

1 **Biological and cultural drivers of oral**
2 **microbiota in Medieval and Post-Medieval**
3 **London, UK**

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5 A. G. Farrer^{1*}, J. Bekvalac², R. Redfern², N. Gully³, K. Dobney^{4,5,6}, A. Cooper¹⁺, and
6 L. S. Weyrich¹⁺

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8 Affiliations:

9 ¹Australian Centre for Ancient DNA, School of Biological Sciences, University of
10 Adelaide, Adelaide, South Australia, Australia

11 ²Centre for Human Bioarchaeology, Museum of London, London, United Kingdom

12 ³School of Dentistry, Faculty of Health Sciences, University of Adelaide, Adelaide,
13 South Australia, Australia

14 ⁴Department of Archaeology, Classics and Egyptology, School of Histories,
15 Languages and Cultures, University of Liverpool, Liverpool, United Kingdom

16 ⁵Department of Archaeology, University of Aberdeen, Aberdeen, United Kingdom

17 ⁶Department of Archaeology, Simon Fraser University, Burnaby, Canada

18 ⁺These authors contributed equally to this work

19
20 Corresponding author:

21 andrew.farrer@adelaide.edu.au

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24 Abstract

25 The trillions of microorganisms that live in association with the human body
26 (microbiota) are critical for human health and disease, but there is a limited
27 understanding of how cultural and environmental factors shaped our microbiota
28 diversity through time. However, biomolecular remnants of the human oral
29 microbiota - recovered from the calcified dental plaque (calculus) of our long-dead
30 ancestors - are providing a new means of exploring this key relationship of our
31 evolutionary history. Here, we correlate extensive experimental, archaeological, and
32 biological metadata with 128 ancient dental calculus specimens from Medieval and
33 Post-Medieval London, UK (1066 – 1853 CE). We identify a significant association
34 between microbiota and oral geography (*i.e.* tooth type and tooth surface), which has
35 confounded ancient microbiota studies to date. By controlling for oral geography,
36 however, we identify the first associations between ancient microbiota and cultural
37 and environmental signatures. We find significant links between ancient British
38 microbiota structure and health, including skeletal markers of stress that may reflect
39 low socioeconomic status. Furthermore, this study provides baseline data to explore
40 factors that drive microbiota differentiation within and between ancient populations
41 and highlights the potential of ancient microbiota to infer detailed health and socio-
42 cultural information about the past.

43 Introduction

44 It is now widely recognized that microbial communities both on and in the
45 human body (microbiota) fulfill key functional roles that include accessing nutrients
46 otherwise inaccessible from food resources, removing dead epithelial cells, restoring
47 tooth enamel, and interacting with the immune system (1–5). Conversely, many
48 diseases have also been linked to alterations in microbiota composition and/or
49 function, including oral disease, arthritis, respiratory disorders, cancer, obesity,
50 mental disorders, and many more (6–10). While several studies have examined how
51 modern commensal bacterial communities respond to changes in lifestyle, diet, and
52 environment (11–13), we know almost nothing about their evolutionary history or
53 how that history shaped our own bio-cultural identity.

54 Bioarchaeological remains contain key data from natural experiments run in
55 the past that can reveal the history of the bacterial communities found in modern
56 populations, including the effects of different living conditions, diet, and disease (14–
57 17). The analysis of such data is now routinely possible using ancient DNA (aDNA)
58 recovered from calcified dental plaque (calculus), which is widespread in the
59 archaeological record (18). Dental calculus is a calcified microbial biofilm on the
60 surface of the teeth (19). This matrix preserves human microbiota in real-time during
61 the life of the individual, forming the only reliable record of its kind in the
62 archaeological record (15,18,20–23). Bacterial DNA recovered from ancient dental
63 calculus has already revealed major changes in human oral microbiota across the
64 major human biocultural transitions of the Holocene/Anthropocene (~10,000 ybp and
65 ~200 ybp, respectively) (18), as well as information on real-time pathogen evolution
66 (22,23). Eukaryotic DNA recovered from ancient dental calculus has also revealed
67 differences in diet between past populations (22,23). However, no detailed, long-
68 term studies of ancient microbiota within a single human population through time
69 exists and, as such, many cultural and environmental factors that shaped our ancient
70 microbiota remain unknown.

71 In addition, several fundamental technical issues remain unresolved in ancient
72 dental calculus research. For example, the influence of oral geography has not been
73 assessed (*i.e.* if differences between the specific sampling location in the mouth
74 significantly impact results). Microbial composition is known to vary between tooth
75 surfaces in modern individuals (24), but this is yet to be explored in detail in studies

76 of ancient oral microbiota. In addition, recent studies of ancient dental calculus have
77 been limited to very small sample sizes across large geographic ranges, likely
78 limiting the ability to detect specific factors that shape ancient microbiota. Lastly,
79 contamination control and detection is a critical issue, especially in samples with low
80 levels of endogenous DNA (such as ancient dental calculus). In such samples,
81 background or contaminant DNA from the laboratory can easily be high enough to
82 drive signals (25). Filtering methods can be used to conservatively remove
83 contaminant species (23); however, there is not currently an accepted systematic
84 approach to wholly assess endogenous oral signal within dental calculus. Such
85 unresolved issues might easily provide false positive results or confound real
86 historical patterns, and these issues will become more important as studies seek to
87 increase resolution.

