1	Contribution of structural recalcitrance to the formation of the deep
2	oceanic dissolved organic carbon reservoir
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12	Running title: Structural recalcitrance dominates DOM persistence
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17 Originality-Significance Statement

There is a lively debate over why the large recalcitrant dissolved organic carbon 18 19 (DOC) reservoir persists in the ocean for extended time periods (up to millennia). Here, we evaluate the relative contributions of structurally recalcitrant DOC and 20 recalcitrant DOC due to dilution to deep oceanic DOC pool by modeling 21 22 transformations of the different DOC components in incubation experiments with different DOC concentration levels. We conclude that the majority of DOC in the 23 24 deep ocean is structurally recalcitrant, which supports the hypothesis that the 25 recalcitrance of RDOC is largely related to its chemical properties rather than its low 26 concentration. This also implies that recalcitrant DOC is independent of from RDOC 27 concentration threshold for bacterial uptake.

28 Summary

29 The origin of the recalcitrant dissolved organic carbon (RDOC) reservoir in the deep 30 ocean remains enigmatic. The structural recalcitrance hypothesis suggests that RDOC 31 is formed by molecules that are chemically resistant to bacterial degradation. The 32 dilution hypothesis claims that RDOC is formed from a large diversity of labile molecules that escape bacterial utilization due to their low concentrations, termed as 33 34 RDOC_c. To evaluate the relative contributions of these two mechanisms in 35 determining the long-term persistence of RDOC, we model the dynamics of both 36 structurally recalcitrant DOC and RDOC_c based on previously published data that describes deep oceanic DOC degradation experiments. Our results demonstrate that 37 the majority DOC (84.5±2.2%) in the deep ocean is structurally recalcitrant. The 38 39 intrinsically labile DOC (i.e., labile DOC that rapidly consumed and RDOC_c) 40 accounts for a relatively small proportion and is consumed rapidly in the incubation 41 experiments, in which 47.8±3.2% of labile DOC and 21.9±4.6% of RDOC_c is 42 consumed in 40 days. Our results suggest that the recalcitrance of RDOC is largely 43 related to its chemical properties, whereas dilution plays a minor role in determining 44 the persistence of deep-ocean DOC.

45 Introduction

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The ocean contains a large amount of recalcitrant dissolved organic carbon (RDOC). This reservoir plays a critical role in global carbon sequestration and potentially affects global climate (Hansell et al., 2009; Jiao et al., 2014). Although we know that RDOC persists in the water column for thousands of years (Brophy and Carlson, 1989; Lancelot et al., 1993; Carlson and Ducklow, 1995; Jiao et al., 2010; Hansell, 2013), the mechanisms underpinning this extraordinary stability remain objects of

controversy (Arrieta et al., 2015b, a; Jiao et al., 2015).

53 Two primary but not mutually exclusive mechanisms have been proposed to explain 54 the persistence of RDOC. The dilution hypothesis postulates that the deep oceanic dissolved organic carbon (DOC) consists of a large diversity of extremely diluted 55 56 molecules, termed RDOC_c. These molecules, although chemically labile, would not be consumed by bacteria because their concentrations could be below the thresholds 57 for direct assimilatory uptake and the metabolic requirements for their degradation 58 cannot be satisfied (Jannasch, 1967; Barber, 1968; Dittmar and Paeng, 2009; 59 60 Kujawinski, 2011; Arrieta et al., 2015b, a). This hypothesis is supported by 61 experimental observations showing that the growth of microbes ceased if their substrates fell below a defined concentration threshold (Jannasch, 1967, 1994; 62 Martens-Habbena et al., 2009; Arrieta et al., 2015b, a). 63

Another explanation for the long-term persistence of RDOC is the structural recalcitrance of the molecules in a specific environmental context, which corresponds to the previously proposed concept of biologically inert RDOC (RDOC_t) (Jiao et al., 2014). This idea is supported by the evidence that the chemical composition of the DOC in the deep ocean, where the RDOC dominates, is different from that of

69 decomposable DOC (Jannasch, 1967; Lancelot et al., 1993; Kaiser and Benner, 2009; 70 Jiao et al., 2010; Benner and Amon, 2015; Walker et al., 2016b). Indeed, solid phase extracted (SPE) DOC from deep ocean generally features higher double-bond 71 72 equivalent values, lower O/C and H/C ratios, as well as a higher degree of 73 unsaturation and rings in molecules, than that from surface waters (Koch et al., 2005; 74 Flerus et al., 2012; Koch et al., 2014; Hansman et al., 2015; Arakawa et al., 2017; Jiao 75 et al., 2018). Although a complete chemical characterization of marine DOM is currently lacking, it was reported that RDOC produced from bacterial consumption of 76 77 simple substrates contains carboxyl and fused alicyclic functional groups, which 78 determines its resistance to biodegradation and refractory nature (Ogawa et al., 2001; 79 Hertkorn et al., 2006; Lechtenfeld et al., 2014). For instance, the carboxyl-rich 80 alicyclic molecules (CRAM), which account for ~8% of the DOC, are one of the most 81 abundant organic components ever identified in the deep ocean (Hertkorn et al., 2006; Hertkorn et al., 2012; Lechtenfeld et al., 2015; Rossel et al., 2015). 82

