

Special Feature for Ecological Research

Running head: Microbial dynamics and culturability

Title:

**Idea Paper: Predicting culturability of microbes from
population dynamics under field conditions**

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21 **Abstract:**

22 Isolation and cultivation of microbes from environmental samples have been
23 fundamental and important for species identification and investigating functions and
24 ecology of target microbes. While cultivation and isolation of microbes are not easy,
25 the natural environment can “culture” any endemic microbes, and thus key
26 information for culturing and isolating microbes may be encoded in the natural
27 population dynamics of microbes. In the present paper, I present the idea that
28 culturability of microbes may be inferred by quantifying dynamic properties of
29 microbes using nonlinear time series analysis, empirical dynamic modeling (EDM).
30 To briefly demonstrate the idea, I analyzed high-frequency, quantitative microbial
31 time series obtained for experimental rice plots. I selected bacterial phyla that
32 included sufficient numbers of microbial taxa, and analyzed 398 microbial taxa using
33 empirical dynamic modeling. The nine phyla analyzed generally followed a similar
34 pattern: many microbial taxa fell into the “Simple” dynamics category, and a small
35 proportion of taxa were categorized in “Simple but nonlinear” or “Nearly random”
36 dynamics categories. The present analysis suggested that many microbes in the study
37 system might be cultivated by modifying a relatively small number of conditions.
38 However, the present idea as well as the result is preliminary, and more precise
39 taxonomic information (i.e., species-level identification) and a culturability dataset
40 will help to validate the idea. If the present idea was found to be valid, *a priori*
41 evaluation of the culturability of microbes would become possible, which would
42 avoid unnecessary costs (labor, time and money) of attempts to cultivate microbes.

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44 **Keywords:** Culturability; DNA; Empirical dynamic modeling; Microbes; Time series

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46 **Research question**

47 Isolation and cultivation of microbial species from environmental samples have
48 been fundamental for identifying microbial species and investigating functions and
49 ecology of a target microbial species. However, unfortunately, the majority of
50 microbes in the environment cannot be cultured easily (e.g., Amann et al. 1995), and
51 improving the recovery of microbes from environmental samples is an important but
52 difficult and labor-, time- and money-consuming work. Therefore, if *a priori*
53 evaluations of the culturability of microbes would become possible, it would
54 contribute to avoiding unnecessary costs and improving the recovery of microbes
55 from environments. For example, a cultivation design that includes a large number of
56 combinations of nutrients in media might not be necessary if the target microbes
57 were classified as “easily cultured”.

58

59 **Value**

60 *A priori* evaluations of the culturability of microbes would contribute to reducing
61 potential costs of cultivation. Also, a framework for quantifying culturability would
62 enable analysis of the relationship between the culturability and genetic factors,
63 which could potentially contribute to understanding why particular microbial
64 species are difficult to culture while others are not.

65

66 **Relevant hypothesis (and approaches)**

67 While isolation and cultivation of microbes are not easy, natural environments can
68 “culture” any endemic microbes, and thus key information for isolating and
69 culturing microbes may be encoded in microbes’ interactions with natural habitats
70 and resultant population dynamics in nature (or under field conditions). Indeed, the
71 diffusion chamber approach, which allows exchanges of chemicals between media
72 and their natural habitat, can simulate the natural environment, and improves the
73 recovery of some microbes from environment (Bollmann et al. 2007; Kaeberlein et al.
74 2002). These findings implied that microbes in nature respond to biotic/abiotic
75 variables in their environment and the interactions among microbes and the
76 environment drive their population dynamics. Therefore, conversely, natural
77 population dynamics of microbes contain integrated information on biotic and
78 abiotic variables and their interactions in the environment. The question is, however,
79 how we can extract the information encoded in the population dynamics of microbes.

80 Information encoded in population dynamics can be extracted using empirical
81 dynamic modeling (EDM) that is based on attractor reconstruction and designed for
82 analyzing nonlinear dynamics (e.g., Sugihara et al. 2012). Core tools of EDM are the
83 simplex projection (Sugihara and May 1990) and S-map (Sugihara 1994), which
84 enable quantifying (1) the best embedding dimension and (2) state-dependence of
85 time series data. Briefly, the best embedding dimension (denoted by E) includes
86 information on how many potential variables might be involved in the process (i.e.,
87 complexity or dimensionality), and state-dependency (quantified by the nonlinear
88 weighting parameter, θ) includes information on how state dependent the process is.
89 A previous study demonstrated that E and θ are effective indices to distinguish
90 random environmental fluctuations from low dimensional, nonlinear dynamics of

