

# Microbiome-aware ecotoxicology of organisms: relevance, pitfalls and challenges

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9

## 10 **Abstract**

11 Over the last 15 years, the advent of high-throughput ‘omics’ techniques has revealed the  
12 multiple roles and interactions occurring among hosts, their microbial partners and their environment.  
13 This microbiome revolution has radically changed our views of biology, evolution and individuality.  
14 Sitting at the interface between a host and its environment, the microbiome is a relevant yet  
15 understudied compartment for ecotoxicology research. Various recent works confirm that the  
16 microbiome reacts to and interacts with contaminants, with consequences for hosts and ecosystems.  
17 In this paper, we thus advocate for the development of a “microbiome-aware ecotoxicology” of  
18 organisms. We emphasize its relevance and discuss important conceptual and technical pitfalls  
19 associated with study design and interpretation. We identify topics such as functionality,  
20 quantification, temporality, resilience, interactions and prediction as major challenges and promising  
21 venues for microbiome research applied to ecotoxicology.

22

## 23 **1. Introduction: the microbiome is relevant to ecotoxicology**

24 The significance of microbes to multicellular organisms is long documented. Because only a  
25 fraction of microorganisms can be isolated in culture, it is the advent of high-throughput sequencing  
26 technologies which ultimately revealed how diverse and numerically abundant they were.  
27 Microorganisms form complex symbiotic communities of eukaryotes, bacteria, archaea, and viruses  
28 referred to as the microbiome (HMP, 2012; McFall-Ngai et al., 2013; Tipton et al., 2019). Over the  
29 last 15 years, the microbiome has been a new frontier in Life Sciences (Alivisatos et al., 2015), and  
30 microorganisms were shown to be involved and even necessary in many host functions including  
31 nutrition, defense, immunity, development and behavior (Rakoff-Nahoum et al., 2004; Archie and  
32 Theis, 2011; McFall-Ngai et al., 2013). A current paradigm assumes that most animals and plants  
33 harbor a microbiome (Rosenberg and Zilber-Rosenberg, 2018). However, some species of comb  
34 jellies and nematomorpha, as well as certain life stages of insects including honeybee larvae, are

35 apparently devoid of a microbiome, suggesting that this may be an over-simplification (Martinson et  
36 al., 2012; Hammer et al., 2019). In humans, the microbiome may represent as many cells as the hosts  
37 and up to 1,000 times more genes, questioning the concept of individuality and the limits of self  
38 (Gilbert et al., 2012; Rees et al., 2018). The holobiont concept, referring to the entity formed by a  
39 host organism and its various microbial associates, arose to encompass the complexity of hosts and  
40 their microbiome (Mindell, 1992; Bordenstein and Theis, 2015). All these discoveries have fueled a  
41 “microbiome revolution” that increasingly spreads through all fields of life sciences, with extensions  
42 to behavioral and human sciences (Blaser, 2014; Cryan and Dinan, 2019).

43         Recent research focuses mostly on the links between hosts and their microbiome, and the  
44 reciprocal influence they exert on each other, revealing its significance to host physiology,  
45 homeostasis, disease, health and fitness (Selber-Hnatiw et al., 2017). Interestingly, most members of  
46 the microbiome are located on epithelia (mucosa, skin...), *i.e.* animal or plant polarized tissues that  
47 separate the inside from the outside of the organism. Sitting at the interface between a host and its  
48 environment, epithelia and their associated microbiome are the hosts buffer and first line of defense  
49 against contaminants and environmental stressors (Barr et al., 2013). Deep-sea hydrothermal vent  
50 mussels are an example of this. They harbor bacteria located in the gill epithelium that oxidize  
51 hydrogen sulfide, a compound toxic to their hosts, and fix carbon that contributes hosts nutrition  
52 (Halary et al., 2008). Toxicology studies investigating the effects of chemical compounds on  
53 organisms examine the accumulation, bio-transformation, elimination and effects in tissues, and are  
54 thus beginning to account for associated microbiome. This paper aims to emphasize the relevance,  
55 pitfalls and promises of host-associated microbiome research for ecotoxicology and advocates for the  
56 emergence of a “microbiome-aware ecotoxicology” of multicellular organisms, *i.e.* an approach that  
57 fully incorporates the microbiome compartment as a dynamic interface interacting with host and  
58 environment (Figure 1).