88 Here, we examine the largest number of ancient dental calculus samples
89 studied to date – from 128 Medieval and Post-Medieval Londoners [1066 – 1853].
90 We focused on a single city in order to remove geographic variation and to explore
91 the localised cultural and environmental factors and how they shaped the British oral
92 microbiota. The dental calculus samples utilized for this study are part of the
93 Museum of London’s (MoL) human osteology collections, which include extensive,
94 additional and detailed information about dating, paleopathology, and cultural
95 context (*e.g.* age, sex, diet, location, oral health, and systemic disease). We go on to
96 combine this extensive archaeological and biological metadata with detailed
97 experimental information in order to explore whether tooth type and location in the
98 oral cavity significantly bias the conclusions drawn.

99 **Methods**

100 *Ethics*

101 Research was conducted after approval from the Human Research Ethics
102 Committee at the University of Adelaide (H-2012-108) following agreement to
103 sample from the MoL Research Committee.

104

105 *Archaeological Context and Site Information*

106 160 samples were collected from nine archaeological sites in a 10-square
107 mile section of London (16km²), which formed a continuous temporal sequence from
108 1066 – 1853 CE (Figure 1; Table S1). The datasets included burial sites and
109 cemeteries of monastic (Bermondsey Abbey, Spital Square, Merton Priory, St. Benet
110 Sharehog, St. Mary Graces, and St. Brides) and laymen (Guildhall Yard), low class
111 (Cross Bones), and upper class (Chelsea Old Church). All individuals were over 18
112 years old and had extensive metadata collated in the MoL's Wellcome Osteological
113 Research Database (WORD), including sex and age estimates, blood disorders,
114 dental and vertebral anomalies and pathologies, and joint disease (Table S1).

115

116 *Sample collection*

117 All sampling was completed at the Museum of London using sterile
118 procedures as previously published (14). Briefly, a facemask and gloves were worn
119 to limit contamination from the researcher. The gloves were changed between each
120 sample to limit cross-contamination. A sterile dental pick was used to remove the
121 dental calculus deposit from a single surface of one tooth per skeleton. Pressure was
122 applied in parallel to the tooth surface to ensure the enamel was not damaged.
123 Calculus fragments were collected in sterile aluminum foil and placed into sterile
124 plastic bags for transport to the aDNA facility at the Australian Centre for Ancient
125 DNA, University of Adelaide, Australia.

126

127 *Sample decontamination, DNA extraction, shotgun library preparation, and* 128 *DNA sequencing*

129 All laboratory work was conducted in the specialized stand-alone aDNA
130 facility at the Australian Centre for Ancient DNA at the University of Adelaide.
131 Samples were processed in a random order to avoid bias. All calculus samples were

132 decontaminated with bleach and UV exposure, as described elsewhere (18). To
133 recover preserved DNA, dry samples were powdered in a sterile tube, immediately
134 following decontamination. A modified silica-based DNA extraction was used on all
135 samples (23). The total volumes of lysis and DNA binding buffers were modified to
136 account for the small sample size: 1.7 mL lysis buffer (1.6 mL 0.5 M EDTA (0.5M);
137 100 μ L SDS (10 %); and 20 μ L proteinase K (20mg/ml)) and 3 mL guanidinium DNA
138 binding buffer. Negative extraction blank controls were included at a ratio of 1:7,
139 control per calculus samples. Next, shotgun libraries were generated without
140 enzymatic damage repair (23). Briefly, 20 μ L of DNA extract was prepared by
141 enzymatic polishing to produce blunt ended fragments prior to ligation of truncated
142 barcoded Illumina adaptors. To maintain sequence complexity, each sample was
143 amplified in triplicate using 13 cycles with HiFi Taq polymerase and full length
144 indexed Illumina adaptors (26). The resulting triplicate amplifications were pooled
145 and purified with the Agencourt AMPure XP system. All samples were then pooled,
146 purified, and then quantified using a TapeStation (Agilent) and quantitative PCR
147 (KAPA Illumina quantification kit) to create a 2 nM sequencing library. In total, 128 of
148 the 161 (79.5%) dental calculus samples yielded high-quality DNA libraries that were
149 sequenced using a high output 2 x 150 bp kit on the Illumina NextSeq.

