How, when, and where relic DNA biases estimates of microbial diversity

JT Lennon[1*], ME Muscarella[1,3], SA Placella[2], BK Lehmkuhl[1]

[1]Indiana University, Department of Biology

6 [2]Michigan State University, W.K. Kellogg Biological Station

[3]Current address: University of Illinois, Department of Plant Biology

* corresponding author: lennonj@indiana.edu

Extracellular or "relic" DNA is one of the largest pools of nucleic acids in the biosphere [1,2]. Relic DNA can influence a number of important ecological and evolutionary

processes, but it may also bias estimates of microbial abundance and diversity, which has implications for understanding environmental, engineered, and host-associated ecosystems.

We developed models capturing the fundamental processes that regulate the size and composition of the relic DNA pools to identify scenarios leading to biased estimates of

biodiversity. Our models predict that bias increases with relic DNA pool size, but only when the species abundance distributions (SAD) of relic and intact DNA are distinct from

one another. We evaluated our model predictions by quantifying relic DNA and assessing its contribution to bacterial diversity using 16S rRNA gene sequences collected from

different ecosystem types, including soil, sediment, water, and the mammalian gut. On average, relic DNA made up 33 % of the total bacterial DNA pool, but exceeded 80 % in

some samples. Despite its abundance, relic DNA had no effect on estimates of taxonomic and phylogenetic diversity, even in ecosystems where processes such as the physical

protection of relic DNA are common and predicted by our models to generate bias. Rather, our findings are consistent with the expectation that relic DNA sequences degrade in

proportion to their abundance and therefore may contribute minimally to estimates of microbial diversity.

When microorganisms die, their DNA leaks into the surrounding environment. The fate of this relic DNA has important implications for evolutionary and ecological processes. For

example, relic DNA can be taken up and incorporated into the genomes of some microorganisms via transformation, thereby serving as a reservoir of genetic information that can confer new

traits and fitness benefits to distantly related taxa [3]. In addition, relic DNA is a high-quality

resource containing carbon, nitrogen, and phosphorus that is consumed by a diverse array of

36   bacteria with consequences for microbial community structure and ecosystem processes [1,4].

Relic DNA also has the potential to bias cultivation-independent estimates of diversity,

38   which are widely used for addressing questions concerning the assembly, biogeography, and

functioning of microbial communities. Microbial DNA extracted from environmental and host-

40   associated samples is not solely derived from metabolically active organisms [5]. A large portion

of the individuals in a microbial community is dormant or dead [6,7]. Although nucleic acids can be

42   temporarily retained in non-viable cells, DNA is ultimately released into the environment when

individuals die from autolysis, senescence, viral infection, or predation [8]. Together, these sources

44   of mortality can create large pools of relic DNA [2,9]. For example, there is an estimated 0.45

petagrams of relic DNA in global ocean sediments, which is 70-fold greater than the amount of

46   DNA contained in intact cells from the same environments [1]. If included in cultivation-

independent approaches, relic DNA could distort our understanding of the ecological and

48   evolutionary processes that regulate the distribution, abundance, and function of microbial taxa.

The processes that regulate relic DNA dynamics vary among ecosystems. In well-mixed

50   microbial habitats, the residence time of relic DNA can be short owing to rapid rates of

hydrolysis, oxidation, and UV-mediated damage [2]. For example, in surface waters of freshwater

52   and marine environments, the extracellular DNA pool turns over in less than one day [10,11] while

plasmid DNA begins to degrade in just minutes [12]. As a result, the size distribution of relic DNA

54   is skewed towards small fragments ranging between 100 and 500 base pairs [13,14]. In more

structured microbial habitats, other factors contribute to the size and turnover of the relic DNA

56   pool. For example, binding with inorganic and organic substances in soils and sediments reduces

the rate of relic DNA degradation [9]. Likewise, biofilms, aggregates, and outer membrane vesicles

58   can protect relic DNA from hydrolytic enzymes [8,15,16]. Alternatively, the heterogeneous

distribution of microorganisms in structured habitats creates "hot spots" of metabolic activity[17]

60   that may increase relic DNA turnover. Collectively, these processes may help explain the

accumulation of relic DNA in a range of ecosystems including ocean sediments [1], the built

62   environment[18], and the mammalian gastrointestinal tract [19].

To date, the documented effects of relic DNA on estimates of diversity are idiosyncratic.

64   Even in samples with large amounts of relic DNA, bias can be non-existent or substantial, and

can lead to the overestimation or underestimation of diversity [20,21]. Such observations reflect the

66   need for a more mechanistic understanding of relic DNA dynamics so that microbial

communities from different ecosystems can be consistently and accurately characterized. We

68   address this issue by developing a theoretical framework that considers the processes regulating

the size and turnover of the relic DNA pool. We emphasize that the direction and magnitude of

70   bias is influenced by sampling from a joint species abundance distribution (SAD) consisting of

sequences from the relic and intact DNA pools. We evaluated our models by quantifying the

72   contribution of relic DNA to the abundance and diversity of bacterial communities in ecosystem

types that are known to have contrasting relic DNA dynamics.

RESULTS

76   We developed an interrelated set of models to identify conditions where relic DNA leads

to biased estimates of microbial diversity. We began with a conceptual model representing the

78   dynamics of intact and relic DNA (Fig. 1a). The size and composition of the intact DNA pool

reflects sequences contained in viable bacteria belonging to different species, which are

80   influenced by births, deaths, and immigration. The size and composition of the relic DNA pool is

determined by the input of sequences associated with the death of bacteria from the intact DNA

82   pool and losses associated with the degradation of relic DNA sequences. With this framework

established, we developed a sampling model to explore how diversity estimates are affected

84   when draws come from a joint distribution of sequences derived from the intact and relic pools

(see Methods). We found that a sufficiently large relic DNA pool is required, but not sufficient to

86   create bias in estimates of species richness, the number of taxa in a sample (Fig. 1b). Critically,

there must also be differences between the species abundance distributions (SAD) of the intact

88   and relic DNA pools for bias to arise. When the relic DNA pool has a more even SAD, sampling

from the total pool (intact + relic) leads to overestimation of true richness (Fig. 1b). Conversely,

90   when the relic DNA pool has a less even SAD, sampling from the total pool leads to

underestimation of true richness (Fig. 1b). Last, we developed a process-based model to identify

92   conditions that lead to relic DNA bias. We determined that relic DNA pool size reaches a stable

equilibrium when $R = m\,I\,/\,d$, where R is the size (number of sequences) of the relic DNA pool,

94   m is the per capita mortality rate of bacteria in the intact DNA pool, I is the size (number of

sequences) of the intact DNA pool, and d is the per capita degradation rate of the relic DNA

96   sequences (see Methods). From this, it can be shown that R increases with the residence time

($\tau = R\,/\,d$) of the relic DNA pool. We also used the process-based model to explore different

98   scenarios that influence the SAD of the relic DNA pool. First, we assumed that the rate at which

relic DNA sequences degrade is equivalent among species. Under this neutral expectation, the

100   shapes of relic SAD and intact SAD are nearly identical. No bias arises under these conditions,

regardless of the relic DNA pool size (Fig. 1c). Next, we simulated protection by reducing the

102   degradation rates of relic DNA sequences belonging to randomly selected species (see Methods).