## 59 **2. The microbiome responds to and interacts with contaminants**

60         Data on effects of environmental contaminants on microbiomes has been published over the  
61 last years, with a focus on animal gut-associated bacteria (Claus et al., 2016; Jin et al., 2017;  
62 Rosenfeld, 2017; Adamovsky et al., 2018). Published studies include controlled exposure of model  
63 organisms to various types of contaminants including pesticides, antibiotics, heavy metals,  
64 xenobiotics, or nanoparticles (Licht and Bahl, 2019; Monroy-Torres et al., 2019). Bacterial  
65 community composition is typically assessed using the 16S rRNA sequence as a taxonomic marker  
66 that identifies Operational Taxonomic Units (OTUs), a commonly used proxy for species. The effect  
67 of exposures on OTU richness and diversity is then evaluated using multivariate statistics to test  
68 whether contaminants interfere with microbiome composition (Ramette, 2007; Evariste et al., 2019).  
69 Many studies include complementary analyses on host parameters of toxicological relevance such as  
70 markers of the immune system, tissue histology and developmental markers.

71         Besides descriptive studies that correlate exposures and microbiome variations, functional  
72 studies directly investigate the interactions between the gut microbiome and contaminants (Figure 2;  
73 Schmidt et al., 2019). It has been shown that microbiomes of mammals guts are able to metabolize a  
74 wide range of xenobiotics (e.g. polycyclic aromatic hydrocarbons, polychlorobiphenyls,  
75 nitrotoluenes) and could protect animals from deleterious effects (Claus et al., 2016). But the  
76 microbiome may also activate some compounds and mediate toxicity to hosts. For instance, the  
77 human colon microbiome was shown to convert polycyclic aromatic hydrocarbons into estrogenic  
78 metabolites with consequences on hormonal equilibrium (Van de Wiele et al., 2005); and the  
79 nephrotoxicity of melamine in rats results from its conversion into toxic cyanuric acid mediated by

80 the bacterium *Klebsiella terrigena* (Zheng et al., 2013). Many pharmaceutical drugs such as  
81 lovastatin or loperamide were also proved to be activated in the small intestine by bacteria-mediated  
82 biotransformations (Lavrijsen et al., 1995; Yoo et al., 2014). Alternatively, xenobiotics can also alter  
83 activities of the gut microbiome (Licht and Bahl, 2019). Besides the obvious example of antibiotics,  
84 many molecules such as epoxiconazole or glyphosate, both pesticides, are known to induce shifts in  
85 microbiome compositions (Xu et al., 2014). Because the microbiome substantially responds and  
86 interacts with contaminants, it must now be considered a key player in toxicology.

### 87 **3. Producing and interpreting microbiome data relevant to ecotoxicology**

#### 88 **3.1. Conceptual pitfalls: identifying a “good” microbiome and a “good” model species**

89 Ecotoxicology studies that address the microbiome rely on a microbial ecology background  
90 and need to consider the caveats associated with this discipline. Experimental investigations to date  
91 have focused mostly on bacteria. However, numerous studies have demonstrated the significance of  
92 Archaea and microbial Eukaryotes, including fungi, to host physiology (Hoffmann et al., 2013;  
93 Richard and Sokol, 2019), and the key role of phages in regulating bacterial populations (Mirzaei and  
94 Maurice, 2017; Bettarel et al., 2018). A comprehensive description of microbiome functioning  
95 requires all components to be accounted for (Rowan-Nash et al., 2019). However, the lack of  
96 universal, easy-to-obtain markers for some groups, notably for viruses, still precludes the  
97 development of systematic analyses that require deep metagenomics and specific expert analysis  
98 pipelines.

