



ORIGINAL ARTICLE

Community-wide transcriptome of the oral microbiome in subjects with and without periodontitis

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Despite increasing knowledge on phylogenetic composition of the human microbiome, our understanding of the in situ activities of the organisms in the community and their interactions with each other and with the environment remains limited. Characterizing gene expression profiles of the human microbiome is essential for linking the role of different members of the bacterial communities in health and disease. The oral microbiome is one of the most complex microbial communities in the human body and under certain circumstances, not completely understood, the healthy microbial community undergoes a transformation toward a pathogenic state that gives rise to periodontitis, a polymicrobial inflammatory disease. We report here the in situ genome-wide transcriptome of the subgingival microbiome in six periodontally healthy individuals and seven individuals with periodontitis. The overall picture of metabolic activities showed that iron acquisition, lipopolysaccharide synthesis and flagellar synthesis were major activities defining disease. Unexpectedly, the vast majority of virulence factors upregulated in subjects with periodontitis came from organisms that are not considered major periodontal pathogens. One of the organisms whose gene expression profile was characterized was the uncultured candidate division TM7, showing an upregulation of putative virulence factors in the diseased community. These data enhance understanding of the core activities that are characteristic of periodontal disease as well as the role that individual organisms in the subgingival community play in periodontitis.

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Introduction

It is now recognized that the study of human health must include the role of the microbiome in maintaining homeostasis of the whole human system. The oral microbiome is one of the best characterized human body sites (Socransky et al., 1998; Paster et al., 2001; Marsh, 2006; Haffajee et al., 2008; Belda-Ferre et al., 2012; Peterson et al., 2013), comprising an extremely complex and highly organized biofilm community (Kolenbrander, 2000; Kolenbrander et al., 2002). More than 600 bacterial species have been identified in the oral cavity (Paster et al., 2001; Dewhirst et al., 2010). Many oral bacterial species have not yet been cultivated, and the only information we possess about them derives from their 16S rRNA phylogenetic affiliation.

Periodontitis is an oral polymicrobial disease caused by the coordinated action of a complex microbial community, which results in inflammation and destruction of the periodontium in susceptible hosts. This widespread disease is responsible for half of all tooth loss in adults: it occurs in moderate form in 39% of American adults and in severe form in 9% of adults, with a prevalence of 70% in adults older than 65 years (Eke et al., 2012).

Despite the limited number of studies, deep 16S rRNA sequencing using next-generation sequencing methods has yielded a wealth of knowledge regarding the genetic composition of the oral microbiome in periodontal health and disease (Griffen et al., 2012; Liu et al., 2012; Segata et al., 2012). Using checkerboard DNA-DNA hybridization technique, periodontitis-associated taxa have been cataloged into groups or 'complexes,' representing bacterial consortia that appear to occur together and that are associated with various stages of disease (Socransky et al., 1998). The 'red complex,' which appears later in biofilm development, comprises three species that are considered the major periodontal pathogens:

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Porphyromonas gingivalis, Treponema denticola and Tannerella forsythia (Socransky et al., 1998; Holt and Ebersole, 2005).

Despite a good understanding of the association between red complex species and periodontitis, we have only limited information on the *in situ* activity of these organisms. Cataloguing the activities of each bacterial species in a community may provide more insight into pathogenesis than simple enumeration of that community's gene content. This is because the community functions as a system, and it is the activities and interactions of the system that control the fate of the microbiome. One approach to describing the functional roles of specific bacteria is to study the community's metatranscriptome in situ. This technique, although indirect, is more direct than phylogenetic analysis. It has been successfully applied to study various environmental microbial communities (Frias-Lopez et al. 2008; McCarren et al. 2010) and, less frequently, to study human-associated microbiomes (Booijink et al., 2010; Porovko et al., 2010; Giannoukos et al., 2012; Xiong et al., 2012). To our knowledge, these methods have never been applied to study the microbiome in periodontitis.

Therefore, the goal of the present study was to characterize *in situ* gene expression patterns of the oral microbiome in periodontal health and disease. We addressed the following questions: What is the function of the expressed genes and particularly those most highly expressed? Do known periodontal pathogens appear to drive a large portion of activity as measured by their gene expression? Among genes that are more highly expressed in periodontitis samples, do virulence factors predominate? Do genes encoding virulence factors tend to be expressed by known periodontal pathogens?

Materials and methods

Study design and subject population

We conducted a cross-sectional comparison of microbial gene expression in subjects with (n=7) and without (n=6) periodontitis who were part of an ongoing study. All aspects of the study protocol were approved by the Institutional Review Board at the Forsyth Institute. The study was described thoroughly to all subjects before obtaining informed consent

Inclusion criteria were as follows: all study subjects were >18 years of age, had \geqslant 15 natural teeth and were in good general health. Periodontally healthy subjects had no pockets >3 mm. Periodontitis subjects had \geqslant 8 teeth with pockets >5 mm and \geqslant 8 teeth with clinical attachment loss >3 mm.

Exclusion criteria were as follows: subjects were excluded if they were cigarette smokers; were pregnant or nursing; received antibiotic or periodontal therapy in the previous three months; had any systemic condition potentially affecting the course

of periodontal disease (for example, diabetes or AIDS); or had any condition requiring antibiotic coverage during periodontal therapy.

