

1	How should we store avian faecal samples for microbiota analyses? Comparing efficacy
2	and cost-effectiveness
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### 19 ABSTRACT

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21 Analyses of bacterial DNA in faecal samples are becoming ever more common, yet we still do 22 not know much about bird microbiomes. These challenges partly lie in the unique chemical 23 nature of their faeces, and in the choice of sample storage method, which affects DNA 24 preservation and the resulting microbiome composition. However, there is little information 25 available on how best to preserve avian faeces for microbial analyses. This study evaluates five 26 widely used methods for preserving nucleic acids and inferring microbiota profiles, for their 27 relative efficacy, cost, and practicality. We tested the five methods (in-situ bead-beating with 28 a TerraLyzer instrument, silica-bead desiccation, ethanol, refrigeration and RNA*later* buffer) 29 on 50 fresh faecal samples collected from captive House sparrows (Passer domesticus). In line 30 with other studies, we find that different storage methods lead to distinct bacterial profiles. 31 Storage method had a large effect on community composition and the relative abundance of 32 dominant phyla such as Firmicutes and Proteobacteria, with the most significant changes 33 observed for refrigerated samples. Furthermore, differences in the abundance of aerobic or 34 facultatively aerobic taxa, particularly in refrigerated samples and those stored in ethanol, puts 35 limits on comparisons of bacterial communities across different storage methods. Finally, the 36 methods that did not include in-situ bead-beating did not recover comparable levels of 37 microbiota to the samples that were immediately processed and preserved using a TerraLyzer 38 device. However, this method is also less practical and more expensive under field work 39 circumstances. Our study is the most comprehensive analysis to date on how storage conditions 40 affect subsequent molecular assays applied to avian faeces and provides guidance on cost and 41 practicality of methods under field conditions.

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43 Key words: Avian faeces, DNA preservation, gut microbiome, House sparrows.

### 45 **1. Introduction**

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47 The gut microbiome is important for host health through its impacts on the immune system 48 (Brisbin et al., 2008; Ruiz-Rodríguez et al., 2009b; Yang et al., 2012), digestion (Dewar et al., 49 2013; Godoy-vitorino et al., 2010; Ruiz-Rodríguez et al., 2009a), development (Barbosa et al., 50 2016; Teyssier et al., 2018; Torok et al., 2011; Videvall et al., 2019) and behaviour (Cryan and 51 Dinan, 2012). While much research on the gut microbiome has focused on mammals, less is 52 known about the causes and consequences of microbiome variation in birds. The applied value 53 of studying avian microbiomes has long been realized in the poultry industry (Oakley et al., 54 2014). However, since the intimate interaction between hosts and their microbiota is thought 55 to have wide-ranging effects on all aspects of host biology, there is tremendous potential for 56 knowledge about the avian microbiome to contribute to research in avian ecology, evolution, 57 and conservation (Hird, 2017; Trevelline et al., 2019).

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59 A growing number of avian studies are capitalizing on this development and investigating 60 interactions between host life-history traits, ecology, and the gut microbiota (Grond et al., 61 2018; Kohl, 2012; Teyssier et al., 2018; Trevelline et al., 2019; van Dongen et al., 2013; 62 Videvall et al., 2019). Faecal sampling is commonly used for representing intestinal microbiota 63 because it is non-invasive. Yet obtaining reliable molecular data from avian faeces is 64 complicated by its chemical composition, as digestive excreta is mixed with urinary products 65 such as uric acid that can degrade DNA or interfere with DNA extraction (Eriksson et al., 2017; 66 Regnaut et al., 2006). The result is that DNA yields from avian faeces are typically low, making 67 amplification difficult and pipelines more sensitive to contamination. The DNA degradation 68 may also be influenced by exposure to ambient conditions, the presence of digested food items, 69 and other natural degradation processes (Hájková et al., 2006). Thus, effective preservation 70 methods are of critical importance. Moreover, faecal microbial communities will change over 71 time with exposure to conditions outside the gut. Effective sampling and storage in the wild 72 can be logistically difficult because methods such as freezing, are impractical under field 73 conditions. Therefore, a key question for many ecological studies, is how to best store and 74 preserve avian faecal samples for downstream molecular work as it affects sampling strategy, 75 experimental design and study costs.