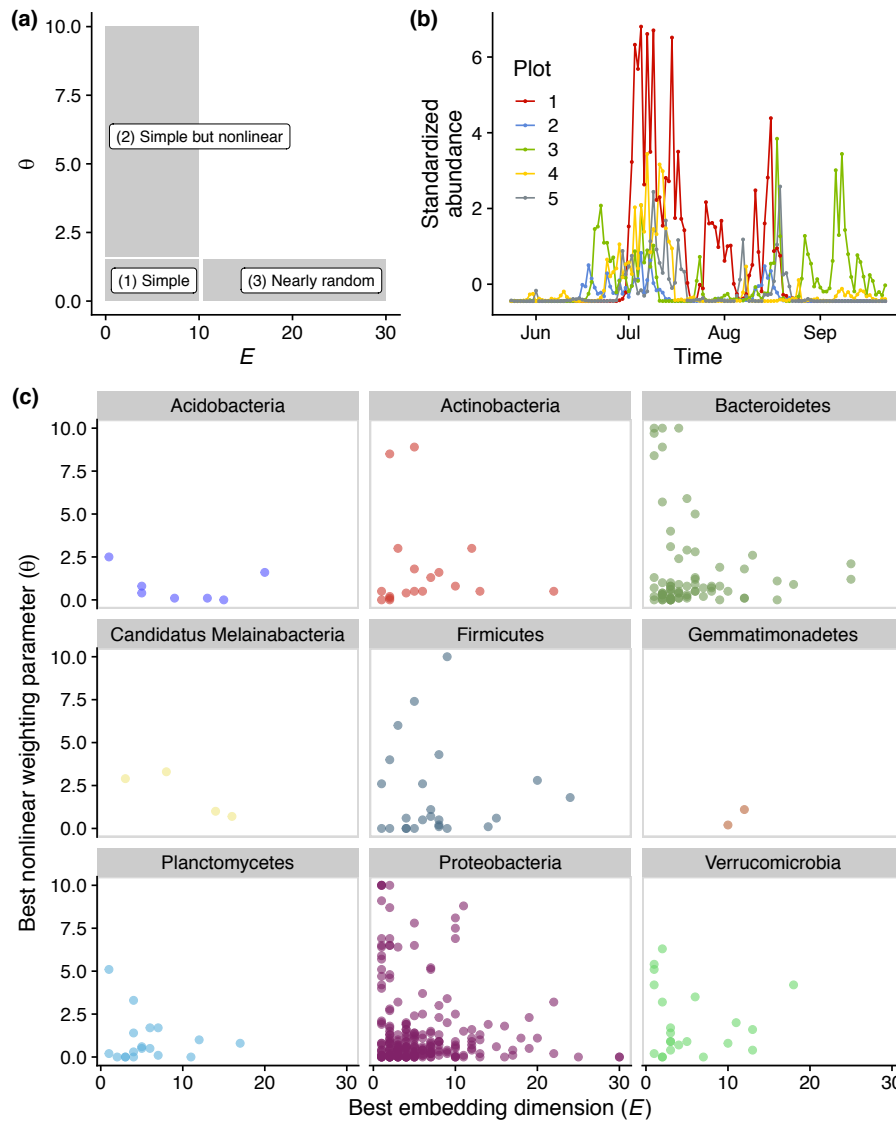
91 organisms (Hsieh et al. 2005). Detailed information on how simplex projection and S-
 92 map are performed is described in previous studies (Sugihara 1994; Sugihara and
 93 May 1990). Also, an overview of EDM is described in Chang et al. (2017).

94

95 **New research idea**

96 In the present paper, I present the idea that culturability of microbes may be
 97 inferred using population dynamics of microbes and EDM. Specifically, I expect that
 98 the complexity (E) and state-dependency (θ) of population dynamics would provide
 99 information on the culturability of microbes.

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103 **Figure 1** | (a) A conceptual explanation of the relationships between E and θ and dynamics
 104 properties. Note that the thresholds between the categories are arbitrarily determined. (b) An
 105 example of time series analyzed in the present study. Changes in abundance (i.e., estimated DNA
 106 copy numbers) of a microbial amplicon sequence variant (ASV) belonging to Bacteroides are
 107 shown. Different colors indicate different rice plots where water samples were taken. (c) The
 108 relationships between E and θ and microbial taxa. Each panel indicates a particular phylum, and
 109 each point indicates the best E and θ of each ASV.