This created a less even relic SAD, resulting in the underestimation of richness in the total DNA

104   pool (Fig. 1c). Finally, we considered the scenario where there are "hot spots" of relic DNA

degradation. When simulating accelerated degradation rates for abundant relic DNA sequences,

106   there was a more even relic SAD that resulted in the overestimation of richness in the total DNA

pool (Fig. 1c).

Variation in relic DNA pool size — We collected samples from different ecosystem types (soil,

110   sediment, water, and mammalian guts) that are known to vary in relic DNA turnover rates [13,22].

Using quantitative PCR with 16S rRNA primers, we quantified intact DNA after treating an

112   aliquot of sample with DNase to remove relic DNA. In addition, we quantified total DNA (intact

+ relic) pool size from a control aliquot of sample that was not treated with DNase (see Methods

114   and Fig. S2). Consistent with previous reports [20,21,23], relic DNA accounted for a substantial but

variable portion of the total bacterial DNA pool. The proportion of relic DNA was normally

116   distributed (mean ± sd: 0.33 ± 0.218) and ranged from 0 to 0.83 across 34 samples obtained from

different ecosystem types (Fig. 2). Even though host and environmental features associated with

118   different habitats are thought to influence relic DNA dynamics, there was only a marginally

significant effect of ecosystem type on the proportion of relic DNA (one-way ANOVA, $F_{3,\,30} =$

120   2.43, $P = 0.084$) [2,20] which likely reflected the contrast between soil and water samples (Tukey's

HSD, adjusted $P = 0.058$). Nevertheless, the large average pool size and range in relic DNA

122   across samples provided us with the opportunity to explore features of our model, specifically

the magnitude and direction of bias that relic DNA should have on estimates of microbial

124   diversity.

126 **Magnitude and direction of bias on microbial diversity** — We sequenced 16S rRNA genes

from the intact and total DNA pools to test whether or not relic DNA biases estimates of within-

128 sample (alpha) bacterial diversity. Despite accounting for a substantial portion of the total DNA

(Fig. 2), relic DNA had no effect on estimates of richness, evenness, or phylogenetic diversity

130 (PD) (Table S1). We expressed each of these alpha-diversity metrics on a per-sample basis as a

ratio (total / intact), where values > 1 represent overestimation bias and values < 1 represent

132 underestimation bias. Relic DNA had no effect on the diversity ratios based on the observation

that the 95 % confidence intervals overlapped with 1.0. Furthermore, the 95 % confidence

134 intervals of the diversity ratios overlapped across ecosystem types, indicating that the

contribution of relic DNA to all measures of diversity was generally low irrespective of the

136 microbial habitat. Finally, using indicator variable multiple regression, we evaluated whether the

magnitude of bias increased with relic DNA pool size, a predicted outcome from some of our

138 simulations (Fig. 1b, c). For all diversity ratios (richness, evenness, and PD), the slopes were not

different from zero ($P \geq 0.65$), the intercepts were not different from 1.0 ($P \geq 0.65$), and the

140 estimates were not affected by ecosystem type ($P \geq 0.26$) (Fig. S5). Moreover, only a very small

amount of variation ($R^2 \leq 0.01$) was explained by relic DNA and ecosystem type in the

142 regression models.

144 **Contribution of relic DNA to community composition** — Based on 16S rRNA gene

sequencing of the intact and total DNA pools, relic DNA had no effect on the compositional

146 (beta) diversity of bacterial communities within or across ecosystem types. First, the intact and

total DNA pools were significantly and very highly correlated with one another when we

148 performed a taxonomically based (Bray-Curtis distances) Mantel test ($P = 0.001$, $r = 0.959$) and

a phylogenetically based (UniFrac distances) Mantel text ($P = 0.001$, $r = 0.996$). Second, we

150   tested for the effect of relic DNA on bacterial community composition for the intact and total

DNA pools using a modified beta-dispersion metric[24]. Specifically, we calculated centroid

152   distance ratios, which directly compared the dispersion between pairs of DNase-treated and

DNase-control subsamples (see Methods and Fig. S7). With this approach, if the centroid

154   distance ratio was > 1, we concluded that relic DNA inflated beta-diversity; if the distance ratio

was < 1, we concluded that relic DNA homogenized beta-diversity. We found that relic DNA

156   had no effect on the centroid distance ratios based on the observation that the 95 % confidence

intervals overlapped with 1.0 (Figs. 4, S8). Furthermore, the 95 % confidence intervals for the

158   centroid distance ratios overlapped across ecosystem types, indicating that the contribution of

relic DNA to taxonomic and phylogenetic beta-diversity was low irrespective of the microbial

160   habitat. Last, we used indicator variable multiple regression to test the prediction that bias in

estimates of community composition would increase with increasing relic DNA pool size (see

162   Fig. S1). The proportion of relic DNA in a sample had no effect on the slopes for centroid

distance ratios regardless of whether they were calculated using taxonomic ($P = 0.63$) or

164   phylogenetic ($P = 0.59$) distance matrices. Moreover, the intercepts for these regression

relationships were not different from 1.0 ($P \geq 0.74$) and the amount of explained variation was

166   low ($R^2 \leq 0.09$) suggesting that the overall effect of relic DNA on beta diversity was negligible

(Fig. S8). The parameters estimates from the regression analyses were not affected by

168   ecosystems type ($P \geq 0.29$).

