



## Towards meaningful scales in ecosystem microbiome research

Journal:	<i>Environmental Microbiology and Environmental Microbiology Reports</i>
Manuscript ID	EMI-2020-1477.R1
Journal:	Environmental Microbiology
Manuscript Type:	EMI - Correspondance
Date Submitted by the Author:	n/a
Complete List of Authors:	Dini-Andreote, Francisco; The Pennsylvania State University - Main Campus, Plant Science Kowalchuk, George; Utrecht University, Ecology and Biodiversity Prosser, James; University of Aberdeen, School of Biological Sciences Raaijmakers, Jos; NIOO-KNAW, Microbial Ecology
Keywords:	species distribution, ecological patterns, microbial communities, microbial ecology

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Manuscripts

1 **Towards meaningful scales in ecosystem microbiome research**

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3 **Article type.** Correspondence

4 **Running title.** The problem of scale in ecosystem microbiomes

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## 20 **Originality-Significance Statement**

21 This is a critical article that articulates on the main issues to be considered in properly addressing the  
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## 28 **Summary**

29 Studies of microbial communities in natural ecosystems have been generally focused on mapping patterns  
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31 microbial diversity and distribution patterns, such census studies often lack a meaningful and explicit  
32 definition of scale. Here, we discuss the importance of scale in environmental microbiology assessments  
33 and consider how patterning ecology can be redirected towards advancing concept and theory formation  
34 in ecosystem microbiome research.

## 36 **Introduction**

37 Increasing attention is being given to mapping 'global' and/or 'cross-continental' patterns of microbial  
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39 *al.*, 2016; Delgado-Baquerizo *et al.*, 2018; Baham *et al.*, 2018). The majority of these studies are founded  
40 on the notion that cataloguing large-scale diversity and distributional patterns of microbes will improve  
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42 regional and global ecosystem processes; e.g. carbon source-sink dynamics and positive or negative

43 microbially mediated effects on global warming (Crowther *et al.*, 2019). These studies have increased  
44 knowledge of, for instance, the ubiquitous versus rare distribution of taxa and functions. However, less  
45 attention has been given to assessing the reliability or value of the often-arbitrary scales used. We argue  
46 that without an explicit consideration of scale and its limitations, extrapolating findings obtained from  
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48 and predict the functional capacity of environmental microbiomes. To stimulate debate on this challenging  
49 topic, we consider three main categorical issues associated with terminology, concept formation and  
50 theory construction.

51

## 52 **The terminology problem**

53 The use of 'global/regional/cross-continental' and 'pattern/distribution/biogeography' terminologies has  
54 been broadly used to refer to large-scale surveys of ecosystem microbiomes. It is important to realize that  
55 data obtained from these assessments are derived from a limited set of samples collected at particularly  
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58 not stated explicitly, it is often implied that such datasets are in fact representative of all sites at that scale,  
59 including non-sampled sites for which information is unavailable. In defence of this approach, it could be  
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61 thousands of kilometres to continental scales. In addition, meaningful or not, correlational outcomes are  
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66 since it is impossible to cover at a fine-scale potential site variation across broad spatial scales (i.e., absence

67 of evidence), whatever patterns emerge from a limited (often scattered) collection of samples will hold for  
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70 explicitly defined/optimal sampling plan. Eliminating these issues – or at least diminishing their importance  
71 – requires the use of proper language, the explicit definition of scale and its limitations, and proper  
72 accounting for both replication and assessment of within-site variation when developing broad-scale  
73 sampling designs.

74

### 75 **Concept formation**

76 The debate about patterns and scale in ecology is not new (see Chave, 2013) but these concepts remain  
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78 (1992) and Chave (2013), the main objective of patterning ecology is to inform theory. Thus, it is critically  
79 important that patterns have some degree of repetition and replicability if they are to have any level of  
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81 therefore essential to infer both small-scale and large-scale patterns, as is the incorporation of nested  
82 designs in accordance with scale to account for autocorrelation (Tedersoo, 2017). Moreover, reliable and  
83 informed prediction can only be achieved through sufficient longitudinal data and/or by clear elucidation  
84 of mechanisms that underlie patterns (Levin, 1992), both of which require a meaningful and explicit  
85 definition of scale. Importantly, the relative influences of distinct mechanisms underpinning taxa  
86 distributions are known to vary as a function of scale, so-called scale-dependency (Dini-Andreote *et al.*,  
87 2015; Chase *et al.*, 2018). Thus, starting with a clean slate and going from processes to patterns, and not  
88 backwards, may represent a more fruitful strategy to advance mechanistic understanding of microbial  
89 distributional patterns.

