<table>
<thead>
<tr>
<th>Title</th>
<th>Metagenomic and metatranscriptomic analysis of saliva reveals disease-associated microbiota in patients with periodontitis and dental caries</th>
</tr>
</thead>
<tbody>
<tr>
<td>Author(s)</td>
<td>Belstrøm, Daniel; Constancias, Florentin; Liu, Yang; Yang, Liang; Drautz-Moses, Daniela Isabel; Schuster, Stephan Christoph; Kohli, Gurjeet Singh; Jakobsen, Tim Holm; Holmstrup, Palle; Givskov, Michael</td>
</tr>
<tr>
<td>Date</td>
<td>2017</td>
</tr>
<tr>
<td>URL</td>
<td><a href="http://hdl.handle.net/10220/45339">http://hdl.handle.net/10220/45339</a></td>
</tr>
<tr>
<td>Rights</td>
<td>© 2017 The Author(s). This article is licensed under a Creative Commons Attribution 4.0 International License, which permits use, sharing, adaptation, distribution and reproduction in any medium or format, as long as you give appropriate credit to the original author(s) and the source, provide a link to the Creative Commons license, and indicate if changes were made. The images or other third party material in this article are included in the article’s Creative Commons license, unless indicated otherwise in a credit line to the material. If material is not included in the article’s Creative Commons license and your intended use is not permitted by statutory regulation or exceeds the permitted use, you will need to obtain permission directly from the copyright holder. To view a copy of this license, visit <a href="http://creativecommons.org/licenses/by/4.0/">http://creativecommons.org/licenses/by/4.0/</a>.</td>
</tr>
</tbody>
</table>
Metagenomic and metatranscriptomic analysis of saliva reveals disease-associated microbiota in patients with periodontitis and dental caries

Daniel Belstrøm1, Florentin Constancias2, Yang Liu3, Liang Yang2, Daniela I. Drautz-Moses2, Stephan C. Schuster2, Gurjeet Singh Kohli2, Tim Holm Jakobsen4, Palle Holmstrup1 and Michael Givskov2,3

The taxonomic composition of the salivary microbiota has been reported to differentiate between oral health and disease. However, information on bacterial activity and gene expression of the salivary microbiota is limited. The purpose of this study was to perform metagenomic and metatranscriptomic characterization of the salivary microbiota and test the hypothesis that salivary microbial presence and activity could be an indicator of the oral health status. Stimulated saliva samples were collected from 30 individuals (periodontitis: n = 10, dental caries: n = 10, oral health: n = 10). Salivary microbiota was characterized using metagenomics and metatranscriptomics in order to compare community composition and the gene expression between the three groups. Streptococcus was the predominant bacterial genus constituting approx. 25 and 50% of all DNA and RNA reads, respectively. A significant disease-associated higher relative abundance of traditional periodontal pathogens such as Porphyromonas gingivalis and Filifactor alocis and salivary microbial activity of F. alocis was associated with periodontitis. Significantly higher relative abundance of caries-associated bacteria such as Streptococcus mutans and Lactobacillus fermentum was identified in saliva from patients with dental caries. Multiple genes involved in carbohydrate metabolism were significantly more expressed in healthy controls compared to periodontitis patients. Using metagenomics and metatranscriptomics we show that relative abundance of specific oral bacterial species and gene expression in saliva associates with periodontitis and dental caries. Further longitudinal studies are warranted to evaluate if screening of salivary microbial activity of specific oral bacterial species and metabolic gene expression can identify periodontitis and dental caries at preclinical stages.

npj Biofilms and Microbiomes (2017) 3:23; doi:10.1038/s41522-017-0031-4

INTRODUCTION

Periodontitis and dental caries are the two most prevalent oral diseases and the primary causes of tooth loss in the western world.1,2 At present, periodontitis and dental caries are mostly diagnosed at late stages of disease, often leading to costly and invasive dental treatment. Therefore, new diagnostic approaches capable of identifying periodontitis and dental caries at preclinical stages, favoring preventive treatment strategies, are urgently needed.

The oral cavity harbors a diverse microbiota comprising more than 700 unique bacterial species.3 The microbiota plays a pivotal role in maintenance of oral homeostasis, as various oral habitats are colonized by characteristic bacterial community profiles organized in local biofilms.4 However, ecological changes, for example in relation to increased sugar intake, insufficiently performed oral hygiene or fluctuations in the immune response can induce structural5–7 and functional alterations8–10 of local oral biofilms. Such alterations may in turn change the relation between the host and the resident microbiota from symbiosis to dysbiosis, thereby fueling initiation and progression of periodontitis and dental caries.