170   **DISCUSSION**

Rarely, if ever, are biological communities completely censused. As a result, estimates of diversity are often based on incomplete sampling, which introduces uncertainty and potential bias[25]. For example, microbiologists using cultivation-independent approaches commonly estimate the diversity of a community using hundreds of thousands of sequences from samples that contain in excess of a billion individuals. Diversity estimation of microbial communities is further complicated by the accumulation and persistence of relic DNA, a pool of sequences that may not accurately reflect the composition of viable microorganisms. We developed a set of models that capture the fundamental processes regulating relic DNA dynamics along with the effects of sampling from a joint species abundance distribution (SAD) that contain sequences from both the relic and intact DNA pools. Our models reveal that two criteria are required in order for relic DNA bias to emerge. First, the relic DNA pool must be large enough so that it diminishes the probability of sampling sequences from the intact DNA pool. Second, the relic SAD must be distinct from the intact SAD. If these conditions are not satisfied, relic DNA has minimal effect on estimates of biodiversity. We then explored the model's expectations by analyzing the intact and total (relic + intact) DNA in ecosystems where relic DNA dynamics are known to vary. Despite making up a large portion of the total DNA, we found that relic DNA had minimal to no effect on estimates taxonomic and phylogenetic diversity, consistent with model scenarios where relic DNA degradation rates are equivalent among species.

Despite sampling a range of habitats with attributes that affect relic DNA turnover, our data suggest that estimates of taxonomic and phylogenetic diversity were unbiased by relic DNA (Figs. 2 and 3) even when it accounted for > 80 % of the total DNA (Figs. S5 and S8). Similar findings have been reported elsewhere. From a survey of North American soils, no bias was detected for taxonomic richness in 30 % of bacterial samples (7 / 31) and 55 % of fungal samples

194   (17 / 31), even when relic DNA made up a substantial portion of the total DNA pool [21].

Likewise, more than 90 % of the bacterial DNA recovered from porcine lung tissues was

196   sensitive to DNase, yet diversity in the enzymatically treated and untreated samples was the

same indicating that large relic DNA pools do not necessarily obscure estimates of microbial

198   diversity [20]. Such findings are supported by our modeling results: relic DNA has no effect on

estimates of microbial diversity when the SADs of the relic and intact DNA pools are equivalent

200   (Fig. 1b). When this assumption is met, the abundance distribution of relic and intact DNA pools

should be identical. Therefore, sampling from the relic DNA pool should not bias estimates of

202   diversity in the intact DNA pool, even if relic DNA makes up a large portion of the total DNA

pool (Fig. 1b).

204   Our simulations identify other ecological scenarios where relic DNA bias may arise. It

has been hypothesized that relic DNA can be protected from degradation inside of aggregates,

206   biofilms, or other complexes that reduce contact between nucleic acids and extracellular

enzymes. Importantly, protection can only increase bias if it alters relic DNA degradation in a

208   species-specific manner. While most studies have emphasized the potential for protection to

overestimate diversity, our model predicts the opposite (Fig. 1b). Protection created a less even

210   relic SAD that resulted in the increased dominance of relic sequences associated with a few

species. Assuming a large relic DNA pool, preferential sampling of these protected sequences

212   will lead to the underestimation or richness, which has been observed in some instances [20]. In

contrast, we identified at least one ecological scenario where relic DNA creates overestimation

214   bias. Microscale heterogeneity in structured habitats can lead to non-uniform distributions of

microorganisms and their metabolic activities. Our simulations indicate that the resulting "hot

216   spots" can degrade relic DNA sequences in a density-dependent matter resulting in a more even

relic SAD. Although not significant, sediment samples trended towards positive bias for richness

218 and PD (Fig. 3), which could potentially reflect "hot spots" of relic DNA degradation. Inflation

bias can also arise when species in the regional pool disperse into habitats for which they are

220 poorly adapted. When these immigrants die, they can enrich the relic DNA pool with sequences

that are dissimilar to sequences found in the local community. A similar effect could arise when

222 dead bacteria are transported across ecosystem boundaries, a phenomenon that occurs, for

example, when marine snow is exported from surface waters to marine sediments [26]. More

224 generally, bias may arise under non-equilibrium conditions where community turnover of the

intact DNA pool is shorter than turnover of the relic DNA pool. Any abiotic or biotic

226 perturbation that removes a substantial amount of living biomass could result in transient

divergence in the composition of sequences in the intact and relic DNA pool. For example,

228 virulent phage can dramatically reduce the abundance of bacterial prey and in the process release

a large pulse of DNA into the environment, which could bias estimates of microbial diversity [27].

230 Last, although not fully explored here, shifts in rank abundance of taxa that are independent of

parameters describing the SAD could also result in biased estimates of microbial diversity. In

232 sum, there are ways to deviate from neutral expectations about the degradation of relic DNA,

which should lead to biased estimates of microbial diversity, but sampling from a range of

234 ecosystems suggests that such conditions are not prevalent in nature.

Our understanding of the microbial biosphere has been transformed by the development

236 and application of molecular-based cultivation-independent techniques. The ability to rapidly

obtain millions of gene sequences and transcripts from a range of environments has yielded

238 valuable insight into the processes that regulate community assembly and function [28,29], and has

also paved the way for the discovery of new metabolisms [30], tests for unifying patterns of

240 biodiversity [31], and an updated tree of life [32]. There are limitations, however, associated with

culture-independent techniques, which include inefficient nucleic-acid extraction methods [32] and

242 "universal" primers that over-represent some taxonomic groups while overlooking others [33].

Sequencing of relic DNA is another important concern, which can potentially lead to the

244 overestimation or underestimation of microbial diversity. However, this bias requires the

decoupling of processes that regulate the compositional turnover of the relic and intact DNA

246 pools. While some recent evidence suggests this can arise [20], our findings suggest that at least in

some ecosystems, relic DNA appears to contribute minimally to the characterization of microbial

248 community structure.

250   **METHODS**

252 **Sampling Model —** We used a set of sampling-based simulations to explore the effects of

mixing intact and relic DNA on estimates of diversity. For each simulation, we defined a

254 regional species pool consisting of 10,000 taxa with a lognormal abundance distribution [34]. The

intact community consisted of 1,000,000 individuals sampled from this regional pool. We then

256 combined this intact community with a relic community at proportions ranging from 0.01 to

0.96. We altered the evenness of the regional pool from which the relic community was sampled

258 by changing the scale parameter of the lognormal distribution. To decrease the evenness of the

relic DNA pool, we increased the scale parameter from 0.98 to 1.8. To increase the evenness of

260 the relic DNA pool, we decreased the parameter from 0.98 to 0.25. After mixing the intact and

relic communities at the defined proportions, we rarified the total community to 10,000

262 observations and calculated richness and Bray-Curtis distances to estimate compositional

differences between the intact and total DNA pools. To estimate contribution of relic DNA to

264 diversity, we calculate ratios (total DNA / intact DNA) for richness and Bray-Curtis distances.