83 Both RDOC_c and RDOC_t are products of the microbial carbon pump (MCP) since they are gradually generated by the successive and repetitive actions of bacteria 84 85 degrading relatively more labile substrates (Jiao et al., 2014). However, the 86 implications that these two pools have for the capacity of the MCP to store carbon by forming RDOC differ dramatically. If the majority of RDOC is RDOC_c, the capacity 87 of the ocean to store carbon as RDOC would be limited by the DOC concentration 88 threshold for bacterial uptake (i.e., RDOC cannot exceed that threshold otherwise it is 89 consumed by microbes). In contrast, if the majority of RDOC is RDOC_t, the capacity 90 of the ocean to store carbon via the MCP would be independent from RDOC 91 92 concentrations. Therefore, the trade-off between these two forms of recalcitrance would profoundly influence the capacity of the ocean to store carbon as RDOC andthus influence the impacts of ocean carbon dynamics on climate change.

In this study, we assess the relative contributions of the dilution and structural 95 96 recalcitrance hypotheses in determining the long-term persistence of DOC by reanalyzing published data from a DOC degradation experiment (Arrieta et al., 97 2015b), in which natural bacterial communities from the deep Pacific and Atlantic 98 99 oceans were incubated with deep-water SPE DOC at different concentrations in about 100 40 days. The dataset includes measurements of SPE DOC utilization and Fourier transform ion cyclotron resonance mass spectrometry (FT-ICR-MS) data of 101 102 hydrophobic SPE DOM (~40% of total seawater DOC). Although the FT-ICR-MS is not totally quantitative due to the matrix effect (i.e., matrix-dependent ion 103 104 suppression), appropriate standardization and experimental designs allow us to 105 identify molecules that are consumed or produced (Arrieta et al., 2015a) and investigate the influence of the matrix effect on results (Osterholz et al., 2015). Here, 106 107 we model the dynamics of DOC components under the framework that i) the initial deep oceanic DOC is composed of structurally recalcitrant DOC (RDOC_t), 108 biodegradable DOC with extremely low individual concentrations that are below the 109 110 corresponding microbial uptake thresholds (RDOC_c) and labile DOC ready for microbial utilization (LDOC); and ii) LDOC is utilized in both controls and 111 concentrated treatments, while RDOC_c can only be partially utilized in concentrated 112 treatments (Figure 1). By leveraging the FT-ICR-MS profiling, which characterizes 113 SPE DOM molecular composition changes, we define a utilization index for each 114 compound by comparing the relative intensity of normalized FT-ICR-MS signal at the 115 start and at the end of the experiments. Then, we calculate the proportion of 116 intrinsically labile DOC (LDOC and RDOC_c) consumed by averaging the utilization 117

index of compounds that are significantly utilized. Ultimately, the contributions of RDOC_c and RDOC_t to the deep oceanic DOC pool are quantified by comparing the

120 observed SPE DOC consumption to the consumption of the intrinsically labile DOC

121 with constraints on the DOC production process.

122

123 **Results and Discussion**

124 The intrinsically labile DOC was rapidly utilized

125 We first estimate the percentage of consumed intrinsically labile DOC in relation to 126 initial intrinsically labile DOC (c_l for LDOC and c_c for RDOC_c as shown in **Table 1**) 127 based on significantly utilized molecules. The percentage of LDOC consumed are 128 estimated to be 54.9±3.3% and 47.8±3.2% for experiments O and P respectively (Table 2), meaning that approximately half of the LDOC was consumed in the 40-day 129 130 incubation. This indicates that the intrinsically labile DOC was quickly consumed, which is consistent with the observations from recent bioassay experiments that 131 132 approximately 75% of labile DOC was consumed in less than 30 days (Lechtenfeld et al., 2015) and supports labile DOC cycles with timescales of days to weeks (Benner 133 and Amon, 2015; Hansman et al., 2015). The propagation uncertainty caused by 134 measurement variability is approximately one order of magnitude lower than the 135 136 estimates, allowing us to rule out measurement uncertainty caused by the FT-ICR-MS 137 data.

The additional removal of deep oceanic DOC, i.e., $RDOC_c$ could be observed within concentrated treatments by comparing the normalized FT-ICR-MS signals at the end of controls and 5-fold concentrated treatments. It shows that approximately 25% of compounds in the $RDOC_c$ pool has been utilized, i.e., 25.2±3.4% for station O and

21.9±4.6% for station P (Table 2). Together with the evidence of microbial growth 142 data shown in Arrieta et al. (2015a), it supports the existence of RDOC_c in the deep 143 144 ocean. The percentage of RDOC_c consumed (21.9% to 25.2%) is smaller than that of LDOC (47.8% to 54.9%), which can be attributed to the fact that molecules from the 145 146 RDOC_c pool are too diluted and have less chance to be encountered by microbes for 147 consumption (Hansell and Carlson, 2014). Analogous to the estimates of c_l with narrow ranges of propagation uncertainty, the propagation of measurement 148 149 uncertainty on c_c is lower than 4.6% as well (**Table 2**). Notably, our method provides 150 the lower limits of c_l and c_c rather than their true values (see Supplementary Methods), implying that there would be more intrinsically labile DOC consumed in 151 reality than in our estimates. 152