150

151 *Bioinformatic analysis and quality filtering*

152 To identify the microbial communities preserved within samples, raw FastQ
153 files were demultiplexed using sabre (<https://github.com/najoshi/sabre>). Bbmerge
154 (<http://sourceforge.net/projects/bbmap/>) was used to merge reads (5 bp overlap),
155 and AdapterRemoval (27) was applied to identify and remove the 5' and 3' barcode
156 and adaptor sequences. Reads greater than 300 bp were discarded, as they likely
157 represent modern contamination (15). Microbial species were identified using
158 MALTX (23,28) against the NCBI nr database (2014), and the resulting information
159 was uploaded and filtered using default LCA parameters in MEGAN5 (29). The
160 identified reads within the samples were then normalized to 129,760 sequences,
161 which was the lowest number of reads observed in any sample. Lastly, laboratory
162 contaminant signal was removed from all samples by filtering any contaminant
163 species observed in the negative controls from the calculus samples, as previously

164 described (23). 464 contaminant taxa (26.4% of all species level identifications) were
165 removed from the overall dataset.

166

167 *Statistical analysis of associated metadata*

168 The oral geography of the samples and the workflow metadata were
169 compared to the microbiota to assess the potential impact of oral geography and
170 sample handling. All sequence reads that could be assigned to organisms for each
171 sample were exported from MEGAN5 and transformed for use in QIIME (V1.8) (30).
172 PERMANOVA tests (anosim; 9,999 permutations) were used to correlate oral
173 sampling location (e.g. upper or lower jaw, tooth type, and tooth surface), sample
174 information (i.e. fragment size and sub- or supra-gingival), and processing details
175 (i.e. date of sampling, extraction, sequencing) with the Bray-Curtis matrix (Table S2).
176 Following this analysis, samples were filtered based on tooth type, tooth surface, and
177 sub- or supragingival calculus. The three largest datasets were taken forward for
178 further analysis: Molar, Lingual/Palatal, Supragingival (n = 36); Premolar,
179 Lingual/Palatal, Supragingival (n = 18); and Incisor, Lingual/Palatal, Supragingival (n
180 = 18). These groupings were then utilized to examine metadata associated with
181 health, socio-cultural, and environmental factors that drive microbiota diversity using
182 a PERMANOVA test (anosim) in QIIME with a cutoff of $p = 0.05$ for significance. A
183 G-test test was used to identify specific species that contributed to the differences
184 observed in each metadata category with a cutoff of $p = 0.05$ for significance. To
185 ensure statistical tests were appropriate, a minimum of five samples per group was
186 used when comparing two metadata categories, while a minimum of three samples
187 per group was enforced when three or more metadata categories were compared.
188 Samples without metadata for a certain category were excluded from those specific
189 tests.

190 Results

191 *Robust oral microbiota recovered from historical London samples*

192 Across the nine archaeological sites included in this study, all samples were
193 dominated by bacterial phyla typical of the modern and ancient human mouth,
194 including Actinobacteria, Firmicutes, and Bacteroidetes (22,23) (Figure 2), consistent
195 with recovery of an oral community structure in these samples. We then examined
196 the impacts of contaminant DNA and laboratory methods on these data. To identify
197 ancient calculus samples that lack true endogenous signal, we looked to identify a
198 signal shared among negative controls that is representative of a laboratory
199 environment. Bray-Curtis dissimilarities were calculated in MEGAN5 for the
200 published ancient calculus control (n=1) (23) and negative controls sequenced within
201 this study (n=10). Negative controls with Bray-Curtis dissimilarities for all negative
202 controls were <0.6 from each other, indicating a consistent microbial signal present
203 in the laboratory. Any calculus sample that fell within four times the mean standard
204 deviation of this group was considered poorly preserved (*i.e.* largely consisted of
205 laboratory contamination) and not fit for downstream analysis. Lastly, PERMANOVA
206 tests indicated that there was no correlation between the microbial community
207 structure of all samples and the date of sampling, DNA extraction, or library
208 preparation (anosim; $p > 0.01$; Table S2). Overall, 128 dental calculus samples from
209 Medieval London appear to contain robust oral microbiota signatures and so were
210 retained for downstream analysis.

211

212 *Oral geography accounts for some variation between individuals*

213 To examine if oral geography impacts the microbiota present in ancient dental
214 calculus samples, we correlated oral microbial diversity in all ancient samples (n =
215 128) with the oral sampling location. The tooth type was significantly correlated with
216 the microbiota in each sample (anosim; $p=0.0001$; Table S2). Tooth type was also
217 the variable that explained the most variation on the PCoA plot calculated from Bray-
218 Curtis values of all samples (Figure 3), accounting for 44.9 % of the variation in the
219 data. We then controlled for tooth type by subsequently processing the samples from
220 each tooth type independently. Within tooth types, we observed that tooth surface
221 (buccal, lingual/palatal, interproximal) was a significant driver of diversity (anosim; p
222 < 0.01; Figure 4; Table S2), except for canines. In the canine data set, samples

223 correlated with lab extraction date and library ID (anosim; $p = 0.0286$ and $p = 0.0105$,
224 respectively), indicating potential introduction of background laboratory contaminants
225 during sample preparation. However, the canine data set also contained the fewest
226 samples ($n=14$), so this correlation may be associated with insufficient sampling
227 depth. For stringency, canine samples were not included in further analysis. This
228 highlights the need to record laboratory information for data processing. Lastly, we
229 examined differences between sub- and supragingival microbiota, as differences
230 between these sampling locations have been reported in modern populations (31).
231 Unfortunately, there were very few samples of subgingival dental calculus in our
232 study (5 of 128 samples; Table S1) limiting our ability to test this. However, to
233 minimize any contributions from this potential bias, the five subgingival samples were
234 removed from downstream analysis. Overall, these data indicate that oral geography
235 can contribute to microbiota diversity within ancient calculus and needs to be
236 controlled for in order to avoid introducing bias.