Saliva is the biological fluid of the oral cavity which is critical for maintenance of oral and general health.11 Therefore, saliva has been intensively investigated for candidate biomarkers associated with oral health and disease.12,13 Saliva is sterile when entering the oral cavity,14 but when sampled, saliva contains a diverse microbiota.15 In healthy oral conditions, the composition of the salivary microbiota is different from that of supragingival and subgingival biofilms.16 On the other hand, the presence of specific bacterial species in saliva such as Porphyromonas gingivalis and Streptococcus mutans has been reported in individuals with periodontitis and dental caries, respectively.17,18 Essentially, these findings suggest that bacteria from local periodontitis and caries lesions may be spilled-over and dispersed into saliva. However, it remains unclear if dispersed bacteria remain metabolically active as they are translocated from the local ecological niche of the biofilms to saliva, which possesses different ecological properties.

So far, only a few studies have reported higher expression of specific bacterial genes to associate with dental caries19,20. In addition, the potential of using comprehensive metatranscriptomic microbial analysis of saliva to identify possible biomarkers has not yet been evaluated.
The aims of the present study were to employ metagenomic and metatranscriptomic analyses to (i) characterize the diversity and community composition of the salivary microbiota, (ii) to utilize these data to determine the microbial activity (DNA/RNA ratio) of each taxon identified and, (iii) to identify differentially expressed microbial genes. The main hypothesis was that the taxonomic composition, as well as microbial activity could be an indicator of the oral health status, and possible differences might potentially be used as proxy biomarkers of oral health and disease.

RESULTS

General information

From a total of 30 saliva samples 768.62 M (mean 25.62 M; range 14.61 M–32.64 M) DNA sequences and 981.48 M (mean 32.72 M; range 26.40 M–38.44 M) RNA sequences were generated out of which 274.17 M (DNA; mean 9.14 M; range 2.58 M–18.52 M) and 148.49 (RNA; mean 4.50 M; range 2.93 M–9.11 M) passed quality control (Supporting material Table S1).

At species level, a significantly higher α-diversity (observed species) was recorded by metagenomics compared to metatranscriptomic analysis (Figs. 1a–b) (p < 0.05). However, no difference in α-diversity was observed between groups using either metagenomics or metatranscriptomic approaches. Metagenomic analysis yielded a significantly different β-diversity between groups (p = 0.027, R² = 0.12), whereas an insignificantly different β-diversity was recorded when employing metatranscriptomic analysis (p = 0.11, R² = 0.10) (Fig. 1c).

A total of 810 to 894 microbial KEGG Orthologs (KO) was identified in salivary microbial metagenomes and metatranscriptomes. These KO were mainly associated to Ribosome, Carbon metabolism, ABC transporter, and Purine metabolism pathways (Supporting material Fig S1A). Within the metabolism category salivary microbiomes were dominated by formate C-acetyltransferase, enolase, 5-methyltetrahydropteroyltriglutamate–homocysteine methyltransferase, and pyruvate oxidase (Supporting material Fig S2A). In all cases, a significantly lower number of KEGG KO and Pathways were identified in the salivary metagenomes as compared to the microbial metatranscriptome (p < 0.05, Supporting material Figs S1B, S2B). In order to characterize the overall community differences in terms of KEGG pathways and KO, Bray–Curtis overall dissimilarities were constructed and no significant discrimination of healthy and disease groups were observed (PERMANOVA p > 0.05 in all case, Supporting material Figs S1 C-D; Fig S2 C-D).

Fig. 1 Microbial diversity. a: Rarefaction curves expressed as number of reads assigned at species level (x-axis) and α-diversity (observed species and Shannon index, y-axis) of salivary microbiota characterized using metagenomics and metatranscriptomics. b: Distribution of α-diversity among groups presented as boxplot (median, lower/upper quartile and standard deviation). c: β-diversity determined by Bray-Curtis dissimilarity and plotted using PCoA. Sample denotation: green: healthy controls, blue: dental caries, red: periodontitis.
Identification of microbial taxa from salivary metagenomes and metatranscriptome

A total number of 12 bacterial phyla, 123 bacterial genera, and 351 bacterial species were identified from the salivary metagenome with a mean number per sample of 11 phyla (10–12), 115 genera (95–123), and 321 species (240–351). Metatranscriptomic bacterial identification yielded a total number of 12 bacterial phyla, 101 bacterial genera, and 234 bacterial species with a mean number per sample of 10 phyla (8–12), 101 genera (79–120), and 262 species (204–324).