90 *“The description of patterns is the description of variation, and the quantification of variation requires the*  
91 *determination of scales”* (Levin, 1992). This raises a critical limitation, which is the extent to which patterns  
92 or variation that emerge from spatially limited designs and lack of temporal components preclude many  
93 studies from appropriate replication. In other words, are these assessments providing global patterns of  
94 biological diversity or intrinsic variation within sample sets collected at sites scattered at a global scale?  
95 For example, samples may be collected across a range of environmental variable(s), and these variable(s)  
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97 ‘how’ and ‘the extent at which’ these variable(s) might physiologically constrain or promote the abundance  
98 or activities of specific taxa. Considering pH, which is often found to be the highest correlate to community  
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101 In addition, millimetre-scale gradients in O<sub>2</sub> in soil particles have been shown to drive community  
102 divergences that can be analogous to those resulting from selection imposed by large-scale abiotic  
103 variables (e.g., pH, salinity, etc.) or anthropogenic disturbances (agricultural practices, land-use conversion,  
104 etc.) at a broader scale (Konopka, 2009; Vos *et al.*, 2013). Moreover, it is worth noting that there are  
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106 relate to the mechanisms structuring these systems and how these relate to organismal dispersal  
107 limitation, spatial connectivity and environmental filtering. For example, one can argue that microbiomes  
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115 therefore, no such thing as a 'gold standard' by which one can define an 'optimum scale' (also called  
116 'characteristic scale', i.e. one that 'maximizes the ratio of deterministic information to stochastic  
117 fluctuations' *sensu* Pascual and Levin, 1999). Rather, there is an urgent need for reproducibility with an  
118 explicit consideration of scale from which patterns emerge, thus paving the way for conceptual and  
119 theoretical developments.

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### 121 **Theory construction**

122 *"Theory is used to classify, interpret and predict the world around us. Without it, microbial ecology is merely*  
123 *the accumulation of situation-bound statements that are of limited predictive ability, providing*  
124 *microbiologists with few insights"* (Prosser *et al.*, 2009).

125 An ecological theory is built upon a contemplative amalgamate of information that aims to explain  
126 generalized patterns in nature. Most importantly, an ecological theory requires an 'explanatory surplus'  
127 (Gillies, 2015), that is, the ability to describe phenomena outside of the immediate realm in which it was  
128 formulated. This means that patterning microbiomes across ecosystems require a clear consideration of  
129 scale that results in consistency, a level of reproducibility and integration of information, as well as  
130 empirical tests and patterning validation assessments. Non-reproducible patterns are likely to result in  
131 obscure noise and/or lead to an overall assessment that microbiomes are stochastically assembled; or that  
132 patterns emerging from a reductionist approach are applicable at broad scales. This would reinforce  
133 assumptions that predicting, monitoring and/or manipulating microbiomes are unrealistic tasks and  
134 promote the argument of context-dependency and absence of realistic trackable mechanisms  
135 underpinning divergence in community assemblages across divergent systems and scales.

136 Two of the main objectives of theory are to inform applied ecology and experimental design. Patterns in  
137 ecosystem microbiomes can promote theory development by providing cross-system assessments using  
138 microbe-centric sampling strategies that also include appropriate measures of the local environmental  
139 complexity. The emerging data, together with other macroscale data, can then be used to develop  
140 validation experiments or environmental assessments that test hypotheses about potential underlying  
141 ecological and evolutionary mechanisms. As a path forward, and as similarly debated in general ecology,  
142 focus should be given to study systems at appropriate and pre-defined scales. This will later inform the  
143 development of models that bridge different scales (Melbourne and Chesson, 2006; Holt and Chesson,  
144 2016; Chesson, 2010; Chase *et al.*, 2018), and avoid the accumulation of an anecdotal collection of datasets  
145 that may confound concept formation and therefore limit, rather than advance, the development of  
146 ecological theory.