237

238 *Microbiota are correlated with disease*

239 We examined links between ancient oral microbiota and evidence of oral and
240 systemic diseases, a wide-range of lifestyle factors, and temporal differences. First,
241 we examined relationships between microbiota (supragingival dental calculus
242 samples within three data sets: the lingual/palatal side of the molar ($n=36$), premolar
243 ($n=18$), and incisor ($n=18$) teeth) and sample metadata, including archaeological site,
244 paleopathology, cultural factors, disease, and period (Table S1). There was no
245 association between microbiota and archaeological site or time period (anosim; $p=$
246 <0.01 ; Tables S4-6), which have been assumed to be driving factors in previous
247 aDNA studies (18,23). In addition, no metadata category significantly explained
248 variation within the incisor group (anosim; $p= <0.01$; Table S4). We also examined
249 factors previously hypothesized to drive microbiota differences in ancient
250 populations, including site location (18) and diet (14), and those modern studies
251 indicate may be present in ancient populations, e.g. rural-urban status (32); none
252 were found to be significant. However, disease factors were related to variation
253 within the molar and premolar groups. Abscesses (oral disease; $p = 0.0052$) and
254 various bone pathologies (*i.e.* porosity, osteophytic lipping, and non-specific
255 periostitis – examples of systemic disease; $p = 0.0308 - 0.0466$) were all

256 significantly linked with microbiota variation within the molar data set (Table S6), and
257 these correlations were investigated further.

258

259 *Oral disease associations with microbiota alteration*

260 The presence of abscesses within the oral cavity was correlated with
261 microbiota within supragingival lingual/palatal dental calculus of molar teeth ($p <$
262 0.0052 ; Table S6). Therefore, we explored the microbial species linked to this oral
263 disease. 30 bacterial species were increased in individuals with dental abscesses,
264 while 20 species were decreased (G-test, $p < 0.05$; Table S7). Notably, *Prevotella*
265 and *Streptococcus* species, which are associated with dental abscesses in modern
266 humans (33), were significantly increased in the ancient individuals with abscesses
267 (178.0% and 330.0% of the mean read count in healthy individuals, respectively;
268 Table S7). However, other species linked with abscess in modern populations, such
269 as *Porphyromonas gingivalis* and *Treponema denticola* (34), were found to be
270 significantly lower in ancient individuals with abscesses (51.0% and 62.2%,
271 respectively). In addition, other potential microbes involved in abscess formation
272 were not significantly different (e.g. *Clostridium*, *Fusobacterium*, and *Bacteroides*
273 species). Individuals with dental abscesses also had significantly increased
274 proportions of archaea, including *Methanobrevibacter* species, which are primary
275 colonizers of the tooth surface (23) and are associated with severe periodontal
276 disease in modern populations (35) (185.5%; Table S7). Together, this suggests that
277 dental abscesses present in ancient Londoners were polymicrobial, similar to
278 modern populations, and left unique signatures in ancient oral microbiota, including
279 the presence of archaea.

280

281 *Relationships between bone pathologies and microbiota composition*

282 Porosity and Osteophytic lipping

283 Observations of bone porosity indicate excessive opening of the pores in
284 bone to allow blood, nerve, and other soft tissue to enter (36). This can be
285 associated with age of the individual (37), as well as specific diseases such as
286 anemia and rickets (38). Osteophytic lipping (outgrowths of bone) form around the
287 joint surfaces and are often present alongside arthritis (39). Within our molar data
288 set, all individuals with bone porosity ($n = 9$) also had osteophytic lipping in joints,

289 while five additional individuals only had osteophytic lipping (Table S1).
290 Consequently, taxonomic differences may not be specifically associated with either
291 trait. Porosity was linked to an increase in 28 microbiota taxa and a decrease in 28
292 taxa (G-test; $p < 0.05$; Table S8). *Prevotella* and *Methanobrevibacter* species, which
293 have previously been associated with bone diseases, such as arthritis (40), were
294 increased in individuals with bone porosity. However, as with abscesses, species
295 linked to oral inflammatory diseases and generally increased systemic inflammation,
296 such as *Porphyromonas gingivalis*, *Treponema denticola*, and *Tannerella forsythia*
297 (6,34), were decreased in individuals with bone porosity (24.6%, 72.3%, and 65.6%
298 respectively). Osteophytic lipping was associated with an increase in 19 microbiota
299 taxa (G-test; $p < 0.05$, Table S9). These taxa are similar to those increased with
300 porosity; for example, *Prevotella* species and *Methanobrevibacter* species are also
301 present in higher abundance in individuals with osteophytic lipping (145.7% and
302 290.1%, respectively). Together, these data suggests that specific oral microbiota
303 communities are found in ancient Londoners with joint disease.