Sample collection

After removal of supragingival plaque, subgingival plaque samples were taken separately from each site using individual sterile Gracev curettes and each sample placed in individual tubes containing 1 ml of RNAlater (Life Technologies, Grand Island, NY, USA). For three periodontally healthy subjects, six separate samples from sites with pocket depths ≤2 mm, no bleeding on probing and no redness were pooled separately (one pool for each subject). For the other three periodontally healthy subjects, only one site was sampled. Samples were initially pooled for fear of insufficient RNA extraction yields. Because we obtained sufficient RNA from single samples, we modified the protocol. For periodontitis subjects, sites with deep pockets (≥6 mm) that bled on probing were sampled. Four subjects had six sites sampled and pooled separately (one pool per subject), whereas the remaining three subjects contributed with one sample each.

Community DNA and RNA extraction

Cells were collected by centrifugation for 10 min at maximum speed in a microcentrifuge. A quantity of 600 µl of mirVana kit (Life Technologies) lysis/ binding buffer and 300 µl of 0.1-mm zirconia-silica beads (BioSpec Products, Bartlesville, OK, USA) were added to the samples. The beads were cleaned and sterilized beforehand with a series of HCl acid and bleach washes. Finally, the beads were treated with diethylpyrocarbonate overnight and autoclaved. Samples were bead beated for 1 min at maximum speed. DNA and RNA were extracted simultaneously following the protocol of mirVana Isolation kit for RNA and ToTALLY RNA kit (Life Technologies) for DNA. Eukaryotic DNA was removed using the MolYsis kit (Molzym GmbH & Co. KG, Bremen, Germany). MICROBioEnrich (Life Technologies) was used to remove eukaryotic RNA and MICROB*Express* to remove prokaryotic rRNA. All kits were used following manufacturer's instructions.

DNA, RNA amplification and Illumina sequencing DNA amplification was performed using the Illustra GenomiPhi V2 amplification kit (GE Healthcare Life Sciences, Piscataway, NJ, USA) according to manufacturer's instructions. RNA amplification was performed on total bacterial RNA using MessageAmp II-Bacteria RNA amplification kit (Life Technologies) following the manufacturer's instructions. Sequencing was performed at the Forsyth Institute and at the Biopolymers Facility Harvard Medical School. At the Forsyth Institute, for RNA sequencing,



Illumina adapter-specific primers (Illumina, San Diego, CA, USA) were used to amplify and selectively enrich for the cDNA generated from enriched mRNA. Quantified libraries were pooled and sequenced using the Miseq v2, 2×150 cycle cartridge (Illumina). The Nextera XT kit was used to generate libraries from amplified DNA. Normalized libraries were pooled and sequenced using the 2×250 Miseq v2 cartridge. At the Biopolymers Facility Harvard Medical School, samples were prepped using NuGen's mRNA-seq kit (Illumina) into next-gen libraries with Illumina adapters. The libraries were then clustered on a single-read flowcell using Illumina's cBot and the sequencing run for 100 cycles single-end for the DNA samples and 50 cycles paired-end for RNA samples on the Illumina HiSeq 2500 (Illumina).

Selection of genomes in databases

Genomes of archaea and bacteria and their associated information were downloaded from the Human Oral Microbiome Database database server (http://www.homd.org/), the Pathosystems Resource Integration Center (PATRIC) ftp server (http://www.patricbrc.org/portal/portal/patric/Home) (Wattam et al., 2013) and the J Craig Venter Institute (http://www.jcvi.org). Viral genomes were downloaded from NCBI (ftp://ftp.ncbi.nlm.nih.gov/genomes/ Viruses/all.fna.tar.gz). A total of 498 genomes from 283 species of bacteria and 2 genomes from 1 archaea species were used in the analysis (Supplementary Table 1). A total of 4000 viral genomes was used to create the viral database.

Comparison of different alignment algorithms on in silico communities

We used the simulator 'wgsim' (https://github.com/ lh3/wgsim), which creates sequences that mimic results obtained by Illumina sequencing, to generate sequences from the oral genomic database consisting of 500 oral microbial genomes (Supplementary Table 1). We generated 200 000 reads with sizes equal to those from the actual results (see Supplementary Table 2 for running parameters). We mapped those reads against the database used in our experiments, comparing the following seven algorithms to identify the one that most accurately recruited the most reads: bowtie2, stampy, bowtie, gassst, bwa, smalt and ssaha2, on the 'in silico' community. After mapping with the different algorithms, we evaluated the results with the script 'wgsim_eval.pl' from the same program.

Short reads sequence alignment analysis

Low-quality sequences were removed from the query files. 'fast_clipper' and 'fastq_quality_filter' from the Fastx-toolkit (http://hannonlab.cshl.edu/ fastx_toolkit/) were used to remove short sequences with quality score >20 in >80% of the sequence. For paired-end sequences, we then selected pairs present in both files. We also removed prokaryotic ribosomal RNA sequences and human sequences (http://hgdownload.cse.ucsc.edu/goldenPath/hg18/ chromosomes) bioinformatically. Cleaned files were then aligned against the bacterial/archaeal database using bowtie2. We generated a gff file to map hits to different regions in the genomes of our database. Read counts from the SAM files were obtained using bedtools multicov from bedtools (Quinlan and Hall, 2010).