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Most research on optimizing faecal microbiome protocols has focused on mammals and
 particularly humans, with much less work on birds and other vertebrates. Results are variable

79 and often contradictory. RNAlater is frequently used to store faecal samples for microbiota 80 analysis (Al et al., 2018; Broquet et al., 2007; Horng et al., 2018; Vlčková et al., 2012; 81 Vogtmann et al., 2017), yet there is evidence that its performance decreases after a period of 82 time at room temperature (Flores et al., 2015), and that the bacterial community profiles differ 83 to those of frozen samples (Choo et al., 2015). Ethanol is also regularly used and has been 84 shown to produce microbial profiles comparable to those obtained with RNAlater (Vogtmann 85 et al., 2017). However, some results when stored at 70% ethanol have shown higher species 86 diversity compared to fresh samples (Horng et al., 2018) with particular disparity in bacterial counts of Enterobacteriaceae (Vlčková et al., 2012) and overall poor performance, showing 87 88 an increase in relative abundance of certain taxa (Song et al. 2016). Previous methodological 89 comparisons have suggested that refrigeration can be used as a practical alternative to freezing 90 for storing faecal samples (Choo et al., 2015; Tedjo et al., 2015; Weese and Jalali, 2014), 91 though Ott et al. (2004) showed significant changes in microbiota diversity in refrigerated 92 samples over time, where the bacterial diversity reduced after 8 and 24 hours. Preserving 93 samples at room temperature might be most practical, however the ability to accurately capture 94 original microbial communities decreases rapidly within the first 24 hours at room temperature 95 (Guo et al., 2016; Tedjo et al., 2015).

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97 To date, no studies have systematically investigated how to optimize sampling and storage of 98 avian faeces for microbiota analysis, to maximize DNA quantity, quality, and cost-99 effectiveness. While much avian microbiome work has focused on commercially important 100 species, such as chickens and turkeys (Waite and Taylor, 2015), the study of avian host-101 microbiota interactions is rapidly growing in ecology and evolutionary biology (Hird, 2017). 102 In this field, microbiota research has covered a range of bird taxa (Lucas and Heeb, 2005; 103 Risely et al., 2018; Videvall et al., 2019). Passerines represent over half of extant birds and are 104 common subjects in field-based avian microbiome research. We therefore focus our 105 methodological optimization on samples from a common passerine, the House sparrow (Passer 106 *domesticus*) as model organism, representative of a large proportion of passerine research. Our 107 aim is to compare five field-compatible sample storage methods (immediate bead-beating with 108 a TerraLyzer instrument, silica-bead desiccation, ethanol, refrigeration and RNAlater), in terms 109 of DNA extraction efficacy and the resultant composition of microbial communities derived. 110 We then present our results in light of the cost and practicality of each method.

- 111
- 112 **2. Methods**

### 114 2.1. Sampling

115 We collected fresh faecal samples from a population of captive House sparrows (Passer 116 domesticus) kept in large groups (100-200 birds per aviary) indoors at the Animal Research 117 facilities, Imperial College London. The house sparrows are descendants from wild birds that 118 have been kept captive since 2005 (see references for husbandry details; Girndt et al., 2018, 119 2017). A clear plastic sheet was placed on the aviary floor after morning feeding time and left 120 there for 180 minutes. Fifty faecal pellets were collected in total - ten biological replicates for 121 each of the five storage methods compared. We assume that each pellet belonged to a different 122 individual due to the large amount of birds in the aviaries. Some variation in pellet size is 123 expected. However, samples had a wet mass of close to 0.05g.