304

305 Periostitis

306 Periostitis is an inflammation of tissue surrounding the bone and often
307 indicative of disease in the underlying bone (41). In living patients, it can be
308 associated with various diseases (including syphilis and skin ulcers) and is linked
309 with trauma (41). Within calculus samples from individuals with periostitis, *Prevotella*
310 species and *Methanobrevibacter* species were again increased (147.8% and 217.7%
311 respectively; Table S10), as observed with porosity and osteophytic lipping.

312 However, several unique bacterial species were linked with this disease, including
313 *Capnocytophaga*, *Clostridia*, *Eubacterium*, *Firmicutes*, *Mogibacterium*, *Neisseria*,
314 and *Pyramidobacter* (G-test; $p < 0.05$; Table S10). This indicates that periostitis may
315 be linked to a unique shift in the human microbiota, which is different to those
316 observed for porosity and osteophytic lipping.

317

318 Schmorl's nodes

319 In contrast to the molar samples, the microbiota of premolar teeth microbiota
320 were correlated with Schmorl's nodes – a vertebral bone pathology (anosim; $p =$
321 0.05; Table S5). While Schmorl's nodes are common, their pathological status is
322 poorly understood. It is thought that they appear in response to herniation of the

323 cartilaginous disc into the vertebral body as a result of direct physical stress (42,43)
324 Methanobrevibacter was increased in individuals with the pathology (129.3%; Table
325 S11). This is consistent with the previous pathologies examined. In contrast,
326 Prevotella was decreased in those with Schmorl's Nodes (75.1%; Table S11). It is
327 worth noting that while each of these bone pathologies is classified independently, it
328 is possible that they have arisen from a similar cause, such as an increase in manual
329 labor within this population (44). The lack of a specific causal factor for these
330 paleopathology could explain why similar associations with the oral microbiota are
331 inferred.

332 Discussion

333 In the largest ancient dental calculus study to date, we investigated the roles
334 that experimental biases can play in ancient dental calculus research and revealed
335 direct associations between ancient human oral and systemic health and the oral
336 microbiota. We show that the oral geography of the mouth is a critical factor when
337 identifying fine-scale factors that impact the human microbiota, as tooth type and
338 tooth surface explained the greatest amount of variation observed in our data set.
339 Once these and other technical factors are controlled, information that underpins the
340 evolutionary history of the oral microbiota can be examined, highlighting the potential
341 of dental calculus to study the impacts of cultural and environmental factors on the
342 oral microbiota and, thus, the evolutionary history of the modern oral microbiota.

343 Within this study, the oral geography of dental calculus accounted for the
344 most significant source of variation within the data, raising questions about
345 conclusions drawn from previous ancient calculus research. Several studies have
346 combined dental calculus samples from different regions within the mouth (22) or not
347 factored in tooth type or tooth surface (23). Our study suggests that there are
348 limitations to these approaches and that ancient dental calculus studies moving
349 forward should account for tooth type and surface when examining correlations
350 between individual metadata and oral microbiota. Even though it markedly reduced
351 the total sample size per group, intrapersonal variation largely masked many of the
352 links between disease and microbiota. Notably, different sampling locations within
353 the mouth also revealed individual associations with skeletal metadata. While limited
354 sample numbers in some tooth groupings may affect these findings, oral microbial
355 communities from different teeth appear to respond differently in states of disease. In
356 future, a study of the oral geography within single ancient individuals, improved
357 bioinformatic methodologies, and increased sample sizes will help understand and
358 account for intrapersonal variation.

359 This study also highlights additional technical factors that may influence
360 ancient oral microbiota diversity, including contamination and experimental
361 metadata. To address potential laboratory contamination, we compared samples to
362 negative blank controls to confirm that samples did not have a microbial signal
363 consistent with the laboratory environment. Conservatively, any microbial species in
364 common with the negative controls was also removed from the samples. Samples

365 were also randomized before undergoing laboratory protocols. This demonstrates
366 the care with which ancient microbiota data must be handled, as researchers must
367 be careful to collect, document, and analyse sampling and experimental metadata to
368 accurately interpret findings in ancient microbiota. In future studies, the collection of
369 detailed metadata will improve the identification of local cultural and environmental
370 factors that influenced the microbiota, and scientists will be better able to identify
371 potential links between oral microbiota and a past individual's life history.

