

1 Harnessing Tissue Engineering Tools to Interrogate 2 Host-Microbiota Crosstalk in Cancer

3 Barath Udayasuryan¹, Tam T. D. Nguyen², Daniel J. Slade², Scott S. Verbridge^{1*}

4 ¹Virginia Tech – Wake Forest School of Biomedical Engineering and Sciences, Blacksburg, Virginia, 24061, USA

5 ²Department of Biochemistry, Virginia Polytechnic and State University, Blacksburg, Virginia, 24061, USA

6 *Lead Contact: Scott S. Verbridge sverb@vt.edu

7 Corresponding Author: Scott S. Verbridge sverb@vt.edu

12 SUMMARY

13 Recent studies have begun to highlight the diverse and tumor-specific microbiomes across multiple
14 cancer types. We believe this work raises the important question of whether the classical
15 “Hallmarks of Cancer” should be expanded to include tumor microbiomes. To answer this
16 question, the causal relationships and co-evolution of these microbiotic tumor ecosystems must
17 be better understood. Because host-microbe interactions should be studied in a physiologically-
18 relevant context, animal models have been preferred. Yet these models are often poor mimics of
19 human tumors, and are difficult to interrogate at high spatiotemporal resolution. We believe that
20 *in vitro* tissue engineered platforms could provide a powerful alternative approach that combines
21 the high-resolution of *in vitro* studies with a high degree of physiological relevance. This review
22 will focus on tissue engineered approaches to study host-microbe interactions and to establish
23 their role as an emerging hallmark of cancer with potential as a therapeutic target.

24 INTRODUCTION

25 The involvement of tumor-specific bacteria, collectively termed the *tumor microbiome*, has
26 garnered significant attention as a key potential regulator of the well-established “Hallmarks of
27 Cancer”. These hallmarks include deregulated proliferation, replicative immortality, genomic
28 instability, evasion of growth suppression, avoidance of immune surveillance, chronic inflammation,
29 angiogenic induction, and the activation of metastatic pathways (Hanahan and Weinberg, 2011).
30 Bacteria and their secreted metabolites have been implicated in influencing most, if not all, of these
31 host factors (Fulbright et al., 2017). While many *in vitro* models have helped elucidate mechanisms
32 related to tumorigenesis, there are only limited models that are amenable to directly investigate
33 host-microbe interactions, and far fewer in the context of cancer. Similarly, though animal models
34 have been an indispensable tool in microbiome studies associated with cancer, they exhibit
35 significant variability in their resident bacterial species and immune profiles when compared to
36 humans. To address these issues, we believe that tissue engineering provides a unique opportunity
37 to bridge the gap between *in vitro* and animal models in analyzing these host-microbe interactions
38 in the tumor microenvironment (TME) that are so critical for tumor progression and therapy
39 response.

40 Tissue engineered models developed from human cells, not only maintain the genetic constitution
41 of the host, but do so in a physiologically relevant three-dimensional structure that consists of
42 multiple, differentiated cell types functioning in synergism as in native tissue. Furthermore, these
43 platforms are amenable to interrogation at high spatiotemporal resolutions that is just not possible
44 in larger animal models. For example, one can use these platforms to study the role of individual
45 bacterial interactions with the host to distinguish correlation from causation in microbial impacts
46 on cancer, which are normally obscured by multiple confounding factors within animal models.

47 Current tissue engineered platforms that have been developed to study host-microbiome
48 interactions are predominantly based on recreating the gut epithelium, which harbors the majority
49 of microorganisms in the human body (Sender et al., 2016). Historically, there have been several
50 challenges limiting the development of these platforms. Each tissue type has its own specific
51 engineering challenges including cellular spatial constraints, physical forces, biochemical cues, and
52 cell growth and differentiation capacities that need to be addressed and hence, a personalized and
53 experimentally-tailored approach is preferred to develop each tissue type. The convergence of
54 technologies from multiple disciplines has enabled the possibility to incorporate advanced sensors
55 and imaging modalities for real-time monitoring of oxygen, pH, barrier permeability, and other
56 biological parameters to recapitulate and analyze host-microbial interactions. In this review, we
57 highlight the latest insights derived from 2D, 3D, and organ-on-chip platforms that have been used
58 to investigate the interactions within the host-microbial consortia, and explore the potential for
59 adapting these platforms to advance our understanding of the tumor microbiome.

60 THE TUMOR MICROBIOME

61 The human microbiome, consisting of trillions of microorganisms co-existing within the human
62 body, has an enormous impact on maintaining health and normal physiology (Brestoff and Artis,
63 2013; Fan and Pedersen, 2020). A dysbiotic microbiome adversely impacts homeostasis which leads
64 to a number of unfavorable outcomes including inflammatory diseases, cardiovascular disease,
65 obesity, diabetes, and can even potentiate cancer initiation and progression (Udayasuryan et al.,
66 2019; Xavier et al., 2020). Studies of the microbiome often characterize microbial compositional
67 alterations in disease conditions (Durack and Lynch, 2019). However, the inherent complexity and
68 variability in these experiments makes it challenging to derive meaningful conclusions on the
69 microbes' direct impact on cancer progression. This is further compounded by the fact that the gut
70 microbiome can play a dual role by being both tumor-promoting and tumor-restricting. A striking
71 recent example within a mouse model revealed that p53 disabling mutations exhibited divergent
72 effects based on the location of the cells within the gut and its spatially-segregated microbial
73 composition, behaving oncogenic distally, but tumor-suppressive proximally (Kadosh et al., 2020).
74 That microbiotic residents may regulate the action of such an archetypal tumor suppressor gene as
75 p53, suggests that we are only scratching the surface of the role of tumor localized microbiomes in
76 cancer.

77 Our understanding of the role of microbes in tumor progression has been advanced by next-
78 generation sequencing (NGS), specifically 16S rRNA sequencing, which has uncovered reproducible
79 microbial signatures within a multitude of tumors. Most recently, Nejman et al, identified distinct
80 intracellular bacteria within cancer and immune cells in 1526 tumor samples (consisting of a mix of
81 flash frozen and Formalin-Fixed Paraffin-Embedded samples) and their adjacent tissues of seven
82 cancer types: breast, lung, ovary, pancreas, melanoma, bone, and brain tumors (Nejman et al.,
83 2020). The number of bacterial species detected in most of the tumors averaged ~ 9. Interestingly,
84 breast tissue was identified as having the most diverse tumor microbiome with an average of 16.4
85 species per sample (Figure 1A-C). In our own work, we previously examined the impact of molecules
86 secreted by a bacterium that had been identified as present in the breast TME (Balhouse et al.,
87 2017). This more recent finding by Nejman et al. provides further evidence that local tumor
88 microbiomes may play a critical role in multiple cancer types. However, critical interactions and
89 modes of tumor regulation have largely only been studied in the context of gastrointestinal cancers.
90 Building tissue engineered models recapitulating the TME of different cancer types that have the
91 ability to support the growth and maintenance of tumor microbiomes will be a key next step to
92 investigate the specific roles of microbes in cancer.

94 The Role of Specific Microbes in Tumorigenesis

95 Many seminal studies have shown that individual microbial species play a role in the onset and
96 progression of multiple cancers. Well known examples include *Helicobacter pylori* in gastric cancer
97 (Correa and Piazuelo, 2011; Uemura et al., 2001) and MALT lymphoma (Farinha and Gascogne,
98 2005), *Salmonella typhi* in gallbladder cancer (Ferreccio, 2012), *Streptococcus bovis* in colon cancer
99 (Boleij et al., 2009), and *Chlamydia pneumoniae* in lung cancer (Zhan et al., 2011). Several

100 associations including the discovery of *Fusobacterium nucleatum* within CRC (Kostic et al., 2012),
101 high abundance of *Acidovorax temporans* in lung cancers with TP53 mutations (Greathouse et al.,
102 2018), and variations in the oral and gut microbiome of melanoma patients undergoing PD-1
103 immunotherapy (Gopalakrishnan et al., 2018), indicate a correlative role of the human microbiome
104 with cancer. In addition, intracellular organisms can also directly impact chemotherapy regimens.
105 For example, Gammaproteobacteria within pancreatic ductal adenocarcinoma (PDAC) can
106 metabolize the chemotherapeutic drug gemcitabine (Geller et al., 2017). The most dramatic
107 consequence is the lowering of patient survival with the presence of these bacteria in the tumor.
108

109 Microbes within the TME can induce a mix of direct and indirect effects to impact tumorigenesis.
110 From prior work largely on colorectal cancer (CRC), it is known that bacteria within tumors can cause
111 chronic inflammation or produce and release toxins that impact the cell cycle and induce DNA
112 damage that leads to tumor initiating or promoting mutations (van Elsland and Neefjes, 2018).
113 Microbes in the TME can also influence tissue remodeling and deregulate mucosal immunity,
114 creating a favorable niche for tumor cells to expand and migrate (Fares et al., 2020). Moreover,
115 bacteria can induce epigenetic alterations upon gaining intracellular access that can activate
116 dormant tumor-promoting genes (Niller and Minarovits, 2016).

117 As a prototypical oncomicrobe that has received significant attention, *Fusobacterium nucleatum*'s
118 involvement in CRC has been extensively characterized in recent years and serves as a prime
119 example to highlight the multiple mechanisms pathogens can use to impact cancer progression.
120 High *Fusobacterium* levels in tumors correlate with decreased patient survival in CRC (Kunzmann et
121 al., 2019; Mima et al., 2016), pancreatic cancer (Mitsuhashi et al., 2015), and esophageal cancers
122 (Yamamura et al., 2016). An oral commensal microbe, it has been associated with periodontitis,
123 gingivitis, and multiple extra-oral diseases. Its surface adhesin, Fap2, targets the host carbohydrate
124 Gal/Gal-NAc (Abed et al., 2016; Parhi et al., 2020) that is overexpressed on many cancers
125 (Shamsuddin et al., 1995) and may explain how these bacteria are found in higher abundance in
126 CRC compared to the adjacent healthy tissue. Strikingly, it was found that *Fusobacteria* can travel
127 intracellularly within a migrating host CRC cell leading to bacterial seeding at distant sites such as
128 the liver (Bullman et al., 2017)(Figure 1D), yet it was unclear if these bacteria were active or passive
129 participants in this process. We more recently provided an early clue as to the answer to this latter
130 question, demonstrating that the cytokines IL8 and CXCL1 are specifically secreted upon *F.*
131 *nucleatum* invasion of HCT116 CRC cells and contribute to enhancing cancer cell migration directly
132 (Casasanta et al., 2020)(Figure 1E). This bacterium is able to induce alterations even at the
133 epigenomic level, where it was discovered that *F. nucleatum* infection, in conjunction with
134 *Hungatella hathewayi*, induces the hypermethylation of tumor suppressor gene promoters in
135 colonic epithelial tissue (Xia et al., 2020) (Figure 1F).

136 **Unanswered Questions**

137 These observations have generated a number of fundamental questions that we believe should be
138 a broad focus of researchers across multiple cancer types beyond the gut, including:

- 139 ▪ Are bacteria seeded early on in tumorigenesis, thereby actively contributing to tumor
140 initiation, or do they arrive at later stages?
- 141 ▪ Are specific features of the TME favorable for bacterial localization or
142 survival/proliferation?
- 143 ▪ How do the bacteria modify the TME?
- 144 ▪ Are there cooperative relationships among multiple bacteria types within the TME, or
145 between tumor and bacteria?
- 146 ▪ What factors govern the bacterial interactions with tumor-associated immune cells?
- 147 ▪ Does elimination of the internalized microbes reverse their effect on the tumor?
- 148 ▪ Can microbial signatures be used to identify the type or stage of the tumor and predict
149 therapy response and toxicity?
- 150 ▪ Why do some tumors host more diverse microbiota than others?
- 151 ▪ Can we harness the tissue or niche-specific bacterial colonization of cancers to enable
152 targeted delivery of therapeutics?

153
154 The consequences of these microbe-microbe and host-microbe interactions may materialize over
155 long time-scales. Nejman et al. suggest that there may be a low level of bacteria in every tissue and
156 bacterial translocation increases after disruption of the epithelial barrier and increased vascular
157 permeability (Nejman et al., 2020). Furthermore, evidence of alterations in metabolic profiles of
158 host cells and internalized bacteria (Kasper et al., 2020), as well as the secretion of inflammatory
159 cytokines, that may dramatically impact the hallmarks of cancer, raise questions on the role of
160 secreted factors in influencing tumor progression. Answers to these questions will help identify
161 novel therapeutic targets and reshape current cancer treatment procedures. However, many of
162 these questions have yet to be addressed directly due to a lack of representative models to study
163 tumor-resident bacteria.

164 In the next section, we discuss the challenges and limitations of existing methods to study host-
165 microbial interactions and how tissue engineering can help model the TME.

166 METHODS TO STUDY HOST-MICROBIAL INTERACTIONS

167 To systematically interrogate host-microbe interactions, targeted questions must be defined in
168 order to select appropriate experimental protocols (Fischbach, 2018). Experiments in animals and
169 humans have generally been limited to overall population/compositional studies via 16S ribosomal
170 RNA gene sequencing and shotgun metagenomics, due to difficulties in isolation and sampling and
171 downstream culture of bacteria (Jovel et al., 2016). Although focusing on mechanistic studies of
172 individual microbes may appear as a low-hanging fruit, microbes exhibit contrasting behaviors when
173 studied within a multi-species community. In fact, many metabolites are only produced in the
174 presence of other microbes (Bertrand et al., 2014). The secreted metabolites themselves may
175 directly affect the tumor viability and proliferation. For example, we have previously shown that
176 *Pseudomonas aeruginosa* found in breast cancer tissue secretes N-(3-oxododecanoyl)-L-
177 homoserine lactone which variably modulates viability of MDA-MB-231 and MCF-DCIS.com cells
178 depending on the specific culture microenvironment (Balhouse et al., 2017).

179 Paradigmatic changes in experimental and conceptual approaches are needed in order to develop
180 a comprehensive understanding of all the factors that influence host-microbiome interactions in
181 cancer. Challenges that have hindered this goal include difficulties in:

182

- 183 ▪ Isolating causal microorganism(s)
- 184 ▪ Preventing over-proliferation of single species in a multi-species model
- 185 ▪ Accurately replicating *in vivo* physiological geometry and biochemical cues
- 186 ▪ Limiting variability in organoid structures and batch-to-batch extracellular matrix (ECM)
187 composition
- 188 ▪ Developing cell culture medium supportive of all non-microbial cells within the model
- 189 ▪ Culturing anaerobic bacteria in oxygenated models.
- 190 ▪ Real-time monitoring of host and microbial cells and their associated biochemical
191 parameters
- 192 ▪ Recapitulating *in vivo* microbial community composition and immune cell interactions

193 Overcoming these bottlenecks is crucial to develop technologies to effectively dissect microbial
194 interactions with the host and to target these therapeutically.

195 Limitations of Animal Models

196 Animal models are frequently employed due to the availability of powerful genetic tools and
197 physiological relevance to humans. Moreover, with recent advances in whole-animal editing, animal
198 models are becoming increasingly “humanized”, and are ideal for long-term compound studies (Hay
199 et al., 2014). However, current animal models (i.e. xenograft tumor mice models) are poor
200 representatives of human biology since they exhibit distinct bacterial compositions and immune
201 profiles compared to humans (Mestas and Hughes, 2004). Furthermore, animal studies are
202 expensive and not as scalable and accessible as other *in vitro* models. More specifically, when it
203

204

205 comes to manipulating signaling molecules or growth factors, there is much less experimental
206 control, making it challenging to add or tune elements that are necessary to mimic a physiological
207 environment. Therefore, it is beneficial to utilize *in vitro* technologies that can provide valuable
208 insights at a fraction of the cost of transgenic animals, and can reduce dependence on animal
209 experimentation at earlier screening or discovery stages of research, or to help dissect specific
210 mechanism in later stages of study.

211 **Embracing Tissue Engineered Models**

212 Tissue engineering evolved as a strategy to build a tissue from the ground up and can prove to be a
213 viable tool to reconstruct physiologically relevant *in vitro* models. These techniques begin from
214 seeding cells in decellularized scaffolds, or ECM based hydrogels. The use of stem cells and induced
215 pluripotent stem cells (iPSCs) catapulted this field with organoid technology. Precise and tunable
216 control with microfluidic devices and microelectromechanical systems have added further control
217 and interrogation options to these technologies. Moreover, engineering principles guide the use of
218 mechanical articulation to simulate biophysical cues which influence the differentiation of cells.
219 However, a number of challenges still remain. While developing tissue engineered models, it is
220 essential to recreate the natural homeostatic environment as well as the tumor-specific
221 microenvironment, specifically for tumor-microbiome studies. Some challenges to this include
222 developing the normal or physiological environment first, before incorporating the pathological
223 tumor element.

224 **Design Considerations for a Complex Tumor Microenvironment**

226 There are a host of factors within the tumor microenvironment that need to be considered for
227 disease modeling. These factors markedly influence the type of microbes that colonize and infect
228 the tumor. **Figure 2** exemplifies the microenvironmental parameters to simulate the gut. There
229 exists complex chemical, pH, nutrient, and oxygen gradients throughout the length of the gut that,
230 for certain bacterial species, determine if the colonies are aerobic or anaerobic. The intestinal walls
231 are composed of several different cell types including enterocytes, enteroendocrine cells, Paneth
232 cells, goblet cells, M cells and Tuft cells, each with unique functions. These cells help establish the
233 epithelial barrier. The barrier itself has a high turnover rate and any compromise to barrier may lead
234 to microbial invasion and dissemination. Vascular and lymphatic networks skirt the walls of the
235 intestine. In addition, cancer associated fibroblasts (CAFs), tumor-associated macrophages (TAMs),
236 stromal cells, and myriad immune cells create a highly intricate and dynamic microenvironment.
237 Immune-host interactions are compartmentalized along the length of the intestinal tract which
238 additionally influences microbial diversity. A major factor to consider is mucus secreted by goblet
239 cells which significantly contributes to bacterial spatial aggregation (Schroeder, 2019).
240 Biomechanical cues arising from peristalsis of the gut and mucociliary flow invariably influence host
241 cell differentiation and the distribution of bacterial colonies. To further add to the tumor milieu, are
242 the microbes themselves, their virulence proteins, and synthesized metabolites, and a medley of
243 host secreted factors, cytokines, and gradients of soluble factors.

244

245 Similar complex microarchitectures can be described for cancers of the breast (Bahcecioglu et al.,
246 2020), pancreas (Ho et al., 2020), lung (Mittal et al., 2016), and other organs. Tissue engineering
247 strategies endeavor to recreate these complex architectures and organ-type specificities to more
248 accurately recapitulate host-microbial interactions in cancer. However, it is essential to note that
249 not all components may be required to study a specific interaction and may additionally impact the
250 reproducibility of studies.

251 In the following sections, we describe the utility of *in vitro* models that recapitulate specific features
252 of the TME and demonstrate their feasibility to study host-microbial interactions.

253 **IN VITRO MODELS**

254

255 There has been a logical evolution of in vitro models of increasing complexity that have enabled the
256 interrogation of host-microbial crosstalk. Despite limitations present in two-dimensional (2D) and
257 3D cell culture systems, both strategies offer fundamental advantages to explore specific aspects
258 influencing the host-microbiome dynamics. Far from comprehensive, the selected models highlight
259 key features and findings in this field.

260 **2D and 2.5D Models**

261 The simplest model used to study host-microbial interactions consists of human cell lines grown as
262 2-dimensional (2D) monolayers and inoculating microbes within the culture medium. An
263 advancement to this system is the hanging basket model which suspends a coverslip consisting of a
264 pre-grown multi-species biofilm of pathogenic bacteria over a monolayer of epithelial cells to study
265 gingival inflammation (Millhouse et al., 2014) (Figure 3A). Using a similar approach to simulate an
266 aerobic-anaerobic interface naturally occurring in the gut, the 'Human oxygen-Bacteria anaerobic'
267 (HoxBan) system (Sadaghian Sadabadi et al., 2015) utilized 50 mL culture tubes with host Caco-2
268 cells attached to a coverslip and positioned over a bacterial culture medium containing
269 *Faecalibacterium prausnitzii* (Figure 3B).