153 The majority of deep oceanic DOC is RDOC_t

154 The observed percentage of SPE DOC consumed (i.e., c_{1X} and c_{5X}) in the incubation periods are 2.1% and 4.1% for controls and 5-fold concentrated treatments, 155 respectively (Table 3; Supplementary Methods). The percentage of RDOC_t in SPE 156 157 DOC is estimated from equation (8) in Methods. Taking results from experiment O as an example (Table 2), LDOC and RDOC_c account for less than 4.4% and 9.1% of 158 SPE DOC, respectively, which implies that $RDOC_t$ represents up to 86.6±1.2% of 159 SPE DOC. The extensive sensitivity tests show that even using a very wide range of 160 parameter values (Table 3), the estimate of the percentage of $RDOC_t$ in SPE DOC 161 changes no more than 2.3%, except for the effect of c_{5X} that yields no more than 9.0% 162 163 of the change. Therefore, our results are robust with respect to the choice of parameters. Repeating the same model process on the dataset from experiment P 164 shows similar results in which the percentage of RDOC_t in SPE DOC is estimated to 165

166 be $84.5\pm2.2\%$. Based on these results, we propose that the majority of the SPE DOC

167 is recalcitrant due to its structurally recalcitrant property rather than the dilution.

168 Since FT-ICR-MS data only provide molecular formulas and no chemical structures, 169 it is impossible to unambiguously designate a specific compound as being refractory 170 due to dilution or chemical structure without any experimental treatments. Here, we 171 have reanalyzed FT-ICR-MS datasets from substrate utilization experiments with 172 concentrated treatments and explored the contributions of both RDOC_c and RDOC_t to long-term persistence of RDOC in the deep ocean. The key rationale of our method is 173 174 that a fraction of RDOC_c becomes bioavailable when increasing the DOC concentrations, the process of which allows the differentiation between compounds 175 with low concentration (i.e., RDOC_c) and structurally recalcitrant compounds (i.e., 176 177 RDOC_t) in each mass peak and thus the quantification of their respective contributions to the peak magnitude. One caveat to use the FT-ICR-MS data is that the 178 efficiency of the SPE method (Bond-Elut-PPL, 1g) is ~40% as shown in Arrieta et al. 179 (2015a) and possibly introduces compositional bias towards hydrophobic compounds 180 while losing other DOC compounds (Coppola et al., 2015; Broek et al., 2017). 181 However, it is shown that the radiocarbon (Δ^{14} C) values and C/N ratios of PPL SPE 182 DOC are statistically indistinguishable from those of bulk DOC at depth, which 183 suggests that PPL SPE DOC is a close representative of the deep oceanic DOC 184 185 dominated by refractory compounds (Broek et al., 2017). Considering the extracted DOC in experiments O and P are from deep ocean, we could ignore the influence of 186 compositional bias caused by the PPL extraction and generalize our conclusion that 187 RDOC_t dominates the deep oceanic DOC. Matrix-dependent ion suppression is 188 189 another issue related to the use of FT-ICR-MS data (Trufelli et al., 2011; Osterholz et al., 2015). In the enrichment experiments, incubation and DOC concentrated 190

191 treatments can result in the changes of both DOC composition and concentration, and 192 therefore lead to the matrix effect. Our model considers matrix-dependent ion suppression to be an inherent assumption, and we evaluate the matrix effect caused by 193 194 DOC composition and concentration changes in incubation experiments (Figure S2). 195 When matrix effect is considered, RDOC_t accounts for 82.1~89.1% and 77.8~87.9% 196 of SPE DOC for experiments O and P, respectively (Table S1), showing that the 197 matrix effect is unlikely to affect our estimates. However, considering that the matrix effect is a complex issue with poorly understood mechanisms, we should pay a great 198 199 attention to sample preparation and the use of calibration approach for the FT-ICR-200 MS community in publishing future datasets (Trufelli, et al., 2010).

Our results are consistent with the size-reactivity model and other studies that link the 201 202 radiocarbon age of organic matter to its chemical composition (Loh et al., 2004; 203 Repeta and Aluwihare, 2006; Walker et al., 2014; Benner and Amon, 2015; Walker et al., 2016a; Walker et al., 2016b). The size-age-composition relationship that organic 204 205 matter size is negatively correlated with radiocarbon age and carbon:nitrogen ratios supports the dominant role of chemical composition in determining the recalcitrance 206 of the RDOC pool (Walker et al., 2016b). In addition, if the majority of deep oceanic 207 DOC is RDOC_c, deep ocean Δ^{14} C calculated from the mass balance model would be 208 209 difficult to reconcile with its observation (Wilson and Arndt, 2017). Most recently, an 210 experimental study using large volume water column (>100 tons) shows that the microbial transformation of LDOC to RDOC in a standing ecosystem can be 211 212 completed in a very short time (a few months) (Jiao et al., 2018), providing further 213 evidence for the dominance of RDOC_t.

214 It is worth noting that the study of Arrieta et al. (2015a) also refers to the RDOC_c fraction by constructing a utilization index of the relative abundance of molecules, 215 which showed that there are more than 70% and 40% of DOC is RDOC_c in the deep 216 217 ocean for experiments O and P, respectively, i.e., no more than 30% and 60% of DOC is RDOC_t. This discrepancy from our study can be attributed to different assumptions 218 219 in the estimation. Their estimation assumes that each molecule is either intrinsically 220 labile or structurally recalcitrant. Therefore the significantly utilized molecules detected by utilization index are only intrinsically labile, i.e., LDOC and RDOC_c. 221 222 However, our method assumes that each observed peak likely contains both intrinsically labile DOC and RDOC_t components, since that the structural isomers 223 with different bioavailability cannot be distinguished with the limit of current FT-ICR 224 225 MS technology (Stubbins et al., 2014; Osterholz et al., 2015).