372 This is the first study to examine ancient oral microbiota variation with respect
373 to a wide-range of individualistic metadata, including sex, age, rural-urban status,
374 religion, time, and location. Despite this wealth of information, this study revealed
375 that disease appears to be the single largest determinant of oral microbiota variation
376 in ancient London. We identified alterations in community structure associated with
377 both oral and systemic disease. In modern populations, oral microbiota have been
378 linked to a wide range of oral and systemic diseases, as diverse as dental caries,
379 gum disease, arthritis, heart disease, pre-term birth, Alzheimer's disease, and mental
380 disorders (5,6,9,34,45). Although past aDNA studies have identified ancient oral
381 pathogens as potential signatures of disease (18,21,22), we identify community level
382 changes linked to abscesses that include both known, abscess-associated
383 pathogens (*Prevotella* and *Streptococcus* species) and lesser known species
384 (*Methanobrevibacter* species). In addition, microbiota from molar teeth correlated
385 with bone disease, such as porosity, osteophytic lipping and peristitis and, in all
386 cases, were demarcated by an increase in oral *Prevotella* species. In modern
387 populations, *Prevotella* species are associated with arthritis (6), although the
388 mechanism underlying this link is not yet known. There may be a direct link between
389 this species and Porosity, Osteophytic Lipping, and Periostitis because oral
390 bacteria can escape the mouth and cause inflammation and lesions elsewhere in the
391 body (46). Alternatively, *Prevotella* species or community structure may serve as an
392 oral marker for overall community alterations that are linked to generalized
393 inflammation in the body. Consequently, causation cannot be inferred, and further
394 studies are needed to investigate mechanistic relationships.

395 Bone diseases linked to greater physical stress and malnutrition are typically
396 seen in ancient individuals with low socio-economic rank (47). In London, low socio-
397 economic status has been previously associated with poor bone health (47). While it
398 is possible that these bone diseases directly led to alterations in microbiota, it is also

399 possible that these relationships between disease and microbiota are explained by
400 greater lifestyle factors, such as low socioeconomic status. In fact, several studies to
401 date have linked socioeconomic status to microbiota community composition (48–
402 50). Our results are consistent with these observations, and may suggest that oral
403 microbiota can be used as a marker for socioeconomic status in the past. Further
404 research is needed to investigate this trend in different human cultures worldwide.

405 Large-scale sampling of ancient dental calculus from a single population has
406 provided an unprecedented opportunity to reveal information about how past
407 lifestyles shape the future of human health. This study reveals the impacts of
408 technical factors and highlights how critical disease was in the past for shaping the
409 evolutionary history of human oral microbiota. In addition, this study provides
410 fundamental baseline evidence to suggest that the oral microbiota can be used as
411 archaeological biomarkers of cultural, health, and environmental change, allowing
412 researchers to gain new and more detailed insights when other archaeological
413 information is unavailable.

414 **Competing Interests**

415 The authors declare no competing interests.

416 Author's Contributions

417 AGF, JB, RR, AC, KD and LSW contributed to study design. AGF, JB, and RR
418 submitted destructive sampling applications, collated metadata, and collected
419 samples. AGF conducted laboratory work and bioinformatic analysis. AGF, NG, KD,
420 AC, and LSW contributed to data interpretation. AGF and LSW outlined and wrote
421 the manuscript, and KD, AC, and LSW edited the manuscript.

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564 Figure and Table Legends

