera found in
150	health were analyzed with a non-parametric Anova Kruskal-Wallis test. Tests were carried
151	out using GraphPad Prism V6 software and were considered significant when p <0.05. The
152	significant patterns of microbial co-occurrence and mutual exclusion were analyzed as
153	described by Faust (18): a compilation of statistical analysis (Spearman, Pearson correlations
154	and Bray Curtis, Kullback-Leibler dissimilarity measures) was used with a threshold set at 0.5.
155	The data matrix was randomized by 100 row-wise permutations. The P values were adjusted
156	by Benjamini-Hochberg false discovery rate (FDR) correction for the number of tests,
157	retaining only p <0.05. Finally, the ratios of genera and pocket depth were controlled for

158 normality followed by Spearman correlation test.

## 159 **RESULTS**

160	Microbial community structure analysis. Using a matrix correlation analysis, the possible
161	clustering of microbiota according to nature of primers used, the site of sampling, or the
162	study investigated was explored. Despite various studies, the analyzed data clustered into
163	five groups according to the clinical status (healthy or CP) or sampling site as shown by the
164	3D PCoA plots (Fig. 1). Healthy sub-gingival samples were, in majority, spread into two main
165	clusters; control samples were clearly separated in two other clusters corresponding to
166	saliva and dentine caries/vagina, while the majority of CP samples were found in a fifth
167	cluster. Two-D beta-diversity analysis showed the precise distribution of the samples in the
168	five clusters (Fig. 2). The search for an association between clusters and primers and/or
169	study type showed that the fourth cluster was associated with V3V4 16S rRNA primers
170	(correlation r=0.537, p<0.001) and with the Kianoush <i>et al.</i> study (15) that has used these
171	specific primers (correlation r=0.608, p<0.001). No other correlation with primers was found.
172	The 2 healthy clusters (1 and 2) were characterized by sub- and supra-gingival samples in
173	similar proportions (Fig. 2, in blue). Focusing on healthy sub-gingival samples, the main
174	difference between the two healthy clusters 1 and 2 was the distribution of samples from
175	the HMP study and from the other studies in the clusters: 225/323 of the HMP study
176	clustered into the healthy cluster 2 while the healthy cluster 1 was richer in samples deriving
177	from the other studies (44/99). Cluster 3 was characterized by saliva as 91% of the saliva
178	samples (258/284) are grouped within this cluster (Fig. 2, correlation r=0.892, p<0.001).
179	Cluster 4 was characterized by samples from dentine caries 73% (80/110) and mid-vagina
180	100% (60/60) (correlation r=0.603 and r=0.638 respectively, p<0.001). Finally, the fifth
181	cluster contained 90% of the CP samples (176/196, correlation r=0.708, p<0.001).

182 It is interesting to note that 10% of the CP samples were found in the 2 healthy clusters
183 (19/196) and contained similar microbiota (analyzed by beta-diversity) as dentine caries
184 and/or mid-vagina (1/196) at the genus level. Conversely, 16% of the healthy sub-gingival
185 samples (69/422) and 17% of the dentine caries samples (19/110) were found in the fifth
186 cluster.

187

188 Microbiota richness and alpha-diversity in sub-gingival samples: Cluster comparison 189 showed that sampling depth (number of reads sequenced) was higher in healthy sub-gingival 190 clusters 1 and 2 than in the fifth cluster. Nevertheless, no significant difference between 191 healthy sub-gingival and CP samples of the fifth cluster was found (Fig.3). The observed 192 richness (S) was lower in the CP samples from fifth cluster than samples of both healthy 193 clusters 1 and 2 and of healthy sub-gingival samples from the fifth cluster (Fig.3). However 194 the Shannon Weaver diversity index showed that the diversity of healthy cluster 2 was 195 significantly higher than the diversity index that is similar in healthy cluster 1 and of all 196 samples from the fifth cluster. 197 198 Patterns of microbial communities in sub-gingival samples (genus level): Genera present in, 199 at least, 95% of all healthy sub-gingival samples or 95% of the CP samples from the fifth 200 cluster are presented in figures 4A and 4B respectively. Results showed that healthy sub-201 gingival samples are dominated by 8 major genera, Fusobacterium, Actinomyces, 202 Streptococcus, Neisseria, Capnocytophaga, Prevotella, Corynebacterium, and Rothia, and by 203 6 minor genera, Leptotrichia, Veillonella, Porphyromonas, Granulicatella, Kingella and 204 Gemella. Associations were found between Fusobacterium and Prevotella, Actinomyces and 205 Rothia and between Leptotrichia and Porphyromonas. Common genera found in CP were