226 Explanations of low DOC consumption in concentrated experiments

Arrieta and coworkers have stated that the 'dilution hypothesis is the primary 227 228 mechanism controlling on the biogeochemical cycling of DOC in the deep ocean' (Arrieta et al., 2015b). It has been pointed out that there is no significant increase in 229 230 DOC consumption (<6%) when increasing DOC concentrations (Jiao et al., 2015). 231 Arrieta and coworkers explained that it would take a longer time for complete consumption of labile substrates and illustrated this process by a simulation study that 232 assumed that DOC utilization depends only on concentration. Here, we carry out a 233 similar simulation but under a more general case that deep oceanic DOC is a mixture 234 of structurally recalcitrant and intrinsically labile DOC. If no RDOC_t is available in 235 the original SPE DOC (Figure 2A), the simulation results are equal to those reported 236 in Arrieta et al. (Arrieta et al., 2015a). The SPE DOC utilized after 40 days is 237 $2.1\pm0.7\%$, $2.6\pm0.9\%$, $3.7\pm1.2\%$ and $5.2\pm1.7\%$ of the initial concentrations for controls, 238

239 2-, 5-, and 10-fold concentrated treatments, respectively. However, if the percentage 240 of RDOC_t in total DOC varies from 10% to 90% (Figure 2B-D), there is no apparent 241 difference in DOC consumption rate among the four scenarios at the 40-day time 242 scale. Based on our conservative estimate of the percentage of RDOC_t in total DOC 243 (84.5%), the DOC utilized after 40 days would be $2.1\pm0.7\%$, $2.5\pm0.8\%$, $3.5\pm1.1\%$ and 244 4.8±1.5% of the initial concentration for controls and concentrated treatments, which 245 are still well within the range of the observations (<6%) and have slight difference from those estimated under the assumption that no RDOC_t exists. 246

Till now, there are two plausible interpretations about the low DOC consumption in 247 248 enrichment experiments. The first one, illustrated in Arrieta et al. (2015a), hypothesizes that a large fraction of DOC in the deep ocean is intrinsically labile 249 250 compounds and attributes the low DOC consumption even under concentrated 251 treatments to the short-term incubation. In other words, the concentrated DOC could 252 be completely consumed if given longer time for incubation (Figure 3A). In contrast, 253 our results show that the percentage of intrinsically labile DOC consumed is high in less than 40-days of incubation (>47.8% for LDOC and >21.9% for RDOC_c), but the 254 small fraction of intrinsically labile DOC results in the small bulk DOC consumption 255 256 (Figure 3B). Interestingly, this discussion is also consistent with the results obtained 257 by a different modeling approach that explored the radiocarbon signature of DOC in relation to the dilution theory (Wilson and Arndt, 2017). While both the models 258 suggest that DOC recalcitrance is largely related to its chemical composition, they 259 also imply that RDOC_c plays a secondary role in explaining the long-term persistence 260 of deep oceanic DOC. 261

262 Conclusions

263 In summary, our results support the hypothesis that RDOC in the deep ocean is 264 dominated by structurally recalcitrant RDOC_t, i.e. at least 84.5% of RDOC in this case. The LDOC and RDOC_c account for a relatively minor fraction of the bulk 265 266 RDOC (<5.0% for LDOC and <10.5% for RDOC_c) and can be rapidly used. The high proportion of RDOC_t implies that the capacity of the ocean to store carbon via MCP is 267 not constrained by a dilution threshold. Further investigations are required to 268 269 understand if the current pool of oceanic RDOC could increase, potentially counteracting the anthropogenically induced increase in atmospheric CO₂. 270

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279 **Conflict of interest**

All authors have seen and approved the final version submitted, and declared noconflict of interest.

282 Supporting Information

Additional Supporting Information may be found in the online version of this article.

284 Supplementary Methods. Descriptive information regarding the methods used

within this study.

286 Supplementary Tables and Figures

Table S1. Matrix effect on estimating the percentage of $RDOC_t$ in SPE DOC (f_t). Here, 287 we consider the matrix effect caused by the incubation process (i.e., change of signal 288 289 intensity from DOC component during incubation experiments) and concentrated 290 treatments (i.e., change of signal intensity due to DOC concentrated treatments). The measurement uncertainty caused by matrix effect (i.e., σ_{ME}) is calculated as the 95th 291 percentile of signal intensity difference by comparing 1X initial and 1X final FT-ICR-292 293 MS, 1X initial and 5X initial FT-ICR-MS, 5X initial and 5X final FT-ICR-MS, 294 respectively. We then investigate the matrix effect on the estimation of f_t according to equation (17) in Supplementary Methods. 295

Figure S1. Histogram distribution of measurement variability of FT-ICR-MS fingerprints of DOC by assembling all molecules together. The measurement variability for each molecule is calculated as the standard deviation of its relative intensities from replicates. The average measurement variability FT-ICR-MS (i.e., σ_m) is chosen as the 95th percentile of this distribution for the downstream propagation uncertainty analysis.

