

***C. elegans* - A biosensor for host-microbe interactions**

Revealing complexity in uncharted functional landscapes

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Abstract

Microbes are an integral part of life on this planet. Microbes and their hosts influence each other in an endless dance dictating how the meta-organism interacts with its environment. While great advances have been made in microbiome research over the past 20 years, the mechanisms by which both host and their microbes interact with each other and the environment are still not well understood. The nematode *Caenorhabditis elegans* has been widely used as a model organism to study a remarkable number of human-like processes. Recent evidence shows that the worm is a powerful tool to investigate in fine detail the complexity that exists in microbe-host interactions. By combining the large array of genetic tools available for both organisms together with deep phenotyping approaches, it has been possible to uncover key effectors in the complex relationship between both parts. In this review, we survey the literature for insightful discoveries in the microbiome field using the worm as a model. We discuss the latest conceptual and technological advances in the field, and highlight the strengths that make *C. elegans* a valuable biosensor tool for the study of microbe-host interactions.

Introduction

31 Microbes are a major force in nature that have shaped this planet since the origin of life¹.
32 Animals and plants have been in close contact with microorganisms for billions of years, affecting each
33 other's biological functions that ultimately carved their evolutionary trajectories²⁻⁴. This shared
34 history led to the acquisition of dependence between the two biological systems, microbes and hosts,
35 that range across several taxonomic levels^{2,4}. The extensive variety of interactions between the host
36 and its microbes raises challenges to our understanding of what an organism or a biological function
37 is in the context of the microbiome⁵⁻⁷, and defies the very same notion that all animals need a
38 microbiome⁸. Nevertheless, microbial communities are key drivers regulating a wide range of
39 biological functions, affecting nutrition, immunity, and many other processes⁹. Yet we are still far from
40 achieving a complete understanding of the intricate relationship between the host and its microbiota.
41 The inherent complexity of this problem can be tackled from different angles using a wide array of
42 tools including theoretical approaches¹⁰⁻¹², classical *in vitro* studies^{13,14}, and the use of animal
43 models^{15,16}. The latter offers the possibility to establish causality links, granting robust interpretations
44 of the real influence of one system upon the other. An increasing number of animal models are
45 currently being used to tackle this problem. In particular, the use of *Caenorhabditis elegans*, a
46 bacterivore nematode, has emerged as a very suitable model that offers strong benefits over its
47 limitations¹⁷⁻¹⁹.

48 In this perspective, we highlight the reasons that make *C. elegans* a supreme model for the
49 study of microbe-host interactions by focusing on the methodologies available and the insightful
50 discoveries made over recent years using the worm as a biosensor of microbial activity.

51

52 **The worm and its microbiota – Parallels to humans**

53 From an evolutionary point of view, *C. elegans* is a suitable model for the identification of the
54 molecular processes involved in pathogenic and commensal interactions, as the mechanisms involved
55 are often conserved in other organisms of interest, including humans²⁰⁻²². At the level of the gut, basic
56 morphological similarities exist between the intestinal cellular structure in humans and worms. Added
57 to the functional similarities in the extraction and absorption of nutrients and the ability to host live
58 microbes, the gut of the worm is a perfectly adequate small but physiologically relevant organ for the
59 analysis of host-microbe interactions. In the wild, *C. elegans* harbours a rich and diverse microbiome
60 that is relatively stable across geography. Like mammals and other animals, the microbiota of *C.*
61 *elegans* is composed of the major phyla such as Firmicutes, Bacteroidetes, Actinobacteria and
62 Proteobacteria²³⁻²⁵, the latter being often the most dominant, mostly from the *Enterobacteriaceae*
63 genus. Likewise, studies have shown great intra-species variations of the intestinal microbiota which
64 varies according to the environment as well as the host's genetics^{19,25,26}. For example,