207 Fusobacterium followed by Streptococcus, Eubacterium, Tannerella and Campylobacter 208 genera. Only one association was found between Eubacterium and Treponema, while 209 Fusobacterium and Treponema presented a negative correlation. 210 211 Dysbiosis calculation - ratio of bacteria: The dysbiosis ratio of genera found mainly in 212 chronic periodontitis (Eubacteria-Campylobacter-Treponema-Tannerella) to the genera 213 found mainly in health samples (Veillonella-Neisseria-Rothia-Corynebacterium-Actinomyces) 214 was significantly different between samples according to their diagnosis. The dysbiosis ratio 215 of healthy sub-gingival samples (from HMP, n=323, r=0.016 and from the other studies, 216 n=99, r= 0.021) obtained a median r=0.018, shallow sites r=0.071 and samples from chronic 217 periodontitis r=1.229 (p<0.001) (figure 5A). 218 Despite achieving a different clustering through beta-diversity, no significant difference was 219 found between the ratios of cluster 1 and 2 according to the clinical status (healthy, shallow 220 and CP). Pooling of samples according to clinical status was done and the resulting ratios 221 were compared to ratios of the fifth cluster as shown in figure 5B. 222 The dysbiosis ratio found in CP samples from the fifth cluster (r=1.510) was significantly 223 higher than the ratios of the majority of samples from cluster 1 and 2 (healthy sub-gingival 224 r=0.015, shallow r=0.052 and CP samples r=0.088), and was also significantly higher than 225 healthy sub-gingival samples (r=0.184) from the same fifth cluster (p<0.001). In clusters 1 226 and 2, the dysbiosis ratio of CP samples was similar to the ratio of shallow sites. These two 227 groups were significantly different from the healthy sub-gingival samples (p<0.05) in the 228 same cluster.

less abundant with 4 major genera, Treponema, Porphyromonas, Prevotella and

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229	Healthy sub-gingival samples (n=69) belonging to the fifth cluster exhibited a dysbiosis ratio
230	(r=0.184) significantly different from the other healthy sub-gingival samples (r=0.015) but
231	also from the majority of the CP samples (fifth cluster, r=1.510). These results confirmed the
232	possible difference of these healthy sub-gingival microbiota (p<0.001) as compared with
233	those of healthy clusters 1 and 2. Their ratio is also not significantly different from the CP
234	sample ratio in cluster 1 and 2, which could be considered "on the mend".
235	
236	Validation of the dysbiosis ratio
237	A different dataset from Bizzarro S. <i>et al.</i> (16), containing well-described samples (pocket
237 238	A different dataset from Bizzarro S. <i>et al.</i> (16), containing well-described samples (pocket depth from 2 to 8 millimeters), was used as an external control to confirm the relevance of
238	depth from 2 to 8 millimeters), was used as an external control to confirm the relevance of
238 239	depth from 2 to 8 millimeters), was used as an external control to confirm the relevance of the dysbiosis ratio of bacteria. The dysbiosis ratio at the genus level was correlated with the

243 simplified ratio of *Porphyromonas, Treponema* and *Tannerella* to *Rothia and* 

244 *Corynebacterium* showed a similar correlation (r=0.659, p<0.001) (Fig. 6B).