270 Transwell-based approaches (sometimes referred to as 2.5D) are widely used to assess migratory
271 and invasive responses of the host cells (Casasanta et al., 2020; Park et al., 2017). This culture format
272 consists of host cells seeded on top of a porous membrane or ECM-deposited membrane, with
273 microbes most commonly introduced to the apical side or conditioned media obtained from
274 infected cells to the lower chamber (Figure 3C). An advantage of Transwells is that they can be used
275 to produce polarized, differentiated, multi-layer epithelial cultures. Moreover, since an epithelial
276 monolayer harbors apical and basolateral compartments, this feature enables independent analysis
277 of secretomes per direction, explaining this model's popularity.

278 Recently, Li et al. introduced a 96-deep well plate-based culturing model (MiPro) that conserved the
279 functional and compositional profiles of individual gut microbiomes (Li et al., 2019). The MiPro setup
280 consists of microbiome samples cultured in a 96-deep well plate in which the plate is covered with
281 a silicone-gel cover and shaken at 500 rpm on a digital shaker (Figure 3D). The authors
282 demonstrated the applicability of this model for high-throughput drug-microbiome interaction
283 studies. More importantly, the MiPro system can be optimized for investigation of host-microbe
284 crosstalk.

285 Though 2D models offer simple low-cost maintenance and high reproducibility, they fundamentally
286 fail to mimic most of the natural 3D structures of tissues, resulting in the inability to induce complete
287 cell differentiation and recapitulate key physicochemical parameters. More specifically, control of
288 cellular bioactivities is insufficient due to cellular over-proliferation and lack of proper nutrient and
289 oxygen gradients which greatly affects the analysis of host-microbe dynamics. These limitations are
290 similar to those that have contributed to the disappointing track record of 2D culture for screening
291 of cancer drugs, due in part to the non-physiological hyperactive metabolism of cells cultured on
292 hard plastic surfaces (Cox et al., 2015).

293 **3D Models**

294 3D cell culture systems provide a relatively young but rapidly maturing approach to addressing the
295 inability of 2D models to reconstitute *in vivo* host-microbiome interactions. Representative
296 examples include the rotating wall vessel, hydrogel scaffolds, organoids and tumor spheroids. While
297 these have been in relatively wide use in tissue engineering broadly, and tumor engineering
298 specifically, we believe there is still great potential to leverage as well as to advance previously
299 developed approaches for tumor-microbiome interaction studies. In this section, we will outline
300 some of the more traditional approaches that have already been used to study host-microbe
301 interactions.

302 The rotating wall vessel (RWV) is a horizontally rotating cell culture chamber with suspended cells.
303 These cells are usually adhered to ECM-coated microcarriers, and rotation results in cell
304 aggregation. This construction induces continuous surface shear stress on the host cells to mimic
305 physiological fluid forces (Barrila et al., 2010)(Figure 3E). Radtke et al. implemented this culture
306 method to provide relevant pathological insights into human enteric *salmonellosis* (Radtke et al.,
307 2010). Using the RWV, Ilhan et al. established a human endometrial epithelial cell (EEC) model. By
308 incubating 13 *Prevotella* clinical strains isolated from the endometrium, vagina, amniotic fluid, and
309 oral cavity with the EEC model, the authors explored species-specific effects of *Prevotella* on
310 physiological and host defense responses in the human endometrial epithelium (Ilhan et al., 2020).

311 By leveraging biocompatible hydrogel materials and 3D bioprinting technologies, host tissue matrix
312 and functionality can be rebuilt with increased accuracy and robustness. Indeed, recent work has
313 shown successful replication of human intestinal epithelium microarchitecture with months-long
314 function *in vitro* (Chen et al., 2015). Silk protein was employed to construct a 3D hollow lumen and
315 to house human intestinal epithelial cells which were supported and nourished by surrounding
316 myofibroblasts (Figure 3F). The authors also showed the applicability of this model to microbiome
317 studies by using it to model *Yersinia pseudotuberculosis* and *Lactobacillus rhamnosus* infections.

318 Organoid technology has greatly advanced the field of tissue engineering. Since organoids more
319 accurately reproduce the complexity of multi-cellular tissue, these systems provide a more precise
320 picture of the host-microbiome interface. Organoids are generated from multiple sources including
321 adult/fetal tissues, embryonic stem cells, induced pluripotent stem cells, and recently, patient-
322 derived cells (Fujii et al., 2016; Shamir and Ewald, 2014; Yao et al., 2020). As organoids are typically
323 embedded in extracellular matrices, they can receive matrix cues to facilitate self-organization and
324 specific lineage commitment, resulting in production of near-native epithelial cell clusters (Bar-
325 Ephraim et al., 2020). This technology has been used to develop infectious models of pathogenic
326 *Helicobacter pylori* by microinjection into the organoid's lumen (Bartfeld et al., 2015; Schlaermann
327 et al., 2016)(Figure 3G). Recently, Pleguezuelos-Manzano et al. exposed human intestinal organoids
328 to genotoxic *pks⁺* *Escherichia coli* by repeated luminal injection over five months (Pleguezuelos-
329 Manzano et al., 2020). The authors were able to identify that colibactin secreted by *pks⁺* *E. coli*
330 directly caused distinct mutations in host epithelial cells; potentially putting individuals that harbor
331 this *E. coli* strain at an increased risk of CRC. Combining organoid technology with the RWV, Barrila
332 et al. demonstrated that incorporation of phagocytic macrophages into this 3D co-culture model
333 revealed the contribution of distinct cell types during host-pathogen interactions of infection,
334 (Barrila et al., 2017).

335 Human intestinal enteroids, developed from Lgr5+ stem cells in the intestinal crypts, are increasingly
336 used to study microbiome and host interactions. Enteroids contain most of the cell types normally
337 found within the intestinal lining and can be grown as 3D spheroids or 2D monolayers based on
338 experimental need. The Enteroid-Anaerobe Co-Culture system (EACC) developed by Fofanova et al.,
339 could recreate the steep oxygen gradients within the platform that are normally observed *in vivo*
340 (Fofanova et al., 2019). This model was constructed by seeding enteroids on Transwells placed
341 within modified gaskets sealed to a gas permeable 24-well plate and placed within an anaerobic
342 chamber. Using the EACC, the authors demonstrated the co-culture of the anaerobes, *Bacteroides*
343 *thetaiotaomicron* and *Blautia* sp. with patient derived enteroids for 24 hours. However, one specific
344 limitation of enteroids is their limited cytokine secretion in response to pro-inflammatory stimuli
345 which may significantly impact microbial induced inflammatory responses. Recent refinements to
346 the cell culture media that support enteroid growth endeavor to address this limitation (Ruan et al.,
347 2020).

348 Finally, tumor spheroid models rely exclusively on cellular aggregation of either homotypic or
349 heterotypic cells (Figure 3H), making this a non-scaffold-based culture method (Costa et al., 2014).
350 Several techniques are available for spheroid production including liquid overlay, hanging drop, U-
351 bottom microplates, microfluidic-based assembly, and spinner flasks (Nunes et al., 2019). Kasper et
352 al. has introduced a tumor spheroid model that promotes the growth of anaerobic bacteria (Kasper
353 et al., 2020). By directly co-culturing 28 *Fusobacterium* clinical isolates, the authors presented a

354 unique model to study intra-tumor anaerobic bacteria and analyze subsequent effects including
355 cancer-related gene expression and metabolomics.

356 While 3D models offer a middle ground between 2D cell cultures and in vivo models and exhibit
357 more accurate physiological response by recapitulating native cell-cell interactions, their
358 applicability can be limiting due to limited differentiation capacities, the inability to provide
359 biomechanical cues, constraints in real-time monitoring of host-microbial interactions, lack of
360 oxygen control, and the long-term maintenance of a stably sustained host-microbiome ecosystem.

361 **MICROFLUIDIC AND ORGAN-ON-CHIP MODELS**

362 Microfluidic organ-on-a-chip models are proving invaluable to dissect microenvironmental factors
363 governing interactions between microbial and human cells, particularly via secreted soluble factors.
364 Fluids are easily manipulated at the microscale allowing for precision tunability and the
365 development of reproducible chemical gradients (Barkal et al., 2017). At this scale, diffusive forces
366 dominate over convective mixing which enables laminar flow profiles to regulate the subtle
367 balances of chemicals and metabolites during infection. The basic template inscribing the fluid flow
368 profile for these devices are constructed through soft lithography techniques predominantly using
369 PDMS (polydimethylsiloxane) polymers. More sophisticated layered structures can be constructed
370 using microporous membranes. Culture medium is typically perfused through the device using a
371 syringe or peristaltic pump to maintain controlled flow rates that are used to manipulate fluid shear
372 stress on cells, which has been shown to directly impact their differentiation and morphogenesis.
373 Air and pressurized gas chambers have also been incorporated within these devices to recreate the
374 physical cyclic compression of the tissue. The following sections describe the current state-of-the-
375 art microfluidic models used to maintain co-cultures of host epithelial cells with microbes.

376 **The HuMiX model**

377 The HuMiX device (Human-microbial crosstalk) mimics the human gut, and allows for the analysis
378 of molecular crosstalk between the microbiome and human colorectal adenocarcinoma enterocytes
379 (Shah et al., 2016). This system consists of three co-laminar microchannels for medium perfusion,
380 human epithelial cell culture, and bacterial consortia. By providing a proximal 0.5-1 mm partition
381 for human and microbial microchambers across a nanoporous membrane, this perfusion bioreactor
382 reproduces a healthy intact epithelial barrier. Furthermore, the HuMiX setup integrates one
383 dedicated inlet and outlet per microchannel, oxygen sensors, and a commercial chopstick style
384 electrode for precise control of physicochemical parameters, accurate monitoring of oxygen
385 concentrations, and valid measurement of transepithelial electrical resistance (TEER). Moreover,
386 Shah et al. demonstrated how the HuMiX enabled recapitulation of transcriptional, metabolic, and
387 immune responses in human Caco-2 cells after co-culturing with probiotic *Lactobacillus rhamnosus*
388 GG.

389 **Gut Chip models**

390 Organ-on-a-chip models, such as the Gut-Chip, have advanced the field by incorporating mechanical
391 stimuli to boost cell differentiation and permit real-time monitoring and assessment of microbial
392 contribution to intestinal disease exacerbation. One such representative biomimetic system (Kim et
393 al., 2016a) contains a microfabricated porous elastic membrane sandwiched between an upper and
394 lower chamber to house intestinal Caco2 cells and endothelial cells (Figure 4 A-B). The device is
395 equipped with two hollow lateral vacuum chambers lining the chambers. By applying a cyclic
396 suction, the design effectively induces peristaltic-like motions and trickling-like flow on the intestinal
397 cells. Most importantly, this cyclic strain induced spontaneous villus morphogenesis of the Caco2
398 cells and differentiated into four lineages of small intestinal cells (absorptive, goblet,
399 enteroendocrine, and Paneth). Using this platform, the authors demonstrated that the addition of
400 lipopolysaccharide (LPS) induced the secretion of the cytokines IL-1 β , IL-6, IL-8, and TNF α into the
401 microvascular chamber replicating chronic inflammatory diseases (Kim et al., 2016a). In a follow-up

402 study, this system was co-cultured with a living microbiome and maintained viability for a week with
403 a mixed population of eight different facultative or obligate anaerobic, probiotic bacteria
404 (*Bifidobacterium breve*, *B. longum*, *B. infantis*, *Lactobacillus acidophilus*, *L. plantarum*, *L. paracasei*,
405 *L. bulgaricus*, and *Streptococcus thermophiles*) (Kim et al., 2016b). Employing the Gut-Chip, Grassart
406 et al. revealed that *Shigella* in effect hijacked the host intestinal microarchitecture and mechanical
407 forces to maximize its infectivity (Figure 4C) (Grassart et al., 2019).

408 The Gut-Chip platform demonstrates high adaptability and feasibility to study specific
409 microenvironmental parameters that may influence host-microbe interactions. For example, to
410 support a more diverse microbial community, the Gut-Chip was modified to enable precise oxygen
411 control. In the 'Anaerobic-intestine-on-a-chip', Jalili-Firoozinezhad et al. established an oxygen
412 gradient across the endothelial and epithelial interface of the Gut-Chip (Jalili-Firoozinezhad et al.,
413 2019). Six oxygen quenched fluorescent particles were embedded within the system to monitor
414 oxygen levels. This enhancement increased the duration for a complex microbiota co-culture with
415 patient-derived intestinal epithelium to at least five days. Furthermore, the introduction of *B. fragilis*
416 to the chip was found to enhance the barrier function of the intestinal lining. Using a
417 complementary approach, the Anoxic-Oxic Interface-on-a-chip (AOI) incorporated the simultaneous
418 flow of anoxic and oxic culture medium through dedicated microchannels to recreate an oxygen
419 gradient within the Gut-Chip (Figure 4D) (Shin et al., 2019). The authors employed TEER and
420 platinum dendrimer-encapsulated nanoparticles to quantify barrier permeability and monitor
421 oxygen gradient, respectively. The AOI chip was used to demonstrate increased viability of two
422 obligate anaerobic bacteria (*Bifidobacterium adolescentis* and *Eubacterium hallii*) in co-culture with
423 the gut epithelium. The Gut-Chip has also been used to model the colonic mucus layer structure
424 and function with the Colon-on-a-chip (Sontheimer-Phelps et al., 2020) (Figure 4F) and has been
425 used to study drug transport and metabolism in the Duodenum Intestine chip (Kasendra et al., 2020)
426 (Figure 4E).

427 Other Microphysiological Models

428 Notably, microphysiological developments of Gut Chip models also give rise to unique physiome-
429 on-chip platforms such as '10-MPS' and 'OrganoPlate' systems (Edington et al., 2018; Trietsch et al.,
430 2017). Edington et al. developed microphysiological systems (MPS) supporting '4-way'
431 (liver/immune, lung, gut/immune, and endometrium), '7-way' (4-MPS supplemented with brain,
432 heart, and pancreas), and even '10-way' (7-MPS with kidney, skin, and skeletal muscles) interactions
433 for weeks-long functional relevance. Together, their multi-MPS systems have enabled high-content
434 preclinical drug screening pipeline, and exhibit increasing appeal for studying the dynamics at host-
435 microbe interface. The OrganoPlate, developed in 2017, showed, for the first time, a comprehensive
436 approach to interrogate culture-perfused epithelia tubules that are exposed to an ECM. The
437 OrganoPlate setup consists of 40 microfluidic channel networks integrated in the bottom of a 384
438 well-plate format, wherein epithelial cells are introduced to a collagen ECM-housing lane adjacent
439 to culture medium lanes. Lanz et al. demonstrated the translational utility of this model through
440 therapy response testing of breast cancer (Lanz et al., 2017). Kramer et al. took it a step further by
441 placing the plate on a tilted rocking platform to create a height difference, subsequently
442 reproducing microfluidic interstitial flow to model intratumoral pressure in pancreatic ductal
443 adenocarcinoma (Kramer et al., 2019).

444 Tumor-on-a-Chip models

445 While tissue or organ-microbiome interaction models have shed light on normal physiological
446 processes, tumor-specific microbiomes have not been widely studied in such platforms. Tumor-on-
447 chip models based on organ-on-chip biomimetic principles hold great potential for recreation of
448 human TMEs and adoption to study host-microbiome crosstalk which may ultimately reveal both
449 similarities as well as differences among different tumor types and stages. In this section, we will
450 outline some models that have aimed to recapitulate the classical hallmarks of cancer, including
451 angiogenic induction, immune interactions, biophysical alterations within the TME, and cell

452 migration and metastasis. While we refer the reader to several excellent review topics that go into
453 more depth (Cox et al., 2015; Ma et al., 2018; Shang et al., 2019; Trujillo-de Santiago et al., 2019;
454 Tsai et al., 2017), here we will focus on some representative examples.

455 Vascularized multi-tissue organ models were the first to incorporate an endothelial layer juxtaposed
456 with an epithelial layer, combined with mechanical stretch. The earliest application of this model
457 was in the design of a lung-on-a-chip that accurately reconstituted the alveolar-capillary interface
458 and its surrounding microenvironment (Huh et al., 2010). This biomimetic lung model expanded the
459 capabilities to model other organs including the gut, breast, and pancreas, as well as in human
460 cancers. Chips that recreated vasculogenesis and angiogenesis have helped elucidate the molecular
461 mechanisms of angiogenic sprouting and serve as a foundation for future vascularized mechanical,
462 biochemical, and cellular studies (Hsu et al., 2013; Nguyen et al., 2013). For instance, the in vitro
463 vascularized microtumors (VMT) system encapsulates some of the complexity of the TME by the
464 addition of an ECM and stromal cells with nutrients perfused through microvessels (Sobrino et al.,
465 2016). Microvasculature embedded within 3D hydrogels are also commonly used in the
466 construction of these devices (Morgan et al., 2013).

467 Immune interactions are ubiquitous within the TME and may influence microbial residents. In a
468 multicellular tumor-on-a-chip platform, Aung et al. demonstrated that cancer cell-monocyte
469 interactions increased T cell recruitment (Aung et al., 2020). Other platforms have studied the
470 effects of macrophages and neutrophils migration and extravasation (Boussommier-Calleja et al.,
471 2016).

472 Several microphysiological devices have also emphasized the importance of the TME in influencing
473 cancer progression and treatment. 3D microengineered models of breast cancer, have revealed
474 insights into how the TME could contribute to an invasive phenotype (Choi et al., 2015). Using an
475 orthotopic lung-on-a-chip, Hassell et al. identified that physical cues from breathing motions could
476 influence lung cancer cell growth, invasion, and response to therapy (Hassell et al., 2017). In
477 addition, the HepaChip® integrated microfluidics and dielectrophoresis to better recapitulate the
478 3D microenvironment of pancreatic cancer and revealed that higher doses of Cisplatin are needed
479 to reduce the viability of Panc-1 pancreatic cancer cells when cultured in a more physiological
480 context (Beer et al., 2017). Other models have also investigated starvation-induced tumor cell
481 adaptations and resulting metabolic profiles that influence the development of necrotic cores in
482 large tumors (Ayuso et al., 2019). Platforms such as the Colorectal-tumor-on-a-chip have further
483 enabled studies in nanoparticle distribution in precision nanomedicine (Carvalho et al., 2019).

484 Finally, recent studies that implicate microbes impacting the metastatic potential of tumor makes
485 chips that study this phenomenon highly relevant (Coughlin and Kamm, 2020). Specific examples
486 include devices that monitor in vitro metastatic breast and brain tumors and their extravasation to
487 secondary tumor sites (Jeon et al., 2013; Xu et al., 2013).

488 Taken together, these microfluidic platforms are proving valuable to model intricate interactions to
489 help better understand the development and progression of cancer. With increased relevance of
490 the tumor microbiome in impacting these models, we believe that the development of an integrated
491 tumor microbiome-on-chip will be especially crucial to advance future studies in this field (Figure
492 5).

493 FUTURE PERSPECTIVES

494 We believe that the future of host-microbiota studies in the context of cancer should focus on the
495 development of next generation platforms to overcome current challenges including stable
496 culturing of user-defined bacterial communities, advancements in precise differentiation and
497 patterning of cells, improved perfusion capabilities, and incorporating immune cells.