110 **How to answer the question through the research approach**

111 I expect three possible dynamics regarding the combinations of E and θ by
112 following Hsieh et al. (2005): (1) “Simple” (i.e., low dimensional dynamics), (2)
113 “Simple but nonlinear”, and (3) “Nearly random” dynamics (Fig. 1a). First, if the
114 population dynamics of a target species show small E and θ , then the number of
115 potential variables might be small and the species responds to cultivation conditions
116 relatively linearly (“Simple” dynamics). In this case, the target microbe might be
117 cultivated by examining a relatively small number of cultivation conditions. Second,
118 if the population dynamics show small E but large θ , then the number of potential
119 variables might be small but the growth of the species is state dependent (“Simple
120 but nonlinear” dynamics). In this case, the target microbe could respond to
121 cultivation conditions nonlinearly, and thus careful considerations of cultivation
122 conditions would be required. Third, if the population dynamics show large E but
123 small θ , then the dynamics are nearly random and the target microbe might have
124 randomly immigrated from outside the system (“Nearly random” dynamics).

125 To briefly demonstrate the idea, I analyzed highly frequent, quantitative microbial
126 time series taken from experimental rice plots established at the Center for Ecological
127 Research, Kyoto University, Japan (Ushio, unpublished). Briefly, water samples were
128 collected from five rice plots every day from 23 May to 22 September in 2017 (610
129 samples in total), and filtered using filter cartridges (pore size = $\phi 0.22 \mu\text{m}$). Then,
130 DNAs were extracted, amplified a portion of the 16S rRNA genes and sequenced
131 using Illumina MiSeq. In the library preparation process, artificial standard DNAs
132 with known concentrations were included to estimate the copy numbers of microbial
133 DNAs (Ushio 2019). The sequences generated were analyzed using the amplicon
134 sequence variant (ASV) approach (Callahan et al. 2016). I selected bacterial phyla that
135 are abundant and contain sufficient numbers of ASVs, which resulted in 398 bacterial
136 ASVs belonging to nine phyla. An example of the microbial time series is shown in
137 Fig. 1b. For the time series selected, simplex projection and S-map were applied to
138 determine the best E and θ using rEDM packages (Ye et al. 2018) of R (R Core Team
139 2019). The whole dataset is being analyzed for different purposes and thus is
140 currently not publicly available.

141 The nine phyla analyzed followed a similar pattern (Fig. 1c): many bacterial ASVs
142 fall into the “Simple” category, and a relatively small proportion of ASVs were
143 categorized as “Simple but nonlinear” or “Nearly random”. This result might
144 indicate that, among the selected bacterial ASVs, many microbes could be cultivated
145 by modifying a relatively small number of conditions, which is contrary to the
146 current general consensus that most environmental microbes are difficult to cultivate
147 and isolate (Amann et al. 1995; but see Martiny 2019). However, technical issues (e.g.,
148 biases generated by an amplicon sequencing approach) may partly contribute to this
149 pattern, and thus careful interpretations of the results are necessary (see debates in
150 the following references: Martiny 2019; Steen et al. 2019).

151 Here I list data that are required to verify my idea more thoroughly. First,
152 microbes should be identified at least at the species level. This is because the
153 culturability would be a property of a species, and thus genus or family level
154 identification is insufficient. The present data analyzed were generated by a short-
155 read marker-gene (amplicon) analysis, and thus the phylogenetic resolution was

156 often not sufficient. Recently introduced long-read sequencing (e.g., full-length 16S
157 rRNA sequencing) would be more appropriate to achieve this goal (e.g., MinION by
158 Oxford Nanopore or Sequel by PacBio). Species-level dynamic properties would be
159 useful to classify and predict previously ignored aspects of microbial properties.
160 Second, accurate information on which microbial species and how the microbial
161 species is cultivated is required. Information on which microbial species is cultivated
162 would be available in public databases, but developing the cultivation method often
163 requires detailed tuning and thus the method as well as its complexity would be
164 difficult to fully document/digitalize. Compiling fundamental information about
165 cultivation methods (e.g., temperature and resources) would help to develop a
166 culturability database, which could also contribute to understanding of the potential
167 habitat range of microbes in nature. Together, if the present idea were validated by
168 overcoming the above limitations, dynamic properties of microbes could be linked to
169 culturability and *a priori* evaluation of the culturability of microbes would become
170 possible.

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179 **Competing interests:** The author has no competing interests.

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