## 245 **DISCUSSION**

246	Many studies have been published since the human microbiome project in 2009, increasing
247	the number of microbiota data available for the research community. However, comparison
248	between studies is challenging, at least at the species level, because of the use of different
249	methods. This issue is a real limitation to understand disease as well as the low number of
250	samples in each study. Additionally, it is more complicated for healthy sub-gingival status
251	that usually represent less than a half of the samples when included in studies (10, 14). This
252	work is a taxon-based analysis at the genus level of sequence reads from several studies.
253	Studying a large number of samples minimized individual variations and overcame technical
254	variations by increasing the effective sample size. Such analysis has already been proposed
255	in a recent microbiota obesity study (8). Studies with described healthy (sulci $\leq$ 3mm) and CP
256	(pocket depth $\geq$ 5mm) samples and available raw sequenced data in data banks were chosen.
257	Data from the HMP resources (two different couples of primers used) were added to
258	increase the number of sub-gingival healthy microbiota data available from 99 up to 422.
259	The different microbiota clustered either by sampling site, such as out-groups used as
260	controls for this study (saliva in cluster 3, dentine caries/vagina both rich in Lactobacillus in
261	cluster 4), or by clinical status, such as sub-gingival samples (healthy samples in cluster 1 and
262	2 and CP samples in the fifth cluster). CP sites can either show a greater microbial diversity
263	and observed richness than healthy sub-gingival sites (12, 19), or present no significant
264	difference in microbial diversity as it has been also reported between health and
265	periodontitis (11). Thus, the high number of samples surpasses the technical variations, at
266	least at the genus level with the primers used in these different studies, and the difference
267	between periodontal health and disease is larger than the technical variations, as described
268	by Kirst et al. (11). No difference was found between the healthy sub- and supra-gingival

269	samples when using beta-diversity analysis at the genus level, as previously described (20).
270	Ninety per cent of CP samples were found in the fifth cluster. To define the fifth cluster as a
271	"periodontitis cluster" by beta-diversity was appealing. However, the fifth cluster also
272	contained healthy sub-gingival samples, indicating that further investigations were necessary
273	to understand and develop prediction markers for chronic periodontitis.
274	A core community is usually identified in publications (genera present in at least 50% of the
275	samples) and provides a basis for disease diagnosis, prevention and therapeutic targets (21,
276	22). However, the genera variability expands as sample size increases, thus limiting its use
277	to establish an easy microbiological marker for dysbiosis. In this work, genera present at a
278	higher prevalence in at least in 95% of the samples were used to determine the genera
279	implicated in health or in favor of the disease. The genera used to calculate the dysbiosis
280	ratio in favor of periodontitis were Treponema, Campylobacter, Eubacterium and Tannerella.
281	These genera were identified at high abundance and high prevalence in CP as compared to
282	healthy samples. They include well-identified species (T. forsythia and T. denticola, C. rectus
283	and <i>E. nodatum</i> ) that are strongly associated with the disease (3, 23-25). It is noted that,
284	while some species such as the newly-cultivated <i>Tannerella</i> clone BU063 (26, 27) that is
285	supposed health-associated, is also found in active periodontal sites (28) and therefore still
286	considered controversial. Despite a significant difference in abundance of the
287	Porphyromonas genus (which includes P. gingivalis that is highly associated with
288	periodontitis) between healthy samples ( $3.35\%$ ) and in CP samples ( $13\%$ ), it was
289	excluded in the first dysbiosis ratio because of its similar prevalence. As the lowest
290	abundance in genus accounted for the CP calculation was Campylobacter, 1.9%, this value
291	was therefore chosen as a cut-off to minimize the number of genera used for the health
292	calculation - Rothia, Corynebacterium, Actinomyces, Veillonella and Neisseria. Neither

293	Capnocytophaga nor Leptotirichia were included because of their high prevalence in
294	periodontitis samples (more than 90%, data not shown). Species belonging to the genus
295	Rothia have been repeatedly described as members of oral communities associated with
296	periodontal health (29-34) or at least being more predominant in health (31). In the same
297	way, Corynebacterium appeared to be more associated with "healthy" sub-gingival biofilm
298	(35, 36). Moreover, Rothia and Corynebacteirum were among the bacteria that showed the
299	greatest increase after periodontal treatment (37), while a study suggested that
300	Corynebacterium might be considered as a putative periodontal protector (38). Veillonella
301	and Actinomyces have been negatively correlated with clinical markers in CP (39) and
302	Neisseria was found in inactive sites (28). The calculated dysbiosis ratio distinguishes clearly
303	healthy sub-gingival samples from CP samples.
304	Shallow samples were divided into two groups that can be easily explained based on the
305	origin of the sampling (healthy sub-gingival sites in mouth presenting chronic periodontitis).
306	Two thirds of the samples had a low ratio (cluster 1 and 2) and can be considered
307	"microbiologically healthy". The remaining one third of the samples (cluster 5) presented a
308	high ratio certainly due to contamination of the sampling sites by bacteria from surrounding
309	CP sites and could be considered as "at risk of periodontitis". Thus, shallow samples may
310	represent an intermediate stage in disease development as proposed by Griffen et al. (12).
311	Healthy sub-gingival samples are divided into 3 different groups. Two of them belonging to
312	clusters 1 and 2 present the same low ratio and describe an absence of dysbiosis. The third
313	group had a higher dysbiosis ratio, similar to shallow sites and CP samples from cluster 1 and
314	2 but significantly inferior to CP or shallow samples from the fifth cluster. As "healthy
315	patients" from HMP are defined as patients with pockets depths <4mm, some of them could
316	have explained this high ratio group, however healthy patients from other studies (19/99)