All simulations and estimations were performed in the R statistic environment (v 3.3.2)[35] using

266 the `vegan` package as well as custom functions.

268 **Process-Based Model** — We developed a set of stochastic simulations to explore how the

processes regulating relic DNA dynamics can give rise to bias in the estimation of microbial

270 diversity. For each simulation, we defined a regional species pool consisting of 4,000 taxa with a

lognormal abundance distribution [34]. We initiated an intact community by randomly sampling

272 20,000 individuals from the regional species pool. Subsequent dynamics were controlled by the

processes depicted in our conceptual model (Fig. 1a). Specifically, we simulated immigration by

274 sampling j individuals from the regional species pool and adding them to the intact community.

We simulated birth by randomly selecting $I \times r$ individuals from the intact community and

276 adding them to the intact community again, where I is the size of the intact community and r is

the per capita birth rate. We simulated death by selecting $(I \times m) + j$ individuals from the intact

278 community and moving them to the relic community (R), where m is the per capita mortality

rate. We simulated degradation by selecting $(R \times d) + j$ individuals from the relic community and

280 removing them, where d is the per capita decay rate of relic DNA.

We simulated three scenarios with the process-based model. First, we considered a

282 neutral scenario. At each time step, we simulated degradation by randomly selecting relic DNA

sequences and removing them. Second, we considered the scenario where some relic DNA

284 sequences have lower rates of degradation owing to protection. Each species was randomly

assigned a degradation susceptibility probability from a beta distribution. The beta distribution is

286 a continuous probability distribution bound by 0 and 1 with two shape parameters. We chose

shape parameters ($\alpha = 0.7$, $\beta = 0.7$) to provide a wide range in susceptibility probability but also

288   have increased probabilities of low (near 0) and high (near 1) susceptibility, which created a

wide U-shaped distribution. Susceptibility probabilities were randomly assigned to species for

290   each model iteration and were used to weight the probability of selecting relic DNA sequences

for degradation at each time step. Last, we considered the scenario where there are "hot spots" of

292   relic DNA degradation, reflecting the clumped distribution of microbial taxa and their associated

metabolic activities. At each time step, we calculated density-dependent selection probabilities

294   by weighting the probability of selecting an individual from the relic community by its species

abundance. Therefore, relic DNA sequencing belonging to more abundant species would be

296   more likely to be selected for degradation. For each of the three scenarios, we ran each

simulation for 10,000 time-steps. We used a constant immigration rate ($j = 2,000$) to maintain

298   intact community diversity, and used equal and constant birth and mortality rates (b and $m = 0.1$)

to maintain intact community size. Likewise, we accounted for immigration during death and

300   degradation to prevent uncontrolled growth. For each set of simulations, we used a range of

decay rates (d), which yielded relic DNA proportions between 0.05 and 0.95 according to the

302   following equation: $d = m/p - m$, where p is the target proportion and m is the fixed mortality

rate (see Supplement for derivation). At the end of each simulation, we determined the effect of

304   relic DNA on diversity estimates by comparing the total community (intact + relic) with the

intact community. All simulations were performed in the R statistic environment.

Sample collection and DNA pools — We collected samples from a diverse set of environmental

308   and host-associated ecosystems. First, we sampled sediments and surface water from lakes near

the Michigan State University, W.K. Kellogg Biological Station (KBS) in Hickory Corners,

Michigan, USA. Soils were sampled from the main sites and surrounding areas at the KBS Long-Term Ecological Research site [36]. We also collected fresh feces as representative gut samples

from cows, dogs, horses, rabbits, and humans. In each of these ecosystem types, we obtained samples from 6 - 8 independent sites. In the laboratory, we applied a DNase treatment to each

sample to remove relic DNA. The procedure is based on methods that have previously been used to quantify relic DNA in marine sediments [37,38], host tissue [37,38], and drinking-water biofilms

[37,38]. The DNase-based method has proven to be effective not only for removing extracellular DNA, but also DNA that is contained inside of dead cells [37,38] while not compromising the

integrity of living cells [37,38]. We determined that the DNase method is efficient (98%) at removing relic DNA that was experimentally added to an environmental sample (see

Supplementary Methods; Fig. S4). Thus, we refer to the DNA remaining in a subsample following enzymatic treatment as "intact DNA" and assume it is derived from viable cells. "Total

DNA" refers to the DNA in subsample that was not treated with DNase (negative control), which reflects the sum of intact DNA and relic DNA.

For aquatic samples, we filtered 250 mL lake water through a 47 mm 0.2 μm Supor PES membrane filter using 10 mm Hg vacuum. We cut the filter in half and randomly assigned one

half of the filter to a DNase treatment and used the other half as the control. We then rolled up each piece of filter using sterile forceps and inserted it into a 2 mL centrifuge tube containing 1.5

mL of pH 7.3 phosphate buffered saline (PBS). After vortexing at room temperature for 5 min, we removed the filter from the tube and centrifuged for 5 min at 10,000 x g. We then discarded

the supernatant and resuspended the pellet in 375 μL of PBS, which was transferred to a 2 mL centrifuge tube. For non-aqueous samples (soil, sediments, and feces), we directly added 0.25 g

of material to a 2 mL centrifuge tube. At this stage in the procedure, aquatic and non-aquatic

samples were identically processed. For each sample in its 2 mL centrifuge tube, we added

334 DNase digestion buffer, which consisted of 382.5 μL of nanopure water, 5 μL of 1 M MgCl$_2$, 2.5

μL of bovine serum albumin (10 mg/mL), and 50 μL of 0.5 M Tris HCl (pH 7.5). For

336 subsamples treated with DNAse, we added 40 μL of a 10 U/μL stock of DNase I (Roche #

04536282001) and 20 μL of nanopure water resulting in a 500 μL final working volume with an

338 80U/μL DNase concentration. (See Supplementary Material and Fig. S3 for description of

experiment to determine minimum DNase concentration needed for relic DNA removal.) For

340 untreated subsamples, we substituted 40 μL of nanopure water for DNase solution. For each

sample, we measured pH using a micro pH probe (Orion 9110DJWP, Thermo Scientific) and

342 adjusted to 7.3 - 7.7, which is in the optimum range for DNase. We then incubated the samples

horizontally on a shaker table at 37 ºC for 60 min. Following this, we transferred each sample to

344 a 15 mL Falcon tube containing 1 mL of 1X hexadecyltrimethylammonium bromide (CTAB)

buffer, which consisted of one part solution 1 (1 g CTAB + 0.58 g NaCl in 10 mL of nanopoure