498 Multiplex devices merging engineering and biology will be critical in this field. Advances on the
499 engineering front will go hand-in-hand with new biological insights. The development of effective

501 biomaterials and scaffolds is critical for recreating physiologically relevant tumor microbiome niches
502 *in vitro*. For example, by making key advancements in collagen hydrogel biomimetic platforms and
503 3D printed platforms, multiple biomaterials and cell types can be patterned (Datta et al., 2020;
504 Murphy et al., 2020) which could include tumor microbiotic participants. Gradients of soluble
505 factors, and even bacteria can be spatially patterned. However, some challenges to overcome are
506 the development of practical bioinks with desired properties, as well as improved mechanical
507 extrusion methods as these can harm cells. Alternative polymers may be used to address the
508 limitations of using PDMS for hypoxic studies.

509
510 Modular approaches need to be compatible with different analytical techniques, and modularity
511 can enable the combination of multiple devices. Automated biosensors can be integrated to
512 continually monitor and measure microenvironmental parameters. Furthermore, the
513 commercialization of these technologies will accelerate scale-up, improve robustness, refine
514 usability, and greatly reduce costs (Ramadan and Zourob, 2020). Advancements in analytical
515 techniques including mass spectrometry for proteomic and metabolite analysis, and epigenetic
516 profiling of ~ 100 cells is now possible using techniques such as MOWChIP (Cox et al., 2019; Zhu et
517 al., 2019) and will complement research in understanding the direct effects of bacteria on
518 tumorigenesis. With respect to visualization, innovative genetic tools are needed to create bacterial
519 mutant strains that express fluorescent proteins for visualization for live microscopy. Advanced
520 imaging technologies such as holographic imaging and light sheet microscopy can improve
521 resolution while live imaging in 3D.

522
523 Although complexity can be limitlessly extended and features added, the strengths of simpler 2D
524 and 3D models shouldn't be overlooked. To recapitulate the large number of variables and features
525 to develop a complex, multi-dimensional TME is a daunting task. However, preliminary and pilot
526 studies based on 2D and 3D culture can inform experimentation in more sophisticated platforms.
527 Population models inevitably will have to be performed with animal models. Observations from
528 individual bacterial species need to be connected to multi-species infection models which may
529 vastly alter the metabolome and infection dynamics. Organoids are increasing in complexity and
530 relevance and will play a huge role in studying immune effects on infected organoids as they are a
531 highly tractable system to study immune regulation (Bar-Ephraim et al., 2020).

532
533 Hypoxia, angiogenesis, metastasis, and immune dysfunction are some of the leading hallmarks of
534 cancer that are impacted by bacterial presence. The relatively immunosuppressed environment
535 within the TME can define multiple cancer stages and therapy response. Without a vascular
536 interface and immune modulation, the engineered models function in isolation greatly diminishing
537 their physiological relevance. Shear stress, pressure, flow rate, oxygen gradients, vascular
538 permeability, and tissue topography greatly affect bacterial localization and pathogenesis.
539 Lymphatic vessels and interstitial fluid pressure are emerging themes in cancer (Munson and Shieh,
540 2014) and provide a route for host cell dissemination. Many microbes are thought to follow a
541 hematogenous route to the tumor and thus, the incorporation of a microvascular and/or lymphatic
542 network is critical to study host bacterial localization and homing. Endothelialized blood vessel
543 models have already been developed for cancer studies which need to be integrated into host-
544 microbial platforms (Sontheimer-Phelps et al., 2019). The ability to develop and maintain hypoxic
545 gradients is especially crucial to study the pathogenic mechanisms used by anaerobic bacteria in
546 tumor cores and deep within the villus structure of the intestine. Real-time monitoring of oxygen
547 concentrations will assist in validation of the mechanistic insights derived from these anaerobic
548 models. The subsequent challenge will be co-culturing both aerobic and anaerobic species within
549 the same platform.

550
551 As a final note, more advanced platforms and high-throughput technology yield enormous amounts
552 of data that must be efficiently excavated. Here, computational techniques, bioinformatics, and
553 mathematical models can help coalesce disparate data and define experimental parameters.
554 Artificial intelligence and machine learning technologies could be more broadly applied to tissue
555 engineering studies where the degree of complexity may be substantial yet defined in a fully
556 deterministic manner (Fetah et al., 2019). New methods may need to be imported from disparate

557 fields such as ecology and evolutionary biology which may have useful quantitative analytical tools
558 to make sense of such complex interacting systems (Cunningham, 2019).

559
560 Ultimately, just as the Hallmarks of Cancer paradigm has led to promising approaches for targeting
561 these hallmarks as therapy, e.g. targeting the altered tumor metabolism, better understanding of
562 tumor-microbe interactions may also reveal new targetable tumor hallmarks. Bacteria can be
563 harnessed to target cancers directly using a Trojan horse approach to deliver drugs to the cells in
564 bacteriotherapy (Suh et al., 2019). While antibiotic therapy has shown promise for inhibiting tumor
565 growth in animal models (Bullman et al., 2017), improved models need to be designed to advance
566 such innovations to use for human patients. Important questions must still be answered, such as
567 will targeting and killing intracellular bacteria help control a tumor once it has advanced beyond a
568 certain point, or are there irreversible phenotypic changes driven by these bacteria that occur early
569 in the history of a given tumor? It is likely that the concept of the tumor microbiome could be too
570 intimately connected to the other hallmarks to be truly considered as distinct. In this case, it may
571 be helpful instead to focus on how tumor microbiota influence each of the other hallmarks. For
572 example, in our own work we have made key observations into mechanisms by which *F. nucleatum*
573 may directly drive metastatic phenotypes (Casasanta et al., 2020).

574 575 CONCLUDING REMARKS

576 The systematic characterization of the tumor microbiome and mounting evidence implicating the
577 role of 'oncomicrobes' in cancer indicate a need to revise our current understanding of the
578 hallmarks of cancer. Shifting from broad integrated microbiome studies to more focused studies
579 that characterize the multiple mechanisms that individual or cohorts of pathogens employ to infect
580 cells requires a conceptual shift to develop versatile experimental techniques to dissect host-
581 microbe crosstalk. Despite significant progress in cancer-focused tissue engineering, current
582 technologies do not completely recreate physiologically relevant systems and hence there is still a
583 preference for expensive and sometimes poorly representative animal models. However, with
584 progress in microfluidic and tissue engineered devices there remains much promise in this field. In
585 summary, recent tissue engineering advances in cancer have resulted in exciting new technologies
586 and biomimetic platforms to characterize host-microbial interactions, thereby opening avenues of
587 thought that could give rise to new paradigms of research and precision medicine.

588 Limitations of the Study

589 Not applicable.

590 Lead Contact

591 Further information and requests should be directed to and will be fulfilled by the Lead Contact,
592 Scott S. Verbridge (sverb@vt.edu).

593 Materials Availability

594 Not applicable.

595 Data and Code Availability

596 Not applicable.

597 Acknowledgements

598 This work has been supported through the NSF Career Award (CBET-1652112 Verbridge), NIH R21
599 Exploratory/Developmental Research Grant (1R21CA238630-01A1), and the Institute for Critical
600 Technologies and Applied Sciences at Virginia Tech. Some figures were created using Biorender
601 (Toronto).

602 Author Contributions

603 Writing – Original Draft: B.U. and T.T.D.N.; Writing – Review and Editing: B.U., T.T.D.N., D.J.S., and
604 S.S.V.; Visualization: B.U. and T.T.D.N.; Supervision: D.J.S. and S.S.V.; Project Administration: D.J.S.
605 and S.S.V.; Funding Acquisition: D.J.S. and S.S.V.

606 **Declaration of Interests**

607 The authors declare no competing interests.

REFERENCES

Abed, J., Emgård, J.E.M., Zamir, G., Faroja, M., Almogy, G., Grenov, A., Sol, A., Naor, R., Pikarsky, E., Atlan, K.A., et al. (2016). Fap2 Mediates *Fusobacterium nucleatum* Colorectal Adenocarcinoma Enrichment by Binding to Tumor-Expressed Gal-GalNAc. *Cell Host Microbe* 20, 215–225.

Aung, A., Kumar, V., Theprungsirikul, J., Davey, S.K., and Varghese, S. (2020). An Engineered Tumor-on-a-Chip Device with Breast Cancer–Immune Cell Interactions for Assessing T-cell Recruitment. *Cancer Res.* 80, 263–275.

Ayuso, J.M., Virumbrales-Munoz, M., McMinn, P.H., Rehman, S., Gomez, I., Karim, M.R., Trusttchel, R., Wisinski, K.B., Beebe, D.J., and Skala, M.C. (2019). Tumor-on-a-chip: a microfluidic model to study cell response to environmental gradients. *Lab. Chip* 19, 3461–3471.

Bahcecio glu, G., Basara, G., Ellis, B.W., Ren, X., and Zorlutuna, P. (2020). Breast cancer models: Engineering the tumor microenvironment. *Acta Biomater.* 106, 1–21.

Balhouse, B.N., Patterson, L., Schmelz, E.M., Slade, D.J., and Verbridge, S.S. (2017). N-(3-oxododecanoyl)-L-homoserine lactone interactions in the breast tumor microenvironment: Implications for breast cancer viability and proliferation in vitro. *PLOS ONE* 12, e0180372.

Bar-Ephraim, Y.E., Kretzschmar, K., and Clevers, H. (2020). Organoids in immunological research. *Nat. Rev. Immunol.* 20, 279–293.

Barkal, L.J., Berthier, E., Theberge, A.B., Keller, N.P., and Beebe, D.J. (2017). Multikingdom microscale models. *PLOS Pathog.* 13, e1006424.

Barrila, J., Radtke, A.L., Crabbé, A., Sarker, S.F., Herbst-Kralovetz, M.M., Ott, C.M., and Nickerson, C.A. (2010). Organotypic 3D cell culture models: using the rotating wall vessel to study host–pathogen interactions. *Nat. Rev. Microbiol.* 8, 791–801.

Barrila, J., Yang, J., Crabbé, A., Sarker, S.F., Liu, Y., Ott, C.M., Nelman-Gonzalez, M.A., Clemett, S.J., Nydam, S.D., Forsyth, R.J., et al. (2017). Three-dimensional organotypic co-culture model of intestinal epithelial cells and macrophages to study *Salmonella enterica* colonization patterns. *NPJ Microgravity* 3.

Bartfeld, S., Bayram, T., van de Wetering, M., Huch, M., Begthel, H., Kujala, P., Vries, R., Peters, P.J., and Clevers, H. (2015). In Vitro Expansion of Human Gastric Epithelial Stem Cells and Their Responses to Bacterial Infection. *Gastroenterology* 148, 126–136.e6.

Beer, M., Kuppalu, N., Stefanini, M., Becker, H., Schulz, I., Manoli, S., Schuette, J., Schmees, C., Casazza, A., Stelzle, M., et al. (2017). A novel microfluidic 3D platform for culturing pancreatic ductal adenocarcinoma cells: comparison with in vitro cultures and in vivo xenografts. *Sci. Rep.* 7.

Bertrand, S., Bohni, N., Schnee, S., Schumpp, O., Gindro, K., and Wolfender, J.-L. (2014). Metabolite induction via microorganism co-culture: a potential way to enhance chemical diversity for drug discovery. *Biotechnol. Adv.* 32, 1180–1204.

Boleij, A., Schaeps, R.M.J., and Tjalsma, H. (2009). Association between *Streptococcus bovis* and Colon Cancer. *J. Clin. Microbiol.* 47, 516.

Boussommier-Calleja, A., Li, R., Chen, M.B., Wong, S.C., and Kamm, R.D. (2016). Microfluidics: A New Tool for Modeling Cancer–Immune Interactions. *Trends Cancer* 2, 6–19.

Brestoff, J.R., and Artis, D. (2013). Commensal bacteria at the interface of host metabolism and the immune system. *Nat. Immunol.* 14, 676–684.

Bullman, S., Pedamallu, C.S., Sicinska, E., Clancy, T.E., Zhang, X., Cai, D., Neuberg, D., Huang, K., Guevara, F., Nelson, T., et al. (2017). Analysis of *Fusobacterium* persistence and antibiotic response in colorectal cancer. *Science* *358*, 1443–1448.

Carvalho, M.R., Barata, D., Teixeira, L.M., Giselbrecht, S., Reis, R.L., Oliveira, J.M., Truckenmüller, R., and Habibovic, P. (2019). Colorectal tumor-on-a-chip system: A 3D tool for precision onco-nanomedicine. *Sci. Adv.* *5*, eaaw1317.

Casasanta, M.A., Yoo, C.C., Udayasuryan, B., Sanders, B.E., Umaña, A., Zhang, Y., Peng, H., Duncan, A.J., Wang, Y., Li, L., et al. (2020). *Fusobacterium nucleatum* host-cell binding and invasion induces IL-8 and CXCL1 secretion that drives colorectal cancer cell migration. *Sci. Signal.* *13*.

Chen, Y., Lin, Y., Davis, K.M., Wang, Q., Rnjak-Kovacina, J., Li, C., Isberg, R.R., Kumamoto, C.A., Mecas, J., and Kaplan, D.L. (2015). Robust bioengineered 3D functional human intestinal epithelium. *Sci. Rep.* *5*, 13708.

Choi, Y., Hyun, E., Seo, J., Blundell, C., Kim, H.C., Lee, E., Lee, S.H., Moon, A., Moon, W.K., and Huh, D. (2015). A microengineered pathophysiological model of early-stage breast cancer. *Lab. Chip* *15*, 3350–3357.

Correa, P., and Piazuelo, M.B. (2011). *Helicobacter pylori* Infection and Gastric Adenocarcinoma. *US Gastroenterol. Hepatol. Rev.* *7*, 59–64.

Costa, E.C., Gaspar, V.M., Coutinho, P., and Correia, I.J. (2014). Optimization of liquid overlay technique to formulate heterogenic 3D co-cultures models. *Biotechnol. Bioeng.* *111*, 1672–1685.

Coughlin, M.F., and Kamm, R.D. (2020). The Use of Microfluidic Platforms to Probe the Mechanism of Cancer Cell Extravasation. *Adv. Healthc. Mater.* *9*, 1901410.

Cox, M.C., Reese, L.M., Bickford, L.R., and Verbridge, S.S. (2015). Toward the Broad Adoption of 3D Tumor Models in the Cancer Drug Pipeline. *ACS Biomater. Sci. Eng.* *1*, 877–894.

Cox, M.C., Deng, C., Naler, L.B., Lu, C., and Verbridge, S.S. (2019). Effects of Culture Condition on Epigenomic Profiles of Brain Tumor Cells. *ACS Biomater. Sci. Eng.* *5*, 1544–1552.

Cunningham, J.J. (2019). A call for integrated metastatic management. *Nat. Ecol. Evol.* *3*, 996–998.

Datta, P., Dey, M., Ataie, Z., Unutmaz, D., and Ozbolat, I.T. (2020). 3D bioprinting for reconstituting the cancer microenvironment. *Npj Precis. Oncol.* *4*, 1–13.

Durack, J., and Lynch, S.V. (2019). The gut microbiome: Relationships with disease and opportunities for therapy. *J. Exp. Med.* *216*, 20–40.

Edington, C.D., Chen, W.L.K., Geishecker, E., Kassis, T., Soenksen, L.R., Bhushan, B.M., Freake, D., Kirschner, J., Maass, C., Tsamandouras, N., et al. (2018). Interconnected Microphysiological Systems for Quantitative Biology and Pharmacology Studies. *Sci. Rep.* *8*, 4530.

van Elsland, D., and Neefjes, J. (2018). Bacterial infections and cancer. *EMBO Rep.* *19*, e46632.

Fan, Y., and Pedersen, O. (2020). Gut microbiota in human metabolic health and disease. *Nat. Rev. Microbiol.* 1–17.

Fares, J., Fares, M.Y., Khachfe, H.H., Salhab, H.A., and Fares, Y. (2020). Molecular principles of metastasis: a hallmark of cancer revisited. *Signal Transduct. Target. Ther.* *5*, 1–17.

Farinha, P., and Gascoyne, R.D. (2005). Helicobacter pylori and MALT lymphoma. *Gastroenterology* 128, 1579–1605.

Ferreccio, C. (2012). *Salmonella typhi* and Gallbladder Cancer. In *Bacteria and Cancer*, A.A. Khan, ed. (Dordrecht: Springer Netherlands), pp. 117–137.

Fetah, K.L., DiPardo, B.J., Kongadzem, E.-M., Tomlinson, J.S., Elzagheid, A., Elmusrati, M., Khademhosseini, A., and Ashammakhi, N. (2019). Cancer Modeling-on-a-Chip with Future Artificial Intelligence Integration. *Small Weinh. Bergstr. Ger.* 15, e1901985.

Fischbach, M.A. (2018). Microbiome: Focus on Causation and Mechanism. *Cell* 174, 785–790.

Fofanova, T.Y., Stewart, C.J., Auchtung, J.M., Wilson, R.L., Britton, R.A., Grande-Allen, K.J., Estes, M.K., and Petrosino, J.F. (2019). A novel human enteroid-anaerobe co-culture system to study microbial-host interaction under physiological hypoxia. *BioRxiv* 555755.

Fujii, M., Shimokawa, M., Date, S., Takano, A., Matano, M., Nanki, K., Ohta, Y., Toshimitsu, K., Nakazato, Y., Kawasaki, K., et al. (2016). A Colorectal Tumor Organoid Library Demonstrates Progressive Loss of Niche Factor Requirements during Tumorigenesis. *Cell Stem Cell* 18, 827–838.

Fulbright, L.E., Ellermann, M., and Arthur, J.C. (2017). The microbiome and the hallmarks of cancer. *PLoS Pathog.* 13.

Geller, L.T., Barzily-Rokni, M., Danino, T., Jonas, O.H., Shental, N., Nejman, D., Gavert, N., Zwang, Y., Cooper, Z.A., Shee, K., et al. (2017). Potential role of intratumor bacteria in mediating tumor resistance to the chemotherapeutic drug gemcitabine. *Science* 357, 1156–1160.

Gopalakrishnan, V., Spencer, C.N., Nezi, L., Reuben, A., Andrews, M.C., Karpinets, T.V., Prieto, P.A., Vicente, D., Hoffman, K., Wei, S.C., et al. (2018). Gut microbiome modulates response to anti–PD-1 immunotherapy in melanoma patients. *Science* 359, 97–103.

Grassart, A., Malardé, V., Gobaa, S., Sartori-Rupp, A., Kerns, J., Karalis, K., Marteyn, B., Sansonetti, P., and Sauvionnet, N. (2019). Bioengineered Human Organ-on-Chip Reveals Intestinal Microenvironment and Mechanical Forces Impacting *Shigella* Infection. *Cell Host Microbe* 26, 435–444.e4.

Greathouse, K.L., White, J.R., Vargas, A.J., Bliskovsky, V.V., Beck, J.A., von Muhlinen, N., Polley, E.C., Bowman, E.D., Khan, M.A., Robles, A.I., et al. (2018). Interaction between the microbiome and TP53 in human lung cancer. *Genome Biol.* 19, 123.

Hanahan, D., and Weinberg, R.A. (2011). Hallmarks of Cancer: The Next Generation. *Cell* 144, 646–674.

Hassell, B.A., Goyal, G., Lee, E., Sontheimer-Phelps, A., Levy, O., Chen, C.S., and Ingber, D.E. (2017). Human Organ Chip Models Recapitulate Orthotopic Lung Cancer Growth, Therapeutic Responses, and Tumor Dormancy In Vitro. *Cell Rep.* 21, 508–516.

Hay, M., Thomas, D.W., Craighead, J.L., Economides, C., and Rosenthal, J. (2014). Clinical development success rates for investigational drugs. *Nat. Biotechnol.* 32, 40–51.

Ho, W.J., Jaffee, E.M., and Zheng, L. (2020). The tumour microenvironment in pancreatic cancer - clinical challenges and opportunities. *Nat. Rev. Clin. Oncol.*

Hsu, Y.-H., Moya, M.L., Hughes, C.C.W., George, S.C., and Lee, A.P. (2013). A microfluidic platform for generating large-scale nearly identical human microphysiological vascularized tissue arrays. *Lab. Chip* 13, 2990–2998.