65 Gammaproteobacteria such as the human pathogen *P. aeruginosa* exploit monosaccharides from the
66 intestinal mucin layer to successfully colonise *C. elegans*²⁷. Whether other commensal, pathobionts
67 and/or pathogenic microbes explore this mechanism for worm colonisation remains to be studied.
68 Historically, the most commonly used bacteria in the context of *C. elegans* work in the laboratory
69 setting has been mono-colonisation with the enterobacteria *E. coli*^{28,29}, an important representative
70 of a human gut commensal. Like mammals, *C. elegans* acquires an active gut microbiota through the
71 oral route where bacteria serve many roles beyond providing nutritional sustenance²⁴. The view that
72 bacteria serve as nutrients only for worms but not for mammals is also misconception. As found in
73 worms, microbes from the environment are often present in the distal gut of mammals³⁰. Similarly to
74 worms, bacterial contents are constantly released into the mammalian gut as one third of the bacterial
75 cells are damaged with impaired membrane polarity and approximately one quarter with severely
76 compromised membrane integrity³¹. Hence, several bacterial products can be found in the host
77 bloodstream such as lipopolysaccharides (LPS)³², gut-commensal proteins²¹ and a wide range of
78 metabolites are either unique to bacteria or co-metabolites of unknown origin. Similar to mammals,
79 a vast literature shows that bacteria supplement the worm host with metabolites that regulate many
80 physiological and metabolic traits or regulate the effects of drugs or nutrients, through mechanisms
81 that are independent of the macronutrient content of bacteria^{24,25,29,33-38}. Therefore, the recent focus
82 on commensal interactions joins past findings where *C. elegans* has been an instrumental workhorse
83 to study microbial pathogenesis³⁹⁻⁴². As new resources become available (e.g. CeMbio) a larger array
84 of bacterial species can now be used to form complex communities covering the major phyla and
85 mimicking the microbial environment in the gut of other organisms^{2,10,43}. The utilisation of a
86 phylogenetically and metabolically diverse microbial community makes it a suitable model to study
87 functional aspects that are equally present in the human gut microbiome^{9,33,44,45}. Interestingly, work
88 in both humans and other organisms including *C. elegans* suggests that the functional capability of the
89 microbiome along with its phylogenetic composition, are a good proxy to link microbial community
90 composition and its regulatory effects on the host^{5,26,46,47}. This advocates for a need to characterise
91 microbiome function with high precision to fully capture its mechanistic links in modulating host
92 molecular and physiological phenotypes.

93

94 **Insights into host-microbiota interactions using worm phenotypes as readouts of bacterial activity**

95 Several reports using *C. elegans* as a model organism published over the last 10 years have
96 provided remarkable insights into host-microbe interactions (Figure 1). For example, several studies
97 using nematode development as a phenotypic readout have shown that its microbiota is essential for
98 the supply of micro-nutrients such as vitamin B2⁴⁸, B6²⁹, B9^{29,49}, B12^{50,51}, iron^{21,31,52} and molybdenum³⁶,

99 as well as reactive oxygen species^{31,53}. The *C. elegans* model has also proven useful in identifying other
100 metabolites at the interface between microbes and host, which regulate adult physiological traits. In
101 particular, *C. elegans* adult survival and longevity phenotypes have identified the production of nitric
102 oxide^{22,54,55} by the Firmicutes *B. subtilis*, and colanic acid^{30,56,57}, methylglyoxal⁵⁸, folate^{49,59} and
103 agmatine²⁸ from the Proteobacteria *E. coli* as key molecules regulating host ageing. Moreover,
104 perturbation of bacterial respiration through coenzyme Q biosynthesis impairment leads to
105 modulation of *C. elegans* lifespan^{60,61}. Using host lipid metabolism as a phenotypic readout, *C. elegans*
106 has also allowed the identification of microbial metabolites that regulate host lipid metabolism
107 through NR5A-Hedgehog signalling⁶². In addition, *C. elegans* has been used as a biosensor to conduct
108 studies on probiotics⁶³. In a recent study, Goya and colleagues found that using a *C. elegans*
109 synucleinopathy model, *B. subtilis* PXN21 inhibits, delays, and reverses α -synuclein aggregation
110 causing Parkinson's disease through alterations in the sphingolipid metabolism pathway of the host⁶⁴.
111 Additionally, Urrutia and colleagues evaluated the effect of various bacteria on another *C. elegans*
112 model of neurodegeneration and found that bacterial production of GABA and lactate conferred 40%
113 of neuroprotection⁶⁵. These studies provide links between bacteria and brain pathologies, offering
114 further mechanistic insights into how the gut microbiota regulates motor deficits and
115 neuroinflammation that were observed in a mouse model of Parkinson's disease⁶⁶. However, the
116 mechanistic basis of this microbiota–brain signalling and its physiological relevance remain largely
117 unknown. Using *C. elegans* olfactory responses (e.g. attractions to odours) as a phenotypic readout, it
118 has been shown that commensal gut bacteria alter olfactory behaviour through the production of the
119 neuromodulator tyramine. In particular, production of this metabolite by commensal *Providencia*
120 bacteria bypasses the requirement for host tyramine biosynthesis and manipulates a host sensory
121 decision promoting fitness of both the host and the microorganism⁶⁷. Likewise, the ability of *C. elegans*
122 to “read” and recognise bacterial small RNAs induces a transgenerational response leading to
123 appropriate behavioural changes in their progeny for pathogen avoidance⁴¹. Further, the use of *C.*
124 *elegans* as a biosensor has highlighted that the microbiota has a fundamental role in defense of the
125 host against pathogens^{68–71} – a mechanism known as colonisation resistance². In particular, using
126 worm survival as a physiological readout, *C. elegans* allowed the identification of cyclic lipopeptides
127 produced by the microbiota that confer resistance against intestinal colonisation by the human
128 pathogen *P. aeruginosa* and by natural pathogens of the worm such as *B. thuringiensis*⁵⁹.