346 water) and one part solution 2 (0.58 g NaCl + 0.82 K$_2$HPO$_4$ + 0.04 KH$_2$PO$_4$ in 10 mL of

nanopoure water). We then added 25 μL of 0.5 M EDTA to each tube and vortexed. We stopped

348 the DNase reaction by incubating the tubes at 75 ºC for 10 min. We began DNA extraction by

adding 1 mL of phenol:chloroform:isoamylalcohol (25:24:1) to the 2 mL tube, which was then

350 vortexed for 10 min. This was followed by centrifuging at 7,100 x g for 10 min. We transferred

the top aqueous layer to a 15 mL Falcon tube and combined it with an equal volume of

352 chloroform:isoamylalcohol (24:1) and then vortexed. After centrifuging the Falcon tubes at

7,100 x g for 5 min, we transferred 400 μL of the top aqueous layer to a new 2 mL centrifuge

354 tube. For clean up, we used the MO BIO PowerLyzer PowerSoil DNA Isolation Kit starting from

Step 9. At Step 21 of the DNA Isolation Kit, we eluted in 50 μL of Solution C6 and centrifuged

356 for 1 min at 10,000 x *g*. We stored extracted DNA at -20 °C for short-term storage or at -80 °C

for long-term storage.

Contribution of relic DNA to bacterial abundance — We used 16S rRNA gene copy numbers

360 generated from quantitative PCR (qPCR) assays to estimate the proportion of relic DNA in a

sample as 1 - (intact DNA / total DNA). Briefly, the previously described qPCR assays[39]

362 consisted of 30 μL reactions containing 1μL of DNA template, 0.5 μL of each primer (10

μmol/L), 14.5 μL of nuclease-free H_2O, and 13.5 μL of 5 Prime 2.5x RealMasterMix SYBR

364 ROX. We amplified a 200 base-pair fragment of the 16S rRNA gene with Eub 338 (forward)

and Eub518 (reverse) primers [40]. PCR assays were performed with an Eppendorf Mastercycler

366 Realplex[2] system using previously reported thermal cycle conditions [40]. We generated qPCR

standards from bacterial genomic DNA (*Micrococcus* sp.) using the TOPO TA Cloning Kit

368 (Invitrogen). We extracted plasmids from transformed cells [41], and used the M13 forward and

reverse primers to generate PCR products. The PCR products were quantified and used to

370 generate a standard curve capturing a range of $10^2 - 10^7$ gene copies/μL. We used a noiseband

threshold for quantification. The coefficients of determination (r^2) for our assays ranged from

372 0.96 and 0.99, while amplification efficiencies fell between 0.93 and 0.99. Based on melting

curve analyses, we found no evidence for primer dimers. All unknown samples, no template

374 controls, and standards were run in triplicate on every plate. The mean coefficient of variation

(standard deviation / mean) for our 16S rRNA qPCR assay was 0.16.

Contribution of relic DNA to bacterial diversity

378 *Community sequencing* — We estimated the contribution of relic DNA to bacterial diversity

using high-throughput sequencing of the 16S rRNA gene. Specifically, we amplified the V4

380   region of the 16S rRNA gene from the intact and total DNA pools of each sample using

barcoded primers (515F and 806R) designed to work with the Illumina MiSeq platform [42]. We

382   cleaned the sequence libraries using the AMPure XP purification kit, quantified the resulting

products using the QuantIt PicoGreen kit (Invitrogen), and pooled libraries at equal molar ratios

384   (final concentration: 20 ng per library). We then sequenced the pooled libraries with the Illumina

MiSeq platform using paired end reads (Illumina Reagent Kit v2, 500 reaction kit) at the Indiana

386   University Center for Genomics and Bioinformatics Sequencing Facility. Paired-end raw 16S

rRNA sequence reads were assembled into contigs using the Needleman algorithm [43]. We

388   obtained a total of 12,916,632 16S rRNA sequences. After quality trimming with a moving

average quality score (window 50 bp, minimum quality score 35), we aligned the sequences to

390   the Silva Database (version 123) using the Needleman algorithm. Chimeric sequences were

removed using the UCHIME algorithm [44]. After this filtering, there was an average (± SEM) of

392   222,701 ± 9,560 sequences per site. We created operational taxonomic units (OTUs) by first

splitting the sequences based on taxonomic class (using the RDP taxonomy) and then binning

394   sequences into OTUs based on 97% sequence similarity. Our depth of sequencing led to a high

degree of coverage across samples (minimum Good's Coverage = 0.98). For phylogenetic

396   analysis, we picked representative sequences for each OTU by using the most abundant unique

sequence. We used FastTree [45] to generate a phylogenetic tree from the representative sequences

398   using the generalized time-reversible model of nucleotide evolution. We calculated phylogenetic

distances using weighted UniFrac distances [46]. All initial sequence processing was completed

400   using the software package mothur (version 1.38.1) [47].

402   *Alpha Diversity* — We estimated the effects of relic DNA on richness, evenness, and

phylogenetic diversity for the intact and total DNA pools with a sample. To estimate the number

404 of OTUs (richness), we used a resampling approach that subsampled each sample to an equal

   number of sequences per sample and summed the number of OTUs that were represented [48].

406 Briefly, we subsampled to 30,000 observations, resampled 999 additional times, and then

   calculated the average richness estimates (± SEM) for each sample. To estimate the equitability

408 in abundance among taxa in a sample (evenness), we used the same resampling approach and

   calculated average evenness estimates (± SEM) using Simpson's Evenness index [49]. To test

410 whether relic DNA affected the phylogenetic diversity within a sample, we subsampled

   communities to 30,000 observations and then calculated Faith's D statistic, which sums the

412 branch lengths for each species found in a sample from the root to the tip of the phylogenetic tree

   [50]. All estimations were performed in the R statistic environment (v 3.3.2)[35] using the vegan, ape,

414 ade4, picante, and plyr packages, along with custom functions.

416 *Beta Diversity* — We estimated the effects of relic DNA on between-sample (beta) diversity by

   comparing the taxonomic and phylogenetic diversity of bacterial communities in the intact and

418 total DNA pools. First, we conducted a Principal Coordinates Analysis (PCoA) on $\log_{10}$-

   transformed relative abundance data to visualize the effects of relic DNA removal (via DNase

420 treatment) on bacterial community composition within and among ecosystem types. The PCoA

   was performed with Bray-Curtis and UniFrac distances to assess taxonomic and phylogenetic

422 effects, respectively. In addition, we used PERMANOVA to test for differences in taxonomic

   and phylogenetic composition based on ecosystem type for the total DNA pool. Second, we

424 conducted a Mantel test to assess the correlation between the community resemblance matrices

   (either Bray-Curtis or UniFrac) represented by the intact and total DNA pools. Last, we tested

426 whether relic DNA altered beta-diversity within an ecosystem type by comparing centroid

   distances. To calculate this metric of sample dispersion, we determined the centroid from a

428 PCoA with either Bray-Curtis or UniFrac distances for the total DNA pool for all sites within a

given ecosystem type. We then measured the Euclidean distances between the centroid and all

430 samples (total and intact) to determine the centroid distances (see Fig. S7).