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125 We tested the most commonly used methods for storing samples under field conditions: (1) 126 Use of Zymo's Terralyzer device ('Terralyzer' treatment). Samples were immediately placed 127 in Zymo BashingBead tubes (with 0.5 & 2mm beads) filled with 500µl of lysis solution, lysed 128 with a TerraLyzer Cell Disruptor instrument (Zymo Research) for 10 seconds and transported 129 to the lab for DNA extraction within one hour of collection. This method is expected to give 130 the most accurate bacterial profiles as bacterial growth within samples is immediately 131 interrupted and DNA is simultaneously stabilised. Therefore, for comparison purposes, this 132 treatment was used as the reference throughout our analyses (2) Desiccation with silica beads 133 ('Dry' treatment). Each sample was placed into a clean cryogenic vial which was then placed 134 inside a plastic vial containing 1.0±0.2g of silica beads; CryoTube cryongenic vial caps were 135 removed, the outer container shut and samples left to dry at room temperature and checked 136 daily for the presence of mould (Regnaut et al., 2006); (3) Immediate submersion in 500µl 96% 137 Ethanol ('Ethanol' treatment). Prior to DNA extraction, samples were placed onto filter paper 138 to absorb most of the ethanol before adding lysis solution for the bead-beating process; (4) 139 Transport back to the laboratory (within 3 hours of collection) on ice in a cool-box before 140 refrigeration (4°C) ('Refrigeration' treatment); (5) Immediate submersion in 500µl RNA*later* 141 Stabilization Solution ('RNAlater' treatment). For DNA isolation, prior to DNA extraction, 142 samples were again dried on filter paper prior to homogenization in lysis buffer.

In all methods except the TerraLyzer treatment, samples were stored in their treatment methodfor one week prior DNA extraction.

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146 2.2. Nucleic acid extraction and DNA quantification

147 Total nucleic acids were isolated from all samples using the Quick-DNA Fecal/Soil Microbe 148 Miniprep Kit (Zymo Research), incorporating minor changes the protocol: samples from all 149 treatments, except the TerraLyzer, were processed in a bead-beater (Retsch MM 440) at 20Hz 150 for eight minutes and all of the supernatant was transferred into Zymo-Spin IV Spin Filters; 151 1000µl Faecal DNA Binding Buffer was used, instead of 1200µl as the protocol suggests; DNA 152 was finally eluted in 40µl rather than 100µl as the original protocol indicates, to maximize 153 DNA concentration. Eluted DNA was stored at 4°C for two weeks, and then at -20°C for a year 154 prior to shipping to the sequencing facility. Total nucleic acid concentration and DNA purity 155 were measured using spectrophotometry (ThermoFisher Scientific NanoDrop 2000); A<sub>260</sub> was 156 used for the concentration calculation while the ratio  $A_{260/280}$  was used for estimating protein 157 contamination and A<sub>260</sub>/A<sub>230</sub> for DNA purity. Double stranded nucleic acid concentration was 158 measured using Fluorometry (ThermoFisher Scientific Qubit 2.0) with a dsDNA High-159 Sensitivity Assay kit.

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### 161 2.3 Microbiota characterization

Bacterial communities were profiled by sequencing the V4-V5 region of 16S rRNA gene using
515F/926F "fusion primers" (Walters et al., 2015). Amplicons (~410 bp) were then sequenced
on a single 2x300-bp Illumina MiSeq sequencing run at the Integrated Microbiome Resource
(IMB) facility. The library preparation and sequencing protocol used is published in Comeau,
Douglas & Langille (2017).

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### 168 2.4. Bioinformatic processing

169 Sequence data was processed using the R package DADA2 (v1.8) (Callahan et al., 2016) to 170 infer amplicon sequence variants (ASVs) (Callahan et al., 2017). First, sequence trimming and 171 quality filtering parameters were chosen and ASVs inferred, then chimeras were removed and 172 taxonomy assigned using the Silva reference database (v128) (Supplementary Information). 173 After the final ASV table was created, taxonomic filtering steps were performed in package 174 *Phyloseq* (v1.22) (McMurdie and Holmes, 2013). We removed taxa assigned as chloroplasts because they are non-informative taxa within this analysis. Abundance filtering was also 175 176 performed for beta diversity analyses, in that taxa present in less than 5% samples were 177 removed from the dataset, to limit the potential influence of contaminants or sequencing 178 artefacts. The R package iNEXT (v2.0) (Hsieh et al., 2016) was used to create sampling 179 completeness curves and decide cut-off parameters for low quality samples. ASV richness

plateaued by approximately 1000 reads, such that any samples with read counts below thisthreshold were excluded.