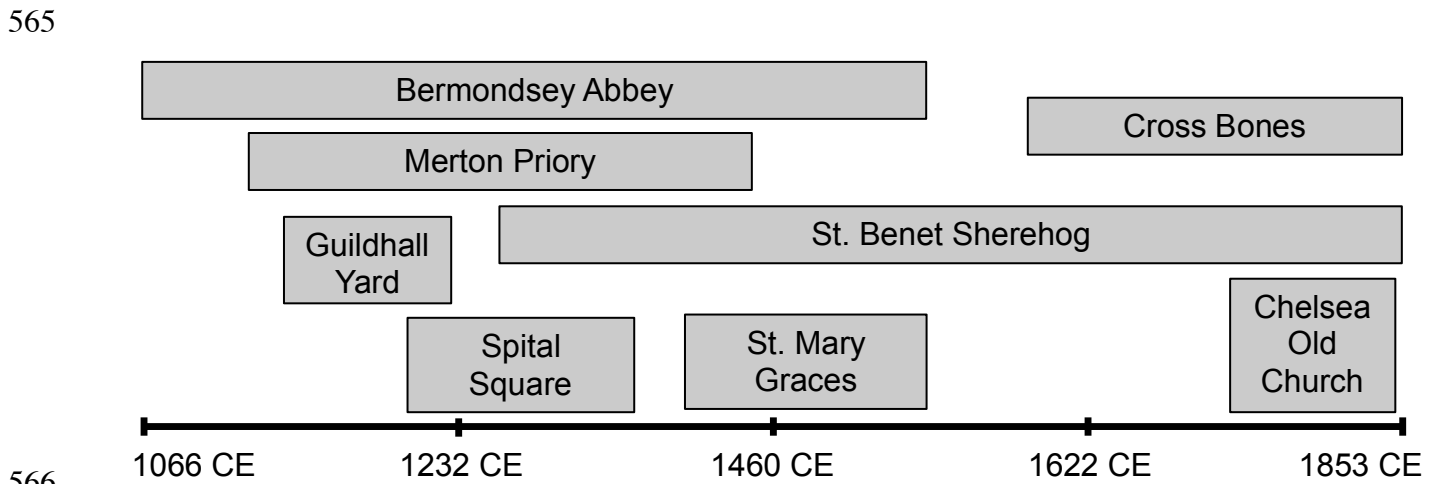
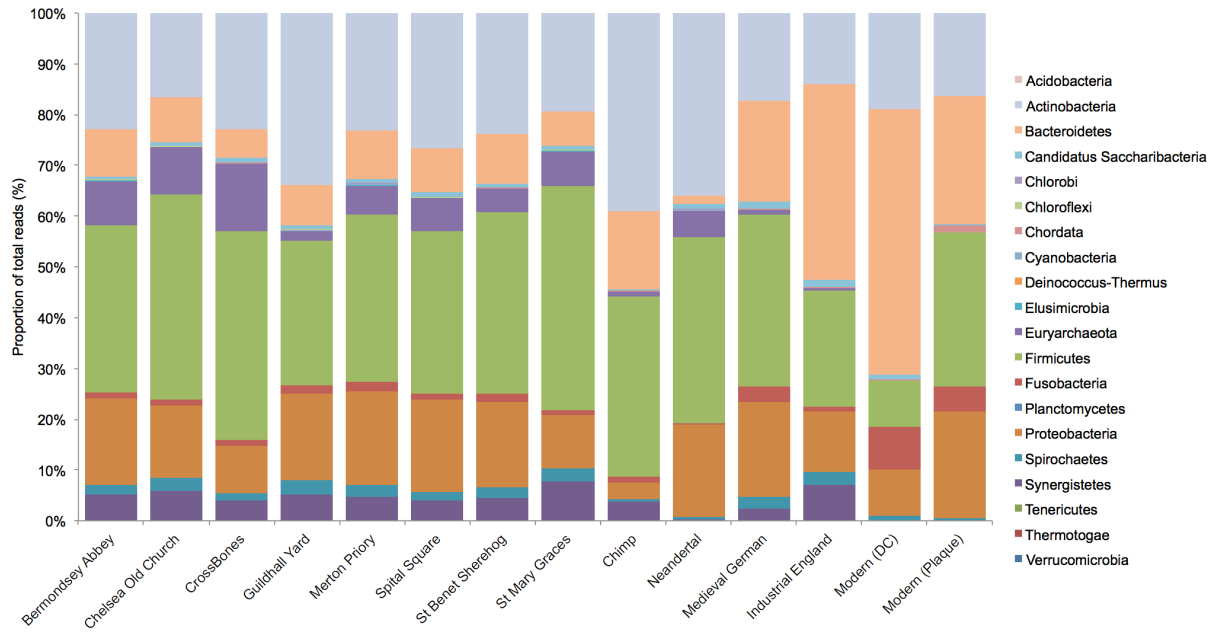
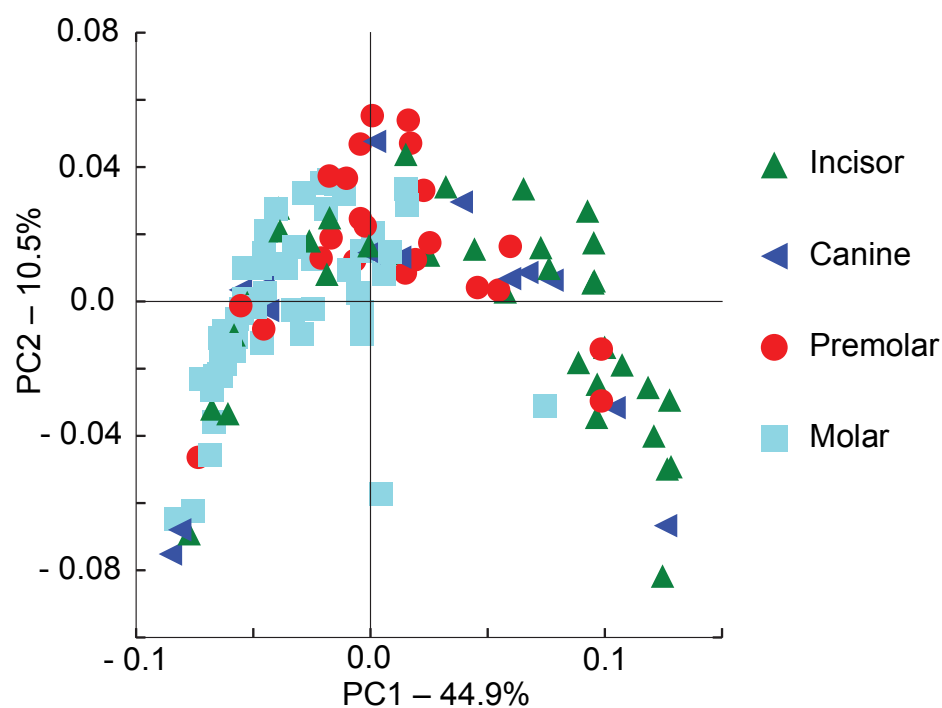


Figure 1: The eight archaeological sites included in this study and the length of time each site was actively used for burials. Samples throughout each of these periods are included in this study.