Phylogenetic analysis of the metagenome and metatranscriptome

Counts from the DNA and RNA libraries were used to determine the phylogenetic composition of the respective libraries. We created a gff file containing information on whole genomes that was used to assign hits to genomes. Counts were normalized by dividing by the total number of reads for each sample and log-transforming the resulting fractions (Ramette, 2007; Brown et al., 2011). Normalized phylogenetic profiles were used for principal coordinates analysis with Manhattan distance performed using the R BiodiversityR package (Kindt and Coe, 2005). In addition, we performed the non-parametric test included in the NOISeq package (Tarazona et al., 2011) on the raw counts assigned to the different species to determine significant changes in counts at the species level.

Differentially expressed genes in subjects with vs without periodontitis

For assessing differential expression in genes within a specific species, we normalized the transcript counts by the relative frequency of the species in the metagenome database. In the case of Gene ontology (GO) and Kyoto Encyclopedia of Genes and Genomes (KEGG) pathways, we did not normalized enrichment analysis by relative abundance, as we were treating the whole community as a single organism.

To identify differentially expressed genes from the RNA libraries, we applied non-parametric tests to the normalized counts using NOISeg's default conditions (k=0.5, pnr=0.2, nss=5, v=0.02, lc=1,replicates = 'biological') and Trimmed Mean of *M*-values ('TMM' option) using the threshold value for significance suggested by the authors (Tarazona et al., 2011; Soneson and Delorenzi, 2013).

We obtained a heat-map representation of the expression profiles for the 506 most highly expressed genes across samples as well as the profile corresponding to 1000 significantly differentially expressed genes according to the NOISeq results that showed the highest differences in expression. Counts for those genes were first normalized according to size and the BiodiversityR package (Kindt and Coe, 2005) was used to χ^2



transform the frequencies (Legendre and Gallagher, 2001). Using the 'heatmap.2' function in R, we clustered samples and represented their heat-maps based on the expression profiles. The clustering function for 'heatmap.2' was 'hclust' selecting 'complete' (complete-linkage) as clustering method.

GO and KEGG pathway enrichment analysis

To evaluate functional activities differentially represented in health or disease, we mapped the differentially expressed genes to known biological ontologies based on the GO project (http://www. geneontology.org/) and the KEGG metabolic pathways database (http://www.genome.jp/kegg/). GO terms and KEGG pathways to which the different open reading frames (ORFs) belong were obtained from the PATRIC database (http://patricbrc.org/portal/portal/patric/Home). GO terms and KEGG pathways from genomes not present in the PATRIC database and whose annotation was obtained from the Human Oral Microbiome Database database or I Craig Venter Institute were acquired using the program blast2GO under the default settings (Götz et al., 2008).

Enrichment analysis on these sets was performed using the R package 'GOseq,' which accounts for biases because of over-detection of long and highly expressed transcripts (Young et~al., 2010). Gene sets with ≤ 10 genes were excluded from the analysis. We used the REVIGO web page (Supek et~al., 2011) to summarize and remove redundant GO terms from the results. Only GO terms with false discovery rate < 0.05 were used. REVIGO plots were obtained for two GO categories: biological processes and molecular functions. The GO project describes a biological process as a recognized series of events or molecular functions with a defined beginning and end, and molecular functions as activities that occur at the molecular level, such as catalytic or binding activities.

Quantification of putative virulence factors

To identify putative virulence factors, we used the Virulence Factors of Pathogenic Bacteria Database (VFDB; http://www.mgc.ac.cn/VFs/). Others have used a similar approach to identify putative virulence factors in genomic islands (Ho Sui *et al.*, 2009). The VFDB contains 1143 virulence factors and 5927 virulence factor-related genes from 75 pathogenic bacterial genera (Chen *et al.*, 2012). We performed a BLAST similarity search of encoded proteins from the genomes in our database against the VFDB, with an *e*-value cutoff of 10^{-25} and identity >99% to exclude distant homologs.

Results

In the comparison of algorithms for recruiting metatranscriptomic reads, 'bowtie2' performed the best using either single-end reads (98.2 reads mapped correctly) or paired-end reads (96.9 reads mapped correctly; Supplementary Table 3).

Phylogenetic comparison of the metagenome and metratranscriptome in health and disease

Sequencing yields ranged from 0.4 to 20 million and 0.9 to 39.0 million for the metagenomes and metatranscriptomes, respectively (Supplementary Table 4). No viral sequences were identified in either set of libraries. Of the 1278494 genes contained in the genomic database, 702106 had $\geqslant 1$ hit in $\geqslant 1$ sample.

Principal coordinates analysis of the phylogenetic profiles based on either the metagenome or metatranscriptome did not clearly distinguish the subjects with and without periodontitis. Nonetheless, six of the seven metatranscriptome of samples with periodontitis clustered along principal coordinates analysis-1, which captured most of the variance (Figure 1).

After ranking the species based on fold changes in hits with vs without periodontitis, P. gingivalis, T. forsythia and T. denticola exhibited higher mean abundance and higher levels of gene expression in periodontitis than in healthy subjects (Figure 2 and Supplementary Table 5). However, the increase in numbers of other species was not always accompanied by an increase in inferred activity and vice versa. For instance Acinetobacter baumannii was significantly more abundant in periodontitis than in healthy subjects but did not have high levels of gene expression in disease, whereas others, such as Bacteroidetes oral taxon 274, Corynebacterium matruchotii or Leptotrichia hofstadii, showed a decrease in relative numbers but represented a high proportion of the metatranscriptome library (Figure 2).