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### 183 2.5. Statistical analysis

184 DNA concentration and purity were compared across treatments using factorial ANOVAs. For 185 alpha diversity analyses, the effect of treatment on microbiota diversity was estimated using 186 the Shannon index calculated by the *breakaway* package (v4.6.8) (Willis and Bunge, 2015). 187 For beta diversity analyses, read counts were normalised using cumulative-sum scaling using 188 the metagenomeSeq package (v1.2) (Paulson et al., 2013). We calculated community 189 dissimilarity matrices (generalised UniFrac and Bray-Curtis dissimilarity) in the packages 190 GUniFrac (v1.1) and vegan (v2.5) (Chen et al., 2012; Dixon, 2003). These dissimilarity 191 matrices were then used in a permutational analysis of variance (PERMANOVA) to examine 192 how storage treatments affected community composition. We used the function betadisper 193 within package vegan (Anderson, 2001) to tests if differences in sample dispersion might 194 influence community composition differences among treatments. Finally, as most gut bacteria 195 are obligate or facultative anaerobes (von Martels et al., 2017), we also evaluated the effects 196 of different storage conditions on the ability to detect anaerobes and aerobes (see 197 Supplementary Information). This gives an insight on possible colonization and outgrowth of 198 aerobes after sample collection. All analyses were carried out in R (version 3.4.4, R Core Team, 199 2014).

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201 2.6. Cost and practicality

202 Cost reflects price in US dollars of sample preservation, including the price of cryogenic vials, 203 buffers, ice and beads, and extra accessories (Table A2). Cost was calculated for projects of 204 100, 500 and 1000 samples. The cost of a TerraLyzer machine was excluded for project 205 expenses as all protocols require and instrument for bead-beating, the difference is whether this 206 is performed in the field (TerraLyzer) or in the laboratory (rest of the protocols). All prices 207 were estimated in March 2019 as displayed online, and do not include discounts for research 208 institutions. To assess the practicality of each method, we developed a time-effort index based 209 on convenience of a process under field conditions using 10 different criteria (Table 1). Each 210 index assigned to a treatment was plotted against cost.

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Table 1. Practicality criteria developed for assessing storage methods for use in the field (top)
and scoring system assessed by single sample for the practicality index (below).

	Criterion	Description
i.	Size	The equipment is large or heavy to carry and
		may require the use of additional boxes for
		transportation
ii.	Temperature sensitivity	The method is sensitive to temperature and
		has to be kept in stable environment
		(fluctuations $< \pm 4^{\circ}$ C)
iii.	Shelf-life	The method or one of its components has to
		be replaced every ≅7 days
iv.	Monitoring	The method requires frequent monitoring of
		external conditions such as temperature and
		humidity (check samples at least once a day)
v.	Sample reorganization	The method requires moving a sample
		between tubes/buffers or reagents from its
		original storing tube.
vi.	Workforce required	The method requires the presence of more
		than one person to help with the storage of a
		sample
vii.	Electricity	The method involves machinery which
		requires access to electricity or needs to be
		charged
viii.	Leak or spillage	The method involves liquid buffers/reagents
		which can spill or leak onto other equipment
		or samples
ix.	Travel restrictions	The method includes components which
		may be restricted when traveling (liquids for
		air travel, dry ice, high concentrations of
		ethanol, lithium-ion batteries, etc.)
x.	Time from source to storage	Time taken from sample collection to
		completion of storage ( $\geq 10$ seconds per
		sample)

Score Description	
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0	Not practical. Six or more of the criteria are met
1	Borderline practicality. Five of the criteria are met
2	Satisfactory practicality. Three or four of the criteria are met
3	Practical. Meets up to two of the criteria