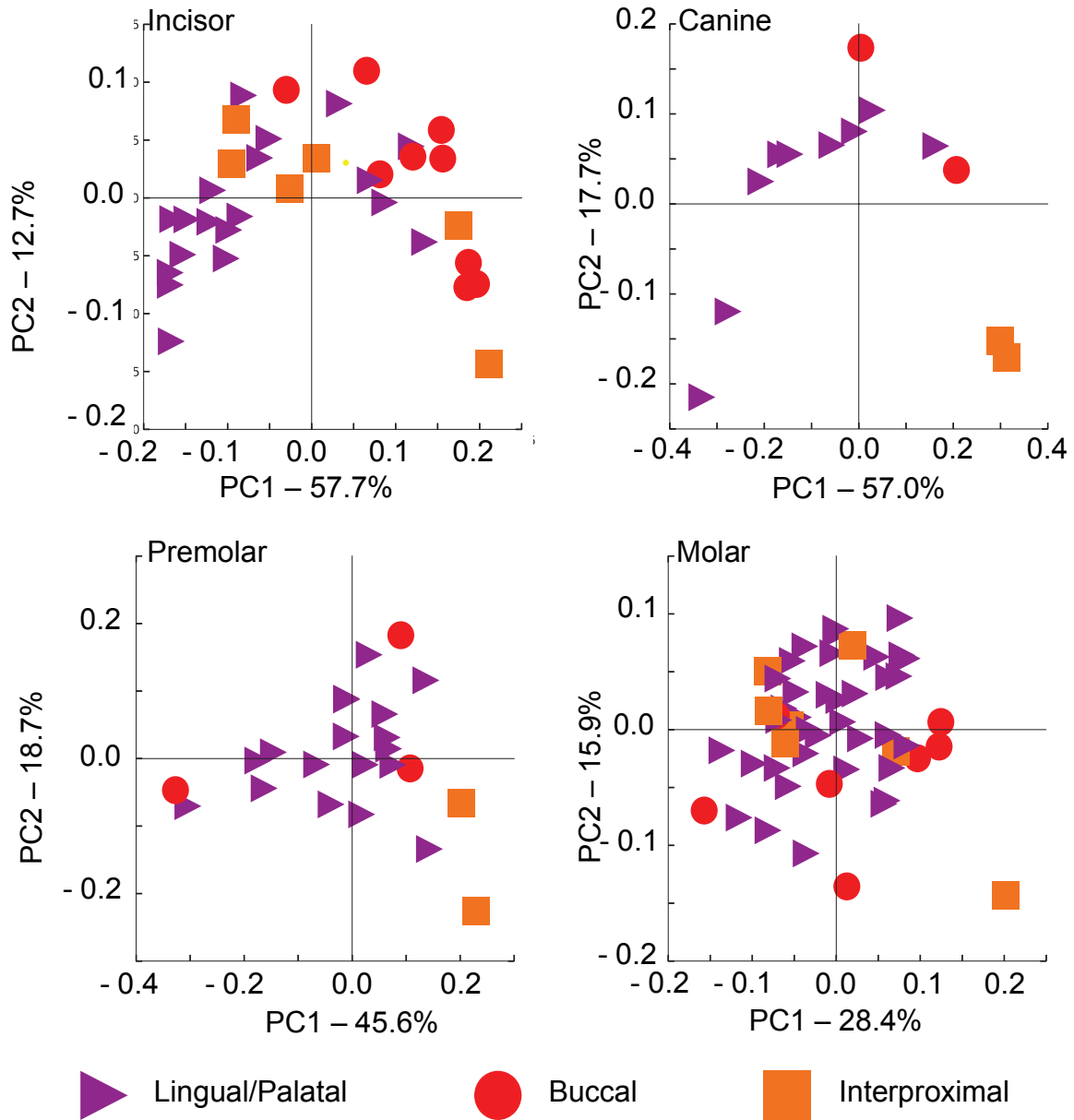
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Figure 2: Stacked bar chart showing Museum of London samples (grouped by archaeological site) and previously analysed samples from Weyrich *et al.* (2017) and Warinner *et al.* (2014) and modern plaque from Belda-Ferre, P. *et al.* (2012)



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576 **Figure 2:** PCoA plot of Bray-Curtis dissimilarity for all dental calculus samples (n =
577 128). Individual samples are labeled by tooth type. The first axis (PC1) indicates the
578 separation of tooth types, notably molar and incisor.



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Figure 3: PCoA plot of Bray-Curtis dissimilarity for all dental calculus samples divided by tooth type. Individual samples are coded by the tooth surface from which the sample was taken.