302 Figure S2. Comparison of signal intensity of FT-ICR-MS to assess DOC composition

303 change during the incubation experiments (A, B, D, E) and the matrix effect due to

304 DOC concentrated treatments (C, F). The diagonal line is shown in the red color in 305 each figure.

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Tables

Meaning	Symbol	Standard value	Source
The percentage of RDOC _t in SPE DOC pool	f_t	-	-
The percentage of RDOC _c in SPE DOC pool	f_c	-	-
The percentage of LDOC in SPE DOC pool	f_l	-	-
The percentage of consumed RDOC _c in relation to initial RDOC _c	C_{c}	-	-
The percentage of consumed LDOC in relation to initial LDOC	c_l	-	-
The percentage of consumed DOC in relation to SPE DOC in controls	$c_{1\mathrm{X}}$	2.1%	Arrieta et al. (2015a)
The percentage of consumed DOC in relation to SPE DOC in 5-fold	$c_{5\mathrm{X}}$	4.1%	Arrieta et al. (2015a)
concentrated treatments			
The new DOC produced per unit of LDOC consumed	p_l	12.5%	(Stoderegger and Herndl, 1998)
The new DOC produced per unit of RDOC _c consumed	p_c	12.5%	(Stoderegger and Herndl, 1998)

Table 2. Propagation of measurement uncertainty of FT-ICR-MS on estimating the percentage of RDOC_t in SPE DOC (f_t). To derive the propagation of measurement uncertainty from equation (17) in **Supplementary Methods**, we set $c_{1X} = 2.1\%$, $c_{5X} =$ 4.1%, $p_l = 12.5\%$, $p_c = 12.5\%$ as standard values and calculate the measurement uncertainty of FT-ICR-MS (σ_m) using the replicates (see **Supplementary Methods**).

434

Dorometer	Estimation	Propagation				
I arameter	(%)	uncertainty (%)				
Exp. O						
c_l	54.9	3.3				
C_c	25.2	3.4				
f_l	4.4	0.3				
f_c	9.1	1.2				
f_t	86.6	1.2				
Exp. P						
c_l	47.8	3.2				
C_c	21.9	4.6				
f_l	5.0	0.3				
f_c	10.5	2.2				
f_t	84.5	2.2				
$egin{array}{c} f_c \ f_t \end{array}$	10.5 84.5	2.2 2.2				

436 Table 3. Sensitivity analysis for each parameter in estimating the percentage of

437 RDOC_t in SPE DOC (f_t).

438

Parameter	Standard Value (%)	Parameter range (%)	f_t range (%)
Exp. O			
c_l	54.9	51.6~58.1 ^a	86.3~86.9
C_{c}	25.2	21.9~28.6	85.2~87.6
$c_{1\mathrm{X}}$	2.1	1.4~2.8	84.6~88.5
$c_{5\mathrm{X}}$	4.1	2.6~5.6	78.8~93.2
p_l	12.5	0.0~25.0	85.9~87.1
p_c	12.5	0.0~25.0	85.0~87.7
Exp. P			
c_l	47.8	44.6~51.0	84.2~84.8
C_{c}	21.9	17.3~26.5	81.7~86.3
c_{1X}	2.1	1.4~2.8	82.6~86.5
$c_{5\mathrm{X}}$	4.1	2.6~5.6	76.3~91.9
p_l	12.5	0.0~25.0	83.7~85.1
p_c	12.5	0.0~25.0	82.7~85.8

439 ^a: the ranges of parameter c_l and c_c are set to be one standard deviation (calculated 440 from the propagation uncertainty analysis) away from their mean values.

441 Figure Legends

442 Figure 1. Schematic representation of the transformation of deep oceanic DOC components in incubation experiments with different DOC concentration treatments. 443 The SPE DOC in the deep ocean is composed of structurally recalcitrant RDOC_t, 444 recalcitrant DOC due to dilution (RDOC_c) and labile DOC ready for microbial 445 utilization (LDOC). The natural bacterial communities collected in the deep ocean are 446 exposed to ambient and 5-fold concentrations of natural DOC collected from their 447 448 original locations. (A) In controls, a fraction of LDOC is consumed by bacteria to 449 generate new DOC and respire back to CO₂ or assimilate to POC. Both RDOC_c and RDOC_t are not consumed during this process. (B) In the 5-fold concentrated 450 treatments, except the LDOC, a fraction of RDOC_c components whose concentrations 451 452 become sufficient for microbial utilization are also utilized and contribute to the SPE 453 DOC decrease. Detailed description of symbols can be seen in Table1.

Figure 2. Expected DOC utilization as a function of DOC enrichment when the RDOC is composed of intrinsically labile DOC (LDOC and $RDOC_c$) and $RDOC_t$. The simulation is the same as in Arrieta et al. (Arrieta et al., 2015a) but assumes that 0% (A), 10% (B), 50% (C), and 90% (D) of RDOC is $RDOC_t$. The 40-day time is denoted as a dotted line. The result of (A) is the same as Figure 1B in Arrieta et al. (Arrieta et al., 2015a) assuming that the DOC utilization is solely limited by concentration. Detailed simulation can be seen in Supplementary Methods.

Figure 3. Two possible interpretations for the small proportion of SPE DOC consumption in a 40-day incubation. (A) A large fraction of the SPE DOC is intrinsically labile and complete consumption of it would take a long time. This is what has been proposed in Arrieta et al. (2015a). (B) According to our estimates, there is a small amount of intrinsically labile DOC in the SPE DOC and it is rapidly utilized.