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#### 148 **Recommendations**

149 This article has aimed to stimulate debate on how scale can be reliable and meaningfully defined and taken  
150 into practice in ecosystem microbiome research. The arguments and recommendations are not intended  
151 to discredit or demerit previous research efforts. We also do not claim to have now closed this debate nor  
152 do we have the ultimate solutions. However, we provide some initial recommendations that we believe  
153 will benefit future research. In brief, to promote conceptual and theory formation, studies should consider  
154 the following recommendations: (1) Explicitly define scale and the extent to which it is representative and  
155 relevant to potentially underlying mechanisms. This should lead to a level of reproducibility that allows for  
156 the proper elucidation and validation of underlying mechanisms and processes that govern community  
157 divergences. (2) Assess variability within local sites and make use of new methods to assist sampling design  
158 (e.g., response-surface methodology, see Albert *et al.*, 2010). This is critical for the development of  
159 experimental plans that enhance sample/site representativeness. Information on local variability can also



160 be used to feed model predictions and indicate the level of uncertainty when attempting to extrapolate  
161 patterns. (3) Account for inaccurate extrapolations by indicating levels of uncertainty in model predictions  
162 and data visualization, as well as acknowledging the existence of sites for which information is unavailable.  
163 Together, these will enhance the clarity of data presentation and promote discussion on challenges and  
164 limitations associated with sampling design. (4) When similar ecosystems are surveyed across large-scale  
165 gradients, studies should consider using replicable spatial designs with consistent sample sizes that  
166 potentially control for autocorrelation of variables. Particularly in the case of autocorrelation and  
167 structured data, models using k-fold cross-validation on geographically-partitioned datasets have been  
168 successfully used in species distribution modelling and has led to much more generalizable model fits (e.g.  
169 Roberts *et al.*, 2017). (5) Since correlational outcomes can be meaningfully applied to generate and test  
170 hypotheses empirically, value should be given to studies that develop prospective experimental designs  
171 aimed at falsifying putative mechanisms. This represents a challenging yet elegant way towards advancing  
172 this fast-moving field of science. In conclusion, despite the various ways in which scale and spatiotemporal  
173 variability can subvert our interpretation of microbial taxa and gene distributional patterns (also see  
174 Armitage and Jones, 2019), we hope new studies in 'global/regional/cross-continental' microbiome  
175 assessments will take these constructive suggestions into account. Collectively, these will help to orient  
176 new experimental designs and enhance precision in data analysis, interpretation and communication. Thus,  
177 paving the way for a more realistic appreciation of the advances and impacts ecosystem microbiome  
178 surveys truly provide.

180 **Conflict of interests**

181 The authors declare no conflict of interest.

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125 An ecological theory is built upon a contemplative amalgamate of information that aims to explain  
126 generalized patterns in nature. Most importantly, an ecological theory requires an 'explanatory surplus'  
127 (Gillies, 2015), that is, the ability to describe phenomena outside of the immediate realm in which it was  
128 formulated. This means that patterning microbiomes across ecosystems require a clear consideration of  
129 scale that results in consistency, a level of reproducibility and integration of information, as well as  
130 empirical tests and patterning validation assessments. Non-reproducible patterns are likely to result in  
131 obscure noise and/or lead to an overall assessment that microbiomes are stochastically assembled; or that  
132 patterns emerging from a reductionist approach are applicable at broad scales. This would reinforce  
133 assumptions that predicting, monitoring and/or manipulating microbiomes are unrealistic tasks and  
134 promote the argument of context-dependency and absence of realistic trackable mechanisms  
135 underpinning divergence in community assemblages across divergent systems and scales.