317	were also included in this group. This result is similar to Zhou Y. et al. where few healthy
318	subjects were detected with signals of disease such as an increase of Treponema (40).
319	Therefore, patients who presented a relatively high ratio as such could be considered at "risk
320	of periodontitis".
321	Conversely, a few CP samples with deep periodontal pockets ( <i>i.e.</i> $\geq$ 5mm) had a low dysbiosis
322	ratio. A hypothesis of appropriate host response (such as a stronger immune response
323	and/or better hygiene) could explain this discrepancy between dysbiosis ratio and diagnosis:
324	these patients might be microbiologically "on the mend", as revealed by both clustering and
325	dysbiosis ratio. Another hypothesis is a sampling concern between the top and base of the
326	periodontal pocket (to be discussed later). To study the microbiota "on the mend"
327	hypothesis and the dysbiosis calculated by the ratio, a recent study presenting a follow-up
328	after treatment with well-defined depth of periodontal pocket metadata was used (16). This
329	study was conducted with a different set of primers (V5V7) and allowed to test the dysbiosis
330	ratio at the genus level on a new set of primers that have not been used to determine the
331	ratio. Consequently, this comparative analysis can be considered as a validation experiment
332	of the ratios. A strong correlation was obtained between the dysbiosis ratio and the pocket
333	depths, thus highlighting the value of calculating the dysbiosis ratio (using the selected
334	genera of our study) as a microbial signature to evaluate the microbiota of chronic
335	periodontitis.
336	A major concern at the beginning of this work was the capacity to identify species with
337	multiple datasets. However the V1V2 and V5V7 primers used in 3 studies were not suitable
338	for species identification. At the genus level, as reported in Bizzarro S. et al. (16), the
339	proposed dysbiosis ratio is a good microbial signature calculated using the online VAMPS
340	software even by using a simplified dysbiosis ratio. Indeed, as Rothia and Corynebacterium

341	are the major healthy genera found, and even if Porphyromonas was found in both health
342	and disease, its abundance increased significantly in disease (from 3.34% to 13%). The result
343	was interesting, where it was found to be similar to the precedent ratio (correlation with
344	pocket depth r=0.659 p<0.001). However, more adjustments were needed as 43 out of 196
345	CP samples presented none of the two healthy genera and a value of 0.1% was attributed for
346	the calculation.
347	Finally, using ratios, some data points still showed discrepancies in predicting the
348	periodontal status. The variability in microbial composition and spatial distribution could
349	explain these results. Deep periodontal pockets in CP patients may present gradients of
350	oxygen tension, pH and nutrients as well as host defense factors from the base of the pocket
351	to the top (opening). This may explain why some genera are typically found at the base of
352	the pocket (Porphyromonas, Treponema) (41, 42). However, the sampling could induce bias
353	even after careful removal of the supra-gingival plaque. Healthy genera may be found
354	predominantly at the top (opening) of the pocket as compared with the genera more closely
355	associated with CP being located at the base of the pocket. Indeed, with the use of NGS
356	analysis, while the architecture of the periodontal pocket has not been clearly studied yet,
357	the importance of the biogeography of microbiome at the micron scale has clearly been
358	shown recently (43).
359	
360	In conclusion, this study aimed to define ratios of bacteria as microbial signature after the
361	analysis of public raw data from different studies, independent of the technical methods
362	used to generate the data. These ratios allow the differentiation of healthy and diseased

363 microbiota in a majority of samples. Standardized protocols of sampling and complete

364 metadata in the public bank are necessary to study dysbiosis in oral health and to improve

the proposed dysbiosis ratios. Adjunction of specific perio-protectors and potential specific
pathogens to the calculation of the dysbiosis could also be promising. Longitudinal studies
are necessary to predict exact pockets microbiologically "on the mend" or sulci with "risk of
periodontitis".