466 **Figure 1.**

A) Controls



B) 5-fold concentration experiments



Figure 2.



Figure 3. 471

А Dilution hypothesis

В	3 Structural recalcitrance hypothesis												
	Bulk	DOC ((t ₀)	В	ulk I	000	C (t ₁))	В	ulk E	000	$C(t_2)$	
					Tin	ne							_

- = intrinsically labile DOC (LDOC+RDOC_c)
 = intrinsically recalcitrant RDOC (RDOC_t)
 = consumed DOC

1 Supplementary Methods

2 Data description

The dataset (http://digital.csic.es/handle/10261/111563) used in this study is taken 3 4 from DOC enrichment experiments (Arrieta et al., 2015a), in which the dilution hypothesis was tested against whether bacterial growth increased with increasing 5 DOC concentrations. The FT-ICR-MS fingerprints of SPE DOC at the beginning and 6 7 at the end of the experiments in controls (1X DOC) and in concentrated samples (5X 8 DOC) were also collected, which allow us to investigate the degradation of thousands 9 of different molecules simultaneously. The FT-ICR-MS datasets, together with the 10 change of SPE DOC concentrations, are used in our study.

11 Model description

12 As shown in the conceptual model (Figure 1), DOC is assumed to be composed of structurally recalcitrant DOC (RDOC_t), recalcitrant DOC due to dilution (RDOC_c) 13 14 and labile DOC ready for microbial utilization (LDOC), with their percentage in DOC pool being f_t , f_c , and f_l , respectively (see **Table 1** for details of variable indicators). 15 Here, LDOC compounds are labile, whose concentrations are above the 16 17 corresponding microbial uptake thresholds in both controls and 5-fold concentrated treatments. RDOC_c compounds are also intrinsically labile, but their concentrations 18 19 are below the corresponding uptake thresholds in controls and above those in 5-fold 20 concentrated treatments. We have

$$f_t + f_c + f_l = 1. (1)$$

21 In controls, only LDOC is utilized, while RDOC_c is not utilized because its 22 concentration is below the microbial uptake thresholds. A fraction of LDOC 23 consumed by bacteria is regenerated to new DOC with the ratio of the regenerated 24 DOC to consumed LDOC being p_l . The rest, i.e., 1 - p_l of consumed LDOC, is respired to CO₂ or assimilated to POC. The remaining components from RDOC_c and 25 RDOC_t pools are not utilized either because of the dilution limit (i.e., RDOC_c) or 26 27 structurally recalcitrance (i.e., RDOC_t). The process describing transformation of 28 DOC components in controls can be represented as follows:

$$f_t + f_c + f_l(1 - c_l) + f_l c_l p_l = 1 - c_{1X},$$
(2)

where c_{1X} denotes the percentage of consumed DOC in relation to SPE DOC in controls and c_l represents the ratio of consumed LDOC to bulk LDOC. By subtracting equation (2) from (1), we have following equation for controls:

$$f_l c_l (1 - p_l) = c_{1X}.$$
 (3)

In 5-fold concentrated treatments, a fraction of $RDOC_c$ compounds become bioavailable because their concentrations are high enough to support the bacterial metabolism. Therefore, except for the consumption of LDOC, which is supposed to be the same as that in controls, $RDOC_c$ is also utilized. Therefore, for 5-fold concentrated treatments,

$$f_t + f_c(1 - c_c) + f_c c_c p_c + f_l(1 - c_l) + f_l c_l p_l = 1 - c_{5X},$$
(4)

where c_{5X} denotes the percentage of consumed DOC in relation to SPE DOC in 5-fold concentrated treatments, c_c represents the ratio of consumed RDOC_c to bulk RDOC_c, and p_c is the new DOC produced per unit of RDOC_c consumed. By subtracting equation (4) from (2), we have

$$f_c c_c (1 - p_c) = c_{5X} - c_{1X}.$$
(5)

41 Equations (3) and (5) explicitly link the degradation of LDOC and RDOC_c 42 components with the SPE DOC utilization. Then f_i and f_c can be derived from these 43 two equations:

$$f_l = \frac{c_{1X}}{c_l(1 - p_l)}.$$
 (6)

$$f_c = \frac{c_{5X} - c_{1X}}{c_c (1 - p_c)}.$$
(7)

Finally, the percentage of RDOC_t in SPE DOC can be estimated from equation (1) as:
45

$$f_t = 1 - \frac{c_{1X}}{c_l(1 - p_l)} - \frac{c_{5X} - c_{1X}}{c_c(1 - p_c)}.$$
(8)

46 The required parameters, i.e., c_{1X} , c_{5X} , p_c , p_l , c_c , and c_l , are estimated in the following 47 section.

48 **Parameter estimation**

49 c_l and c_c : Following the idea proposed by Arrieta and coworkers, a utilization index is 50 derived from FT-ICR-MS data, which is defined by subtracting the remaining relative 51 contribution at the end of controls from the initial relative contribution for each mass 52 peak, as follows

$$I^{l} = \frac{H^{0} - (1 - c_{1X})H^{1X}}{H^{0}},$$
(9)

where H^0 and H^{1X} are the normalized FT-ICR-MS signal of SPE DOC at the beginning and at the end of controls, respectively. The term $(1 - c_{1X})H^{1X}$ can be interpreted as the DOC normalized relative peak magnitude (Lechtenfeld et al., 2014). A positive I^l indicates the net utilization (bacterial usage minus regeneration) of the corresponding intrinsically labile compound. When I^l is negative, it means that the regeneration of this compound is more than its consumption. As a result, I^l provides the lower boundary of the compound production but no information about the decrease of this compound. Therefore, c_l is estimated by averaging the utilization index I^l on molecules whose index values are positive, i.e.,

$$c_l > c_l^* = \frac{1}{N_l} \sum_{l^l > 0} I^l,$$
(10)

62 where N_l denotes the number of molecules whose index values is positive.