### 216 **3. Results**

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218 *3.1. DNA extraction assessment* 

219 In total 50 DNA extracts were obtained from 50 faecal pellets (~0.05g each). The mean nucleic 220 acid concentration by spectrophotometry (NanoDrop) was  $36ng/\mu l \pm 1$  SE. TerraLyzer samples 221 had the highest mean concentration ( $41ng/\mu l \pm 3$  SE) while the refrigeration ( $4^{\circ}C$ ) method 222 presented the lowest mean concentration ( $30ng/\mu l \pm 2$  SE). As expected, double stranded DNA 223 concentrations measured by Fluorometry (Qubit), were lower than the spectrophotometry 224 (NanoDrop) measures (Table 2); the mean concentration was 0.22ng/µl ±0.01 SE, and DNA 225 concentration was not significantly predicted by storage method ( $F_{4,45}$ = 1.0, p= 0.133). Average 226 values for protein contamination in the samples  $(A_{260/280})$  were outside the range of 1.8-2.0 227 (1.29±0.02 SE) regarded as indicative of low protein contaminant content (Table 2). Overall, 228 the  $A_{260/280}$  ratio was not significantly predicted by storage method (F<sub>4,45</sub>= 1.32, p= 0.275); but 229 ethanol had the highest protein contamination compared to TerraLyzer samples. DNA purity 230 ratio  $(A_{260/230})$  was below 1.8 in all samples (mean 0.24±0.01 SE), possibly suggesting a high 231 concentration of contaminants (Table 2); and it did not show significant differences with 232 respect to treatment ( $F_{4,45}=0.77$ , p=0.546).

# Table 2. DNA concentration, protein contamination and purity of house sparrow faecal sample DNA extractions for each method tested. Mean±SE is shown in all cases.

Treatment	DNA conc.	dsDNA conc.	Protein	DNA purity
	(Spectrophotometry,	(Fluorometry,	contamination	$(A_{260}/_{230})$
	ng/µl)	ng/µl)	$(A_{260}/_{280})$	
TerraLyzer	41 ±3	0.24 ±0.01	1.22 ±0.05	0.27 ±0.03
Dry	37 ±2	0.23 ±0.02	1.32 ±0.01	0.26 ±0.02
Ethanol	34 ±3	0.18 ±0.02	1.39 ±0.08	0.25 ±0.04
4°C	30 ±3	0.95 ±0.70	1.29 ±0.05	0.21 ±0.02
RNAlater	38 ±3	$0.20 \pm 0.02$	1.25 ±0.04	$0.20 \pm 0.03$

### 236 *3.2 Microbiota profiles*

Only 38 of 50 samples (76%) were included in 16S rRNA microbiota profiling. Of these, 17 (45%) satisfied quality filtering parameters during the bioinformatic pipeline (100% TerraLyzer, 71% Dry, 33% ethanol, 43% ice and 100% RNA*later*). A total of 851,284 sequence reads were obtained following quality filtering, comprising 22,402  $\pm$ 5,748 SE raw reads per sample. Read count was not significantly predicted by treatment (Kruskal-Wallis chisquared= 7.22, df= 4, p= 0.124).

243

All treatments differed in Shannon diversity compared to the TerraLyzer treatment, though

the direction varied (estimated sigma $^2_u = 17.15$ , p= 0.00), except for the samples stored

dried (p=0.18), though these samples also presented the highest variability in diversity (Fig.

247 1a).

248 Overall, treatment had a strong and significant effect on microbial community composition

249 (PERMANOVA on weighted UniFrac,  $F_{4,16}=2.74$ ,  $R^2=0.47$ , p=0.007), and we didn't find

different levels of dispersion within treatment (betadisper,  $F_{4,12}$  = 0.50, p = 0.73; Fig. A1). The

treatment that had the most similar community composition to TerraLyzer on average was

252 RNAlater, however samples from this treatment, also had the highest variation in community

253 composition (Fig. 1a); the storage method that produced an average composition most

distinct from that of the TerraLyzer was ethanol with a mean Bray-Curtis distance of 0.95

255 (Fig. 1b).