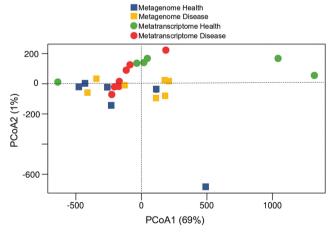


Figure 1 Principal coordinate analysis (PCoA) of phylogenetic profiles from metagenomes and metatranscriptomes. Ordination graph for the first two axes of a PCoA of the phylogenetic profiles obtained from the metagenome and metatranscriptome of the oral microbial communities in healthy sites and sites with severe periodontitis. The % on the axes represents the variance explained by each of the coordinates.



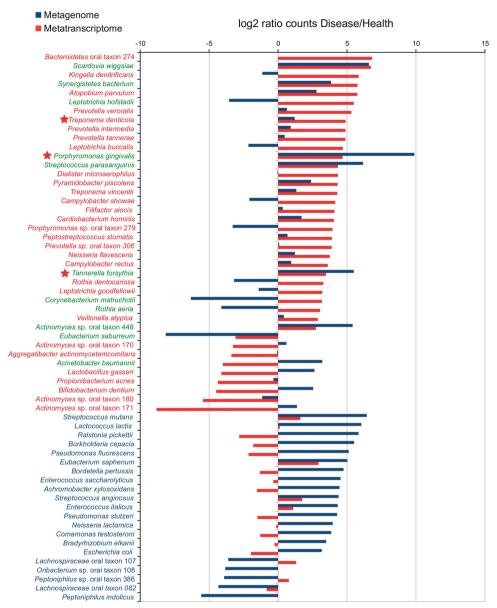


Figure 2 Rank distribution of statistically significant relative increase in number of hits for the metagenome and metatranscriptome results. The ratio of counts in disease vs health was log2 transformed and plotted according to ranks. The statistical significance was calculated using the non-parametric test implemented in the program NOISeq as described in the Materials and methods section. Only species with significant differences, according to NOISeq, either in metagenomic or metatranscriptomic counts, are presented. In green, species with statistical differences in both metagenome and metatranscriptome. In blue, species with statistical differences in metagenomic counts. In red, species with statistical differences in metatranscriptome counts. Red star indicates major periodontal pathogens previously assigned to the 'red-complex.'

We compared phylogenetic assignment of expression levels of differentially expressed genes in the non-pooled vs pooled samples after trimmed mean of M-values normalization and we observed similar numbers for the species with higher differences in expression (Supplementary Figure 1). All members of the 'red- complex' were represented in both set of samples showing similar numbers of differentially expressed genes. Furthermore, other organisms that we found important both in numbers as well as in expression of virulence factors (for example, Corynebacterium matruchotii, see Supplementary Figure 1) showed also similar patterns in both set of samples.

Global patterns of gene expression in health and disease: enrichment of GO terms and KEGG metabolic pathways

In total, NOISeq analysis identified 91 113 differentially represented genes without normalizing for species abundance and 123317 differentially expressed genes after normalization by relative abundance of the different species in the community. We analyzed the expression profiles of the most highly represented genes across samples (Figure 3a) as well as the profiles for the most highly differentially represented genes in our libraries (Figure 3b). Although both profiles separated clusters from

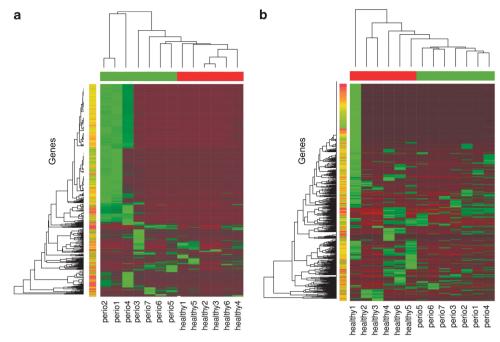


Figure 3 Heat maps of gene expression profiles in the metatranscriptomes. (a) Profiles of the 1212 most highly expressed genes across samples. Genes were first normalized by gene length and χ^2 transformed before analysis. Most highly expressed genes were selected according to the sum of normalized counts across samples. (b) Profiles of the 1000 most highly differentially expressed genes across samples. Genes were selected from the NOISeq results and counts were χ^2 transformed before analysis. Color bars in red correspond to healthy samples and in green to periodontitis samples.

healthy and periodontitis samples, the profiles for the most highly differentially expressed genes better distinguished the two subject groups. Expression profiles of the most highly differentially represented genes seem to be highly similar across samples in the periodontitis group. If confirmed, this observation would suggest that periodontitis might be characterized by a core of specific microbial activities (Figure 3b).

A total of 146 GO terms were overrepresented and 42 underrepresented in healthy vs periodontitis subjects (Supplementary Table 6). Using the same package, 'GOseq,' but mapping to KEGG pathways instead of GO categories, we identified 14 pathways overrepresented and 18 underrepresented in disease (Supplementary Table 6). There were no meaningful differences in results for pooled and un-pooled samples. In particular, the themes that emerged from GO analysis were the same regardless of pooling of samples (Supplementary Figure 2). The differences we observed when splitting the samples into the two groups may be due in part to the reduction in the number of samples per group. In the whole analysis, we compared six healthy vs seven periodontitis samples, whereas in Supplementary Figure 2 we compared three pooled and four non-pooled samples separately, reducing the statistical power of the analysis.