Twenty most abundant bacterial genera based on metagenomic and metatranscriptomic bacterial identifications are displayed in Fig. 2a. As seen in Fig. 2a, the three most predominant genera based on both identification methods were *Streptococcus*, *Prevotella*, and *Veillonella* constituting approx. 70% of all assigned DNA and RNA reads. *Streptococcus* was significantly more abundant based on metatranscriptomic identification (50%) than metagenomic identification (25%, \( p < 0.05 \)).

The predominant 20 bacterial species based on metagenomic and metatranscriptomic bacterial identifications are displayed in Fig. 2b. While the salivary metagenomes were dominated by equal proportions of *Streptococcus sp.* and *Rothia sp.*, the salivary metatranscriptomes were characterized by high proportions of *Streptococcus sp.* but very low proportions of *Rothia sp.*

**DISCUSSION**

The purpose of this study was to characterize and compare salivary microbiota in oral health and disease using metagenomic and metatranscriptomic approaches. The primary finding was that both metagenomic and metatranscriptomic analyses identified higher salivary abundance of specific oral bacterial species in periodontitis and dental caries patients. Furthermore, salivary expression of specific metabolic genes was significantly different in patients with periodontitis and caries compared to healthy controls, and high activity (RNA/DNA ratio) of *F. alocis* and *Neisseria sp.* was associated with periodontitis. To the best of our knowledge, this is the first study to perform comprehensive simultaneous metagenomic and metatranscriptomic characterization of the salivary microbiota.

*Streptococcus* was identified as the most predominant bacterial genus by use of both metagenomic and metatranscriptomic analyses (Figs. 2a–b). These findings are in line with 16S-metabarcoding and metaproteomic analyses performed on the same samples,\(^3,17,21\) and underline the importance of *Streptococcus* in the composition of the salivary microbiota. Moreover, metagenomic and metatranscriptomic analyses yielded significant differences in characterization of the salivary microbiota (Figs. 2a–b). Remarkably, this could be acknowledged at genus level where *Streptococcus* accounted for approx. 25% of all assigned DNA reads compared to 50% of all assigned RNA reads. This finding was further supported, as phylum *Firmicutes*, which includes the genus *Streptococcus*, was the most active bacterial phylum in saliva as determined by the RNA/DNA ratio (Fig. 4a). Interestingly, this observation aligns with metagenomic and metatranscriptomic data extracted from a batch of supragingival samples, in which major differences in bacterial metagenomic and metatranscriptomic profiles were observed.\(^5,8\) Thus, this observation suggests that some bacterial genera, including *Streptococcus* may constitute the more active part of the salivary microbiota, while other members of the salivary microbiota, for example *Rothia sp.* might be largely inactive.

Based on relative abundance of the taxa at metagenomic level, multiple periodontitis-associated and caries-associated bacterial species were identified (Fig. 3b). These findings are in line with previous studies employing various contemporary, molecular methods for bacterial identification in saliva,\(^17,18,20,22,24\) including 16S-metabarcoding identification performed on the same saliva samples.\(^7\) Subgingival colonization of bacterial species such as *P. gingivalis*, *T. forsythia*, *P. micra*, and *F. alocis* is considered strongly associated with periodontal disease activity.\(^6,25,26\) Furthermore, *S. mutans* and *Lactobacillus sp.* are considered key etiological agents in the cariogenic supragingival biofilm based on their proficient capability of sugar fermentation.\(^7,27\) These specific and well-known disease-associated bacterial species were identified in higher relative abundance in saliva samples from patients with periodontitis and dental caries. Essentially, this finding reinforces the concept that pathogenic bacteria are shed from local periodontitis and dental caries. Essentially, this finding reinforces the concept that pathogenic bacteria are shed from local periodontitis and dental caries.
caries lesions into saliva. This may increase salivary abundance of these bacteria in patients with periodontitis and dental caries. However, low levels of \textit{P. gingivalis} and \textit{S. mutans} are also occasionally identified in saliva from orally healthy individuals\textsuperscript{28,29}, indicating that in some individuals these bacteria may be part of the resident oral microbiota.