AEM

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- 501 Author Contributions (names must be given as initials)
- 502 VM, FBH and MBM conceived and designed the research. VM realized the sampling. SLGD
- 503 realized the molecular biology. VM, SLGD, EB, LAA, BM, SBF, BM, performed the experiments
- 504 (bioinformatics analysis, statistics) and wrote the manuscript. MBM supervised the project.

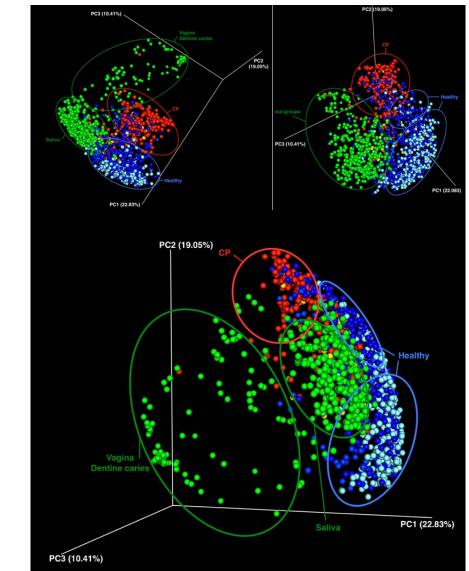
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#### 506 **Competing interests**

507 The authors declare no competing financial interests.

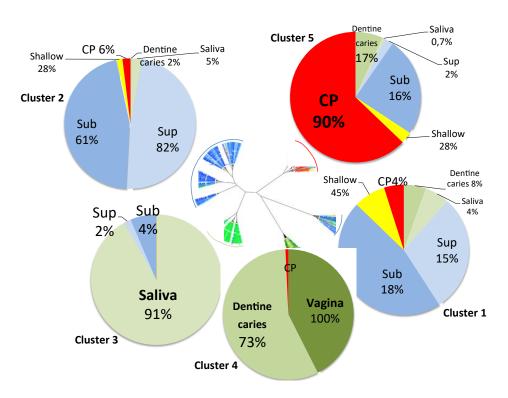
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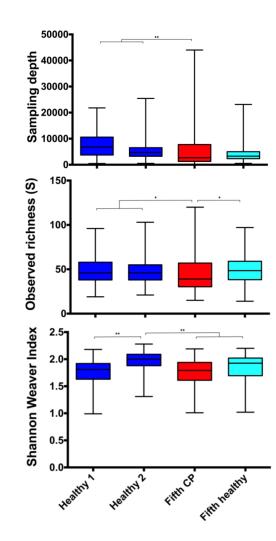


- 510 Fig. 1: Different views of 3D PCoA plots illustrating the beta-diversity of bacterial
- 511 populations as a function of sampling site and diagnosis. Light blue: supra-gingival samples;
- 512 dark blue: healthy sub-gingival samples; green: out-groups as saliva, mid-vagina and dentine
- 513 caries; red: chronic periodontitis CP).
- 514



516 Fig. 2: Unrooted tree displaying genus Bray Curtis beta-diversity clustering microbiota and 517 pie charts related to sample origin within each cluster. The tree was realized using Figtree 518 software v1.4.2. Distribution of microbiota in each cluster is represented by pie charts with 519 different colors according to sampling sites (supra-gingival: Sup in light blue, saliva: light 520 green, dentine caries: green and mid vagina: dark green) and diagnosis for sub-gingival 521 samples (healthy: Sub in dark blue, shallow in yellow and chronic periodontitis in red). 522 Written percentages correspond to the number of samples from a given specific sampling 523 site in a given cluster to the total number of samples from the same specific sampling site.

524





526 Fig. 3: Alpha diversity index. Microbiota sampling depth, observed richness (S, number of

527 different taxa per sample) and diversity (Shannon Weaver index) comparisons in sub-gingival

528 samples between samples of healthy clusters 1 and 2 (blue) and samples of cluster 5, either

529 with chronic periodontitis (red) or sub-gingival healthy samples (light blue). \*p<0.05,