129 In addition to identifying key mechanisms and metabolites involved in microbe-host interactions, the
130 worm has also been a great model for revealing novel host mechanisms. For example, it reveals the
131 molecular processes on the host that interprets signals from bacteria which play an essential role in
132 mediating various physiological functions. A groundbreaking study by Liu and colleagues identified the

133 organismal pathways that survey and defend mitochondria against toxic by-products of several
134 members of the microbiota⁷². Another example which illustrates this connection is the production of
135 biofilm by *B. subtilis* leading to increased life expectancy of the worms. This is the result of the
136 production of communication molecules involved in bacterial quorum sensing and nitric oxide (NO).
137 These molecules then trigger a dietary restriction-like response which involves the DAF-2, DAF-16 and
138 HSF-1 signalling pathways that regulate lifespan^{54,55}. Additionally, several bacterial metabolites have
139 been shown to impact host lifespan and healthspan by acting on mitochondrial function, activating
140 the unfolded protein response, remodeling the host lipid response, and interfering with insulin-like
141 and dietary restriction-related pathways⁷³⁻⁷⁵.

142 Over the years, the worm has provided unique insights into health and disease phenotypes by
143 elucidating key molecular mechanisms occurring in both entities of this complex meta-organism.

144

145 **An experimental pipeline to explore mechanisms of host-microbe interactions**

146 Interactions in the host-microbe system are a two-way road, where each part is sensing and
147 reacting to signals from its counterpart in a continuous loop. Interventions in the host and/or its
148 microbes transforms how the entire system reacts to stimuli coming from a set of environmental
149 conditions. Being able to characterise how both microbes and host perform these tasks is therefore
150 essential to capture the mechanisms underlying their interaction. Molecular and synthetic biology
151 tools to modify and study with great precision most layers of biological information exist for the worm
152 host, the microbial community, and its environment (Figure 2). Thus, the worm and its microbiota
153 provide an excellent system where each variable can be modified while concomitantly allowing for
154 deep phenotyping⁷⁶.

155

156 **1) The *C. elegans* host**

157 This semi-transparent nematode has a short life cycle (approximately 3 days) and lifespan
158 (mean of approximately 18 days), that together with its simple and cost-effective handling allows the
159 study of a wide range of evolutionarily conserved processes and diseases in humans⁷⁷. The
160 superpower of the worm model lies in its simple genetics and amenability for high-throughput
161 screening. As fully reviewed in ⁷⁸, “The power of any genetic model organism is derived, in part, from
162 the ease with which gene expression can be manipulated”. Therefore, *C. elegans* is a great genetic
163 model organism due to the wide range of molecular tools available to modify its genome, the low cost
164 and simplicity associated with generating mutants and a fully supportive community of researchers
165 that widely distribute their reagents. For example, thousands of genetically modified strains are

166 readily available from the Caenorhabditis Genetic Center and the National BioResource Project Lab.
167 The creation of mutant strains is also very well documented, both for the knockout and knockdown of
168 genes and the generation of transgenic lines⁷⁸⁻⁸¹. Precisely regulating the expression of host genes is
169 essential for causally linking the influence of an environmental cue with a phenotype. In particular,
170 unbiased forward (e.g. random mutagenesis⁸²⁻⁸⁵) and reverse genetic screens^{79,86} have been powerful
171 tools in revealing novel genes linking genotype to phenotype. For example, in the context of host-
172 microbe interactions, an RNAi-based approach has provided remarkable insights into the host
173 mechanisms regulating microbial aversion⁸⁷. The transparent nature of worms allied to the simple
174 generation of transgenic fluorescent reporter lines permits the study of how microbes regulate gene
175 expression profiles in a spatial (tissue-specific), temporal and quantitative manner⁸⁸. The combination
176 of these unique features and molecular techniques has been employed to identify microbial signalling
177 and metabolic pathways, including the respective metabolites, regulating host lipid metabolism²⁸.
178 Furthermore, the combination of these approaches with forward and reverse genetic methods has
179 been instrumental in identifying the host genetic networks underlying lipid metabolism regulation
180 mediated by bacterial vitamin B12⁵¹. Overall, these tools allow researchers to modify and study with
181 precision the role that host genetics plays in determining a molecular or physiological phenotype
182 driven by microbes.