Figure 1. Microbial community diversity and composition differences for the five tested
treatments. a) Estimated Shannon diversity of ASVs for each of the five treatments. Points
and error bars indicate mean diversity estimates and confidence intervals respectively.
(b) Bray-Curtis distance in community composition between samples in the TerraLyzer
treatment and those analysed with other treatments. Points and error bars in both plots
indicate means and standard deviations for each comparison, respectively.

264

265 Across all storage conditions, the dominant phyla detected were Firmicutes and Proteobacteria, 266 but the ratio of relative abundance between these two differed significantly among treatments 267 (Kruskal-Wallis chi-squared = 141.47, df = 4, p = 0.00). A pairwise Wilcoxon rank sum test was 268 applied to detect differences of relative abundance of the eight most abundant phyla among 269 treatments; the greatest differences between the Terralyzer samples and the rest, were seen for 270 refrigerated samples (pairwise Wilcoxon test p=0.00) with a considerably higher proportion 271 of Bacteroidetes, SBR1093, Thaumarchaeota and Actinobacteria (Fig. 2a). Also, refrigerated 272 samples had higher relative abundances of Flavobacteriales (2%), Rhizobiales (3%),

Salinisphaerales (0.7%), SAR11\_clade (2.5%) and from other unassigned orders (13%),
compared to the rest of the treatments (Fig. 2b).

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A total of 101 ASVs were identified to genus level and included in the analysis of respiration 281 282 type/aerotolerance (Table A1). The proportion of detected genera that were either obligate or 283 facultative anaerobes (expected in the gut) was similar in TerraLyzer, refrigerator and 284 RNA*later* treatments. However, refrigeration revealed proportionally more aerobic genera than 285 the other treatments. Samples stored dried and in ethanol presented substantially lower relative 286 abundance of obligate anaerobic genera compared to facultatively aerobic bacteria (Table 3). 287 This result suggests that storage methods may differ in the extent to which they allow 288 aerotolerant or aerobic bacteria to multiply post-collection.

	Aerobic	Anaerobic	Facultative	Unclassified
TerraLyzer	1.5	35.5	62.8	0.2
Dry*	0.0	3.2	96.6	0.1
Ethanol	1.6	5.6	92.6	0.2
4°C	3.1	28.3	51.8	16.8
RNAlater	1.5	36.4	61.8	0.3

\*0.09% rounding error in Dry treatment

290 Table 3. Relative abundance (%) of bacterial genera classified by their cellular 291 respiration, found in different sample storage conditions.

292 293

#### 294 3.3 Cost and practicality

295 According to the cost analysis ethanol is the cheapest method per sample (\$0.75 USD) and the 296 use of ice with additional refrigeration to keep samples refrigerated at 4°C is the most expensive method per sample (\$8.16 USD, Table A2), but as the size of the project increases, 297 298 refrigeration becomes the cheapest method (\$379.6 USD for 1000 samples), and the use of 299 TerraLyzer (in situ bead-beating) method the most expensive (\$1482 USD for 1000 samples; 300 Fig. 4a). If the practicality of using each method in the field is analysed together with the cost 301 of a 100-sample project, then the methods with the best price-practicality ratio are ethanol and 302 RNAlater. The refrigeration method is the most affordable storage method, however, is also 303 the least practical to perform in field work conditions (Fig. 4b).





Figure 4. Cost and practicality of five sample storage treatments. a) Total costs of projects
 using different number of samples: 100, 500, 1000. b) Practicality and costs for a 100-sample
 project. In the practicality index, "0" is the least practical treatment, and "3" the most
 practical.