- 530 \*\*p<0.01.
- 531

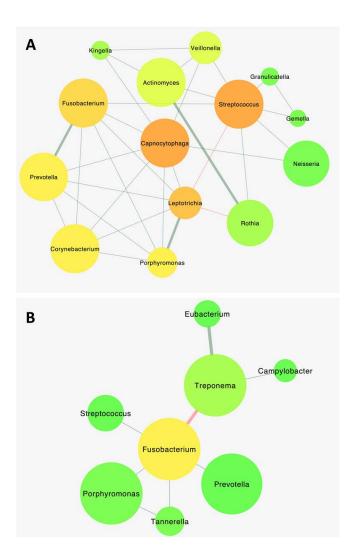
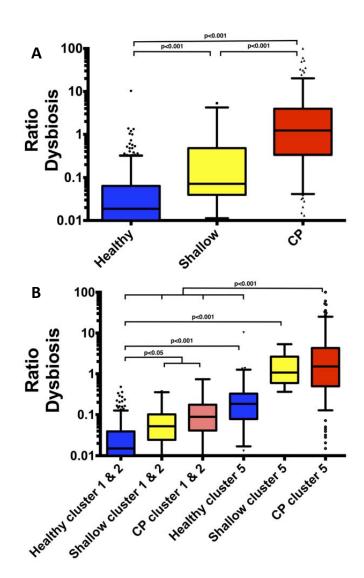




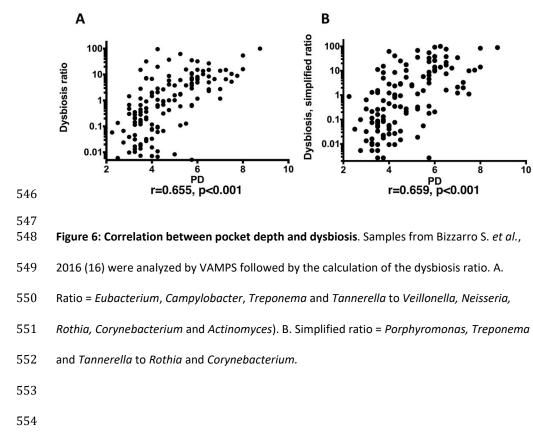
Fig. 4: Patterns of sub-gingival microbial communities. A: Patterns of genera present at
least in 95% of all healthy sub-gingival samples. B: Patterns of genera present at least in 95%
of all chronic periodontitis (CP) samples from the cluster 5. Edges represent one (thin line) or
2 to 3 (thick line) significant correlation between genera (green: positive; red: negative).
Node colors represent the number of partners ranging from one (green) to 7 (dark orange).
Node sizes represent the abundance of each taxon.







- 542 Tannerella to Veillonella, Neisseria, Rothia, Corynebacterium and Actinomyces. A. Between
- 543 healthy, shallow and chronic periodontitis samples (CP) from all clusters. B. Between clusters
- 544 1&2 and Cluster 5 for healthy, shallow and CP samples.
- 545