99 Besides, OTUs composition provides only a partial description of the real microbial diversity.  
100 Indeed, a single 16S rRNA-based OTU can encompass a diversity of distinct bacterial genotypes,  
101 potentially quite different in term of their respective functional phenotypes and responses to a  
102 stimulus (Andam, 2019; Hanafiah and Lopes, 2020). The genus *Vibrio* for example includes strains  
103 with very different lifestyles, including commensals, light-producing mutualists of the squid  
104 *Euprymna scolopes*, and pathogens of numerous metazoans, that all display almost identical 16S  
105 rRNA sequences (Sawabe et al., 2007). The relative abundances and dynamics of these different  
106 phenotypes, can thus not be monitored using 16S rRNA.

107 Another major difficulty is the general lack of baseline knowledge regarding microbiomes of  
108 toxicology model species, for which very little-to-no data is available regarding wild populations  
109 (Uenishi et al., 2007; Hird, 2017; Shinohara et al., 2019). Besides, organisms used in tests are often  
110 sourced from rearing facilities, and have thus experienced domestication, a process documented to  
111 lead to massive changes in bacterial microbiome compositions. In vertebrates, changes include  
112 overall bacterial species richness decrease and shift in taxa abundances due to dietary, social and  
113 environmental conditions of captivity (reviewed in Hird, 2017; Alessandri et al., 2019). Effects in  
114 other taxa are less documented and less clear-cut. In the silkworm for example, domestication is  
115 associated with higher bacterial diversity (Chen et al., 2018). The representativity of model species in  
116 ecotoxicology versus their wild relatives thus remains to be evaluated in the light of their  
117 domestication history. Interestingly, humans are no exception to this trend, and gut microbiomes in  
118 industrialized societies greatly differ from the recent ancestral microbiome and from that of  
119 contemporary traditional populations (e.g. hunters-gatherers). Changes in diets, sanitation and  
120 medical practices have led to a functional shift from fiber to mucus degraders, high frequency of  
121 antibiotic resistance, loss of particular taxa (e.g. Spirochaetes) and overall diversity decrease (Blaser,  
122 2016; Sonnenburg and Sonnenburg, 2019). This is assumed to result in non-optimal microbiomes  
123 associated with increased risk of chronic diseases. Lack of knowledge, along with inter-individual

124 variability (discussed below, 3.2), undermines the identification of the ‘normal’, balanced  
125 microbiome composition, i.e. the eubiotic state. This compromises the proper diagnosis of a  
126 dysbiosis (an ‘abnormal’, unbalanced) state upon exposure to contaminants (Figure 2). Indeed,  
127 although many factors can cause dysbiosis that may lead to health issues, it is not easy to establish  
128 what a healthy/eubiotic microbiome is (Iebba et al., 2016). Recently, authors insisted that dysbiosis  
129 due to stressors is first of all the destabilization of the stable eubiotic state. Changes in abundances of  
130 certain beneficial taxa are evident signs of dysbiosis, but interestingly, increased inter-individual  
131 variability in microbiome composition could be another signature of dysbiosis (Zaneveld et al.,  
132 2017).

133 In between the relative simplicity of most invertebrate-associated microbiomes in which a  
134 few OTUs are usually dominant (e.g. Raymann et al., 2017) and the extreme complexity of mammal-  
135 associated microbiomes (with hundreds to thousands of OTUs), teleost fish and their tens to a few  
136 hundred bacterial OTUs offer an interesting intermediate, besides their relevance to the monitoring  
137 of aquatic ecosystems (Llewellyn et al., 2014). Choosing a model thus involves addressing different  
138 levels of microbiome complexity, functionality and domestication history. Whether current models in  
139 toxicology are relevant to microbiome-aware ecotoxicology studies needs to be evaluated.