The salivary microbiota has been reported to primarily consist of bacteria shed from the tongue, throat and tonsils\textsuperscript{16}. The turnover time of the salivary microbiota is extremely high due to continuous swallowing\textsuperscript{15}. Bacterial activity is ultimately determined by the level of substrates and local environmental conditions\textsuperscript{30,31}. Thus, bacterial activity of the salivary microbiota

**Fig. 2** Relative abundance. \textbf{a}: Top 20 predominant microbial genera. \textbf{b}: Top 20 predominant microbial species based on metagenomic and metatranscriptomic analyses
Fig. 3 Periodontitis and caries-associated salivary microbiota. a: Venn diagram displaying number and proportion of microbial species in salivary core and non-core microbiome based on metagenomic approach. b: Bacterial species with significantly higher relative abundance in oral health, dental caries or periodontitis. c: Venn diagram displaying number and proportion of microbial species in salivary core and non-core microbiome based on metatranscriptomic approach. d: Bacterial species with significantly higher relative abundance in oral health, dental caries or periodontitis. Sample denotation: green: healthy controls, blue: dental caries, red: periodontitis.
may relate to remnants of the phenotypic profile expressed in the local environment. Functional analysis of metatranscriptomic data highlighted significant differences in salivary gene expression in patients with periodontitis and caries compared to orally healthy controls. Saliva from orally healthy controls was associated with an up-regulation in functional activity of carbohydrate metabolism compared to periodontitis patients (Fig. 5a). Furthermore, saliva from caries patients expressed up-regulation of carbohydrate metabolism and down-regulation of lipid-metabolism compared to orally healthy controls (Fig. 5b). Thus, data from this study clearly demonstrate bacterial metabolic activity of the salivary microbiota at the community level, which is surprising, when considering the high turn-over rate of the salivary microbiota. Unfortunately, the sequencing depth was not sufficient for comparisons of species-specific metabolic activity between groups. However, as approx. 50% of all RNA reads could be taxonomically assigned to Streptococcus species it is likely that a substantial part of salivary metabolic activity derived from Streptococcus species, which is further supported by the fact that Streptococcus species are proficient in carbohydrate metabolism. Notably, 16S-metabarcoding analysis of the same samples showed significantly lower relative abundance of Streptococcus species in samples from periodontitis patients compared to healthy controls, which suggests that metatranscriptomic profiles observed in this study probably associates with differential functional activity of salivary Streptococcus species in oral health and disease.

Taxonomic assignment of metatranscriptomic data of the salivary microbiota was employed as a proxy marker of microbial activity. Notably, all the usual suspects of traditional periodontal pathogens (P. gingivalis, T. forsythia, P. micra, and F. alocis) and cariogenic bacteria (S. mutans and Lactobacillus sp.) were identified with significantly higher relative abundance at metatranscriptomic level (Fig. 3d). This suggests that bacteria shed from local diseased lesions sustain microbial activity when dispersed into saliva. This is further demonstrated by the higher salivary RNA/DNA ratio of F. alocis and Neisseria sp. in patients with periodontitis. Consequently, higher salivary bacterial activity of the usual suspects of periodontitis (P. gingivalis, T. forsythia, P. micra, and F. alocis) and dental caries (S. mutans and Lactobacillus sp.) may possess a potential for use as a biomarker of the respective diseases. Further studies with additional sequencing depth and larger sample size (including supra and subgingival samples)
would allow the comparison of salivary and local diseased lesions' microbiota. This would further strengthen the finding of the present study.

In conclusion, this study is the first to demonstrate differences in salivary metabolic gene expression in oral health and disease based on metatranscriptomic characterization of the salivary microbiota. Future longitudinal studies are warranted to reveal whether salivary screening of metabolic gene expression can identify oral diseases at preclinical stages.

METHODS

Study population and sample collection

The study population and oral examination has been presented elsewhere in detail. In brief, 10 patients each, with chronic periodontitis, or manifest dental caries and orally healthy controls were enrolled in April 2015 at Copenhagen University, School of Dentistry. Periodontitis was defined as bleeding on probing ≥ 25% of total sites + minimum two teeth with clinical attachment level ≥ 4 mm + minimum two teeth with probing depth ≥ 6 mm and dental caries was defined as manifest untreated caries ≥ 3 surfaces. All 30 participants signed informed consent. The study was completed in accordance with all relevant guidelines and procedures, approved by the regional ethical committee (H-15000856) and reported to the Danish Dental Authorization (2015-S4-0970).