63 Compared with controls, a fraction of $RDOC_c$ compounds are utilized in 5-fold 64 concentrated treatments and the consumption of LDOC is supported to be the same. 65 Thus, we can also identify the utilization index of $RDOC_c$ by comparing the relative 66 contribution of FT-ICR-MS at the end of controls and 5-fold concentrated treatments:

$$I^{c} = \frac{(1 - c_{1X})H^{1X} - (1 - c_{5X})H^{5X}}{(1 - c_{1X})H^{1X}},$$
(11)

67 where H^{5X} is the normalized FT-ICR-MS signal of SPE DOC at the end of 5-fold 68 concentrated treatments. Let N_c be the number of mass peaks with positive values of 69 I^c , we calculate c_c as

$$c_c > c_c^* = \frac{1}{N_c} \sum_{I^c > 0} I^c.$$
 (12)

The terms c_l^* and c_c^* in equations (10) and (12) could possibly provide the lower limits of c_l and c_c . This is because 1) the defined utilization index (i.e., I^l and I^c) calculates the proportion of consumed labile DOC to SPE DOC, which would be smaller than the ratio of consumed labile DOC to labile DOC (i.e., c_l and c_c); 2) due to the dynamical transformation process of DOC, the observed utilization index actually represents a net decrease of intrinsically labile compounds that subtract the DOC regeneration from its consumption.

*c*_{1X} *and c*_{5X}: Because no DOC consumption data are reported for Stations O and P (Arrieta et al., 2015a), the DOC consumption information from Stations K, L, and N are used as the proxies of that for both Stations O and P. The DOC consumption data from Station M is excluded because they are anomalous due to the observation of a second phase of intense growth (Arrieta et al., 2015a). The percentage of consumed SPE DOC is approximated to be 2.1±0.7% for c_{1X} (average ± SE) and 4.1±1.5% for c_{5X} (average ± SE).

84 p_c and p_i : A portion of intrinsically labile DOC consumed by bacteria is converted 85 into bacterial-derived DOC, representing approximately 25% of the respired carbon (Stoderegger and Herndl, 1998). Considering that the bacterial respired carbon 86 87 amounts to approximately 35.0% ~ 99.0% of the assimilated organic carbon (Del 88 Giorgio and Cole, 1998), we have the ranges of the newly produced DOC per unit of LDOC or RDOC_c consumed from 0.0% to 25.0%, and set the mean value of their 89 90 range (i.e., 12.5%) as their standard value. The bacterial-derived DOC can be 91 progressively used during the incubation, which leads to an even lower amount of 92 newly produced DOC per unit of consumed intrinsic DOC in reality.

93 Error propagation analysis

94 Uncertainty propagation for f_l , f_c and f_t derived from the measurement variability of 95 FT-ICR-MS fingerprints of SPE DOC is estimated as follows. Except for the controls 96 at the beginning of the experiments, each scenario has two or three replicates for the 97 FT-ICR-MS fingerprints of SPE DOC. We first obtain the distribution of measurement variability by assembling the standard deviation of intensity for each molecule across all scenarios and determine its 95% confidence level as the measurement variability σ_m . The confidence level is chosen to allow us to investigate an exaggerated uncertainty of measurement variability on the estimation of f_t . Applying the error propagation formula, which is the function of measurement variability (Glover et al., 2011), to equation (10):

$$\sigma^{2}(c_{l}) = \frac{1}{N_{l}^{2}} \sum_{I^{l} > 0} (1 - I^{l})^{2} \left\{ \frac{\sigma^{2}(H^{1X})}{(H^{1X})^{2}} + \frac{\sigma^{2}(H^{0})}{(H^{0})^{2}} \right\}$$
(13)

104 Based on equation (6), the error for f_l propagated from the c_l is estimated as

$$\sigma^{2}(f_{l}) = \left(\frac{c_{1X}}{1 - p_{l}}\right)^{2} \frac{1}{c_{l}^{4}} \sigma^{2}(c_{l}).$$
⁽¹⁴⁾

105 Assuming that all molecules share the same measurement variability, i.e., $\sigma_m^2 = \sigma^2(H^{1X}) = \sigma^2(H^0)$, by plugging equation (13) into equation (14), we have

$$\sigma^{2}(f_{l}) = \left(\frac{c_{1X}}{1 - p_{l}}\right)^{2} \frac{\sigma_{m}^{2}}{N_{l}^{2} c_{l}^{4}} \sum_{I^{l} > 0} (1 - I^{l})^{2} \left\{\frac{1}{(H^{1X})^{2}} + \frac{1}{(H^{0})^{2}}\right\}.$$
(15)