### 140 **3.2. Technical pitfalls: performing the right experiment to detect effects**

141 Most studies using controlled microcosms are monitoring various compartments that are  
142 potential sources of microbial diversity (e.g. food for animals, water for aquatic organisms). When  
143 scaling up to more holistic approaches such as mesocosms or the natural environment, potential  
144 sources of microorganisms dramatically increase, requiring the investigation of multiple  
145 compartments (e.g. food, prays, parasites, water, particles, sediments).

146 Fifteen years of human gut microbiome research revealed the high level of intra- (between  
147 body regions or life stages) and inter-individual heterogeneity in community compositions (HMP,  
148 2012; Rothschild et al., 2018). Although less documented, high levels of intra- and inter-individual  
149 variation are reported in other taxa including fish (e.g. Atlantic cod, salmon, rainbow trout, zebrafish)  
150 (Star et al., 2013; Llewellyn et al., 2014; Lowrey et al., 2015; Duperron et al., 2019; Evariste et al.,  
151 2019). In the zebrafish for example, gut-associated communities become increasingly different from  
152 those in the environment, and inter-individual variation increases across development (Stephens et  
153 al., 2016). Skin-associated communities are different on different body regions in the rainbow trout  
154 (Lowrey et al., 2015). These examples emphasize the importance of replication levels, and of  
155 addressing the exact same life stage and tissue region in all individuals.

156 Sex-differentiated responses to compounds are commonly reported in ecotoxicology studies,  
157 for example in medaka fish exposed to cyanotoxins (Le Manach et al., 2016). Sex also influences  
158 microbiome composition in various vertebrates, probably due to hormones and sex-specific immunity  
159 responses (Haro et al., 2016). It also affects microbiome responses. In lab-reared stickleback fed  
160 different diets, diet induced changes in some bacterial taxa abundances, but effects on bacteria in  
161 males were uncorrelated with effects observed in females, supporting that diet effects were clearly  
162 sex-specific (Bolnick et al., 2014). Authors measured similar sex-specific diet effects in mice and  
163 humans. Sex-specific effects on microbiome responses to contaminants are also documented.  
164 Exposure to silver nanoparticles was for example shown to modify the gut microbiome structure in  
165 male zebrafish, but not in females (Ma et al., 2018). Experimental design should thus carefully  
166 consider confounding factors of which sex is an important one.

## 167 **4. The roads less traveled: challenges in microbiome-aware ecotoxicology**

### 168 **4.1. Functionality and integration**

169 One major finding of the Human Microbiome Project was that despite high levels of intra-  
170 and inter-individual variation in the taxonomic compositions of bacterial communities, the functions  
171 they performed, as encoded by the metagenome, were highly conserved (HMP, 2012). Similar  
172 functions are thus performed by taxonomically distinct microorganisms. This concept known as  
173 functional redundancy is now recognized as key to the resistance and resilience of microbial  
174 communities (Allison and Martiny, 2008; Moya and Ferrer, 2016). Because of this, and the fact that  
175 closely related bacteria can display markedly different functionalities, community composition is not  
176 a reliable predictor of functions. Predictive tools for functional profiling based on composition (e.g.  
177 PICRUSt (Langille et al., 2013)) thus suffer limitations, and identity and functions should ideally be  
178 investigated in tandem. Functional capabilities can be evaluated through metagenomic sequencing,  
179 but genes and functions that are expressed at specific time points are better evaluated by  
180 metatranscriptomic or metaproteomic approaches. Metabolomics, which map metabolites, are  
181 another important tool that profiles ongoing metabolisms, and thus informs functions (Bundy et al.,  
182 2009; Gao et al., 2017) although, as for all of the above, the improvement of databases supporting  
183 metabolite identifications will be critical (Gertsman and Barshop, 2018). The integration of these  
184 approaches in multi-omics appears challenging, yet particularly promising for revealing the causal  
185 role of the microbiome and mechanisms involved in contaminants metabolism and toxic effects  
186 (Rohart et al., 2017; Koh and Bäckhed, 2020). This will be key in integrating the microbiome in  
187 adverse outcome pathways (AOPs) and risk assessment (figure 2 and Adamovsky et al., 2018).