#### 556 Table 1: Sub-gingival microbiota used in this study

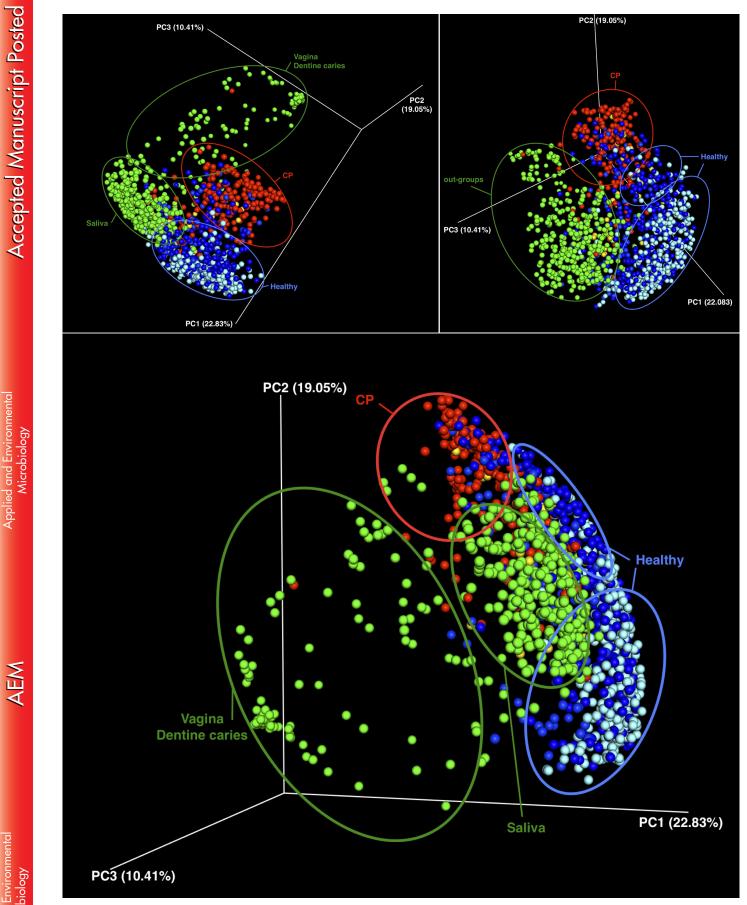
		Number of sub-gingival microbiota				
Authors	Accession Number	Health	Shallow <sup>#</sup>	СР		16S rRNA
Autions				Diagnosis	Follow up after treatment	gene regions
Abusleme <i>et al.</i> (10)	SRA051864	10		44		V1V2
Kirst <i>et al.</i> (11)	PRJNA269205	25		25		V1V3
Griffen <i>et al.</i> (12)	SRP009299	29	29	29		V1V2
Gimen <i>et ul.</i> (12)	3KP009299	29	29	29		and V4
Camelo-Castillo <i>et</i>	MG-RAST: ID	22		60		V1V3
al. (14)	12161	22		00		1115
Zhou <i>et al.</i> (40)	SRA062091	13		18		V1V3
Meuric et al.	In progress <sup>\$</sup>			24		V3V4
HMP (7)		119				V1V3
HMP (7)		204				V3V5
Bizzarro <i>et al.</i> (16)	PRJNA289294			37*	110*	V5V7

557 \* CP microbiota from patients with a follow up treatment used to confirmed the dysbiosis

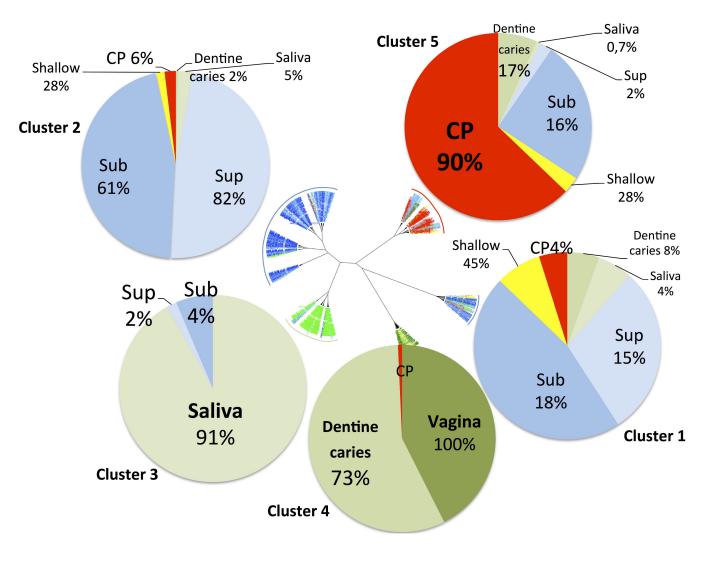
558 ratios hypothesis (16).

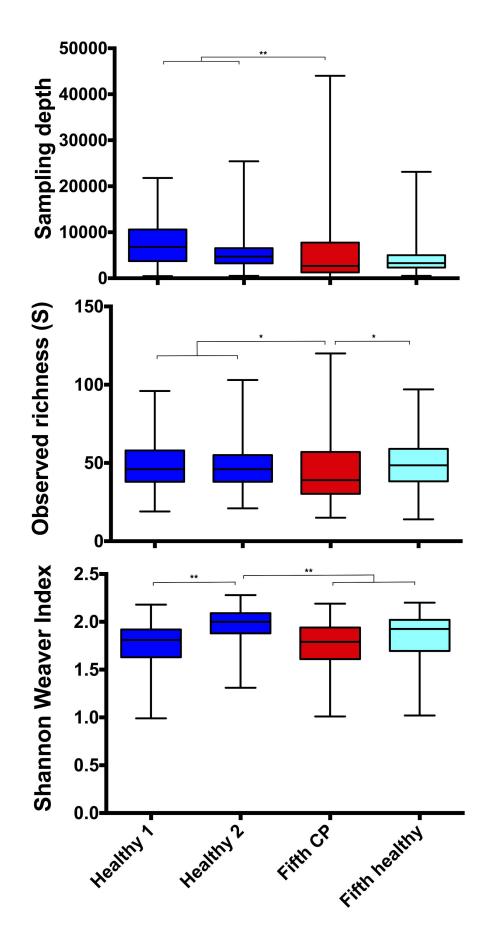
<sup>\*</sup>Site defined as healthy in patients with periodontitis (12).