432 **ACKNOWLEDGMENTS**

We acknowledge constructive feedback from KD Webster, RZ Moger-Reischer, EK Hall, NI

434 Wisnoski, and WR Shoemaker. We thank S. Kuenzel and B.K. Lehmkuhl for technical support.

This work was supported by the USDA National Institute of Food and Agriculture Grant 2011-

436 67019-3022 (JTL), the National Science Foundation Dimensions of Biodiversity Grant 1442246

(JTL), and US Army Research Office Grant W911NF-14-1-0411 (JTL). All code and data used

438 in this study can be found in a public GitHub repository

(https://www.github.com/LennonLab/relicDNA) and the NCBI SRA.

442 **REFERENCES**

1 Dell'Anno, A. & Danovaro, R. Extracellular DNA plays a key role in deep-sea ecosystem
444 functioning. *Science* **309**, 2179-2179 (2005).

2 Torti, A., Lever, M. A. & Jorgensen, B. B. Origin, dynamics, and implications of
446 extracellular DNA pools in marine sediments. *Mar. Genomics* **24**, 185-196 (2015).

3 Redfield, R. J. Genes for breakfast - the have-your-cake-and-eat-it-too of bacterial
448 transformation. *J. Hered.* **84**, 400-404 (1993).

4 Lennon, J. T. Diversity and metabolism of marine bacteria cultivated on dissolved DNA.
450 *Appl. Environ. Microbiol.* **73**, 2799-2805 (2007).

5 Klein, D. A. Microbial communities in nature: a postgenomic perspective. *Microbe* **2**,
452 591-595 (2007).

6 Luna, G. M., Manini, E. & Danovaro, R. Large fraction of dead and inactive bacteria in
454 coastal marine sediments: Comparison of protocols for determination and ecological
significance. *Appl. Environ. Microbiol.* **68**, 3509-3513 (2002).

456 7 Lennon, J. T. & Jones, S. E. Microbial seed banks: the ecological and evolutionary
implications of dormancy. *Nat. Rev. Microbiol.* **9**, 119-130 (2011).

458 8 Vorkapic, D., Pressler, K. & Schild, S. Multifaceted roles of extracellular DNA in
bacterial physiology. *Curr. Genet.* **62**, 71-79 (2016).

460 9 Levy-Booth, D. J. *et al.* Cycling of extracellular DNA in the soil environment. *Soil Biol.
Biochem.* **39**, 2977-2991 (2007).

462 10 Paul, J. H., Jeffrey, W. H., David, A. W., Deflaun, M. F. & Cazares, L. H. Turnover of
extracellular DNA in eutrophic and oligotrophic freshwater environments of southwest
464 Florida. *Appl. Environ. Microbiol.* **55**, 1823-1828 (1989).

466   11   Dell'Anno, A. & Corinaldesi, C. Degradation and turnover of extracellular DNA in marine sediments: Ecological and methodological considerations. *Appl. Environ.*

468   *Microbiol.* **70**, 4384-4386 (2004).

      12   Deflaun, M. F. & Paul, J. H. Detection of exogenous gene-sequences in dissolved DNA

470   from aquatic environments. *Microb. Ecol.* **18**, 21-28 (1989).

      13   Deflaun, M. F., Paul, J. H. & Jeffrey, W. H. Distribution and molecular-weight of

472   dissolved DNA in subtropical estuarine and oceanic environments. *Mar. Ecol. Prog. Ser.*

          38, 65-73 (1987).

474   14   Beebee, T. J. C. Identification and analysis of nucleci acids in natural freshwaters. *Sci.*

          Total. Env. **135**, 123-129 (1993).

476   15   Romanowski, G., Lorenz, M. G. & Wackernagel, W. Adsorption of plasmid DNA to mineral surfaces and protection against DNase-1. *Appl. Environ. Microbiol.* **57**, 1057-

478   1061 (1991).

      16   Schwechheimer, C. & Kuehn, M. J. Outer-membrane vesicles from Gram-negative

480   bacteria: biogenesis and functions. *Nat. Rev. Microbiol.* **13**, 605-619 (2015).

      17   DeAngelis, K. M., Lindow, S. E. & Firestone, M. K. Bacterial quorum sensing and

482   nitrogen cycling in rhizosphere soil. *FEMS Microbiol. Ecol.* **66**, 197-207 (2008).

      18   Emerson, J. B. *et al.* Schrodinger's microbes: Tools for distinguishing the living from the

484   dead in microbial ecosystems. *Microbiome* **5**, 86 (2017).

      19   Sun J, Wang P. & Shi, Y. High fat diet decrease diversity of extracellular DNA of mice

486   gut microbiota. *FASEB J* **30**, ib204 (2016).

Pezzulo, A. A. *et al.* Abundant DNase I-Sensitive Bacterial DNA in Healthy Porcine Lungs and Its Implications for the Lung Microbiome. *Appl. Environ. Microbiol.* **79**, 5936-5941 (2013).

Carini, P. *et al.* Relic DNA is abundant in soil and obscures estimates of soil microbial diversity. *Nat. Microbiol.* **2**, 16242 (2016).

Morrissey, E. M. *et al.* Dynamics of extracellular DNA decomposition and bacterial community composition in soil. *Soil Biol. Biochem.* **86**, 42-49 (2015).

Corinaldesi, C., Danovaro, R. & Dell'Anno, A. Simultaneous recovery of extracellular and intracellular DNA suitable for molecular studies from marine sediments. *Appl. Environ. Microbiol.* **71**, 46-50 (2005).

Anderson, M. J., Ellingsen, K. E. & McArdle, B. H. Multivariate dispersion as a measure of beta diversity. *Ecol. Lett.* **9**, 683-693 (2006).

Magurran, A. E. *Measuring Biological Diversity*. (Wiley-Blackwell, 2004).

Azam, F. & Long, R. A. Oceanography - Sea snow microcosms. *Nature* **414**, 497-498 (2001).