### 188 **4.2. Quantification**

189 Microbiome-aware ecotoxicology should identify contaminant threshold values relevant to  
190 microbiomes. Indeed, microbial communities may shift rapidly and non-linearly between contrasting  
191 alternative, more or less stable states provided some parameters reach threshold values. The existence  
192 of yet-undescribed tipping points is for example hypothesized to explain the existence of bimodal  
193 distributions of abundances of certain bacteria in the human gut (Lahti et al., 2014; van Nes et al.,  
194 2016), and may be a trigger of dysbiosis. To identify tipping points in ecotoxicology, studies should  
195 examine dose-dependent responses and chronic exposure to low doses as done for toxicological  
196 effect on host traits, for example the determination of non-observable adverse effect limit (NOAEL).

197 Microbiome composition assessments also need to become more quantitative. Indeed,  
198 metabarcoding datasets produce taxa relative abundances tables, and their variations. In these, an  
199 increase in one group thus cannot be properly interpreted, as it may as well represent a lower  
200 decrease relative to other groups in a globally shrinking population. Bacterial densities in guts of  
201 distinct lineages of rainforest ants were for example shown to vary by orders of magnitude based on  
202 qPCR quantifications; interestingly, absolute abundance variations were better correlated with habitat  
203 (arboreal or terrestrial) and trophic position than actual community compositions (Sanders et al.,  
204 2017). Antibiotics, which are reported to affect relative abundances in bacterial communities, act first  
205 by affecting the total number of bacteria present, as was clearly demonstrated for streptomycin and  
206 sancomycin, this being their major influence on the microbiome (Willing et al., 2011; Vieira-Silva et  
207 al., 2019). Absolute abundances are relevant to our understanding of the environment-host-  
208 microbiome continuum, and should thus be informed whenever possible, for example using  
209 quantitative PCR (Tkacz et al., 2018). Quantifying bacteria within organisms is however challenging,



210 as demonstrated by the very different estimations of bacteria-to-human cell ratios found in the  
211 literature (Sender et al., 2016).

### 212 **4.3. Temporality and resilience**

213 The nature and amplitude of variations are important aspects of microbiome response to  
214 contaminants. Composition of communities and diversity indices are still the main endpoints of most  
215 studies. However, the dynamics of these variations during and after exposure are certainly as  
216 important. In humans, microbiome dynamics are individual-dependent (Flores et al., 2014).  
217 Dynamics inform resilience, evaluating whether variations have long-term effects on the microbiome,  
218 or whether it fully recovers and returns to a naive, pre-exposure stable state (Figure 2). Antibiotic  
219 exposure was for example shown to affect human gut bacterial communities for several months post-  
220 exposure, and similar effects can be expected with many contaminants (Dethlefsen and Relman,  
221 2011; Francino, 2015). Whether iterative exposure to some contaminants may lead to habituation,  
222 and thus become less influential to microbiomes over time, is also an important issue. Finally, how  
223 microbiome resilience itself may be affected by environmental factors (e.g. temperature, pH,  
224 interactions, seasonality) remains to be investigated.