GO biological processes related to flagellar motility, peptides transport, iron acquisition and betalactam degradation were overrepresented in disease, as was biosynthesis of the lipid A component of endotoxins from Gram-negative bacteria (Figures 4a and b). GO biological processes underrepresented in disease included potassium transport and polysaccharide biosynthesis. From gene set enrichment analysis based on KEGG pathways, which represents a completely different ontology, the overrepresented pathways were related to broad activities, such as energy metabolism (for example, Aminoacyl-tRNA biosynthesis, nitrogen metabolism), carbohydrate metabolism (for example, glycolysis/gluconeogenesis, citrate cycle, metabolism), bacterial motility and chemotaxis, and genes that function in electron transport (for example, photosynthesis and porphyrin and chlorophyll metabolism; Supplementary Table 6). Biosynthesis of secondary metabolites and antibiotics, metabolism of terpenoids and alkaloids and lipid metabolism accounted for most of the KEGG pathways underrepresented in the periodontitis samples (Supplementary Table 6).

Expression profiles of putative virulence factors in major periodontal pathogens

The three members of the red complex showed high expression of metalloproteases and peptidases (Supplementary Table 7, in red). In addition, proteins involved in iron metabolism represented an important fraction of upregulated putative virulence factors in these periodontal pathogens. These included ferrous iron transport protein B, iron compound ABC transporter ATP-binding



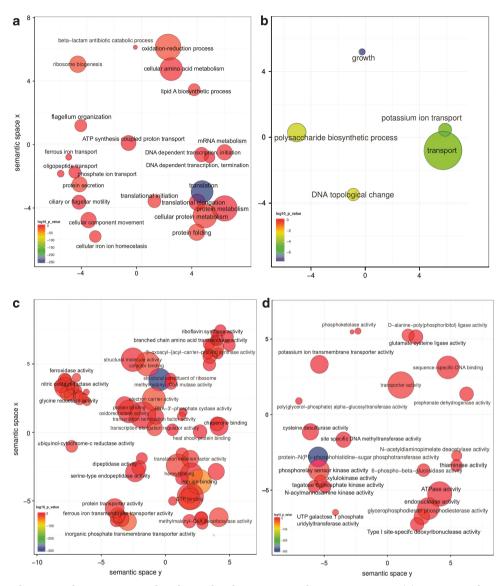


Figure 4 GO enrichment analysis summarized and visualized as a scatter plot using REVIGO. (a) Summarized GO terms related to biological processes in periodontitis. (b) Summarized GO terms related to biological processes in health. (c) Summarized GO terms related to molecular function in periodontitis. (d) Summarized GO terms related to molecular function in health. GO terms are represented by circles and are plotted according to semantic similarities to other GO terms (adjoining circles are most closely related). Circle size is proportional to the frequency of the GO term, whereas color indicates the log10 P value (red higher, blue lower).

proteins, ferric enterobactin transport ATP-binding protein FepC and iron chelate uptake ABC transporter FeCT permease proteins (Supplementary Table 7, in green). In periodontitis subjects, there was higher expression of *P. gingivalis* hemolysin genes and P. gingivalis and T. denticola genes involved in vitamin B12 import (for example, vitamin B12 ABC transporter permease component BtuC).

Also more highly expressed in periodontitis samples were genes encoding other proteins not involved directly in iron metabolism, such as ATPdependent Clp protein, ATP-binding subunit ClpA, ClpB protein; the chaperone ClpB protein, described as involved in cellular invasion in P. gingivalis; GroEL; protein export cytoplasm secA ATPase RNA helicase and RNA polymerase sigma factor RpoD.

Most of the genes needed to synthesize flagella in T. denticola were overrepresented in the periodontitis metratranscriptome. These included flagellar basal-body FlgC, FlgF and FlgG, flagellar hook proteins FlgE and FlgK, flagellar motor rotation proteins MotA and MotB, flagellar motor switch proteins FliG and FliM, flagellin protein FlaA, flagellar M-ring protein FliF, flagellum-specific ATP synthase FliI and proteins involved in flagella biosynthesis (FleN, FlhA-B, FlhF Supplementary Table 7). Of the three major periodontal pathogens, T. denticola was the only one exhibiting over-representation of proteins involved in oligopeptide transport (OppA, OppD and OppF).

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Expression of virulence factors and virulence factorrelated genes in the oral community

A total of 7846 hits of putative virulence factors from 255 species were identified in the genes periodontitis overrepresented samples, in although for only 64 species did this include ≥40 virulence factors. Members of the 'red-complex' as well as Aggregatibacter actinomycetemcomitans, an organism associated with localized juvenile periodontitis, over-expressed a large number of the putative virulence factors in the periodontal samples (Figure 5). Nonetheless, surprisingly enough, a set of organisms that have been previously associated only with diseases other than periodontitis or with periodontally healthy sites, such as Neisseria spp., Corynebacterium matruchotii, Rothia dentocariosa, Veillonella parvula, Actinomyces spp. and others, over-expressed a large number of putative virulence factors that could have importance in the evolution of the disease (Figure 5). Moreover, a large number of different homologs from different ORFs within each genome were all over-expressed. For instance, C. matruchotii showed upregulation of seven different ORFs with homology to the Fe⁽³⁺⁾ ions import ATP-binding protein FbpC and 10 ORFs homologs of iron (III) ABC transporter ATP-binding proteins in the VFDB. Moreover, C. matruchotii upregulated 20 different ORFs that are annotated as putative hemagglutinin hemolysin-related proteins (Supplementary Table 8, in green).