183 *C. elegans* provides an excellent opportunity to perform microbiota studies at a scale only
184 outcompeted by *in vitro* unicellular high-throughput screening¹⁷. For example, several worm
185 phenotypes can be monitored in a high-throughput manner. This includes survival readouts when
186 worms are challenged with pathogens or xenobiotic compounds^{20,39,40,68,89-93}, measurements of adult
187 lifespan^{30,32,54,55,57,64,70}, arrested development^{21,22,32,36}, impaired fertility^{54,55,69,94,95} resistance to
188 stress^{55,70}, and altered behaviour^{96,97}. The measurement of such phenotypes in a high-throughput
189 manner allows the study of individual or communities of microorganisms that colonise its intestine in
190 regulating several traits of host physiology. Recently, the high-throughput capacity of this model has
191 experienced a giant leap in the current state of the art - deep phenotyping⁷⁶. This has been possible
192 through a new set of hardware and software tools that allow the manipulation, recording and imaging
193 of worms grown in hundreds of conditions in parallel. These include the use of flow cytometry-like
194 stations as COPAS⁹⁸ to measure worm physical variables such as axial length and optical density; the
195 'WorMotel' from Fang-Yen⁹⁹ and the Lifespan Machine from Fontana¹⁰⁰ that measure lifespan; diverse
196 tracking platforms from Brown^{96,101}, Nollen¹⁰², Rex¹⁰³, Driscoll¹⁰⁴ or Goodman¹⁰⁵ labs that can assess
197 motor-related phenotypical variables including behavior; or fully automated workstations from
198 Pincus¹⁰⁶ or Lu⁹⁷ that monitor long-term behavior and healthspan, among many other examples¹⁰⁷⁻¹¹⁰.
199 These technologies allow worms to be phenotypically characterised with a high degree of

200 reproducibility by often affordable imaging systems¹¹¹. Exploring the transparent nature of *C. elegans*,
201 fluorescence imaging can also be performed in a high-throughput manner, further extending the
202 screening capabilities of the worm as a biosensor. Fluorescence information can be obtained at
203 different scales, from studying protein dynamics in whole animals^{108,112} to processes in particular
204 specific cell lineages¹¹³. Hence, by using worm molecular or physiological phenotypes as readouts of
205 bacterial activity, it is possible to capture in fine detail several molecular mechanisms at the host level
206 involved in these complex host-microbe interactions. As a consequence, this leads to the generation
207 of a large amount of complex phenotypic data which requires the use of computational tools to
208 analyse and extract meaningful information. To solve this issue, machine learning and deep learning
209 algorithms are now being used in addition to commonly used statistical techniques (e.g. PCA or
210 correlation¹¹⁴) to uncover hidden features and trait prediction¹¹⁵ from complex datasets.

211 Thus, *C. elegans* provides unique opportunities as a model organism for studying host-microbe
212 interactions, as several layers of information can be edited and studied at a depth covering a vast and
213 uncharted functional landscape.

214

215 **2) The microbes**

216 A major problem of current microbiome research is the excess of data available implying
217 causation between microbiota physiology and host function drawn simply from correlation data. One
218 potential solution to this problem is the creation of synthetic communities which may offer a plausible
219 strategy to dissect causality in complex host-microbiota interactions. To increase the likelihood of
220 success one should consider creating a simplified microbial community that represents both
221 phylogenetically and functionally the complex microbiome that it is attempting to represent. It should
222 also be designed taking into consideration the model organism that will host the community. Finally,
223 and most importantly, whether the mock-community and their host are adequate models to address
224 the scientific question at hand. The groups of Félix, Samuel, Schulenburg and Shapira²³⁻²⁵ have
225 importantly contributed towards this goal. As a result of their in-depth meta-analysis of the natural
226 microbiome of *C. elegans*, they have created CeMbio⁴³ - a simplified natural worm microbiota mock-
227 community. Some of its key features include a set of strains easily culturable that colonise the worm
228 gut and that distinctly affect *C. elegans* physiological traits and life trajectories. These strains have fully
229 sequenced genomes, diagnostic PCR primers and characterised metabolic network models allowing
230 modelling with ease. Computer modeling of metabolism and experimental characterisation of
231 bacterial physiology allow to study not only the impact of bacterial functional diversity but also the
232 role of the environment on host functions. Zimmermann and colleagues have led this integrative
233 approach, bringing together the use of phenomic microarrays (Biolog) technology¹¹⁶, to assess

234 metabolic competences of selected bacteria, metagenomics and computer modeling, to reconstruct
235 metabolic networks at the community level and study ecological interactions between members of
236 the community. Their work has shown that host physiology and fitness is dependent on the nutritional
237 landscape for microbe-microbe interactions¹¹⁷.