### 225 **4.4. Interactions and prediction**

226 The holobiont is more than just the sum of its parts (Bordenstein and Theis, 2015). With  
227 dozens-to-thousands distinct coexisting microbial taxa, and many more if phages are considered, an  
228 animal's gut or skin is a whole ecosystem in which multiple interactions among members and with  
229 the host influence its functioning. These as well as interactions with the environment, including the  
230 contaminants and microorganisms occurring there, need to be accounted for. For example, a new  
231 method coupling spatial imaging of metabolites and bacterial genotypes, MetaFISH, was developed  
232 to characterize interactions occurring between symbiotic bacteria and the gill epithelial cells of  
233 hydrothermal vent mussels. It allowed the identification of metabolites located at the host-symbiont  
234 interface on tissue sections at the micrometer scale (Geier et al., 2020). Such a method is promising  
235 to monitor small-scale interactions between contaminants, the microbiome and the host and further  
236 explore causality (Koh and Bäckhed, 2020). Co-occurrence networks that are based on positive or  
237 negative correlations between the occurrence of microorganisms, functions, and environmental  
238 parameters also help in exploring interactions and formulating hypotheses (reviewed in Faust and  
239 Raes, 2012). A strong relationship between the presence of a contaminant and that of certain bacterial  
240 taxa can suggest an ability to metabolize the former, which can then be tested (Claus et al., 2016).  
241 Changes in the network structure itself can indicate microbial successions in time series experiments  
242 or dysbiosis, and may support modelling approaches (Hoffmann et al., 2007; Zhou et al., 2010).

### 243 **5. Conclusion: what can the microbiome do for ecotoxicology and vice versa?**

244 By analogy with the famous essay written by Dobzhansky (1973), it is tempting these days to  
245 suggest that “Nothing in Biology makes sense except in the light of the microbiome”. Ecotoxicology  
246 is no exception to this trend, and must not lag behind other disciplines that have embraced the  
247 microbiome revolution. However, the microbiome is not just another ecotoxicological endpoint, but a  
248 peculiar and complex biological compartment that exhibits its own ecological, metabolic, functional  
249 and thus ecotoxicological rules (Evariste et al., 2019). Instead, a microbiome-aware ecotoxicology of  
250 organisms needs to develop (Figure 1). This involves questioning, and not only transferring, classical  
251 toxicology protocols and model organisms' relevance to microbiome studies. Close cooperation  
252 between microbial ecologists and ecotoxicologists is needed. They have a lot in common: the

253 complexity of microbiomes and their response mirrors that of contaminants and their interactions;  
254 and both domains start with reductionist approaches, and strive to scale up to holistic approaches that  
255 encompass ecosystems full complexity and produce real-life-relevant data.

256 A major challenge is to move on from observing correlations to addressing causality, and  
257 ultimately explain processes, e.g. demonstrate mitigating effects of the microbiome at the population  
258 level in a given ecosystem. Repeatability is a key point, which involves inter-studies comparisons  
259 and meta-analyses for which tools are becoming available (e.g. Amplicon Sequence Variants for  
260 OTU clustering (Callahan et al., 2017)). Microbiome features including taxa or functions may  
261 become bioindicators of contamination, as recently proposed in stream ecosystems (Simonin et al.,  
262 2019). Modelling interactions between environment, contaminants, microbiomes and hosts will  
263 become tractable, with a certain level of predictive power (Gould et al., 2018). No doubt the dialogue  
264 between disciplines will result in mutual enrichment, and will allow to make the most of the  
265 microbiome revolution applied to ecotoxicology.

266

### 267 **Conflict of Interest**

268 The authors declare that the research was conducted in the absence of any commercial or financial  
269 relationships that could be construed as a potential conflict of interest.

### 270 **Author Contributions**

271 All authors (SD, SH, AG and BM) have contributed to the writing of the manuscript and the  
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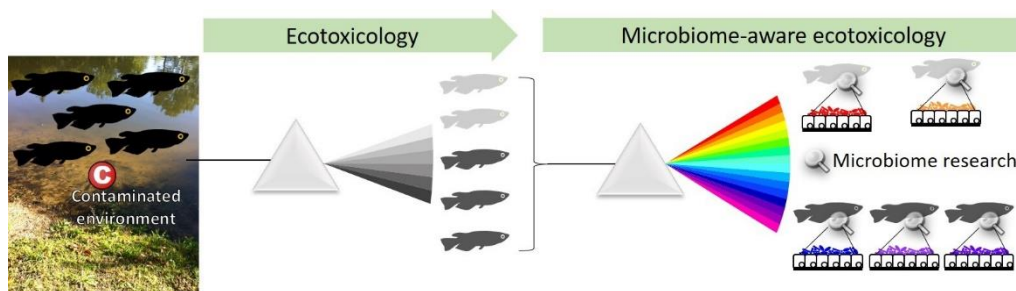
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## 515 Figures