There were several other pathways for which genes were overexpressed in periodontitis samples in a large number of species. Second in number only to those related to iron metabolism were genes in lipopolysaccharide biosynthesis pathways (Supplementary Table 8, in red). Other such pathways or groups included: oxidative stress tolerance (trigger factor, superoxide dismutase and alkyl hydroperoxide reductase C protein AhpC); specific regulatory proteins (for example, phosphate regulon transcriptional regulatory protein PhoB SphR, RNA polymerase sigma factors RpoD/RpoD and translation elongation factor Tu); specialized protein transport systems (for example, zinc ABC transporters ZnuA and ZnuC and ABC-type multi-drug transport systems ATPase and permease components); oligopeptide ABC transporters; nitrate reductase proteins (alpha, beta, delta and gamma subunits); serine-proteases (for example, MucD/ AlgY, DegP/HtrA); hemolysins; large numbers of proteins involved in flagellar biosynthesis (for example, FlaA-B, FlaG, FleN, FlgB-I, FlgK-L, FlhA-B, FlhF, FliD-I, FliK-N, FliP-S, MotA and MotB) and chaperones such as GroEL. Organisms expressing high numbers of putative virulence factors expressed high levels of a putative ABC transporter that acts as an efflux pump of the anticancer drug daunorubicin (DrrA), as well as the corresponding daunorubicin-resistant ABC transporter ATPase subunit.

The role of non-cultivable organisms in health and disease: the case of candidate division TM7 Despite the sequence of candidate division TM7 genome being incomplete, we identified differentially expressed genes, including cell division proteins (for example, chromosomal replication initiator proteins DnaA, FtsA and FtsH) and chaperones (GroEL, GroES, DnaJ and Hsp20). In addition, several putative virulence factors were also more highly expressed in periodontitis samples. These were related to iron acquisition (for example, hemagglutinin-related proteins and an exoprotein involved in heme utilization of adhesion), pili biosynthesis (PilB and PilC) and oligopeptide transport (OppA; Supplementary Table 9). OppA was also highly overrepresented in the periodontopathogen

Discussion

T. denticola.

To gain insight into the functional phenotype of periodontal dysbiosis, we used a combined metagenomic/metatranscriptomic approach. This allowed us to infer functional differences between the subgingival microbiota of periodontal health and disease and generate hypotheses regarding the role of the different community members in controlling the fate of the biofilm. We took advantage of both next generation sequencing techniques and the fact that a large number of genomes from oral isolates have been sequenced, although not all (Chen et al., 2010). Moreover, the amplification methods used for metagenomic and metatranscriptomic analysis do not introduce bias in the final relative proportions of hits (Teles et al., 2007; Hoeijmakers et al., 2011).

In a recent deep sequencing analysis using 16S rDNA, members of the genus Prevotella, Fusobacterium, Treponema, Selenomonas and Porphyromonas were more abundant, whereas Actinomyces and Streptococcus were less abundant in samples from periodontitis compared with healthy subjects (Griffen et al., 2012). We observed similar patterns using metagenomic analysis, although with a slight increase in numbers of Streptococcus in periodontitis samples. Moreover, we observed high numbers of members of the genus Neisseria in the diseased samples and of candidate division TM7 in samples from both healthy and periodontitis sites. Two recent studies have reported a higher number of these organisms in oral samples (Liu et al., 2012; Segata et al., 2012) contradicting previous studies that had shown high prevalence but with very low abundances (Brinig et al., 2003; Podar et al., 2007), probably due to an improvement in detection techniques. We also looked at the presence of viruses and archaea, which have been previously associated with disease (Lepp et al., 2004; Slots, 2010). Archaea accounted for a very small fraction of the total community, which agrees with other reports on the abundances of these



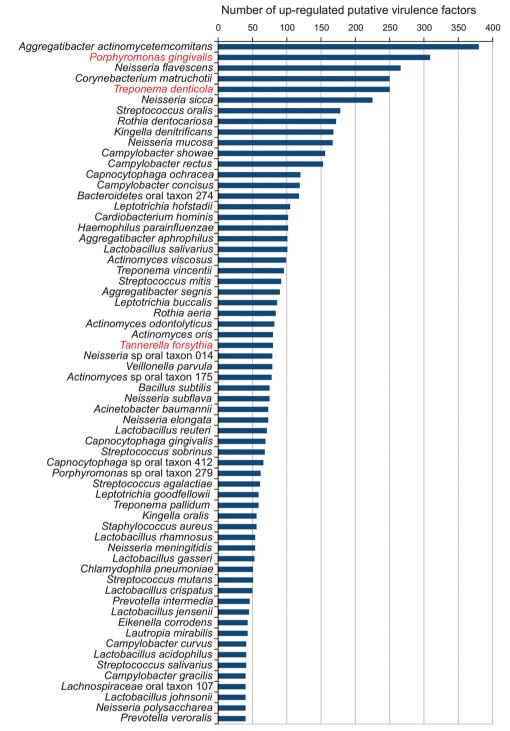


Figure 5 Ranked species by the number of upregulated putative virulence factors in the metatranscriptome. Putative virulence factors were identified by alignment of the protein sequences from the different genomes against the Virulence Factors of Pathogenic Bacteria Database as described in the Materials and methods section. Number in the graph are the absolute number of hits for the different species for the putative virulence factors identified. In red are the members of the 'red-complex.'

organisms in oral samples (Lepp et al., 2004) and we did not detect any viral sequence in either the metagenome or the metatranscriptome.