310

### 311 4. Discussion

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313 Results of this study show that faecal sample storage method affects the microbial community 314 detected in downstream analysis. Three major findings derive from the current study. First, 315 microbial composition is determined by storage method; relative abundances of certain phyla 316 change across treatments, especially on refrigerated and ethanol samples; this could be driven 317 by the differentiated proportion of aerobes and anaerobes, indicating selective detection rates. 318 Second, the efficiency on faecal DNA quality (concentration and purity) is not determined by 319 the storage of faeces, and it does not reflect microbiome composition results. Third, treatments 320 that include the use of RNAlater and ethanol meet important criteria such as being low-cost 321 and are highly practical under field conditions, however they do not necessarily reliably store 322 the microbial composition of house sparrow faeces. Together, these results suggest that 323 knowing the caveats associated with each storage method are crucial during design, analyses 324 and interpretation of avian microbial results.

325

326 The evidence here confirms that each treatment alters microbial communities by affecting the 327 relative abundances in great magnitude; thus, care should be taken when comparing values across studies using different protocols, especially when incorporating metrics such as Shannon 328 329 index. The most abundant phyla across all samples were Proteobacteria and Firmicutes, which 330 is consistent with what was previously reported for House sparrows (Kohl et al., 2019; Mirón 331 et al., 2014); however, we found higher relative abundance of Proteobacteria in samples stored 332 dried. This result suggests that consideration should be given to differences in abundance at 333 certain taxonomic levels that have undergone this type of storage, particularly those involving 334 Proteobacteria and Actinobacteria which are able to grow at a range of temperatures (Weese 335 and Jalali, 2014).

336

Furthermore, changes observed on microbial abundances at order level, particularly from the
ones stored at chilled temperature can be attributed to oxygen exposure resulting in bacterial
degradation (Ott et al., 2004). The ability to detect total aerobes and anaerobes from different

340 storage conditions can be used as a proxy of the global effect on storage methods (Fouhy et al., 341 2015); we found that a greater proportion of aerobes were recovered following refrigeration, 342 suggesting that oxygen-tolerant bacteria are thriving after collection, driving biases on the 343 community composition. We also found that the levels of total anaerobic and facultative 344 bacteria in RNA*later* samples were similar to the ones detected in samples processed using the 345 TerraLyzer, which suggests that immediate submersion in buffer solution following collection 346 enables the recovery of comparable types of microbiome. Remarkably, the recovery rates of 347 taxonomic groups in RNA*later* are not comparable to those found in samples processed by the 348 TerraLyzer.

349

Encouragingly, inter-individual variations were smaller than variation between methods, suggesting consistency in sampling within each method applied; therefore, as long as the same preservation method is used across a study, unbiased comparisons can be made between samples. Having said this, there will always be methodologic or biologic related biases as established by Hallmaier-Wacker *et al.* (2018) and (Pollock et al., 2018); this highlights the need for proper validation and standardization for each sample type and the use of blank control samples, to assess the limitations in protocols and datasets.

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358 Going forward, numerous studies have suggested that inadequate storage can result in reduced 359 DNA quantity and quality and addressing this issue will ensure effective and accurate genotyping (Murphy et al., 2007; Soto-Calderón et al., 2009). However, this study shows that 360 361 adequate storing protocols are not enough to achieve high quality avian gut microbiome 362 profiles. Faecal extracts are characterized by low DNA concentration and high degradation 363 (Dai et al., 2015; Demay et al., 2013), and sparrow samples analysed here are no exception. 364 Avian DNA concentrations and purity are consistently lower compared to those reported for 365 mammal faeces DNA (Bubb et al., 2011; Costa et al., 2017; Horng et al., 2018). This suggests 366 that further studies should focus on the implementation of methodologies that improve DNA 367 recovery from avian faeces beyond sampling optimization.

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These analyses represent the first attempt to test how storage methods of bird faeces affect microbiome research. We are still in search of the best methodologies, however the sole focus on the storage protocol will not resolve other difficulties associated with working with avian faeces, such as high uric acid content. Until then, other factors can be taken into account such as cost and practicality under field conditions. The present study allows to choose theaffordability of the equipment and reagents used for each protocol.