238 Yet, to functionally characterise the contribution of each microbe in maintaining the
239 homeostasis of the community and their role in regulating host physiology, one needs to go deeper in
240 the functional characterisation of each microbial member. For this purpose, the availability of a wide
241 range of technologies for bacterial modification is necessary, such as the random insertion of
242 transposons in the genome^{34,118–120} and the use of bacteriophages³⁵. These techniques allow the
243 creation of mutant libraries from several species giving us an opportunity to study and identify new
244 genes regulating bacterial function. For example, such approaches have led to the identification of
245 thousands of bacterial genes belonging to Proteobacteria and Bacteroidetes of previously unknown
246 functions¹²¹. One of the most notorious examples is the construction of the Keio library, which
247 contains a set of precisely defined monogenic deletions of all the non-essential genes of *E. coli* K-12
248 (3,985 genes out of a total of 4,288)¹²². These transposon and deletion bacterial libraries are now being
249 used in mono-colonisation experiments to causally link the effects of bacterial genes from a single
250 species with a wide range of phenotypes from diverse hosts^{51,123–127}. For *C. elegans*, the Keio library
251 has been screened for *E. coli* genes involved in the regulation of lifespan^{30,49}, to show the dependence
252 of the host development on its microbiota for micronutrients (e.g. folate, iron and molybdenum^{31,32,36})
253 and to infer the role of the microbiota in drug action^{29,128} (e.g. fluoropyrimidines). *B. subtilis*, a human
254 probiotic bacterium, is another commonly used bacterial strain for mono-association studies with *C.*
255 *elegans*. The Gross lab has constructed two ordered, barcoded, antibiotic-resistance-marked single-
256 gene deletion libraries, comprising 3968 and 3970 genes respectively, allowing the study of gene and
257 pathway function genome-wide of a Gram-positive bacteria on host physiology¹²⁹. To date, this
258 resource has not been used in combination with a *C. elegans* host but may well provide an important
259 tool to expand the possibilities of this microbiome model beyond its current potential. While loss-of-
260 function libraries are more widely used and studied with *C. elegans*, gain-of-function libraries are also
261 available. For example, the ASKA library, made up of single strains containing multicopy vectors
262 overexpressing any gene of *E. coli*³⁷, was used with *C. elegans* and led to the identification of the
263 bacterial metabolite methylglyoxal in regulating host lifespan⁵⁸.

264 New tools are also being developed or implemented to study the gut environment in *C.*
265 *elegans*. For example, RNAseq has been used to study the effects of the gut environment and genetics
266 of *C. elegans* in bacterial gene expression and metabolic pathways of *E. coli* inside the gut¹³⁰. The
267 authors found that active metabolism of bioactive lipids in the gut may regulate potential host-

268 microbial interactions. A similar observation was recently made in a mammalian model, where
269 sphingolipids produced by the microbiota enter and regulate host lipid metabolism¹³¹. Interestingly,
270 the authors observed an increase in aerotaxis-related genes by bacteria growing in the gut compared
271 to *in vitro* growth, suggesting that the gut of *C. elegans* may in fact be anaerobic. While this is an
272 interesting observation, which could further expand the usefulness of this model, measurements of
273 oxygen tension inside the worm gut are required. New synthetic biology microbial reagents are also
274 being designed to expand this toolset. This includes the development of bacterial biosensors capable
275 of detecting molecules in the gut of worm¹³², bioluminescent bacteria to evaluate bacterial survival in
276 the gut¹³³, and the development of optogenetic tools in bacteria to control bacterial metabolism and
277 indirectly regulate host physiology⁵⁷.

278

279 **3) The environment**

280 Recent research in humans shows that the environment dominates over host genetics in shaping the
281 microbiota^{134,135}. All the aforementioned tools provide a robust framework where the role of the
282 microbiota on host physiology can be captured at the mechanistic level. The scalability of the current
283 worm methods allows the set-up of systematic studies where environmental perturbations can be
284 added as important variables of investigation.

285 The inclusion of drugs as an additional factor produces a complex interaction landscape
286 between microbes, drugs and host. For example, host physiology may be affected as a result of
287 modified drug pharmacokinetics through direct microbial biotransformation^{38,136-138}, or alternatively
288 by the indirect effects resulting from the action of drugs on microbial community structure and
289 function^{13,14,38,139}. Levodopa, a medicine to treat Parkinson's disease, is used as a carbon source by
290 bacterial taxa containing tyrosine decarboxylase activity, thus reducing its efficacy¹⁴⁰. For the latter,
291 examples in cancer research show that chemotherapy treatments often lead to intestinal disorders
292 following an overall reduction in microbial abundance¹⁴¹ or through increased drug toxicity after re-
293 activation by microbial enzymes^{142,143}. In addition, drugs can limit the biological functions of some
294 taxa, arresting their growth and allowing other disease-associated taxa to out-compete. For example,
295 colonisation by *C. difficile* is prevented by colonisation resistance properties of the faecal microbiota.
296 Thus, weakening microbial colonisation resistance by widespread use of antibiotics in the clinical
297 setting is a major risk factor for *C. difficile* associated morbidity¹⁴⁴. *C. elegans* offers a reliable platform
298 to investigate these complex relationships between host, microbes and drugs. Using the worm as a
299 biosensor for host-microbe-drug interactions showed that doxorubicin, a commonly used anticancer
300 drug, is metabolised by human gut bacteria such as *K. pneumoniae*, *E. coli* and *R. planticola* among
301 others¹³⁷. Fluoropyrimidines (5-FU) are essential anti-cancer chemotherapy for colorectal cancer with