Huh, D., Matthews, B.D., Mammoto, A., Montoya-Zavala, M., Hsin, H.Y., and Ingber, D.E. (2010). Reconstituting Organ-Level Lung Functions on a Chip. *Science* 328, 1662–1668.

Ilhan, Z.E., Łaniewski, P., Tonachio, A., and Herbst-Kralovetz, M.M. (2020). Members of Prevotella genus distinctively modulate innate immune and barrier functions in a human three-dimensional endometrial epithelial cell model. *J. Infect. Dis.*

Jalili-Firoozinezhad, S., Gazzaniga, F.S., Calamari, E.L., Camacho, D.M., Fadel, C.W., Bein, A., Swenor, B., Nestor, B., Cronce, M.J., Tovaglieri, A., et al. (2019). A complex human gut microbiome cultured in an anaerobic intestine-on-a-chip. *Nat. Biomed. Eng.* 3, 520–531.

Jeon, J.S., Zervantakis, I.K., Chung, S., Kamm, R.D., and Charest, J.L. (2013). In vitro model of tumor cell extravasation. *PLoS One* 8, e56910.

Jovel, J., Patterson, J., Wang, W., Hotte, N., O'Keefe, S., Mitchel, T., Perry, T., Kao, D., Mason, A.L., Madsen, K.L., et al. (2016). Characterization of the Gut Microbiome Using 16S or Shotgun Metagenomics. *Front. Microbiol.* 7, 459.

Kadosh, E., Snir-Alkalay, I., Venkatachalam, A., May, S., Lasry, A., Elyada, E., Zinger, A., Shaham, M., Vaalani, G., Mernberger, M., et al. (2020). The gut microbiome switches mutant p53 from tumour-suppressive to oncogenic. *Nature* 1–6.

Kasendra, M., Luc, R., Yin, J., Manatakis, D.V., Kulkarni, G., Lucchesi, C., Sliz, J., Apostolou, A., Sunuwar, L., Obrigewitch, J., et al. (2020). Duodenum Intestine-Chip for preclinical drug assessment in a human relevant model. *ELife* 9, e50135.

Kasper, S.H., Morell-Perez, C., Wyche, T.P., Sana, T.R., Lieberman, L.A., and Hett, E.C. (2020). Colorectal cancer-associated anaerobic bacteria proliferate in tumor spheroids and alter the microenvironment. *Sci. Rep.* 10, 5321.

Kim, H.J., Li, H., Collins, J.J., and Ingber, D.E. (2016a). Contributions of microbiome and mechanical deformation to intestinal bacterial overgrowth and inflammation in a human gut-on-a-chip. *Proc. Natl. Acad. Sci.* 113, E7–E15.

Kim, H.J., Lee, J., Choi, J.-H., Bahinski, A., and Ingber, D.E. (2016b). Co-culture of Living Microbiome with Microengineered Human Intestinal Villi in a Gut-on-a-Chip Microfluidic Device. *J. Vis. Exp. JoVE*.

Kostic, A.D., Gevers, D., Pedamallu, C.S., Michaud, M., Duke, F., Earl, A.M., Ojesina, A.I., Jung, J., Bass, A.J., Tabernero, J., et al. (2012). Genomic analysis identifies association of *Fusobacterium* with colorectal carcinoma. *Genome Res.* 22, 292–298.

Kramer, B., de Haan, L., Vermeer, M., Olivier, T., Hankemeier, T., Vulto, P., Joore, J., and Lanz, H.L. (2019). Interstitial Flow Recapitulates Gemcitabine Chemoresistance in A 3D Microfluidic Pancreatic Ductal Adenocarcinoma Model by Induction of Multidrug Resistance Proteins. *Int. J. Mol. Sci.* 20.

Kunzmann, A.T., Proença, M.A., Jordao, H.W., Jiraskova, K., Schneiderova, M., Levy, M., Liska, V., Buchler, T., Vodickova, L., Vymetalkova, V., et al. (2019). *Fusobacterium nucleatum* tumor DNA levels are associated with survival in colorectal cancer patients. *Eur. J. Clin. Microbiol. Infect. Dis.* 38, 1891–1899.

Lanz, H.L., Saleh, A., Kramer, B., Cairns, J., Ng, C.P., Yu, J., Trietsch, S.J., Hankemeier, T., Joore, J., Vulto, P., et al. (2017). Therapy response testing of breast cancer in a 3D high-throughput perfused microfluidic platform. *BMC Cancer* *17*.

Li, L., Abou-Samra, E., Ning, Z., Zhang, X., Mayne, J., Wang, J., Cheng, K., Walker, K., Stintzi, A., and Figeys, D. (2019). An in vitro model maintaining taxon-specific functional activities of the gut microbiome. *Nat. Commun.* *10*, 4146.

Ma, Y.-H.V., Middleton, K., You, L., and Sun, Y. (2018). A review of microfluidic approaches for investigating cancer extravasation during metastasis. *Microsyst. Nanoeng.* *4*, 1–13.

Mestas, J., and Hughes, C.C.W. (2004). Of Mice and Not Men: Differences between Mouse and Human Immunology. *J. Immunol.* *172*, 2731–2738.

Millhouse, E., Jose, A., Sherry, L., Lappin, D.F., Patel, N., Middleton, A.M., Pratten, J., Culshaw, S., and Ramage, G. (2014). Development of an in vitro periodontal biofilm model for assessing antimicrobial and host modulatory effects of bioactive molecules. *BMC Oral Health* *14*, 80.

Mima, K., Nishihara, R., Qian, Z.R., Cao, Y., Sukawa, Y., Nowak, J.A., Yang, J., Dou, R., Masugi, Y., Song, M., et al. (2016). *Fusobacterium nucleatum* in colorectal carcinoma tissue and patient prognosis. *Gut* *65*, 1973–1980.

Mitsuhashi, K., Noshio, K., Sukawa, Y., Matsunaga, Y., Ito, M., Kurihara, H., Kanno, S., Igarashi, H., Naito, T., Adachi, Y., et al. (2015). Association of *Fusobacterium* species in pancreatic cancer tissues with molecular features and prognosis. *Oncotarget* *6*, 7209–7220.

Mittal, V., El Rayes, T., Narula, N., McGraw, T.E., Altorki, N.K., and Barcellos-Hoff, M.H. (2016). The Microenvironment of Lung Cancer and Therapeutic Implications. In *Lung Cancer and Personalized Medicine: Novel Therapies and Clinical Management*, A. Ahmad, and S.M. Gadgeel, eds. (Cham: Springer International Publishing), pp. 75–110.

Morgan, J.P., Delnero, P.F., Zheng, Y., Verbridge, S.S., Chen, J., Craven, M., Choi, N.W., Diaz-Santana, A., Kermani, P., Hempstead, B., et al. (2013). Formation of microvascular networks in vitro. *Nat. Protoc.* *8*, 1820–1836.

Munson, J.M., and Shieh, A.C. (2014). Interstitial fluid flow in cancer: implications for disease progression and treatment. *Cancer Manag. Res.* *6*, 317–328.

Murphy, S.V., De Coppi, P., and Atala, A. (2020). Opportunities and challenges of translational 3D bioprinting. *Nat. Biomed. Eng.* *4*, 370–380.

Nejman, D., Livyatan, I., Fuks, G., Gavert, N., Zwang, Y., Geller, L.T., Rotter-Maskowitz, A., Weiser, R., Mallel, G., Gigi, E., et al. (2020). The human tumor microbiome is composed of tumor type-specific intracellular bacteria. *Science* *368*, 973–980.

Nguyen, D.-H.T., Stapleton, S.C., Yang, M.T., Cha, S.S., Choi, C.K., Galie, P.A., and Chen, C.S. (2013). Biomimetic model to reconstitute angiogenic sprouting morphogenesis in vitro. *Proc. Natl. Acad. Sci.* *110*, 6712–6717.

Niller, H.H., and Minarovits, J. (2016). Patho-epigenetics of Infectious Diseases Caused by Intracellular Bacteria. In *Patho-Epigenetics of Infectious Disease*, J. Minarovits, and H.H. Niller, eds. (Cham: Springer International Publishing), pp. 107–130.

Nunes, A.S., Barros, A.S., Costa, E.C., Moreira, A.F., and Correia, I.J. (2019). 3D tumor spheroids as in vitro models to mimic in vivo human solid tumors resistance to therapeutic drugs. *Biotechnol. Bioeng.* **116**, 206–226.

Parhi, L., Alon-Maimon, T., Sol, A., Nejman, D., Shhadeh, A., Fainsod-Levi, T., Yajuk, O., Isaacson, B., Abed, J., Maalouf, N., et al. (2020). Breast cancer colonization by *Fusobacterium nucleatum* accelerates tumor growth and metastatic progression. *Nat. Commun.* **11**, 3259.

Park, G.-S., Park, M.H., Shin, W., Zhao, C., Sheikh, S., Oh, S.J., and Kim, H.J. (2017). Emulating Host-Microbiome Ecosystem of Human Gastrointestinal Tract in Vitro. *Stem Cell Rev. Rep.* **13**, 321–334.

Pleguezuelos-Manzano, C., Puschhof, J., Rosendahl Huber, A., van Hoeck, A., Wood, H.M., Nomburg, J., Gurjao, C., Manders, F., Dalmasso, G., Stege, P.B., et al. (2020). Mutational signature in colorectal cancer caused by genotoxic pks + *E. coli*. *Nature* **580**, 269–273.

Radtke, A.L., Wilson, J.W., Sarker, S., and Nickerson, C.A. (2010). Analysis of Interactions of *Salmonella* Type Three Secretion Mutants with 3-D Intestinal Epithelial Cells. *PLoS ONE* **5**.

Ramadan, Q., and Zourob, M. (2020). Organ-on-a-chip engineering: Toward bridging the gap between lab and industry. *Biomicrofluidics* **14**, 041501.

Ruan, W., Engevik, M.A., Chang-Graham, A.L., Danhof, H.A., Goodwin, A., Engevik, K.A., Shi, Z., Hall, A., Rienzi, S.C.D., Venable, S., et al. (2020). Enhancing responsiveness of human jejunal enteroids to host and microbial stimuli. *J. Physiol.* **598**, 3085–3105.

Sadaghian Sadabad, M., von Martels, J.Z.H., Khan, M.T., Blokzijl, T., Paglia, G., Dijkstra, G., Harmsen, H.J.M., and Faber, K.N. (2015). A simple coculture system shows mutualism between anaerobic faecalibacteria and epithelial Caco-2 cells. *Sci. Rep.* **5**.

Schlaermann, P., Toelle, B., Berger, H., Schmidt, S.C., Glanemann, M., Ordemann, J., Bartfeld, S., Mollenkopf, H.J., and Meyer, T.F. (2016). A novel human gastric primary cell culture system for modelling *Helicobacter pylori* infection in vitro. *Gut* **65**, 202–213.

Schroeder, B.O. (2019). Fight them or feed them: how the intestinal mucus layer manages the gut microbiota. *Gastroenterol. Rep.* **7**, 3–12.

Sender, R., Fuchs, S., and Milo, R. (2016). Revised Estimates for the Number of Human and Bacteria Cells in the Body. *PLOS Biol.* **14**, e1002533.

Shah, P., Fritz, J.V., Glaab, E., Desai, M.S., Greenhalgh, K., Frachet, A., Niegowska, M., Estes, M., Jäger, C., Seguin-Devaux, C., et al. (2016). A microfluidics-based in vitro model of the gastrointestinal human–microbe interface. *Nat. Commun.* **7**, 11535.

Shamir, E.R., and Ewald, A.J. (2014). Three-dimensional organotypic culture: experimental models of mammalian biology and disease. *Nat. Rev. Mol. Cell Biol.* **15**, 647–664.

Shamsuddin, A.M., Tyner, G.T., and Yang, G.Y. (1995). Common expression of the tumor marker D-galactose-beta-[1-->3]-N-acetyl-D-galactosamine by different adenocarcinomas: evidence of field effect phenomenon. *Cancer Res.* **55**, 149–152.

Shang, M., Soon, R.H., Lim, C.T., Khoo, B.L., and Han, J. (2019). Microfluidic modelling of the tumor microenvironment for anti-cancer drug development. *Lab. Chip* **19**, 369–386.

Shin, W., Wu, A., Massidda, M.W., Foster, C., Thomas, N., Lee, D.-W., Koh, H., Ju, Y., Kim, J., and Kim, H.J. (2019). A Robust Longitudinal Co-culture of Obligate Anaerobic Gut Microbiome With Human Intestinal Epithelium in an Anoxic-Oxic Interface-on-a-Chip. *Front. Bioeng. Biotechnol.* 7, 13.

Sobrino, A., Phan, D.T.T., Datta, R., Wang, X., Hachey, S.J., Romero-López, M., Gratton, E., Lee, A.P., George, S.C., and Hughes, C.C.W. (2016). 3D microtumors in vitro supported by perfused vascular networks. *Sci. Rep.* 6, 31589.

Sontheimer-Phelps, A., Hassell, B.A., and Ingber, D.E. (2019). Modelling cancer in microfluidic human organs-on-chips. *Nat. Rev. Cancer* 19, 65–81.

Sontheimer-Phelps, A., Chou, D.B., Tovaglieri, A., Ferrante, T.C., Duckworth, T., Fadel, C., Frismantas, V., Sutherland, A.D., Jalili-Firoozinezhad, S., Kasendra, M., et al. (2020). Human Colon-on-a-Chip Enables Continuous In Vitro Analysis of Colon Mucus Layer Accumulation and Physiology. *Cell. Mol. Gastroenterol. Hepatol.* 9, 507–526.

Suh, S., Jo, A., Traore, M.A., Zhan, Y., Coutermash-Ott, S.L., Ringel-Scaia, V.M., Allen, I.C., Davis, R.M., and Behkam, B. (2019). Nanoscale Bacteria-Enabled Autonomous Drug Delivery System (NanoBEADS) Enhances Intratumoral Transport of Nanomedicine. *Adv. Sci.* 6, 1801309.

Trietsch, S.J., Naumovska, E., Kurek, D., Setyawati, M.C., Vormann, M.K., Wilschut, K.J., Lanz, H.L., Nicolas, A., Ng, C.P., Joore, J., et al. (2017). Membrane-free culture and real-time barrier integrity assessment of perfused intestinal epithelium tubes. *Nat. Commun.* 8, 262.

Trujillo-de Santiago, G., Flores-Garza, B.G., Tavares-Negrete, J.A., Lara-Mayorga, I.M., González-Gamboa, I., Zhang, Y.S., Rojas-Martínez, A., Ortiz-López, R., and Álvarez, M.M. (2019). The Tumor-on-Chip: Recent Advances in the Development of Microfluidic Systems to Recapitulate the Physiology of Solid Tumors. *Materials* 12.

Tsai, H.-F., Trubelja, A., Shen, A.Q., and Bao, G. (2017). Tumour-on-a-chip: microfluidic models of tumour morphology, growth and microenvironment. *J. R. Soc. Interface* 14, 20170137.

Udayasuryan, B., Slade, D.J., and Verbridge, S.S. (2019). Microfluidics in Microbiome and Cancer Research (John Wiley & Sons, Ltd).

Uemura, N., Okamoto, S., Yamamoto, S., Matsumura, N., Yamaguchi, S., Yamakido, M., Taniyama, K., Sasaki, N., and Schlemper, R.J. (2001). *Helicobacter pylori* infection and the development of gastric cancer. *N. Engl. J. Med.* 345, 784–789.

Xavier, J.B., Young, V.B., Skufca, J., Ginty, F., Testerman, T., Pearson, A.T., Macklin, P., Mitchell, A., Shmulevich, I., Xie, L., et al. (2020). The Cancer Microbiome: Distinguishing Direct and Indirect Effects Requires a Systemic View. *Trends Cancer* 6, 192–204.

Xia, X., Wu, W.K.K., Wong, S.H., Liu, D., Kwong, T.N.Y., Nakatsu, G., Yan, P.S., Chuang, Y.-M., Chan, M.W.-Y., Coker, O.O., et al. (2020). Bacteria pathogens drive host colonic epithelial cell promoter hypermethylation of tumor suppressor genes in colorectal cancer. *Microbiome* 8, 108.

Xu, Z., Gao, Y., Hao, Y., Li, E., Wang, Y., Zhang, J., Wang, W., Gao, Z., and Wang, Q. (2013). Application of a microfluidic chip-based 3D co-culture to test drug sensitivity for individualized treatment of lung cancer. *Biomaterials* 34, 4109–4117.

Yamamura, K., Baba, Y., Nakagawa, S., Mima, K., Miyake, K., Nakamura, K., Sawayama, H., Kinoshita, K., Ishimoto, T., Iwatsuki, M., et al. (2016). Human Microbiome Fusobacterium Nucleatum in Esophageal Cancer Tissue Is Associated with Prognosis. *Clin. Cancer Res.* 22, 5574–5581.

Yao, Y., Xu, X., Yang, L., Zhu, J., Wan, J., Shen, L., Xia, F., Fu, G., Deng, Y., Pan, M., et al. (2020). Patient-Derived Organoids Predict Chemoradiation Responses of Locally Advanced Rectal Cancer. *Cell Stem Cell* 26, 17-26.e6.

Zhan, P., Suo, L., Qian, Q., Shen, X., Qiu, L.-X., Yu, L., and Song, Y. (2011). Chlamydia pneumoniae infection and lung cancer risk: a meta-analysis. *Eur. J. Cancer Oxf. Engl.* 1990 47, 742–747.

Zhu, B., Hsieh, Y.-P., Murphy, T.W., Zhang, Q., Naler, L.B., and Lu, C. (2019). MOWChIP-seq for low-input and multiplexed profiling of genome-wide histone modifications. *Nat. Protoc.* 14, 3366–3394.

Main Figure Titles

Figure 1. Characterization of the tumor microbiome and its functional relationship with carcinogenesis

Figure 2: A multitude of interacting factors within the complex tumor microenvironment of CRC.

Figure 3. Techniques used to study host-microbiome crosstalk.

Figure 4: The Gut Chip and models derived from it.

Figure 5: Incorporating salient features from Tumor-on-a-chip models to develop a Tumor-Microbiome-on-a-chip model.

Figure 6: Summary of current models.

Figure Legends

Figure 1. Characterization of the tumor microbiome and its functional relationship with carcinogenesis (A) Nejman et al. characterized the cancer microbiome profile of 1526 human tumors across seven different tumor types. The presence of bacteria was assessed by bacterial 16S rDNA qPCR. (B) There is high diversity of microbial species across the tumor types. (C) Nejman et al. also characterized the prevalence of 19 bacterial species across the different tumor types. Reproduced from Nejman et al., 2020 (D) Bullman et al. demonstrated persistence of *F. nucleatum* in patient-derived xenografts over a period of 204 days. Reproduced with permission from Bullman et al., 2017 (E) Casasanta et al. showed *Fusobacterium nucleatum* induced IL-8 and CXCL1 secretion from HCT116 colorectal cancer cells, driving cell migration *in vitro* using *F. nucleatum* conditioned media and that could be blocked by inhibiting bacterial internalization. Reproduced with permission from (Casasanta et al., 2020). (F) Xia et al. showed increased positive correlation of interactions involving *Fusobacterium nucleatum* and *Hungatella hathewayi* and the genes MLH1, APC, PTEN, and CDX2 in colorectal cancer (CC-BY-4.0 (<https://creativecommons.org/licenses/by/4.0/>)).