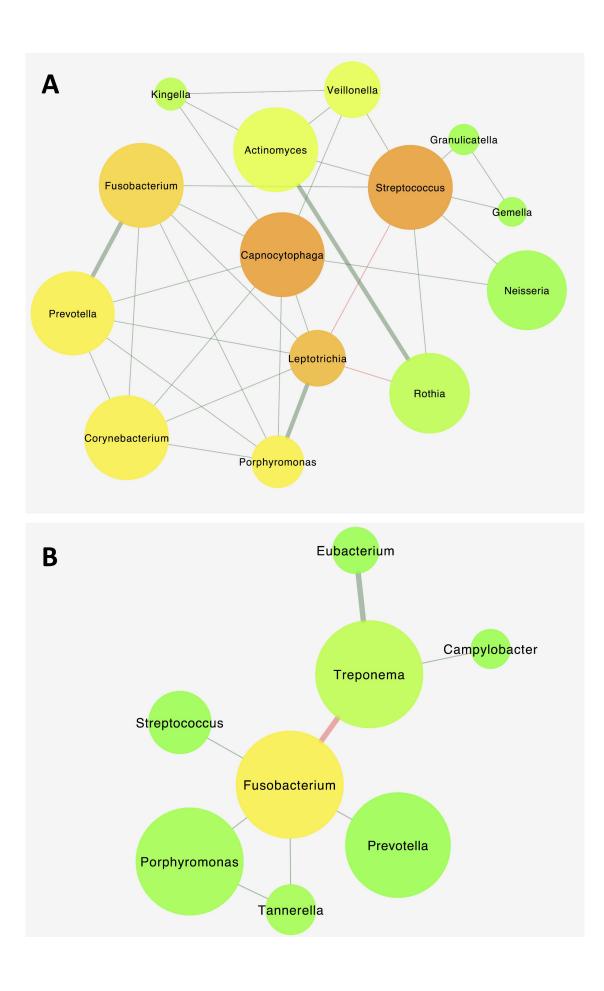
560 <sup>\$</sup> Data available on VAMPS for the reviewers, dataset: "Y\_Hemoparo"



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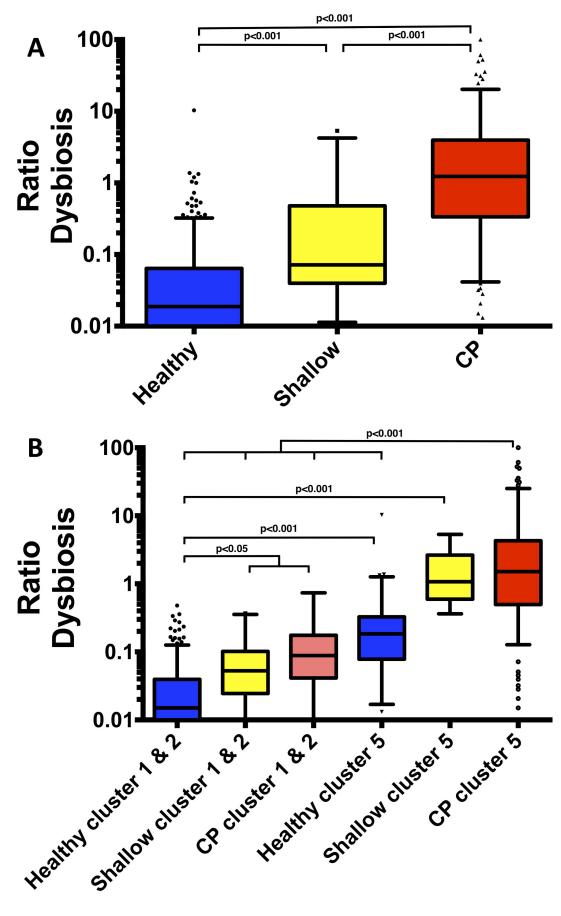






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