302 highly dependent patient efficacy. To investigate the role of microbes in anti-cancer drug toxicity, our
303 group in parallel with the Walkout and O'Rourke labs developed a three-way (microbe-drug-host)
304 high-throughput screening method to explore the role of microbial genetics in the toxicity of
305 fluoropyrimidines on *C. elegans*^{29,128,145} development, reproduction and survival. The relative
306 contribution of each *E. coli* gene in mediating drug toxicity on the host was obtained to perform a
307 genome-wide systematic analysis of the pathways and processes involved in the mediation of drug
308 effects. These three independent studies show that microbes can bolster or suppress the effects of
309 fluoropyrimidines through metabolic drug interconversion involving bacterial vitamin B6, B9, and
310 ribonucleotide metabolism, and highlight the value of these approaches to unravel the mechanistic
311 complexity that exists in such interactions.

312 Nutrition is a key element at the interface between microbes and host dictating the fitness of
313 the entire meta-organism. Given the immense complexity of nutrition¹⁴⁶, mapping the biological
314 response of a host and its associated microbes to the different types of chemical components is a
315 challenging task. Historically, the microbial nutritional landscape has often been studied with the well-
316 established microbial phenotyping technology from Biolog¹¹⁶. This technology allows the investigation
317 of hundreds of metabolites covering all major nutrient classes (e.g. sugars, fatty acids, amino acids)
318 on regulating microbial growth phenotypes. Zimmermann and colleagues recently applied this
319 technology to investigate how different microbes from the natural microbiome in *C. elegans*
320 metabolised a diverse range of nutrients¹¹⁷. They showed that specific nutritional requirements by
321 members of the worm's microbiota dictate the nature of their interaction (e.g. competitive,
322 commensal) in a complex microbial community and their role in regulating worm physiology. Using
323 the same technology, our lab developed a high-throughput four-way microbe-drug-nutrient-host
324 screening approach to investigate how 337 dietary elements affect the efficacy of metformin on host
325 physiology in a bacterial-dependent manner. Metformin is the most widely used drug for type 2
326 diabetes and a potential treatment for ageing or age-related disease. Research spanning from worms
327 to humans shows that metformin acts indirectly through the microbiota to regulate distinct host
328 phenotypes and diseases¹⁴⁷⁻¹⁵⁰. Using a nutrient systems approach, we discovered that *E. coli*
329 integrates signals from both metformin and the diet into a signalling cascade that affects the
330 expression of the master nutrient regulator Crp, which in turn, indirectly regulates host physiology
331 through modified arginine-derived metabolites²⁸. Recently, a study by the O'Rourke lab investigated
332 the role of amino acids in microbe-drug-host interactions to discover that dietary serine enhances
333 fluoropyrimidine anticancer chemotherapy without altering pro-drug activation by *E. coli*¹⁴⁵. Overall,
334 the current use of *C. elegans* as a biosensor of bacterial activity is one of the ultimate state-of-the-art

335 models to reveal novel mechanisms at the interface between drugs-nutrients-microbes and host
336 physiology.

337

338 **Future outlook**

339 The vast complexity present in the human microbiome may only be fully understood with
340 careful and systematic navigation of all the potential physical and biological constraints that exist in
341 the gut. For example, this model can be further extended to probe the effect of a wide variety of
342 environmental conditions including new drugs/xenobiotics or pH fluctuations. As scalability of this
343 system continuously grows from additional technological, biological, and computational tools with
344 seamless integration, new layers of complexity will be captured. Yet, despite the important
345 contributions achieved using this model, the microbiome *C. elegans* research field is still in its infancy.
346 To conquer the vast unexplored complexity that exists in host-microbiome interactions, *C. elegans*
347 research boundaries will have to be expanded to include larger microbial communities in complex but
348 defined nutritional environments. The immense value of mimicking specific human microbiome
349 conditions through the addition of further layers of complexity to this highly scalable system will
350 permit the discovery of key principles in host-microbe interactions.

351 Past research on *C. elegans* as a model for the study of host-microbe interactions gives us
352 hope. It has permitted the discovery of many bacterial effectors on host physiology and the respective
353 host mechanisms. In particular, some of the most exciting discoveries using worms as biosensors of
354 complex phenotypic traits mediated by microbes have been extended to diverse host organisms^{28,56},
355 and identified mechanisms that are conserved across taxa. The quote “You have evolved from worm
356 to man, but much within you is still worm” by the German philosopher Friedrich Nietzsche has often
357 been used to capture with simplicity the use of *C. elegans* as a valuable model organism for human
358 disease processes. And once again, now in the study of complex host-microbe interactions, this simple
359 model organism continues to enlighten and surprise us. What the future holds will lead to pioneering
360 discoveries in one of the most extraordinary and complex problems that biology faces today.