Figure 2: A multitude of interacting factors within the complex tumor microenvironment of CRC. The gut microenvironment is composed of several different cell types including epithelial, endothelial, and immune cells. Thousands of microbial species co-exist within the microenvironment. The host cells maintain the barrier integrity through multiple host defense strategies whereas the microbes utilize multiple virulence strategies to target the host. There are variations in oxygen levels within the folds of the microvilli and the tumor specific microenvironment can exhibit changes in local ECM composition. Vascular networks can be hijacked by tumor cells via angiogenic signaling which can enable cancer cell metastasis. Furthermore, many alterations can occur upon bacterial intracellular invasion that could contribute to tumor formation. (Figure panels adopted from (Barkal et al., 2017))

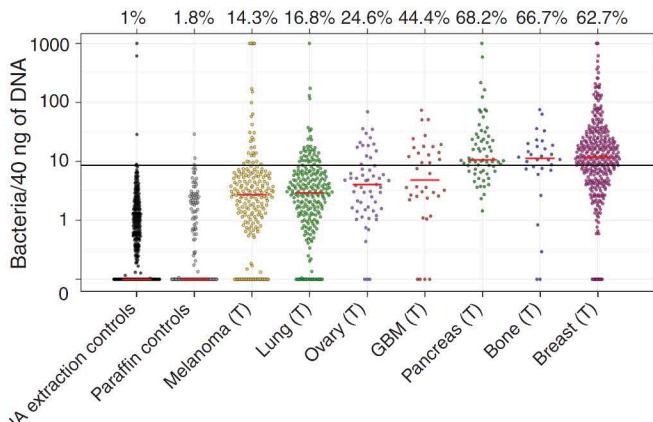
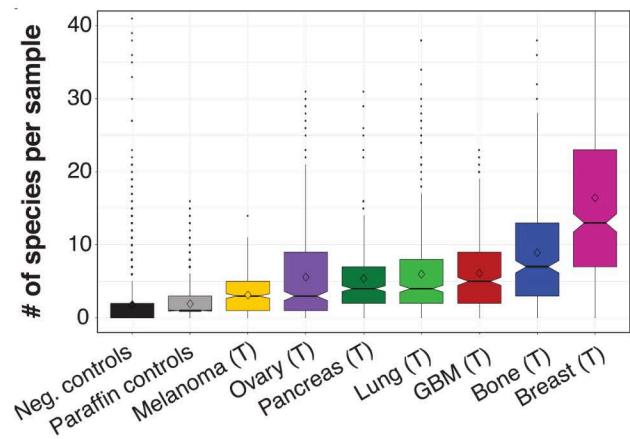
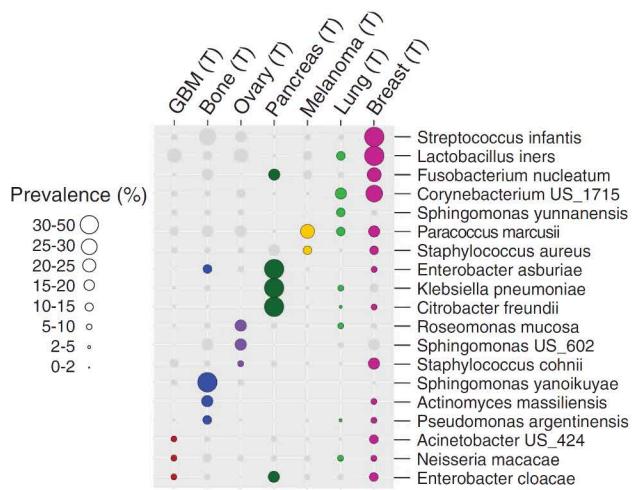
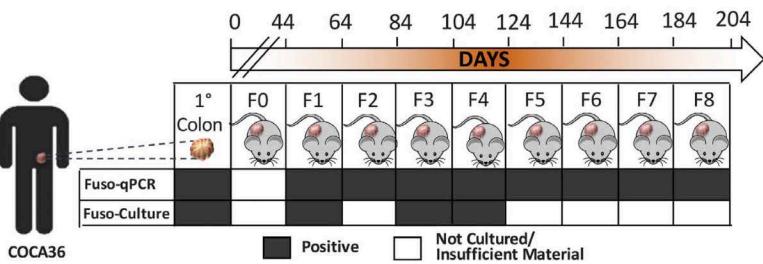
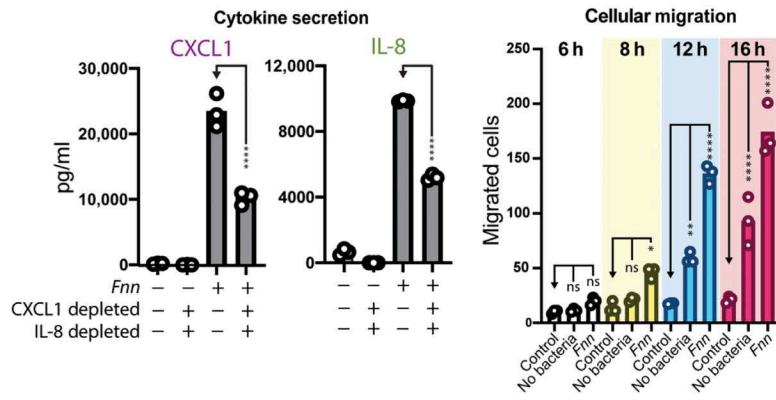
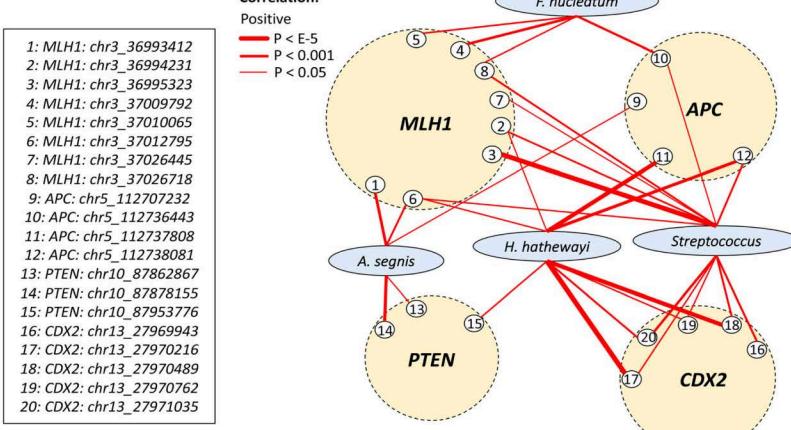
Figure 3. Techniques used to study host-microbiome crosstalk. 2D *vitro* static models include: (A) the Hanging Basket model used to study gingival inflammation (Millhouse et al., 2014), (B) the HoxBan model which co-cultured *Faecalibacterium prausnitzii* in medium overlaid with Caco-2 cells attached to a coverslip (Sadaghian Sadabadi et al., 2015), (C) the Transwell model that studied the migration of host HCT116 cells in response to conditioned media from *Fusobacterium nucleatum* infected HCT116 cells (Image reproduced with permission from Casasanta et al., 2020), and (D) The MiPro model utilizes a shaking 96-deep well microplate format to culture microbiome samples. 3D host-microbe culture systems include: (E) the rotating wall vessel (RWV) that induces continuous fluid rotation, ultimately enabling formation of 3D aggregates to model *Salmonella enterica* infection (Radtke et al., 2010), (F) hydrogel embedding, where a 3D porous silk scaffolds was used to reconstitute human intestinal model (Chen et al., 2015), (G) 3D organoid models to mimic host-pathogen interactions. (Schlaermann et al., 2016)., CC-BY-4.0

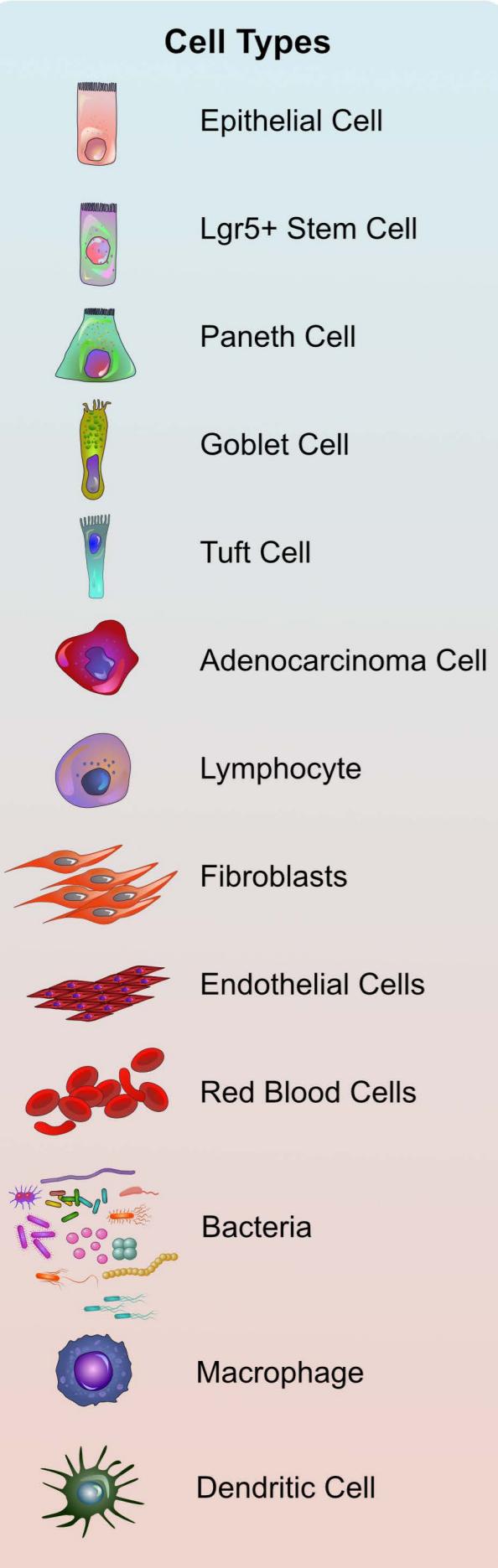
(<https://creativecommons.org/licenses/by/4.0/>), and (H) tumor spheroids, used to co-culture intra-tumor Fusobacteria with colorectal tumor spheroids in a microplate platform (Kasper et al., 2020).

Figure 4: The Gut Chip and models derived from it. **(A)** The first human-gut-on-a-chip (The Gut Chip) was used to demonstrate that probiotic gut microbiota can protect against enteroinvasive *E. Coli* (EIEC)-induced, immune cell-associated injury on chip with the presence of PBMCs (immune cells) (Reproduced with permission from Kim et al., 2016a). **(B)** Schematic of the Gut Chip. **(C)** *Shigella*-WT-GFP (green) infections in the Intestine Chip is dependent upon flow and stretch (Reproduced with permission from Grassart et al., 2019). **(D)** The Anoxic-Oxic interface (AOI) on a chip by generating an oxygen gradient by balancing the flow rates of anoxic and oxic culture medium. The authors co-cultured *B. adolescentis* with the 3D epithelial cells on the AOI Chip which show significantly increased viability and demonstrate enhanced barrier integrity quantified using TEER (Shin et al., 2019) CC-BY-4.0 (<https://creativecommons.org/licenses/by/4.0/>). **(E)** The Duodenum Intestine-chip shows multi-lineage differentiation the human intestine and show the expression of markers specific for each differentiated intestinal type (Kasendra et al., 2020) CC-BY-4.0 (<https://creativecommons.org/licenses/by/4.0/>). **(F)** The Colon-on-chip was used to study PGE2-induced mucus layer swelling on the chip (Sontheimer-Phelps et al., 2020) CC-BY-4.0 (<https://creativecommons.org/licenses/by/4.0/>).

Figure 5: Incorporating salient features from Tumor-on-a-chip models to develop a Tumor-Microbiome-on-a-chip model. Several tumor-on-a-chip platforms investigate the different aspects of the hallmarks of cancer. The technologies used to build these devices can be adopted to create a tumor-microbiome-on-a-chip that incorporates the necessary elements to characterize tumor-associated microbes.

Figure 6: Summary of current models. The strengths of current models, their utility in tumor-microbiome studies, and advancements needed to augment their current capabilities.

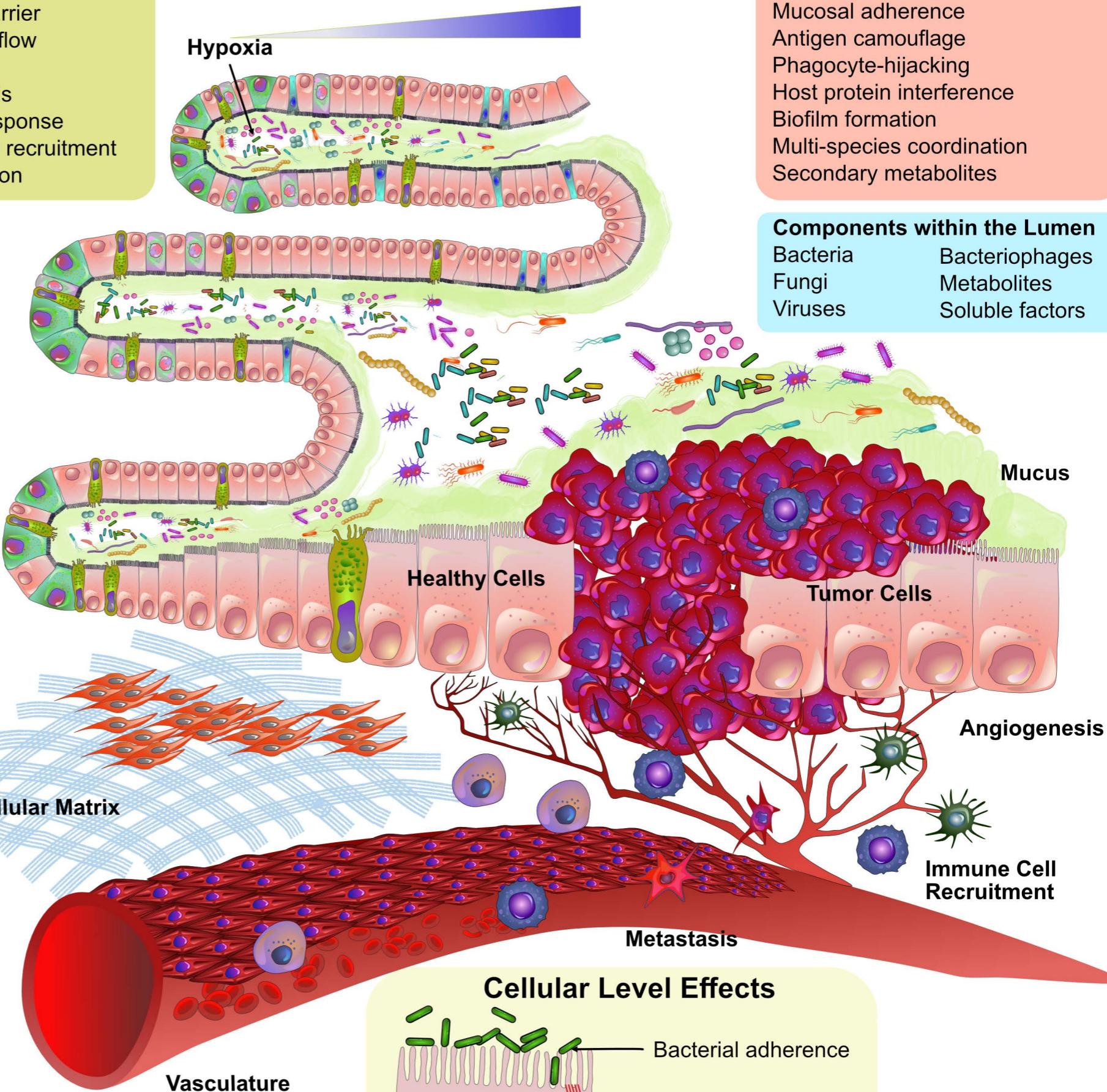
A**B****C****D****E****F**



Host Defense Strategies

- Epithelial barrier
- Mucociliary flow
- Fluid flow
- Phagocytosis
- Cytokine response
- Immune cell recruitment
- NET formation