Corinaldesi, C., Tangherlini, M., Luna, G. M. & Dell'Anno, A. Extracellular DNA can preserve the genetic signatures of present and past viral infection events in deep hypersaline anoxic basins. *Proc. R. Soc. Lond. Biol.* **281** (2014).

Lindstrom, E. S. & Langenheder, S. Local and regional factors influencing bacterial community assembly. *Environ. Microbiol. Rep.* **4**, 1-9 (2012).

Aanderud, Z. T., Jones, S. E., Fierer, N. & Lennon, J. T. Resuscitation of the rare boisphere contributes to pulses of ecosystem activity. *Front. Microbiol.* **6** (2015).

Beja, O. *et al.* Bacterial rhodopsin: Evidence for a new type of phototrophy in the sea. *Science* **289**, 1902-1906 (2000).

Locey, K. J. & Lennon, J. T. Scaling laws predict global microbial diversity. *Proc. Natl. Acad. Sci. U.S.A.* **113**, 5970-5975 (2016).

Feinstein, L. M., Sul, W. J. & Blackwood, C. B. Assessment of bias associated with incomplete extraction of microbial DNA from soil. *Appl. Environ. Microbiol.* **75**, 5428-5433 (2009).

Klindworth, A. *et al.* Evaluation of general 16S ribosomal RNA gene PCR primers for classical and next-generation sequencing-based diversity studies. *Nucleic Acids Res.* **41**, e1 (2013).

Shoemaker, W. R., Locey, K. J. & Lennon, J. T. A macroecological theory of microbial biodiversity. *Nat Ecol Evol* **1,** 107 (2017).

R Core Development Team R: A language and environment for statistical computing, reference index version 2.8.1. R Foundationfor Statistical Computing, Vienna, Austria. ISBN 3-900051-07-0, URL http://R-project.org. (2009).

Hamilton, S. K., Doll, J. E. & Robertson, G. P. *The ecology of agricultural landscapes: Long-term research on the path to sustainability*. (Oxford University Press, 2015).

Dell'Anno, A., Bompadre, S. & Danovaro, R. Quantification, base composition, and fate of extracellular DNA in marine sediments. *Limnol. Oceanogr.* **47**, 899-905 (2002).

Villarreal, J. V., Jungfer, C., Obst, U. & Schwartz, T. DNase I and Proteinase K eliminate DNA from injured or dead bacteria but not from living bacteria in microbial reference systems and natural drinking water biofilms for subsequent molecular biology analyses. *J. Microbiol. Methods* **94**, 161-169 (2013).

532   39   Lau, J. A. & Lennon, J. T. Evolutionary ecology of plant-microbe interactions: soil

          microbial structure alters selection on plant traits. *New Phytol.* **192**, 215-224 (2011).

534   40   Fierer, N., Jackson, J. A., Vilgalys, R. & Jackson, R. B. Assessment of soil microbial

          community structure by use of taxon-specific quantitative PCR assays. *Appl. Environ.*

536   *Microbiol.* **71**, 4117-4120 (2005).

      41   Sambrook, J. & Russell, D. W. *Molecular cloning: a laboratory manual.* (Cold Spring

538   Harbor Laboratory Press,, 2001).

      42   Caporaso, J. G. *et al.* Ultra-high-throughput microbial community analysis on the

540   Illumina HiSeq and MiSeq platforms. *ISME J.* **6**, 1621-1624 (2012).

      43   Needleman S.B. & C.D., W. A general method applicable to the search for similarities in

542   the amino acid sequence of two proteins. *J. Mol. Biol.* **48**, 443-453 (1970).

      44   Edgar, R. C., Haas, B. J., Clemente, J. C., Quince, C. & Knight, R. UCHIME improves

544   sensitivity and speed of chimera detection. *Bioinformatics* **27**, 2194-2200 (2011).

      45   Price, M. N., Dehal, P. S. & Arkin, A. P. FastTree 2-Approximately Maximum-

546   Likelihood Trees for Large Alignments. *PLOS ONE* **5** (2010).

      46   Lozupone, C. A. & Knight, R. Species divergence and the measurement of microbial

548   diversity. *FEMS Microbiol. Rev* **32**, 557-578 (2008).

      47   Schloss, P. D. *et al.* I Introducing mothur: open-source, platform-independent,

550   community-supported software for describing and comparing microbial communities.

          Appl. Environ. Microbiol. **75**, 7537-7541 (2009).

552   48   Muscarella, M. E., Jones, S. E. & Lennon, J. T. Species sorting along a subsidy gradient

          alters community stability. *Ecology* (2016).

554   49   Smith, B. & Wilson, J. B. A consumer's guide to evenness indices. *Oikos* **76**, 70-82

(1996).

556   50   Faith, D. P. onservation evaluation and phylogenetic diersity. *Biol. Conserv.* **61**, 1-10,

(1992).

 FIGURE CAPTIONS

562 **Figure 1 | Modeling relic DNA dynamics. a** The amount of relic DNA in a microbial

environment is determined by inputs associated with the mortality of viable individuals with

564 intact DNA and by losses associated with the degradation of relic DNA. If the diversity of

sequences contained in the relic DNA pool is sufficiently different from that in the intact DNA

566 pool, then relic DNA may bias estimates of microbial biodiversity (as indicated by different

colored boxes) when sampling from the total (intact + relic) DNA pool. **b** We developed a

568 sampling-based simulation model to explore the effects of mixing intact and relic DNA on

estimates of diversity. We populated intact and relic communities with individuals from a

570 lognormal species abundance distribution (SAD). We altered the diversity of the relic

community by changing the scale parameter of the lognormal distribution describing the SAD.

572 We then sampled and mixed the intact and relic communities so that the relic contribution to

total community ranged from 0.01 to 0.96.**c** To gain mechanistic insight into how bias arises, we

574 created a stochastic process-based model that captures features that influence relic DNA

dynamics, including: immigration, birth, death, degradation (**a**). We simulated a range of

576 degradation rates to achieve relic DNA pool sizes with proportions ranging between 0.05 and

0.95. To explore how degradation alters the SAD of the relic community, we explored three

578 scenarios. First, we simulated a neutral scenario where relic DNA sequences produced by

different species degrade at the same rate. Second, we simulated conditions where physical,

580 chemical, or biological processes slow the degradation rate of relic DNA belonging to some

species via protection. Third, we simulated "hot spots" where there more abundant relic DNA

582 sequences experience faster rates of relic DNA degradation, a condition that may arise in

structured habitats where there are patchy distributions of individuals and their metabolic

584   products (i.e., enzymes). We ran simulations for 10,000 time steps and then sampled the intact

and relic communities. To quantify bias in diversity (**b**, **c**), we calculated "richness ratios" which

586   reflect the number of species in the total DNA pool (intact + relic) divided by the number of

species in the intact DNA pool in a simulation. When values richness ratios = 1, relic DNA has

588   no effect on estimates of diversity; when richness ratios > 1, relic DNA overestimates true

diversity; and when richness ratios < 1, relic DNA underestimates true diversity.