Expression profiles of the most highly differentially expressed genes seem to be highly similar across samples in the periodontitis group. If confirmed, this observation would suggest that periodontitis might be characterized by a core of specific microbial activities.

Testing of over-representation of GO terms and KEGG metabolic pathways in the microbial community as a meta-organism gives an overall view of the



microbial activities under different states of health. The results pointed to several functional signatures characteristic of periodontitis. Iron acquisition was abundantly overrepresented in the periodontitis samples. Iron is an essential enzymatic cofactor and microbial pathogens have developed a large number of strategies to acquire it (Brochu et al., 2001; Rhodes et al., 2007; Nairz et al., 2010). Also transport proteins oligopeptide were expressed. Some important oral bacteria, such as Parvimonas micra, Fusobacterium nucleatum and P. gingivalis, use peptides more efficiently than free amino acids (Tang-Larsen et al., 1995) and it is plausible that the degree of efficiency in using peptides may influence the ecology of the periodontal biofilm. Lipid-A biosynthesis was also overrepresented in disease. Lipopolysaccharide is a key factor in the development of periodontitis (Fives-Taylor et al., 1999; Jain and Darveau, 2010) and high levels of lipopolysaccharide and lipid A from P. gingivalis have been reported to delay neutrophil apoptosis and provide a mechanism to modulate the restoration and maintenance of inflammation in periodontal tissues (Hiroi et al., 1998; Preshaw et al., 1999; Murray and Wilton, 2003).

The above-described metatranscriptomic differbetween subjects with and without periodontitis are consistent with the current understanding of periodontal pathogenesis. A novel observation was the relatively abundant expression in periodontitis samples of genes related to the response to antibiotics in general, and beta-lactam antibiotic degradation processes in particular. Betalactamase-producing bacteria have been frequently isolated from the oral cavity (Handal et al., 2004; Rams et al., 2012) and beta-lactamase activity has been observed in adult periodontitis at low-level enzymatic activity but with high prevalence (Van Winkelhoff et al., 1997). Beta-lactamase activity seems to be a frequent phenomenon in samples from polymicrobial diseases (Brook, 2009) and has been detected in specimens of clinical abscesses and mixed infections, such as empyema (Bryant et al., 1980), cerebrospinal fluid (Boughton, 1982) and abscesses (Brook, 1986). What role this enzymatic activity has on the homeostasis of the community is an open question given that none of the samples came from patients being treated with antibiotics at the time of sampling. Nonetheless, the presence and prevalence of beta-lactamase genes in the subgingival microbial community, even in the absence of antibiotic, has public health implications given the rise of resistance among a large number of bacterial pathogens. Our findings suggest that the target of beta-lactamase activity may be a different molecule, which may hamper efforts to eliminate beta-lactamase genes from the genetic pool of the human microbiome.

We also found an under-representation of antibiotic biosynthesis genes in periodontitis samples, consistent with the expression of antibiotics as a potential mechanism of homeostasis in healthassociated microbiota. Further, the production of beta-lactamase might represent a mechanism that confers protection to some species in the subgingival microbiota against antibiotics released by others.

P. gingivalis, T. denticola and T. forsythia are considered major periodontal pathogens and appear in great numbers later in biofilm development when the polymicrobial infection is well established (Socransky et al., 1998; Ximenez-Fyvie et al., 2000; Holt and Ebersole, 2005). Consistent with their postulated role, in subjects with severe periodontitis, we observed relatively high levels of expression of proteolysis-related genes. Gingipains or fimbrial genes of P. gingivalis have been described as essential in the early stages of colonization (Imamura, 2003; Enersen et al., 2013); however, we observed no over-representation of these genes in our results, may be due to the fact that we are analyzing severe periodontitis samples and no early stages of the disease.

All members of the 'red-complex' exhibited an over-representation of RNA-encoding proteins involved in invasion of cells, including ClpB, a general protection factor against stress, GroEL and DnaK, which have also been associated with cellular invasion (Rodrigues and Progulske-Fox, 2005; Xia et al., 2007; Yuan et al., 2007), and ferritin, linked to invasion of cells by *P. gingivalis* (Rodrigues and Progulske-Fox, 2005; Xia et al., 2007). GroEL has been identified as a virulence factor in the periodontal pathogen *Actinobacillus actinomycetemcomitans* (Goulhen et al. 1998).

T. forsythia has been detected intra-cellularly in buccal and crevicular epithelium of patients with periodontitis (Rudney et al., 2005) and has been shown to invade epithelial cells in vitro (Inagaki et al., 2006; Mishima and Sharma, 2011). We observed over-representation in periodontitis of a homolog of the cell-surface-associated protein sialidase NanH (locus VBITanFor42681_2003) required for attachment to epithelial cells (Honma et al., 2011).