136 Two of the main objectives of theory are to inform applied ecology and experimental design. Patterns in  
137 ecosystem microbiomes can promote theory development by providing cross-system assessments using  
138 microbe-centric sampling strategies that also include appropriate measures of the local environmental  
139 complexity. The emerging data, together with other macroscale data, can then be used to develop  
140 validation experiments or environmental assessments that test hypotheses about potential underlying  
141 ecological and evolutionary mechanisms. As a path forward, and as similarly debated in general ecology,  
142 focus should be given to study systems at appropriate and pre-defined scales. This will later inform the  
143 development of models that bridge different scales (Melbourne and Chesson, 2006; Holt and Chesson,  
144 2016; Chesson, 2010; Chase *et al.*, 2018), and avoid the accumulation of an anecdotal collection of datasets  
145 that may confound concept formation and therefore limit, rather than advance, the development of  
146 ecological theory.

147

148

#### 149 **Recommendations ~~moving forward~~**

150 This article has aimed to stimulate debate on how scale can be reliable and meaningfully defined and taken  
151 into practice in ecosystem microbiome research. The arguments and recommendations are not intended  
152 to discredit or demerit previous research efforts. We also do not claim to have now closed this debate nor  
153 do we have the ultimate solutions. However, we provide some initial recommendations that we believe  
154 will benefit future research. In brief, to promote conceptual and theory formation, studies should consider  
155 the following recommendations: (i) Explicitly define scale and the extent to which it is representative  
156 and relevant to potentially underlying mechanisms. This should ~~then~~ lead to (ii) a level of reproducibility  
157 that (iii) allows for the proper elucidation and validation of underlying mechanisms and processes that  
158 govern community divergences. (2) Access variability within local sites and make use of new ~~We suggest~~

159 ~~that accessing variability within local sites and introducing new~~ methods to assist sampling design (~~for~~  
160 ~~instancee.g.~~, response-surface methodology, see Albert *et al.*, 2010). This is critical for ~~can lead~~ the  
161 development of experimental plans ~~that enhance~~that enhance sample/site representativeness.  
162 Information on local variability can also be used to feed model predictions and indicate the level of  
163 uncertainty when attempting to extrapolate patterns. (3) Accounting Account for inaccurate extrapolations  
164 by indicating levels of uncertainty in model predictions and data visualization, as well as acknowledging the  
165 existence of sites for which information is unavailable. Together, these, can will enhance the clarity of data  
166 presentation. This will improve and promote discussion ~~of on~~ challenges and limitations associated with  
167 sampling design, ~~thus avoiding overarching assumptions and statements and the noise these may echo.~~ (4)  
168 ~~Moreover, w~~hen similar ecosystems are surveyed across large-scale gradients, ~~the studies should~~  
169 consider use using of replicable spatial designs with consistent sample sizes that potentially control for  
170 autocorrelation of variables ~~can be beneficial.~~ It is worth noting, particularly in the case of autocorrelation  
171 and structured data, ~~that~~ models using k-fold cross-validation on geographically-partitioned datasets have  
172 been successfully used in species distribution modelling and has led to much more generalizable model fits  
173 (e.g. Roberts *et al.*, 2017). (5) Since correlational outcomes can be meaningfully applied to generate and  
174 test hypotheses empirically, value should be given to studies that develop prospective experimental  
175 designs aimed at falsifying putative mechanisms. This represents a challenging yet elegant way towards  
176 advancing this fast-moving field of science.! In conclusion, despite the various ways in which scale and  
177 spatiotemporal variability can subvert our interpretation of microbial taxa and gene distributional patterns  
178 (also see Armitage and Jones, 2019), we hope new studies in 'global/regional/cross-continental'  
179 microbiome assessments will take ~~into account~~ these constructive suggestions into account. Collectively,  
180 these will help to orient new experimental designs and enhance precision in data analysis, interpretation  
181 and communication. Thus, paving the way for a more realistic appreciation of the advances and impacts  
182 ecosystem microbiome surveys truly provide.-

183 Finally, since correlational outcomes can be meaningfully used to generate and test hypotheses empirically,  
184 switching gears towards developing prospective experimental designs for falsifying putative mechanisms  
185 represents another elegant way towards advancing this fast-moving field of science.

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186 **Conflict of interests**

187 The authors declare no conflict of interest.

188

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