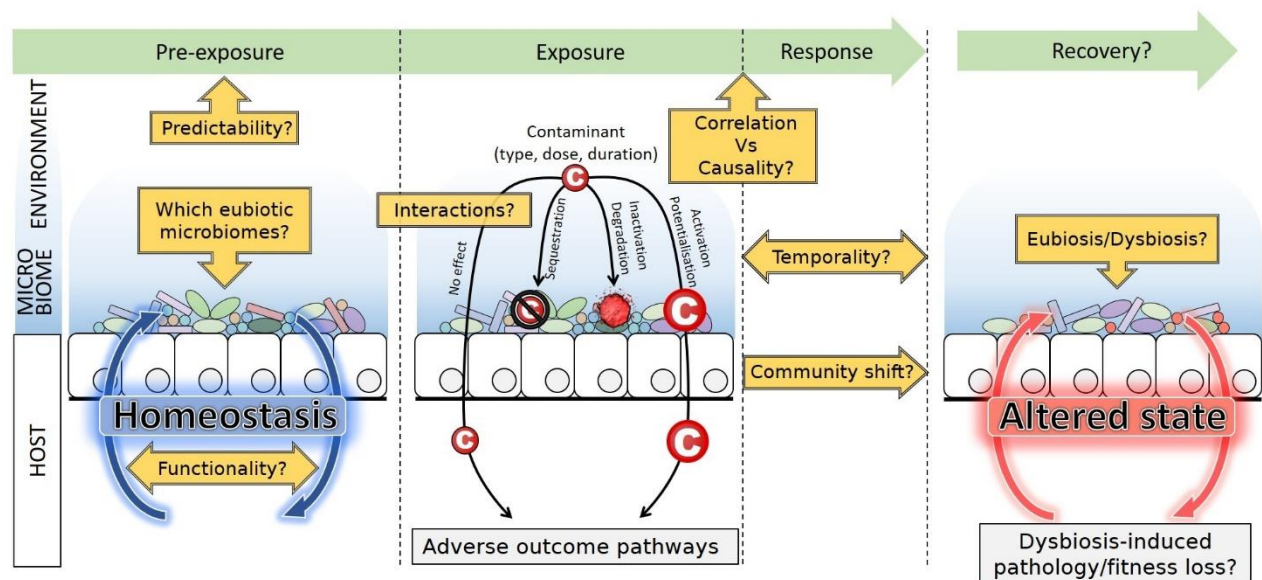
516



517 **Figure 1:** Ecotoxicology studies the effect of chemicals on organisms at the population level. A  
 518 microbiome-aware ecotoxicology perspective acknowledges the importance of associated  
 519 microorganisms in their hosts biology, at the level of individuals as well as populations. Indeed, the  
 520 microbiome is interacting with hosts, environment and contaminants and may be linked with health  
 521 status (here, light versus dark grey). The microbiome thus needs to be integrated as an element of the  
 522 system, and protocols to investigate ecotoxicological effects at each level need to be adapted.

523

524



525

526 **Figure 2:** Sitting at the interface between environment and host, the microbiome may interact with  
527 contaminants. Sequestration, inactivation and degradation mitigate potential effects on host health,  
528 while activation or potentialisation reinforce the effect of contaminants. Microbiome composition,  
529 abundance and functions respond to exposure, and dysbiosis can occur. Post-exposure recovery leads  
530 to a new stable state, identical or altered compared to the pre-exposure state. Future lines of research  
531 are emphasized. Although contaminants may alter the host health through adverse outcome  
532 pathways, dysbiosis itself may also induce pathology and fitness loss, difficult to disentangle from  
533 each other.

534