375 The use of the TerraLyzer has not been widespread, however, we showed the use of such an 376 instrument to be useful a preliminary bead-beating step to break tissues in the field and increase 377 optimal storage. Such a device is easy to use as it ensures a good bead-motion. In particular, 378 the TerraLyzer becomes cost-effective when used for multiple eDNA studies under field 379 conditions. Applying the two-step silica desiccation method has demonstrated to be useful on 380 recovering microbiome communities similar to those on control samples (Bhagavatula and 381 Singh, 2006), nevertheless this method requires special attention and extra care when handling 382 and monitoring the samples, and climatic variables should also be considered when working in 383 humid and hot environments. Freezing is not possible under field conditions, unless there is 384 access to electricity or liquid nitrogen. This study substituted it by placing the samples on ice 385 and refrigerating them and, similarly, to freezing the samples, the practicality of this method 386 was low. The treatment that involves the use of a buffer (RNAlater) has the best 387 cost/practicality ratio, as does the use of ethanol, however careful attention must be paid to 388 these methods as they might be underrepresenting the original microbial community.

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### **390 5.** Conclusions

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392 The results shown provide guidelines to aid researchers embarking a microbial project on wild 393 bird populations. We further advise other to perform a pilot study to determine which storage 394 approach is optimal for them, as this will be dependent not only on their objectives, but also 395 on the practicality and cost-efficiency of each approach. The optimization of the sampling 396 protocols should take into account the environments from which samples will be collected, the 397 length of time the sample will be in storage for, and the size of the project. Importantly, we 398 show that regardless of the method chosen, consistency of storage within project is a prime 399 practice to achieve replicable and reliable results for microbial ecology.

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- 633 Appendix A. Supplementary Information
- 634

### 635 *Methodology for bioinformatics*

636 The first step for the bioinformatics followed a pipeline from the R package DADA2 used to 637 evaluate the quality and size of raw reads, this enabled to choose the cut-off parameters for the 638 trimming and filtering of the sequences; we used the standard filtering parameters and 78% of 639 the sequences survived this step. Next, an error model was calculated for the specific dataset, 640 then, to reduce computational time, we dereplicated the sequences by eliminating redundant 641 comparisons and allocating abundances of each "unique sequence". Amplicon sequence 642 variants (ASVs) were then inferred and spurious ones were further reduced by overlapping 643 reads, this step removes substitution and indel errors, but not chimeras, therefore, a simple 644 phase on identifying and removing chimeras was applied. At this point was possible to classify 645 sequence variants taxonomically.

646

### 647 Methodology for determining bacteria respiration type

648 We selected the lowest taxonomic level –Genus- in order to have the highest resolution on the 649 identity of each taxa. We created a search strategy for each taxon to find the respiration type: 650 we used Google Scholar, PubMed and the book "The Prokaryotes. prokaryotic Biology and 651 Symbiotic Associations" (Rosenberg et al., 2013). Once the respiration type was identified, 652 each ASV was labelled with either aerobic, anaerobic, facultative or not-identified. This 653 labeller allowed to know the relative abundances of each type of respiration found in each 654 treatment (Table A1). In the cases where no information was found for a specific Genus, the taxon was not considered for the analysis. A total 101 ASV were included for this part of the 655 656 analysis.

657

### **Table A2. Material costs (USD\*) per sample for each sample storage treatment**

Treatment	Tubes		Medium		Extras		Total
TerraLyzer	BashingBead	0.54	Lysis	0.93	-	0.0	1.48
	tubes		solution				
	(0.5/2mm)						
RNAlater	1.5 ml	0.36	RNAlater	0.44	-	0.0	0.80
	CryoTubes						
Ethanol	1.5 ml	0.36	Ethanol	0.39	-	0.0	0.75
	CryoTubes		(90%)				

Refrigeration	1.5 ml	0.36	Ice	1.3	Cool box	6.5	8.16
(4°C)	CryoTubes						
Desiccation	1.5 ml	0.36	Silica	0.35	Plastic vials	0.46	1.13
(dry)	CryoTubes		beads				

\*Prices to March 2019.

## 661 Figure A1. Differences in the faecal microbiota between treatments. Bray-Curtis distances