361

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366

367 **FIGURE LEGENDS**

368

369 **Figure 1.** *Timeline of research publications providing remarkable insights into C. elegans-microbe*
370 *interactions.* Each publication entry has been classed (by position and colour) regarding the main
371 functional landscape explored in the work. Worm as a biosensor reveals links between: (black) host
372 and/or microbial genetics; (red) drugs-microbe and host; (green) nutrition-microbe and host; (blue)
373 microbe-microbe host relationships. Selected publications have been highlighted in bold.

374

375 **Figure 2:** *An experimental pipeline to explore host-microbe interactions.* A modular, scalable, layered
376 and flexible workflow to explore the complex landscape of host-microbe interactions. Green - Multiple
377 experimental options exist for *C. elegans* molecular or phenotypic readouts, including the use of large
378 mutant libraries for host genetics. Blue - High-throughput screening of single or complex microbial
379 communities and metabolic network modeling tools are available for worm-microbiota studies. Red -
380 Drug and dietary interventions can be incorporated to study the influence of environmental cues on
381 host-microbe interactions. Black - Multi-omics and computational analyses can be applied to
382 individual or combined entities of the worm-microbe model for in-depth molecular characterization.
383 Orange - The evolutionary conserved nature of the worm-microbe model permits testing and further
384 validation of findings in more complex models.

385

386

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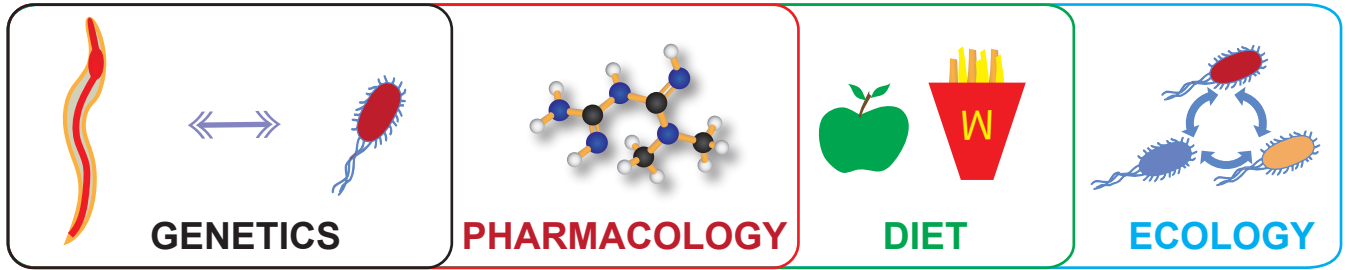
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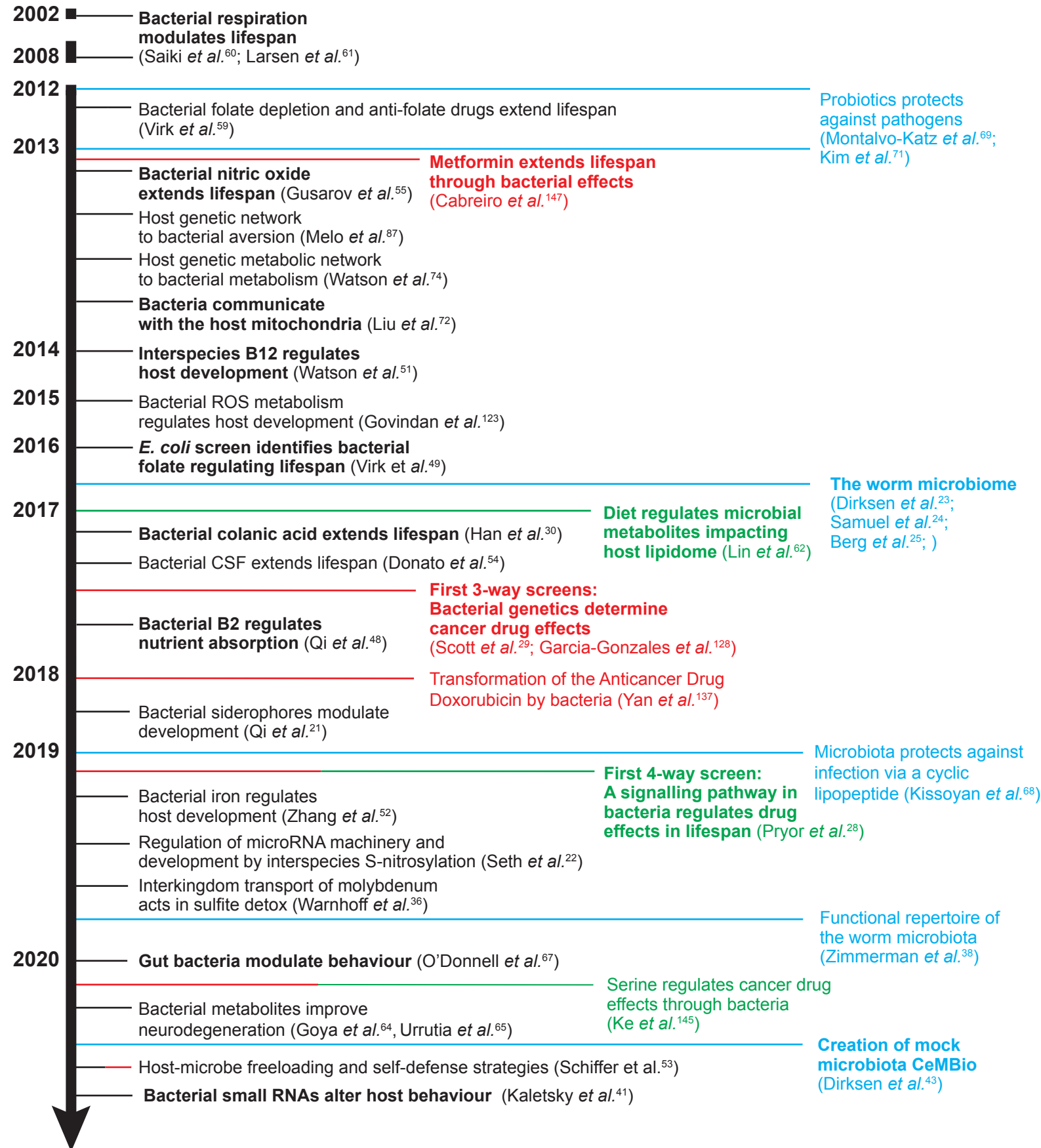
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Functional Landscape