Stimulated saliva samples (n = 30) were collected between 8:00 a.m. and 11:00 a.m. according to a standardized protocol. Immediately after collection each saliva sample was divided in four aliquots of 1 mL each. Two of these aliquots have previously been used for analysis of bacterial 16S metabarocoding targeting rRNA gene and metaproteomics. The remaining aliquots were allocated for this study, one each for metagenomics and metatranscriptomics. RNAlater (Life Technologies, Denmark) was added to the aliquot allocated for metatranscriptomics and all aliquots were immediately stored at −80°C until further processing.

Metagenomic and metatranscriptomic library preparation and sequencing

Frozen saliva samples were thaw on ice and centrifuged to collect debris. Lysing Matrix B tube (MP Biomedicals) was used to lyse the debris (10 m, 30 s, twice). DNA and RNA (i.e., total RNA) were extracted using MasterPure™ Complete DNA and RNA Purification Kit (Epientre, Chicago, IL, USA). For DNA samples, RNase was used to remove RNA contamination. For RNA samples, DNase was used to remove DNA contamination.

RNA quality, quantity and residual DNA were determined using a 2100 Bioanalyzer (Agilent Technologies, Santa Clara, CA, USA), as well as Invitrogen RiboGreen and PicoGreen assays (ThermoFisher Scientific, Waltham, MA, USA), respectively. Library preparation was performed according to the TruSeq Stranded mRNA protocol (Illumina, San Diego, CA, USA) with the following modifications: The oligo-dT mRNA purification step was omitted and instead, 200ng of total RNA were directly added to the Elution2-Frag-Prime step. The PCR amplification step, which selectively enriches for library fragments that have adapters ligated on both ends, was performed according to the manufacturer's recommendation but the number of amplification cycles was reduced to 12. Each library was uniquely tagged with one of Illumina's TruSeq RT RNA barcodes to enable library pooling for sequencing. RNA libraries were then sequenced in six lanes of Illumina Hiseq2500 rapid runs at a read-length of 100 bp paired-end, generating an average of 32.71 M reads per sample (26.40–38.44 M). DNA quality, quantity and depth were determined using a 2100 Bioanalyzer and Invitrogen's PicoGreen assay, respectively. Library preparation was performed according to Illumina's TruSeq Nano DNA Sample Preparation protocol. The samples were sheared on a Covaris S220 (Covaris, Woburn, MA, USA) to ~ 450 bp, following the manufacturer’s recommendation, and uniquely tagged with one of Illumina's TruSeq LT DNA barcodes. DNA libraries were then sequenced in 2 lanes of an Illumina Hiseq2500 rapid run at a read-length of 250 bp paired-end, generating an average of 25.62 M (16.46–32.64 M) reads per sample.

Sequence data was deposited in GenBank (Sequence Read Archive) and is available under the BioProject (PRJNA396840).

Read preprocessing

Illumina adaptors, 5’ or 3’ bases with quality scores lower than 20, as well as read pairs having a mate with any ambiguous base (i.e., N) or shorter than 150 or 50 bp for DNA and RNA sequences, respectively, were removed using cutadapt in paired-end mode (version 1.10, -a -A --max-n 0 -n 1 -q 20,20 --quality-base 33 –minimum-length 150/50 0 6). Overall, quality-trimmed reads represented 98.1% (97.5–98.6%) of DNA and 99.3% (99.0–99.4%) of RNA reads.

RNA reads were then subjected to sortmeRNA (version 2.0), using provided rRNA databases and default parameters for in silico depletion of ribosomal RNA. Ribosomal RNA sequences represented an average of 82.9% of the quality-trimmed RNA reads (71.1–88.4%). Human reads were then removed from metagenomic and metatranscriptomic dataset by aligning DNA and mRNA reads to the human genome (hg38, ftp://ftp.ncbi.nlm.nih.gov/ genomes/all/GCF/000/014/36.3_GRCh38.p10/GCF_0000140536.3_GRCh38.p10_Genomic.fna.gz) using Bowtie2 (version 2.2.9, –un --sensitive-local). Overall, around 9.1 M (2.3–9.2 M) DNA and 4.9 M (2.9–9.1 M) RNA reads were used after quality filtering, RNA removal and host (i.e., human) contamination for further analysis. Prior to further analysis, 2.26 M reads were randomly selected per sample to obtain equivalent sequencing depth and allow robust comparison.