Oxygen Gradient



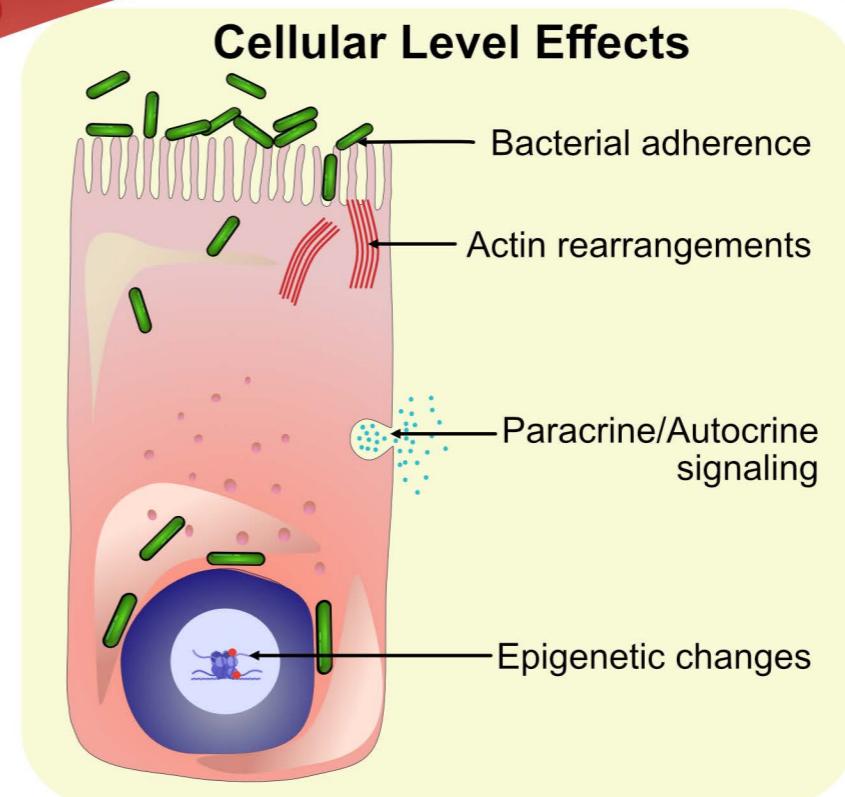
Bacterial Virulence Strategies

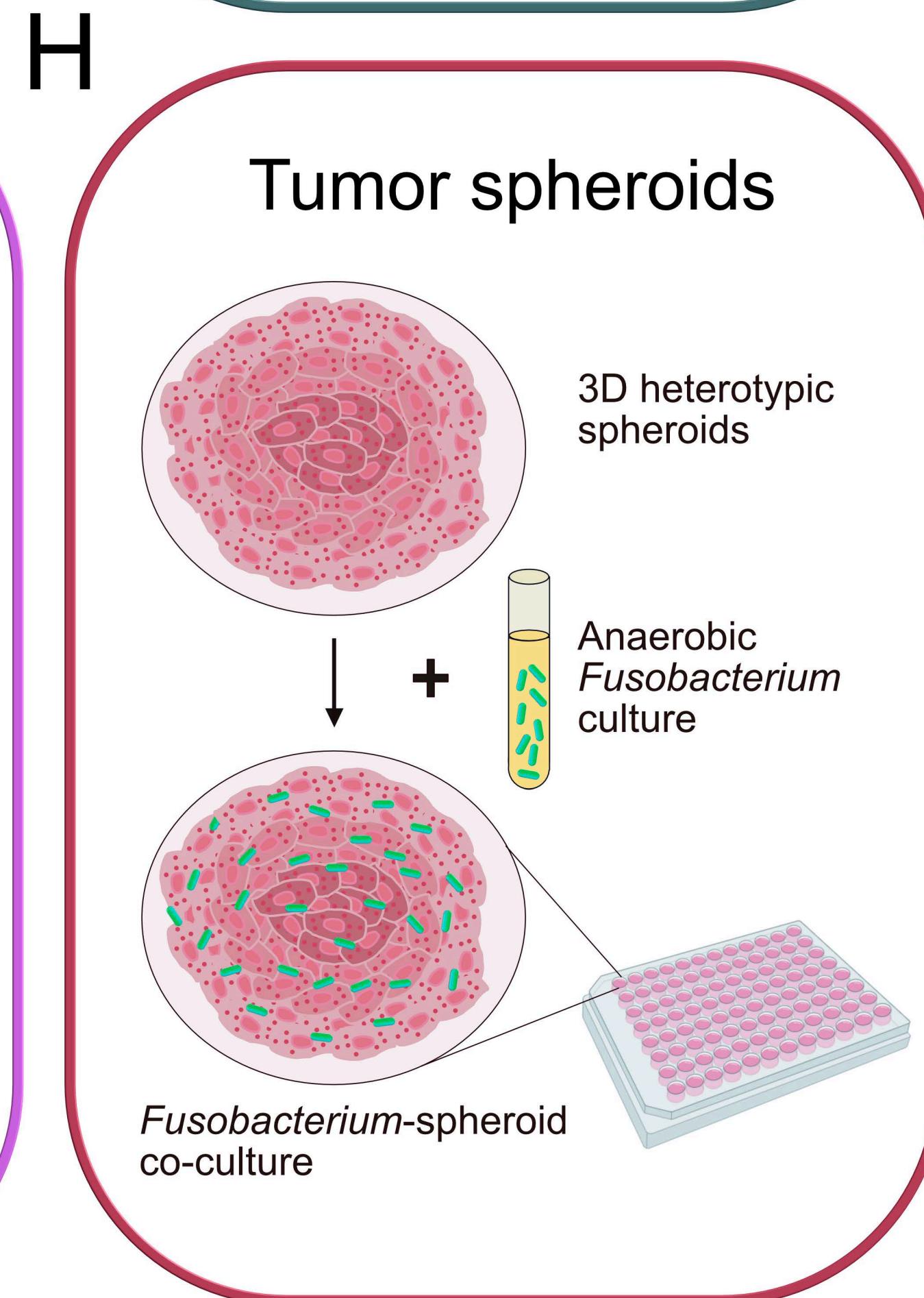
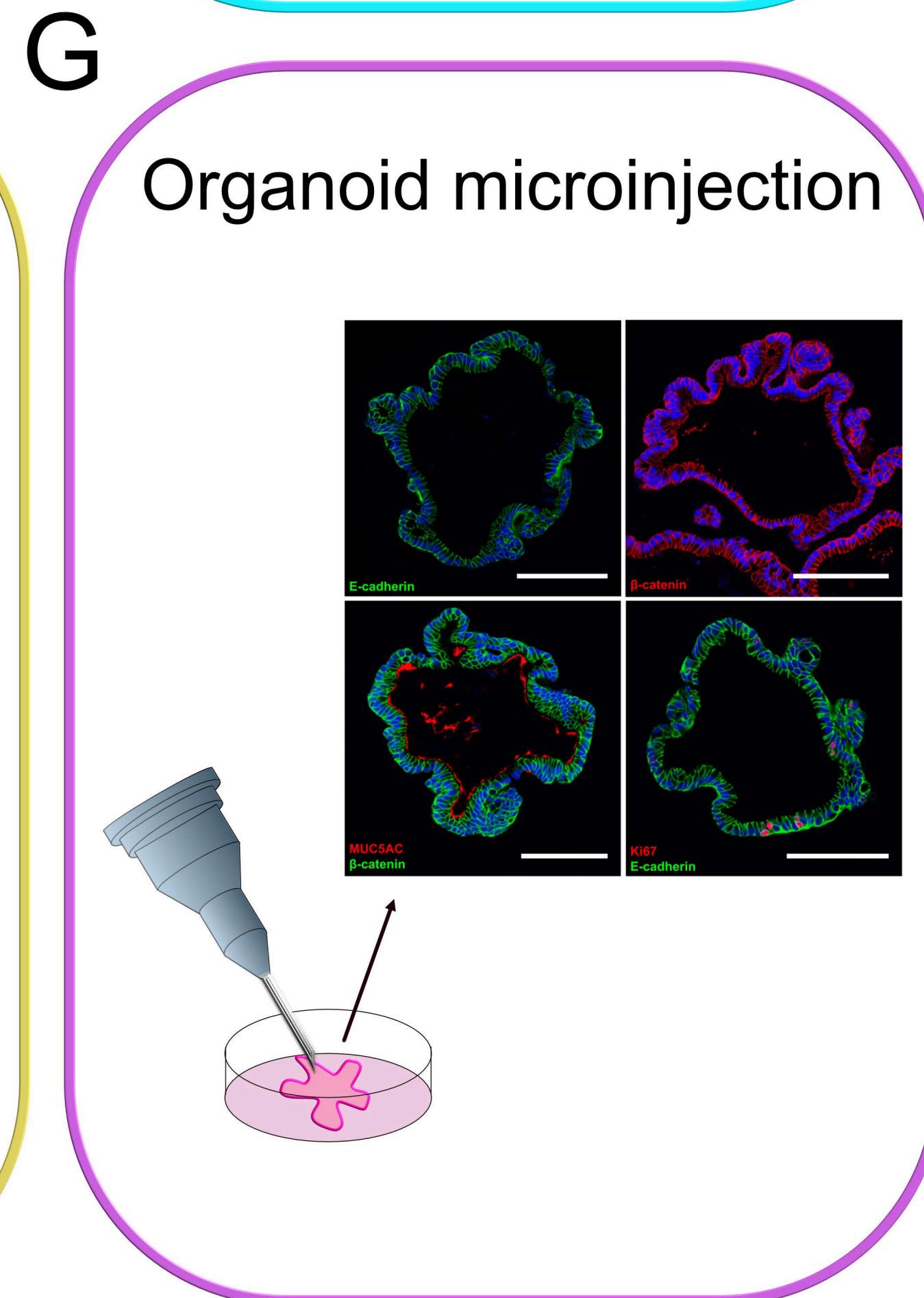
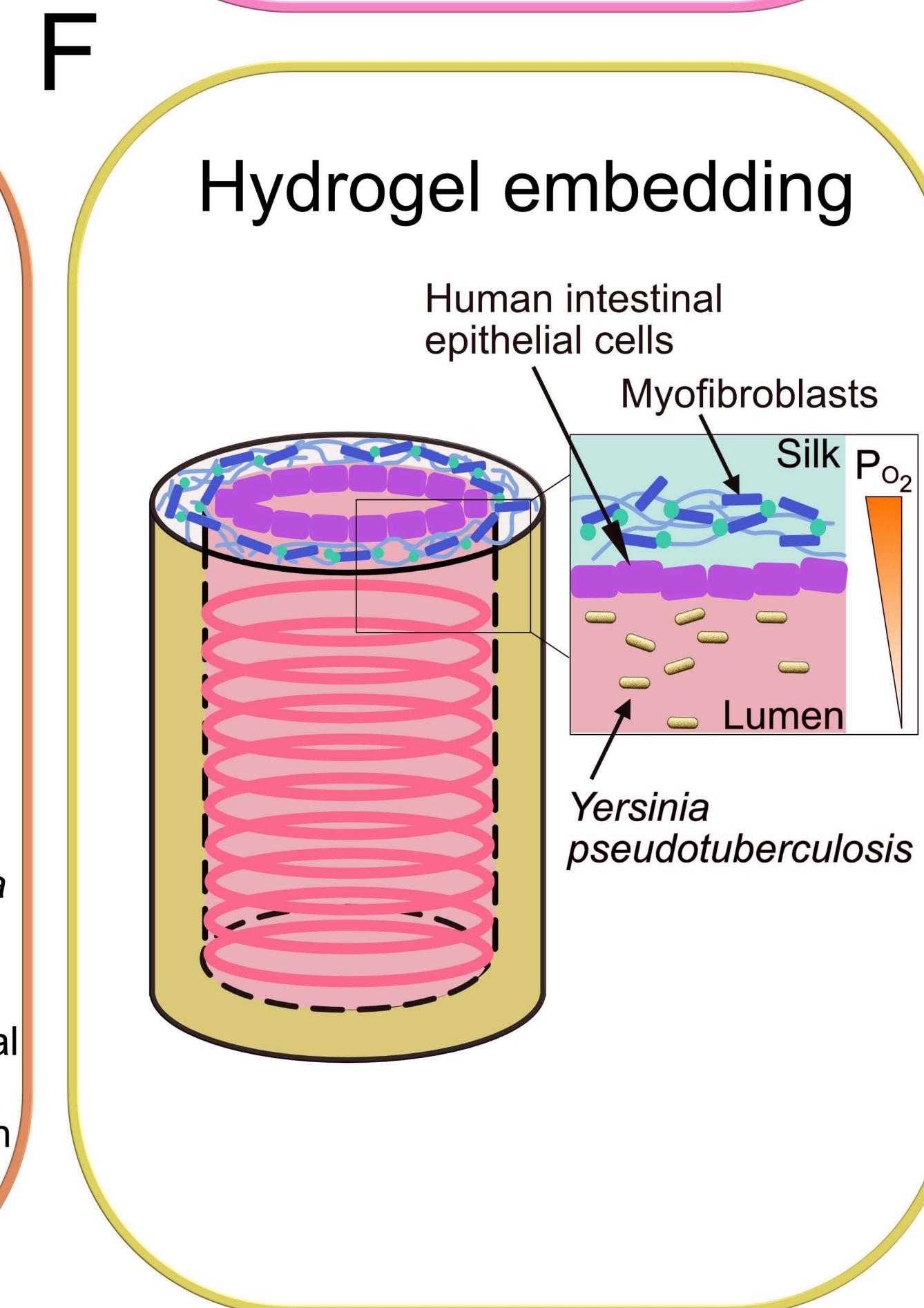
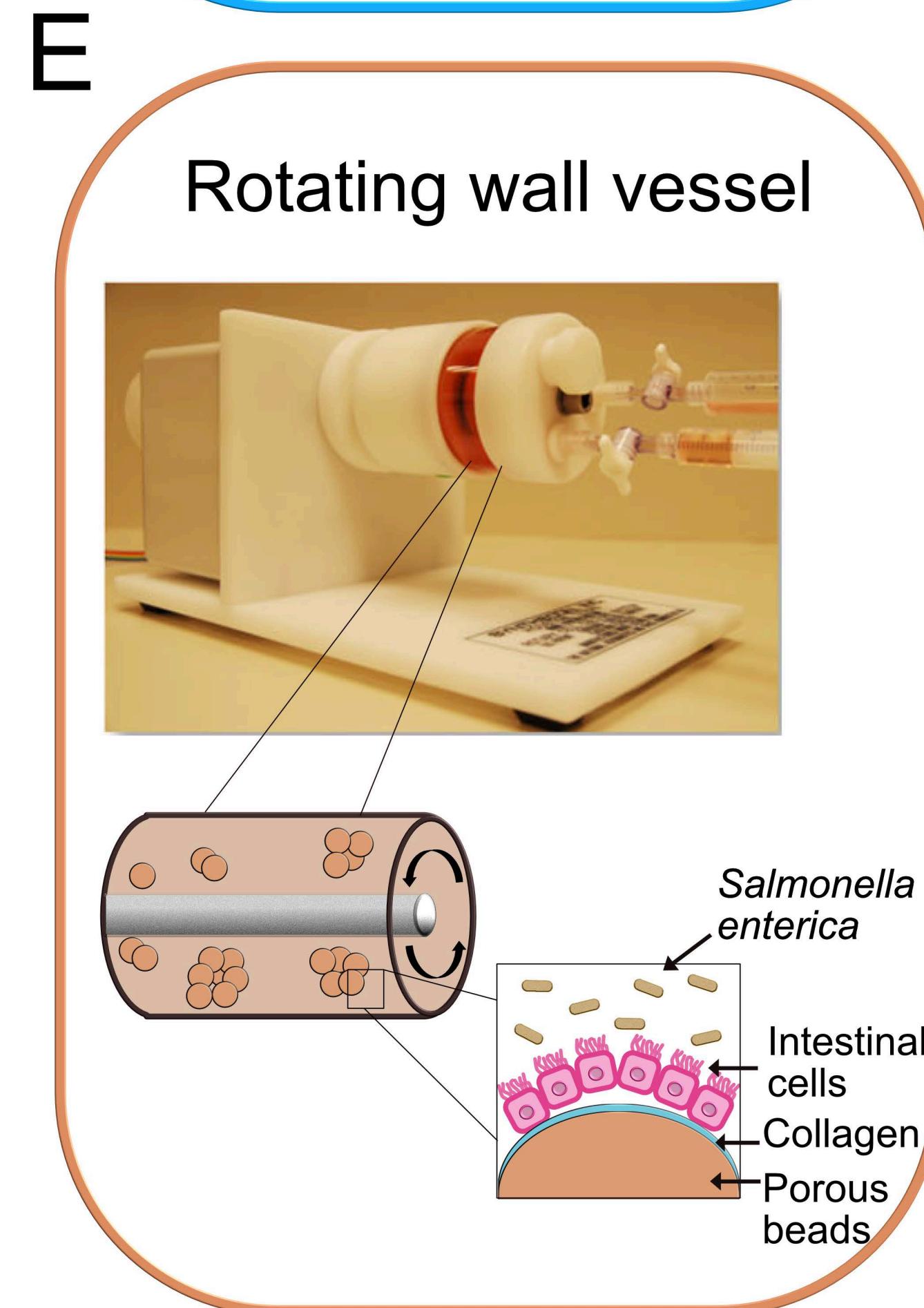
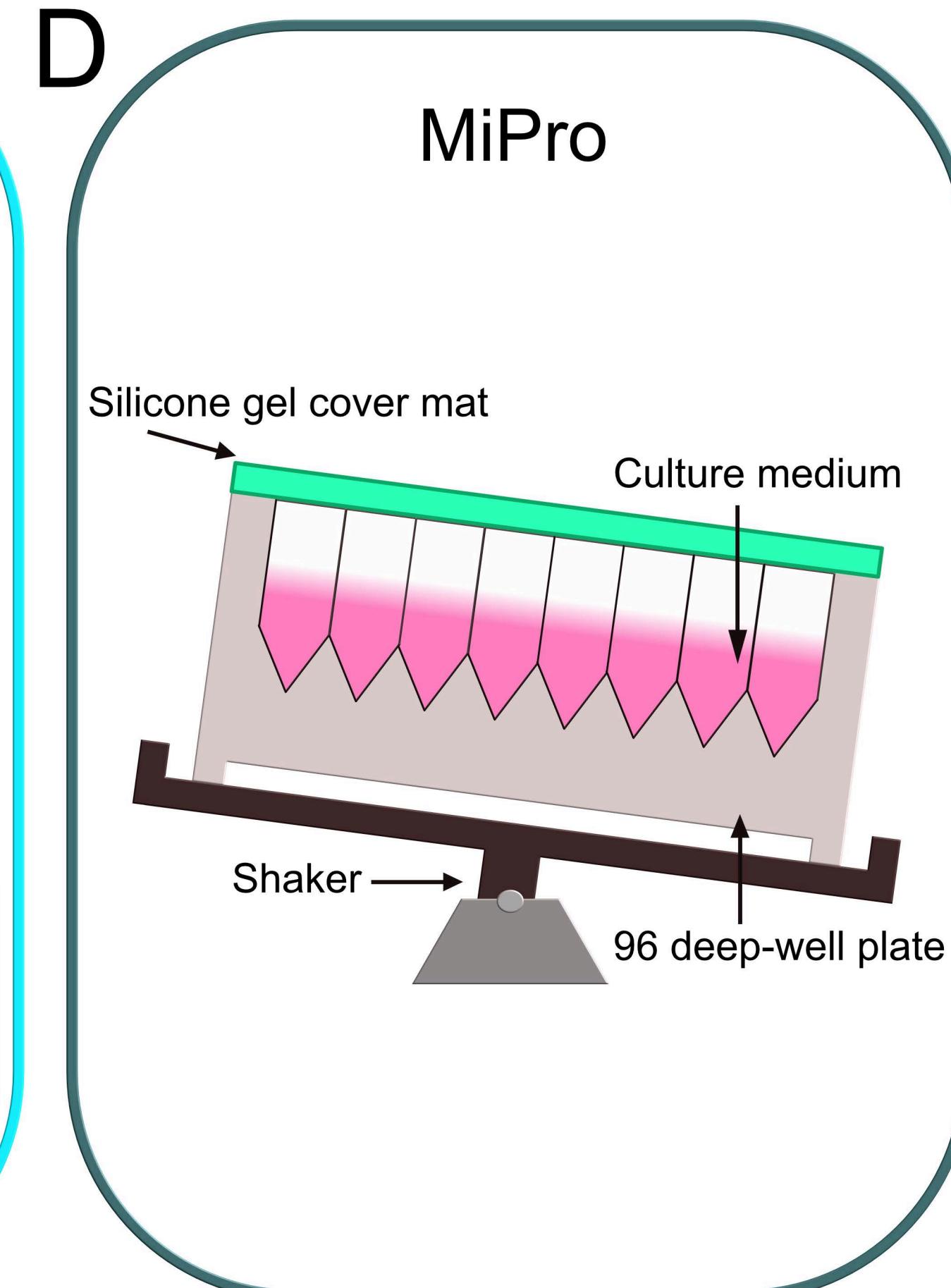
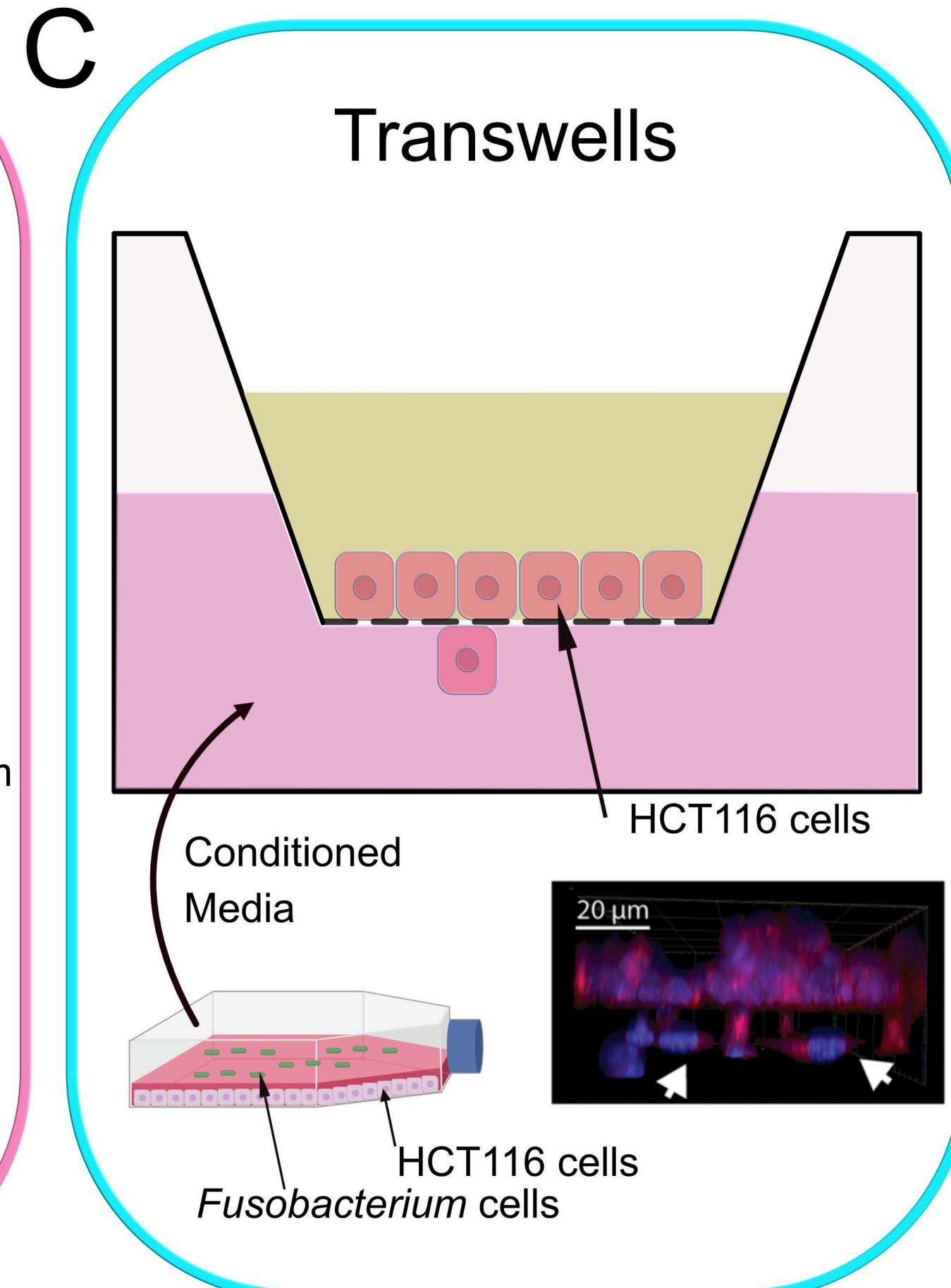
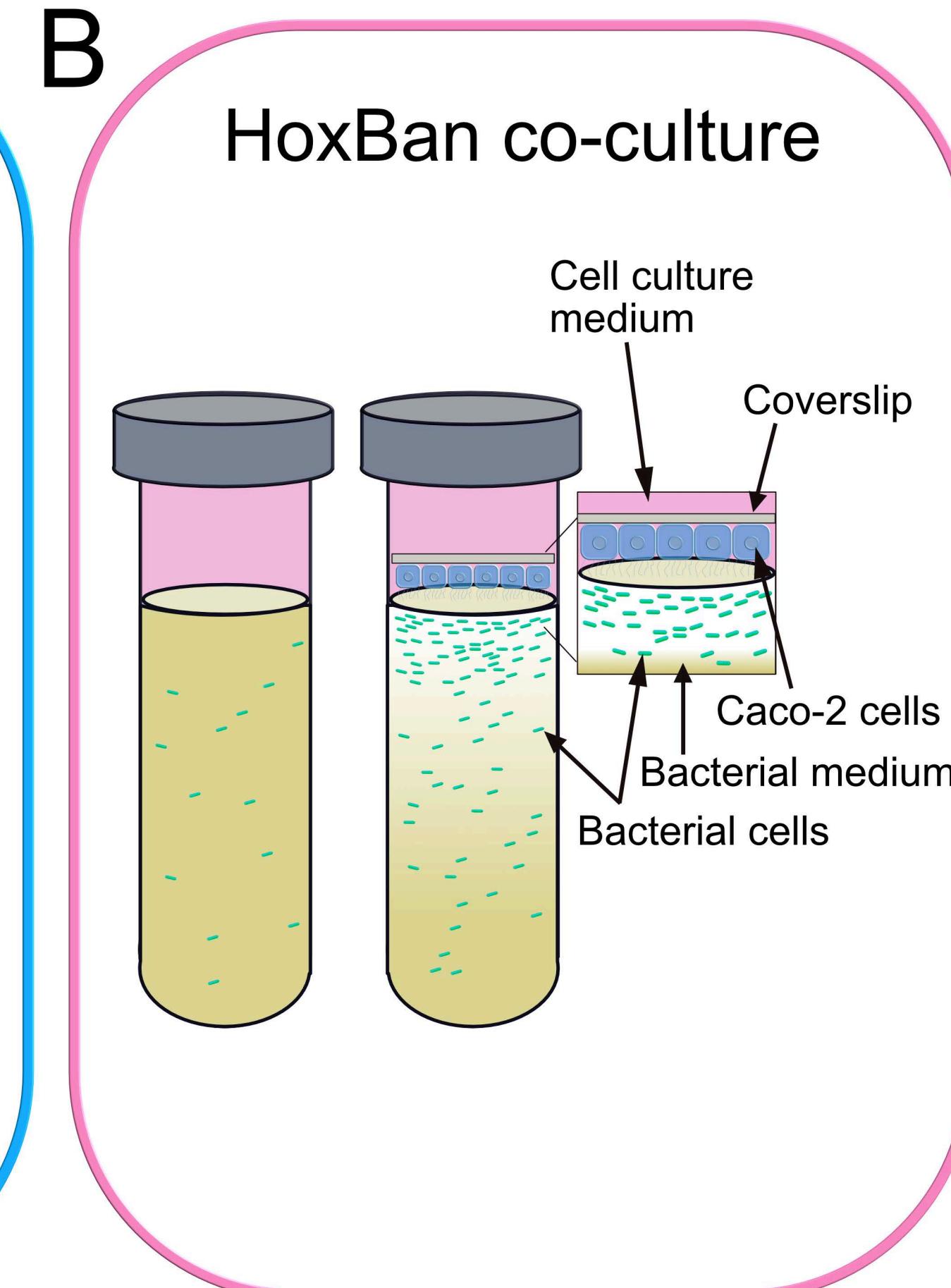
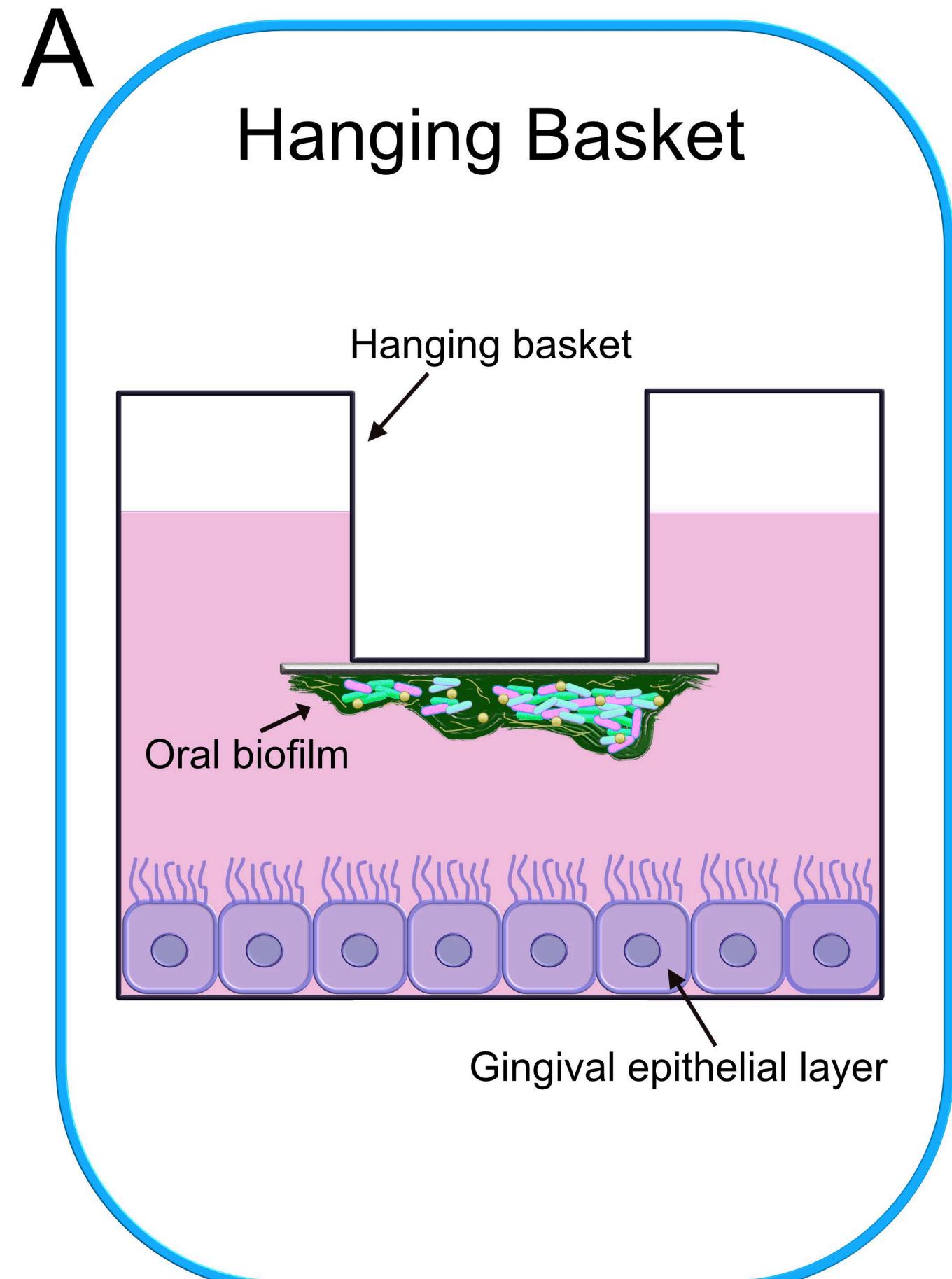
- Mucosal adherence
- Antigen camouflage
- Phagocyte-hijacking
- Host protein interference
- Biofilm formation
- Multi-species coordination
- Secondary metabolites