Figure 2 | Proportion of bacterial relic DNA in different ecosystem types. We quantified the

592   amount of intact DNA in a sample after removing relic DNA with a DNase treatment. We then

estimated the proportion of relic DNA as 1 - (intact DNA / total DNA), where the total DNA

594   concentration was quantified without DNase treatment. Relic DNA constituted an appreciable

portion of the total DNA pool, but was not affected by the ecosystem type from which the

596   sample was collected (gut, soil, sediment, and water). Grey symbols are the observed data and

black symbols represent means ± 95% confidence intervals.

Figure 3 | Bias of relic DNA on within-sample bacterial diversity in different ecosystem

600   **types. a-c** We tested for the effects of bias caused by relic DNA by calculating diversity ratios

for (**a**) richness, (**b**) evenness, and (**c**) phylogenetic diversity. The ratios reflect the diversity of

602   the total DNA pool (intact + relic) divided by the diversity of the intact DNA pool. Relic DNA

did not bias any measures of diversity in any of the ecosystem types. Richness was calculated as

604   the number of operational taxonomic units (97% sequence similarity of the 16S rRNA gene),

evenness was calculated using Simpson's evenness index, and phylogenetic diversity was

606   calculated using Faith's *D* index. Grey symbols are the observed data and black symbols

represent means ± 95% confidence intervals.

Figure 4 | Bias of relic DNA on the among-sample bacterial diversity in different

ecosystem types. a-b We tested for the effects of bias caused by relic DNA by calculating a

beta-diversity ratio based on centroid distances. Centroid distances were estimated after

performing Principle Coordinates Analyses (PCoA) using taxonomic (**a**) and phylogenetic (**b**)

distance metrics (Bray-Curtis and UniFrac, respectively). The centroid distance ratio was

calculated on each sample within an ecosystem type and reflects the composition of the total

DNA pool (intact + relic) relative to the intact DNA pool. Relic DNA had no effect on beta-

diversity for any of the ecosystem types sampled. Grey symbols are the observed data and

black symbols represent means ± 95% confidence intervals.

 Fig. 1.

Figure 1 | A model of relic DNA dynamics. Modeling relic DNA dynamics. **a** The amount of relic DNA in a microbial environment is determined by inputs associated with the mortality of viable individuals with intact DNA and by losses associated with the degradation of relic DNA. If the diversity of sequences contained in the relic DNA pool is sufficiently different from that in the intact DNA pool, then relic DNA may bias estimates of microbial biodiversity (as indicated by different colored boxes) when sampling from the total (intact + relic) DNA pool. **b** We developed a sampling-based simulation model to explore the effects of mixing intact and relic DNA on estimates of diversity. We populated intact and relic communities with individuals from a lognormal species abundance distribution (SAD). We altered the diversity of the relic community by changing the scale parameter of the lognormal distribution describing the SAD. We then sampled and mixed the intact and relic communities so that the relic contribution to total community ranged from 0.01 to 0.96. **c** To gain mechanistic insight into how bias arises, we created a stochastic process-based model that captures features that influence relic DNA dynamics, including: immigration, birth, death, degradation (**a**). We simulated a range of degradation rates to achieve relic DNA pool sizes with proportions ranging between 0.05 and 0.95. To explore how degradation alters the SAD of the relic community, we explored three scenarios. First, we simulated a neutral scenario where relic DNA sequences produced by different species degrade at the same rate. Second, we simulated conditions where physical, chemical, or biological processes slow the degradation rate of relic DNA belonging to some species via protection. Third, we simulated "hot spots" where there more abundant relic DNA sequences experience faster rates of relic DNA degradation, a condition that may arise in structured habitats where there are patchy distributions of individuals and their metabolic products (i.e., enzymes). We ran simulations for 10,000 time steps and then sampled the intact and relic communities. To quantify bias in diversity (**b, c**), we calculated "richness ratios" which reflect the number of species in the total DNA pool (intact + relic) divided by the number of species in the intact DNA pool in a simulation. When values richness ratios = 1, relic DNA has no effect on estimates of diversity; when richness ratios > 1, relic DNA overestimates true diversity; and when richness ratios < 1, relic DNA underestimates true diversity.

612   **Fig. 2.**

Figure 2 | Proportion of bacterial relic DNA in different ecosystem types. We quantified the amount of intact DNA in a sample after removing relic DNA with a DNase treatment. We then estimated the proportion of relic DNA as 1 - (intact DNA / total DNA), where the total DNA concentration was quantified without DNase treatment. Relic DNA constituted an appreciable portion of the total DNA pool, but was not affected by the ecosystem type from which the sample was collected (gut, soil, sediment, and water). Grey symbols are the observed data and black symbols represent means ± 95% confidence intervals.

Figure 3 | Bias of relic DNA on within-sample bacterial diversity in different ecosystem types. a-c We tested for the effects of bias caused by relic DNA by calculating diversity ratios for (**a**) richness, (**b**) evenness, and (**c**) phylogenetic diversity. The ratios reflect the diversity of the total DNA pool (intact + relic) divided by the diversity of the intact DNA pool. Relic DNA did not bias any measures of diversity in any of the ecosystem types. Richness was calculated as the number of operational taxonomic units (97% sequence similarity of the 16S rRNA gene), evenness was calculated using Simpson's evenness index, and phylogenetic diversity was calculated using Faith's *D* index. Grey symbols are the observed data and black symbols represent means ± 95% confidence intervals.

618   **Fig. 4.**

Figure 4 | Bias of relic DNA on the among-sample bacterial diversity in different ecosystem types. a-b We tested for the effects of bias caused by relic DNA by calculating a beta-diversity ratio based on centroid distances. Centroid distances were estimated after performing Principle Coordinates Analyses (PCoA) using taxonomic (**a**) and phylogenetic (**b**) distance metrics (Bray-Curtis and UniFrac, respectively). The centroid distance ratio was calculated on each sample within an ecosystem type and reflects the composition of the total DNA pool (intact + relic) relative to the intact DNA pool. Relic DNA had no effect on beta-diversity for any of the ecosystem types sampled. Grey symbols are the observed data and black symbols represent means ± 95% confidence intervals.