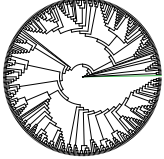


Timeline

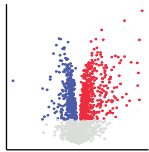


EXPERIMENTAL TOOLS

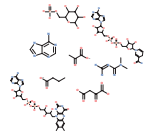
Genomics



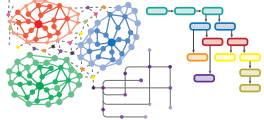
Proteomics



Metabolomics



Computational analysis



HOST

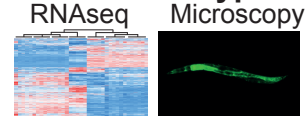
C. elegans

Genetic tools

- Mutants by CRISPR/Cas9
- Forward and reverse genetic screens
- Tissue-specific transgenesis

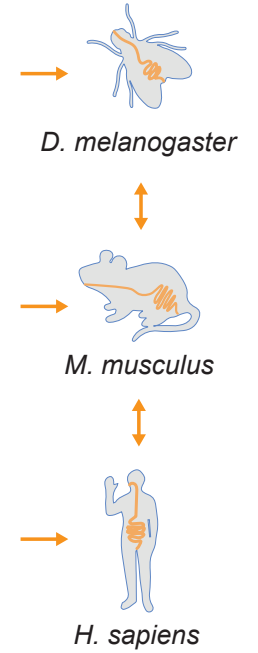


Phenotypic readouts



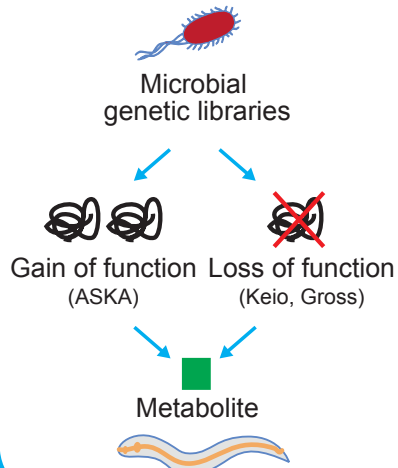
- Life/healthspan
- Development
- Reproduction
- Behaviour

TRANSLATION

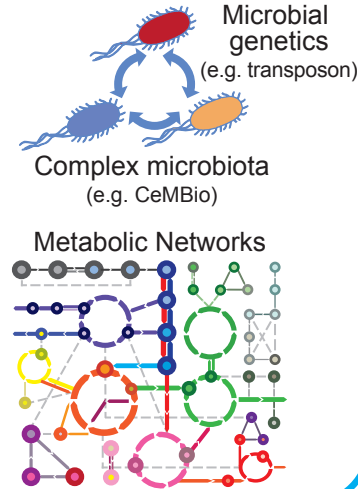


MICROBIOTA

Single bacterial species



Complex bacterial communities



INTERVENTIONS