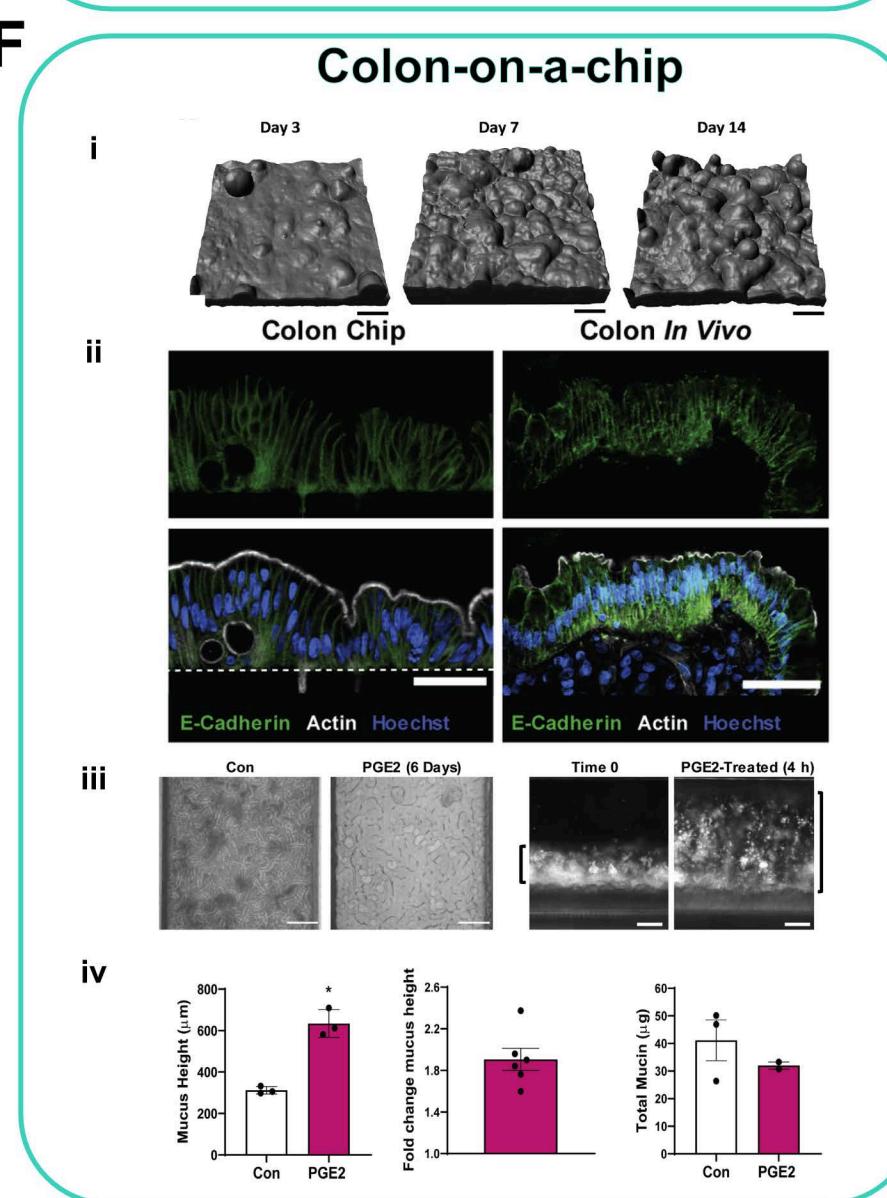
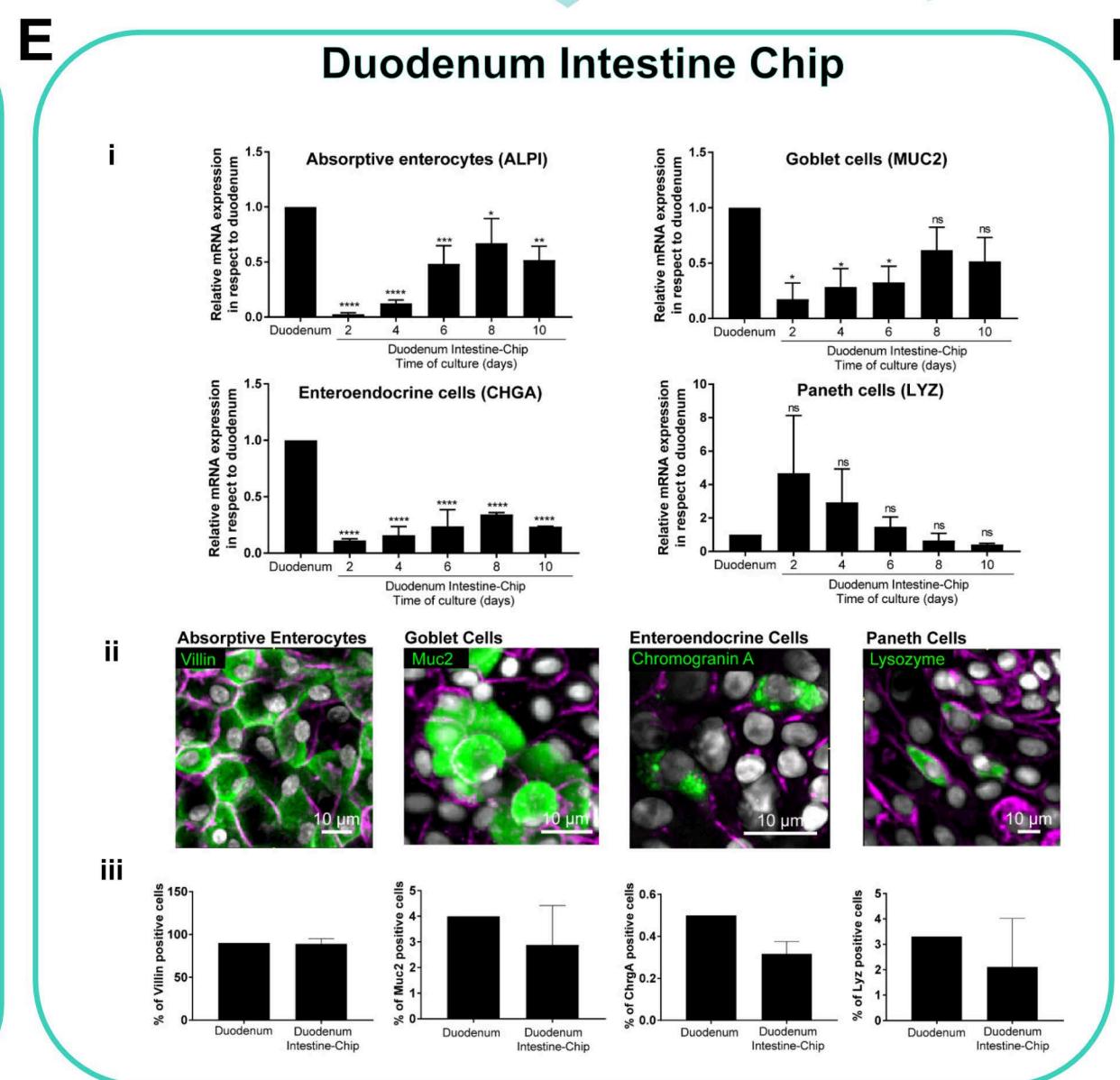
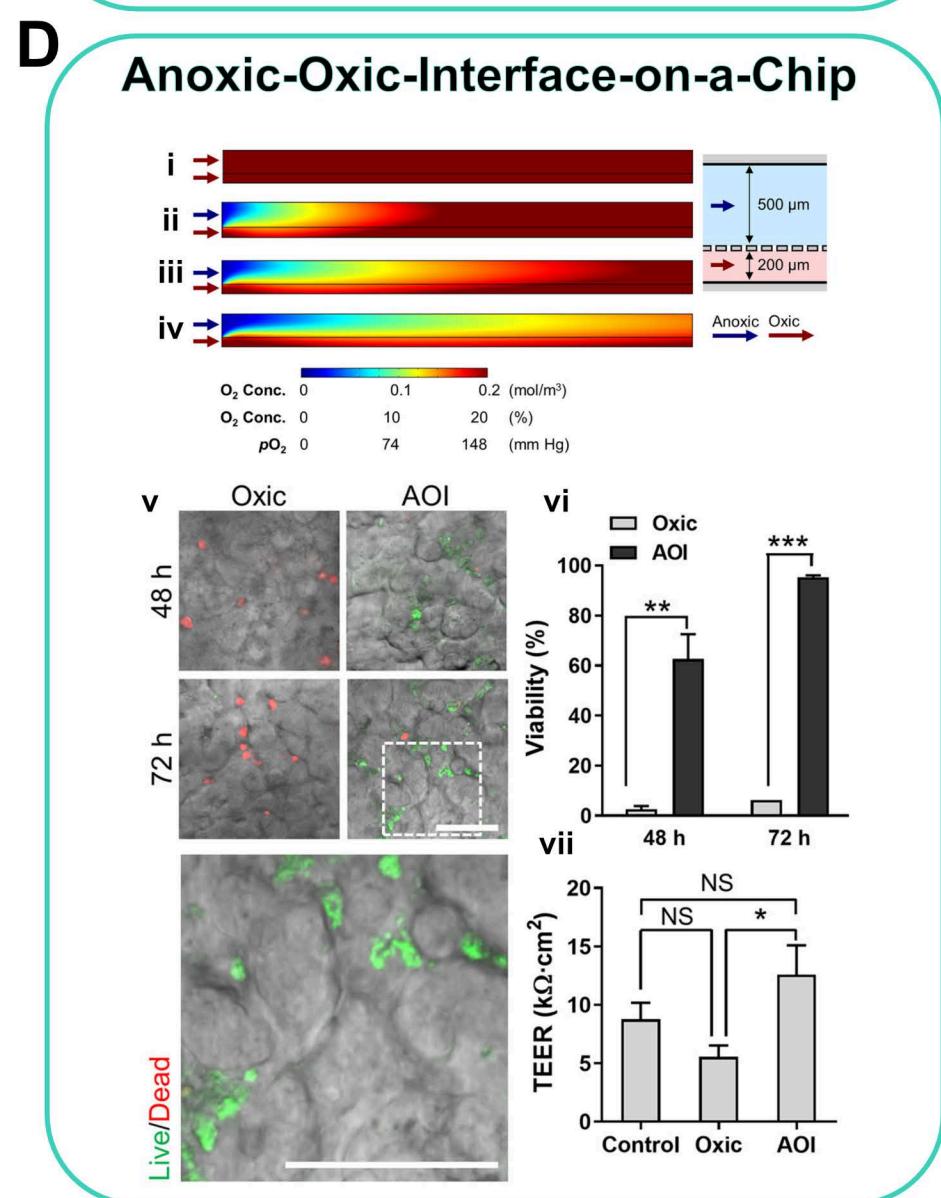
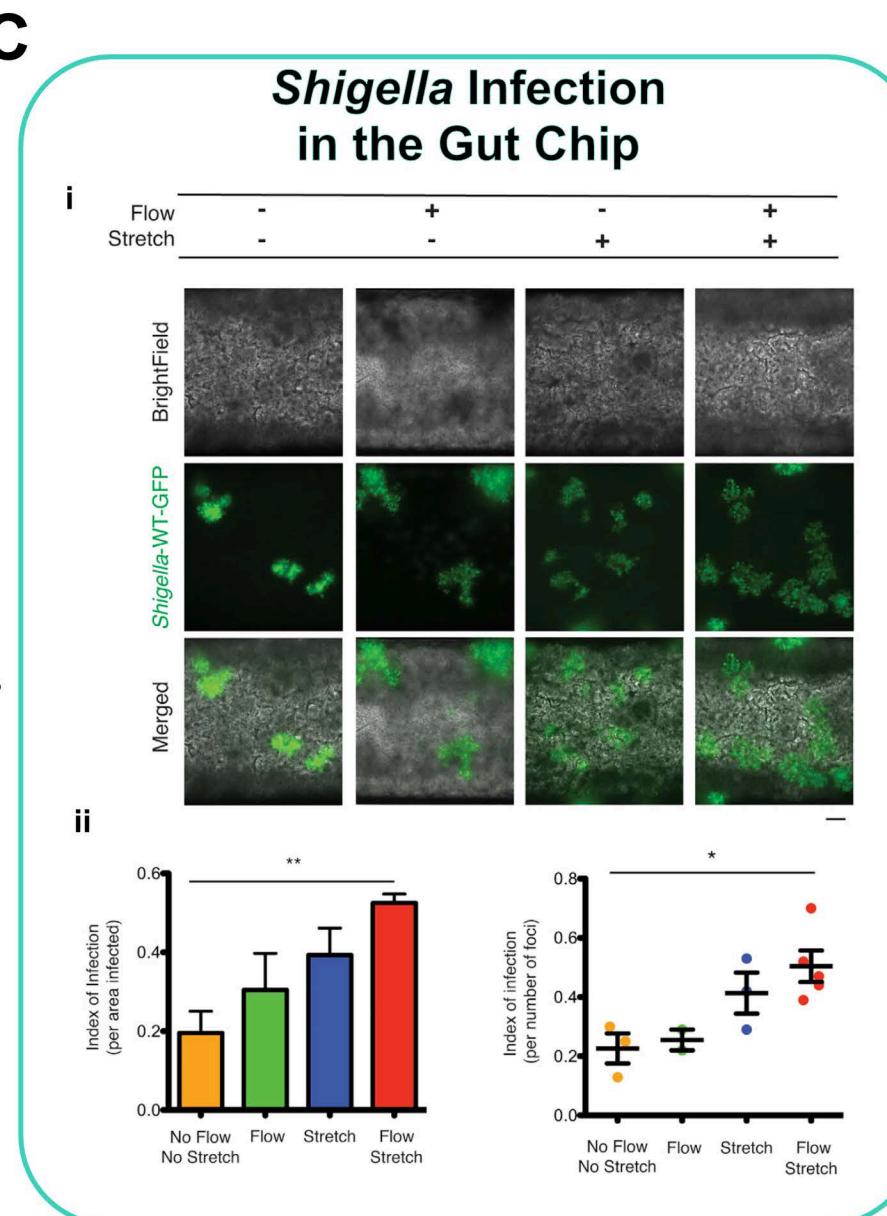
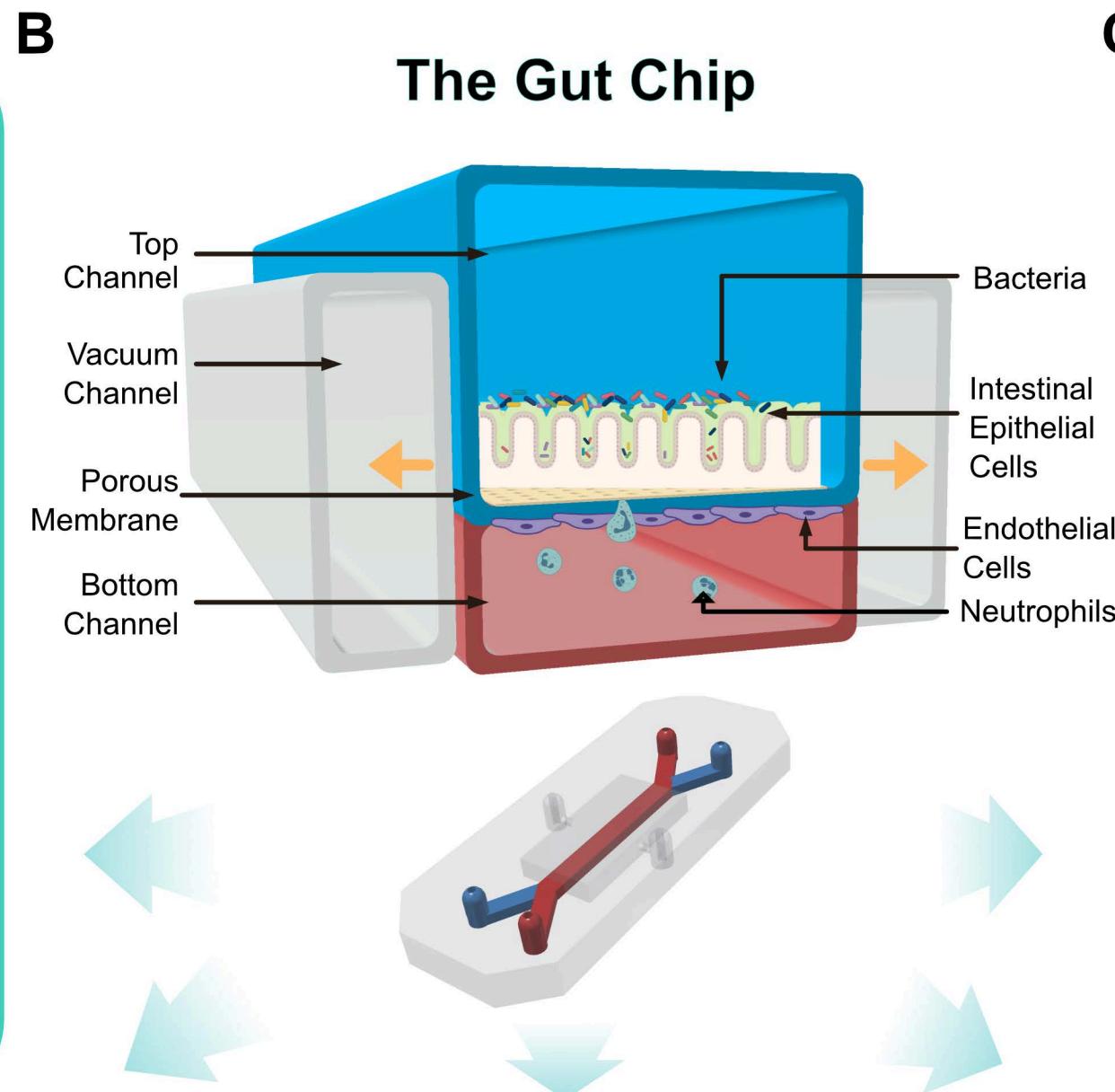
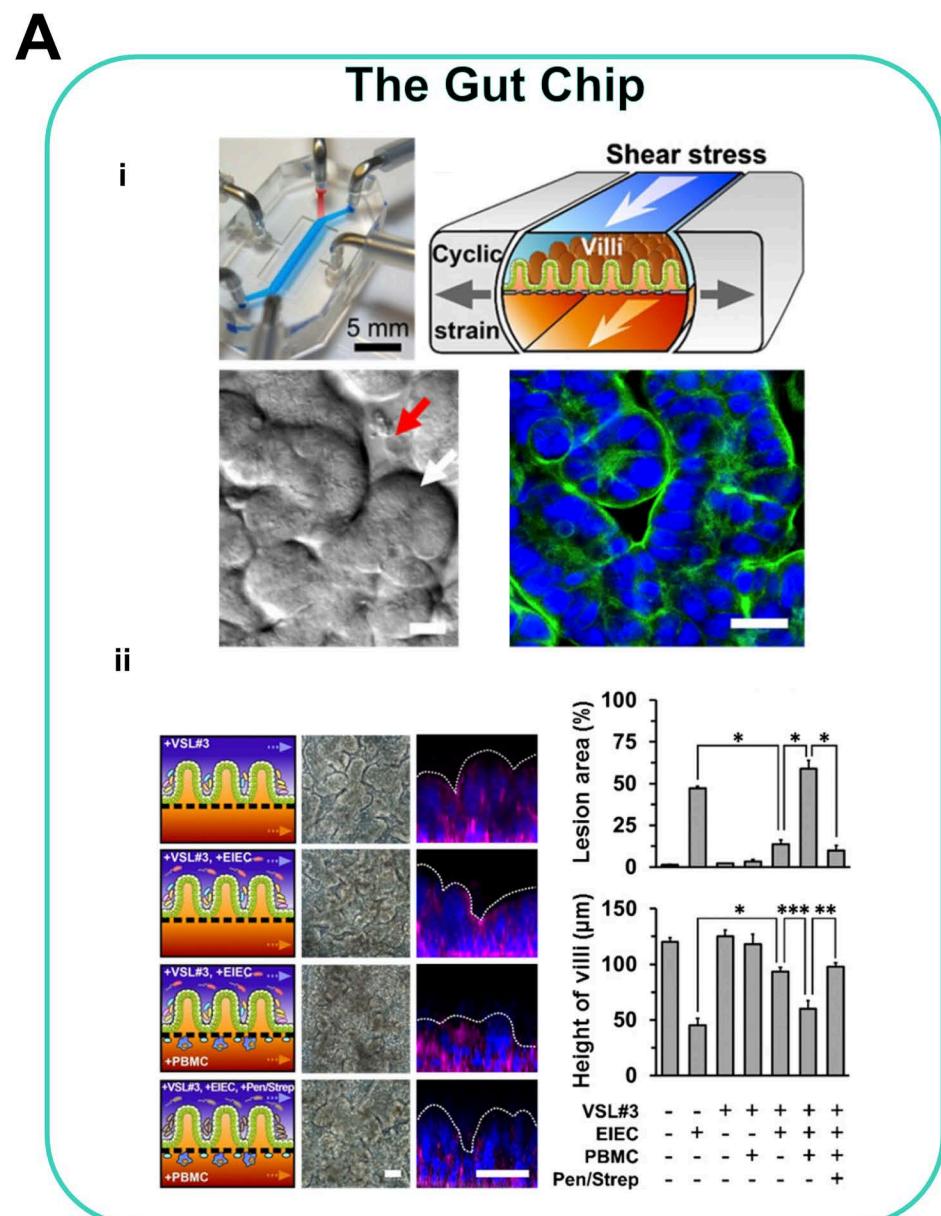
Components within the Lumen