T. forsythia and T. denticola over-expressed homologs to internalins, surface proteins that are key in invasion of mammalian cells by Listeria monocytogenes (Lecuit et al., 1997; Cossart et al., 2003), and could have a similar role in those periodontal pathogens.

Finally, we identified as upregulated most of the genes necessary to synthesize flagella in *T. denticola* as well as genes necessary for chemotaxis that could direct flagellar movement. Flagella are essential virulence factors in *T. denticola*, the lack of which prevents it from penetrating periodontal tissue (Lux et al., 2001). Taken together, these observations strongly support the idea that in deep bleeding pockets the members of the 'red-complex' are actively invading host cells.

Recently, gaining traction is the idea of the 'pathogenic microbial community' (Berezow and

Darveau, 2011) or 'the community as pathogen' (Relman, 2012), which postulates that the integrated actions of the components of the microbial community would result in disease. In part to address this hypothesis, we examined the possible role of organisms not previously recognized as periodontal pathogens. A large number of organisms in the oral community were expressing putative virulence factors. Members of the genus Streptococcus such as S. mitis, generally associated with health, were expressing homologs of virulence factors from known pathogenic streptococci such as; neuraminidase A (Jedrzejas, 2001), zinc metalloprotease ZmpB (Bender and Weiser, 2006) and the saliva-binding protein adhesin B (Herbert et al., 2004).

Surprisingly, C. matruchotii, R. dentocariosa, Veillonella parvula and Neisseria sicca were expressing a large variety of putative virulence factors. Although C. matruchotii and R. dentocariosa have occasionally been associated with disease (Sigueira et al., 2000; Thomas et al., 2012), in general, they (and other Gram-positive organisms) have been considered associated with health of the periodontium (Kolenbrander, 2000; Gross et al., 2010; Liu et al., 2012).

Some of the observed associations point to possible synergistic interactions among community members. Specifically, siderophores such as entorobactins and ABC-type cobalamin Fe3+-siderophores were highly over-represented in the periodontitis samples, yet members of the 'redcomplex' do not appear to produce them (Brochu et al., 2001; Rhodes et al., 2007). The production of these molecules by the accompanying community may facilitate iron acquisition by the 'red-complex.'

Finally, genes involved in stress tolerance were highly over-represented in a large fraction of the community and their upregulation may be related to the production of reactive oxygen species during disease (Wright et al., 2011; Bullon et al., 2012).

One important unanswered question in microbiology is the role of uncultured organisms in the activities of the oral community. Candidate bacterial phylum TM7 has no cultivated members and has been detected in high number in oral samples (Brinig et al., 2003; Kumar et al., 2003; Ouverney et al., 2003; Colombo et al., 2009; Liu et al., 2012; Casarin et al., 2013; Kim and Lee, 2013). We still do not know if TM7 has a role in the progression of periodontitis but we identified several homologs to virulence factors expressed in TM7 that were associated with periodontitis. These included the oligopeptide ABC transporter periplasmic oligopeptide-binding protein (OppA), which has been identified as a virulence factor in group A Streptococci (Wang et al., 2005); a hemagglutinin-related protein (locus Div80122 0315); exoproteins involved in heme utilization or adhesion (locus VBICanDiv80122 0453); an homolog of LemA (locus VBICan-Div80122_0048), a widely conserved two-component regulatory system that has been identified as regulating virulence factors and toxin production in Pseudomonas syringae (Kitten et al., 1998).

Also differentially expressed more highly in periodontitis samples were TM7 genes encoding a chitinase (locus VBICanDiv80122 0016) and a putative ROK-family transcriptional regulator (locus VBICanDiv80122 0244), which regulates chitinase expression in the Gram-positive pathogen *Listeria* monocytogenes at the post-transcriptional level (Larsen et al., 2010). Representatives of both Grampositive and Gram-negative bacterial pathogens encode chitinases that support infection of nonchitinous mammalian hosts (Frederiksen et al., 2013), possibly by suppressing host innate immunity (Chaudhuri et al., 2013). Finally, we also observed over-representation of genes involved in pili biosynthesis. Although type IV pili may facilitate the adherence of bacteria to epithelial cells, it has been suggested that in the case of TM7 it may be involved in gliding motility (Marcy et al., 2007). These comparative expression studies provide only circumstantial evidence of a pathogenic role of TM7 members; however, they also demonstrate the usefulness of in situ metatranscriptomic analysis for investigating the possible roles of uncultured organisms and for developing new hypothesis to be tested in the laboratory.

Conclusions

Our findings highlight the usefulness of metatranscriptomic analysis for elucidating the potential importance of virulence factor-expressing organisms not previously linked to pathogenesis, also including previously uncultured organisms. Moreover, these results support the idea that the whole community, and not just a selected group of pathogens, promote virulence factors during disease development. The data presented here illustrate the value of using metatranscriptomic analysis to examine microbial activities in whole human microbial communities and as a hypothesis-generating strategy based on results obtained under in situ conditions.

Data accessibility

The data sets used in these analyses were deposited at the Human Oral Microbiome Database under the submission number 20130522 (ftp://ftp.homd.org/ publication_data/20130522/).

Conflict of Interest

The authors declare no conflict of interest.

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