Taxonomic and functional community profiling

In order to characterize the taxonomic composition of metagenomic and metatranscriptomic dataset we used a similar approach as illot et al. 2016. Subsampled sequences were aligned against the NCBI non-redundant (NR) protein database (ftp.ncbi.nlm.nih.gov/blast/db/FASTA/nr.gz, March 2016) using DIAMOND (version 0.7.10.59) with default parameters. The lowest common ancestor approach implemented in MEGAN6 (L.C.A., version CE 6.5.5), ms 50 (minimum alignment bitscore = 50) –supp 0 –suf 25 was used to assign reads at the phylum, genus and species levels for the metagenomic and metatranscriptomic dataset, respectively. In order to characterize the molecular level function of the salivary microbiome, aligned reads were also assigned to KO functional orthologs using GT to KEGG KO functional identifier mapping file as implemented in MEGAN6.

Statistical analysis

Taxonomic and functional count tables from metagenomic and metatranscriptomic dataset were exported from MEGAN6 and imported as single phyloseq objects (phyloseq, R). Taxonomic and functional tables were filtered to remove taxa accounting less than 100 sequences in total and observed in less than 3 samples (filter_taxa function). Tables were then rarefied to an even sequencing depth using rarely_even_depth function.

Microbial communities were characterized using alpha-diversity indices (i.e., number of observed taxa and Shannon diversity indices) and beta-diversity (e.g., Bray-Curtis dissimilarity) at the Phylum, Genus and Species levels on both metagenomic and metatranscriptomic datasets.

Differences in alpha and beta diversity between healthy control, caries and periodontitis groups dataset were tested using ANOVA and PERMANOVA statistical tests for alpha and beta-diversity, respectively (i.e., Adonis, PERMANOVA functions). Gaussian distribution of ANOVA residuals and multivariate spread was tested to ensure ANOVA and PERMANOVA assumptions (i.e., shapiro.test and betadisper). Statistical significance of taxa-group association was tested using one-sided signassoc function from the indicspecies R-package. Sidak’s correction was applied for multiple testing. RNA/DNA log ratios were determined for the relative abundance of taxa identified in metagenomic and metatranscriptomic dataset at phylum and species level. Due to skewed distribution of RNA/DNA log ratios, non-parametric Kruskal-Wallis H test was applied to test differences between groups.

Differential expression analysis of identified KO functional orthologs between healthy controls and patients with periodontitis and caries were conducted on unrerared functional counts using DESeq2 standard meta-/transcriptomic pipeline which includes size factor normalization and standardization procedures as well as negative binomial model testing. Values were adjusted for multiple testing using FDR correction—using the following parameters: test = "wald" and fitType = "parametric".

Data availability statement

Sequence data that support the findings of this study have been deposited in GenBank (Sequence Read Archive) and is available under the BioProject (PRJNA396840).
ACKNOWLEDGEMENTS
This study was supported by SCELSE core funding.

AUTHOR CONTRIBUTIONS
Daniel Belstrom (D.B.), Palle Holmstrup (P.H.), and Michael Givskov (M.G.) designed the study. D.B. and Tim Holm Jakobsen (T.H.J.) collected and stored the samples. Yang Liu (Y.L.), Liang Yang (L.Y.), Daniela I. Drautz-Moses (D.I.D.M.) and Stephan C. Schuster (S.C.S.) performed DNA and RNA sequencing. Florentin Constancias (F.C.), Gurjeet Singh Kohli (G.S.K.H.) and D.B. performed the bioinformatics analyses. D.B., F.C., and M. G. wrote the manuscript, which P.H., T.H.J., Y.L., L.Y., D.I.D.M., S.C.S., and G.S.H. critically revised. All authors read and approved the final version of the manuscript.

ADDITIONAL INFORMATION
Supplementary Information accompanies the paper on the npj Biofilms and Microbiomes website (https://doi.org/10.1038/s41522-017-0031-4).

Competing interests: The authors declare that they have no competing financial interests.

Publisher’s note: Springer Nature remains neutral with regard to jurisdictional claims in published maps and institutional affiliations.

REFERENCES

Open Access This article is licensed under a Creative Commons Attribution 4.0 International License, which permits use, sharing, adaptation, distribution and reproduction in any medium or format, as long as you give appropriate credit to the original author(s) and the source, provide a link to the Creative Commons license, and indicate if changes were made. The images or other third party material in this article are included in the article’s Creative Commons license, unless indicated otherwise in a credit line to the material. If material is not included in the article’s Creative Commons license and your intended use is not permitted by statutory regulation or exceeds the permitted use, you will need to obtain permission directly from the copyright holder. To view a copy of this license, visit http://creativecommons.org/licenses/by/4.0/.

© The Author(s) 2017