107 Similarly, the error for f_c propagated from the FT-ICR-MS fingerprints of SPE DOC is 108 estimated as

$$\sigma^{2}(f_{c}) = \left(\frac{c_{5X} - c_{1X}}{1 - p_{c}}\right)^{2} \frac{\sigma_{m}^{2}}{N_{c}^{2} c_{c}^{4}} \sum_{I^{c} > 0} (1 - I^{c})^{2} \left\{\frac{1}{(H^{5X})^{2}} + \frac{1}{(H^{1X})^{2}}\right\}.$$
(16)

109 With the simplistic assumption that the error propagations of f_l and f_c are independent,

110 the error propagation of f_t could be represented as

$$\sigma^2(f_t) = \sigma^2(f_l) + \sigma^2(f_c). \tag{17}$$

111 Sensitivity analysis

112 Analyses are conducted to investigate the sensitivity of f_t to each parameter over wide 113 ranges. Six parameters in equation (8) are considered. Specifically, the ranges of c_l and c_c are set to be one standard deviation, which is calculated from the error propagation analysis, away from their base values. The ranges of c_{1X} and c_{5X} are set to be one standard deviation away from their base values as well. Both p_l and p_c are varied from 0.0% to 25.0% to cover most likely range.

118 Calculate the consumption of SPE DOC as a function of DOC enrichment

In these simulations, we explore how the utilization rate of SPE DOC changes as a 119 function of DOC enrichment given that DOC is composed of intrinsically labile DOC 120 (LDOC and RDOC_c) and RDOC_t. Our simulation is based on the simplification that 121 122 all components in the RDOC_c pool are available for microbial utilization in 123 concentrated experiments, which could possibly lead to overestimation of the SPE DOC utilization rate. Consistent with simulations in Arrieta et al. (2015b), we 124 consider four different DOC concentration levels (controls and the corresponding 125 two-, five- and ten-fold concentrations of nature DOC). Let x(t) be the concentration 126 of intrinsically labile compounds at day t. Assuming that the utilization rate is a 127 128 function of concentration and cell abundance, we have

$$\frac{dx(t)}{dt} = x(t)K_sS,\tag{18}$$

129 in which K_s is the Monod substrate affinity constant and *S* the maximum cell 130 abundances observed in the experiment. The solution of equation (18) can be 131 represented as:

$$x(t) = x(0)e^{-K_s St}.$$
 (19)

132 Let z(0) be the initial DOC concentration:

$$x(t) = z(0)\gamma e^{-K_s S t},\tag{20}$$

in which $\gamma = f_l + f_c$. z(0) is set to 30 (in arbitrary units), and the cell abundances *S* for controls, two-, five- and ten-fold concentration treatments are 55, 67, 97, and 139 ×10⁶ prokaryotes L⁻¹, respectively (Arrieta et al., 2015b). The substrate affinity constant *S* is calculated by reconciling the proportion of consumed SPE DOC at t = 40in equation (20) to that of the observed values in the controls (i.e., c_{1X}). Therefore,

$$K_s = \frac{\ln(1 - c_{1X}/\gamma)}{-40 \times 55 \times 10^6}.$$
(21)

By plugging (21) into (20), the proportion of consumed SPE DOC at *t* day is given as:

$$c(t) = \frac{x(t) - x(0)}{z(0)} = \gamma(e^{-K_s S t} - 1).$$
(22)

139 In our simulation, γ ranges from 0 to 1. All SPE DOC is RDOC_t when $\gamma = 0$ and is

140 RDOC_c/LDOC when $\gamma = 1$. We also conducted sensitivity analysis to investigate the

141 change of c(t) by varying c_{1X} from 1.4% to 2.8% in equation (21).

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1 Supplementary Tables

Table S1. Matrix effect on estimating the percentage of $RDOC_t$ in SPE DOC (f_t). Here, 2 3 we consider the matrix effect caused by the incubation process (i.e., change of signal intensity from DOC component during incubation experiments) and concentrated 4 5 treatments (i.e., change of signal intensity due to DOC concentrated treatments). The measurement uncertainty caused by matrix effect (i.e., σ_{ME}) is calculated as the 95th 6 7 percentile of signal intensity difference by comparing 1X initial and 1X final FT-ICR-MS, 1X initial and 5X initial FT-ICR-MS, 5X initial and 5X final FT-ICR-MS, 8 9 respectively. We then investigate the matrix effect on the estimation of f_t according to 10 equation (17) in Supplementary Methods.

		Exp. O	Exp. P
$\sigma_{\!M\!E}$	1X initial, 1X final	0.084	0.073
	1X initial, 5X initial	0.074	0.040
	5X initial, 5X final	0.066	0.068
Max. of σ_{ME}		0.084	0.073
Range of f_t , %		82.1~89.1	77.8~87.9

11 Supplementary Figures



Figure S1. Histogram distribution of measurement variability of FT-ICR-MS fingerprints of DOC by assembling all molecules together. The measurement variability for each molecule is calculated as the standard deviation of its relative intensities from replicates. The average measurement variability FT-ICR-MS (i.e., σ_m) is chosen as the 95th percentile of this distribution for the downstream propagation uncertainty analysis.





Figure S2. Comparison of signal intensity of FT-ICR-MS to assess DOC composition change during the incubation experiments (A, B, D, E) and the matrix effect due to DOC concentrated treatments (C, F). The diagonal line is shown in the red color in each figure. The close to the diagonal line means that the overall FT-ICR-MS signal pattern is similar among different treatments.