Bacteria	Bacteriophages
Fungi	Metabolites
Viruses	Soluble factors

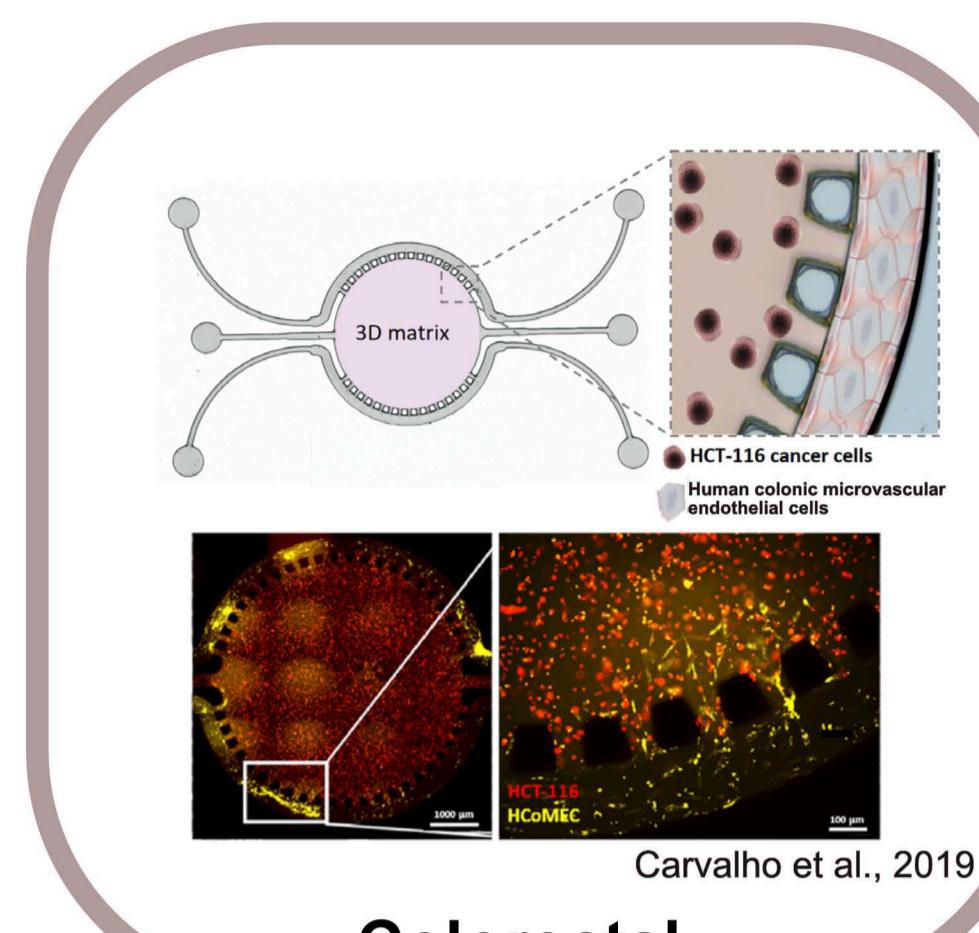
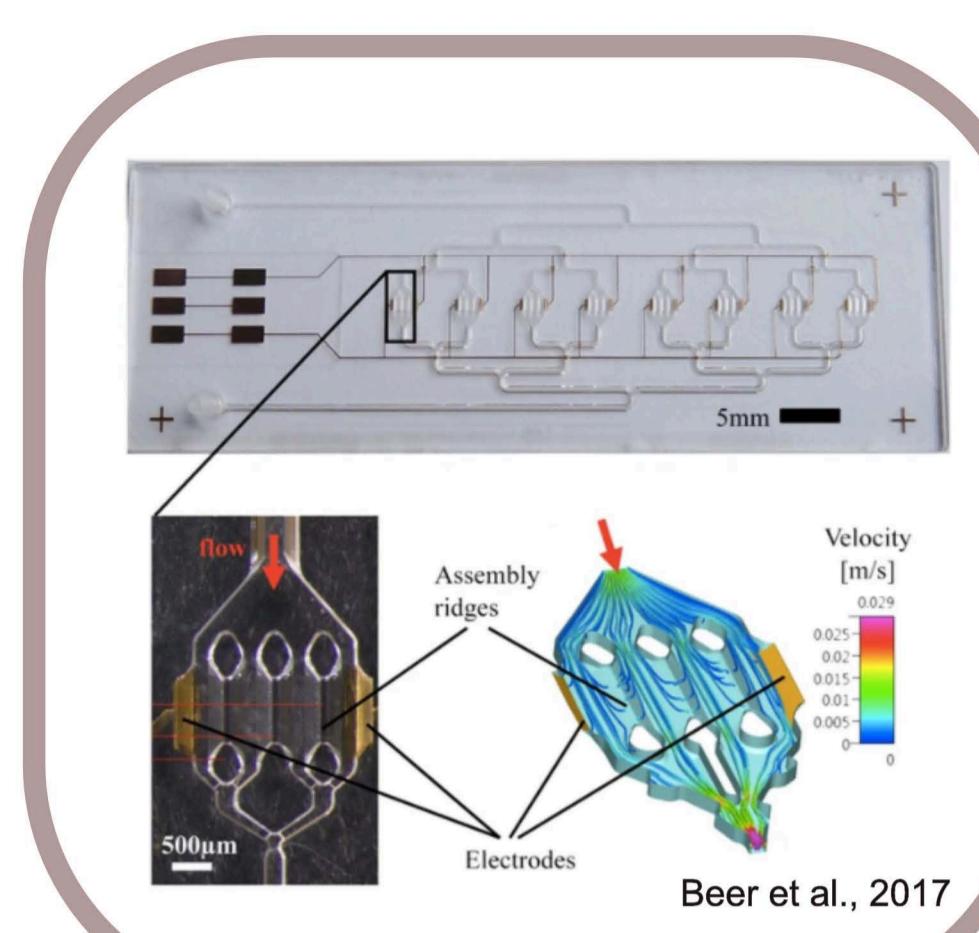
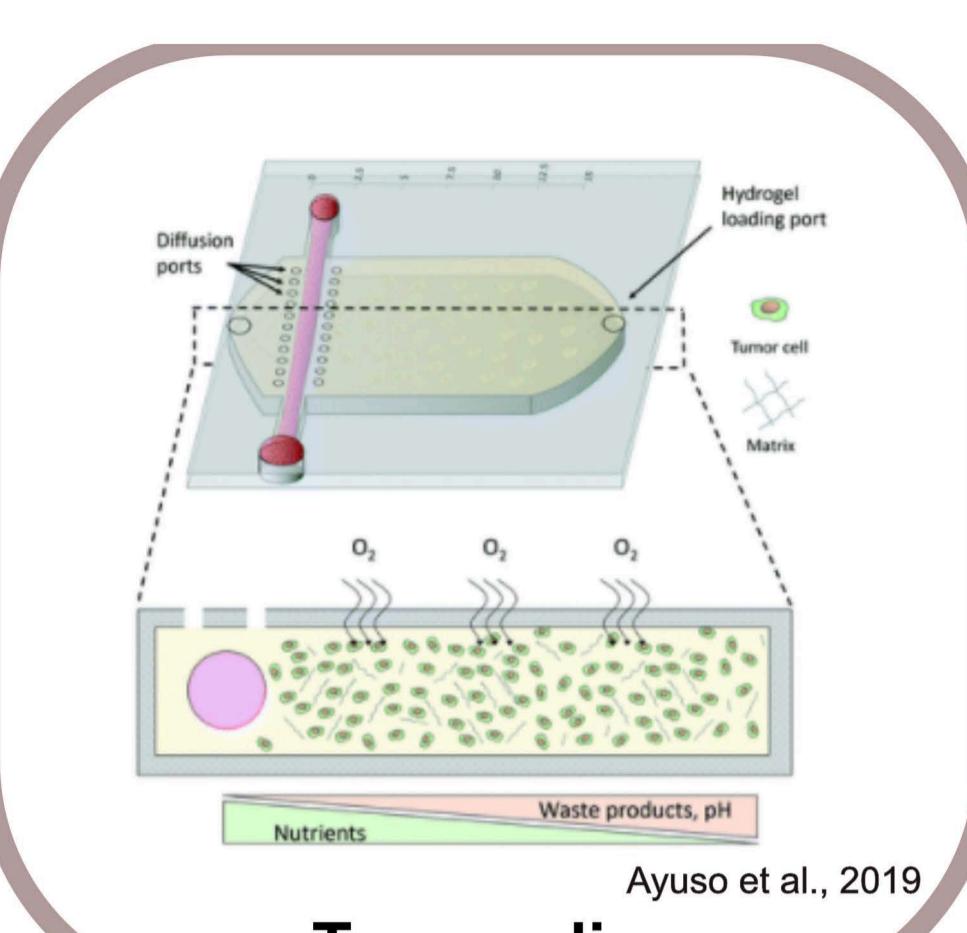
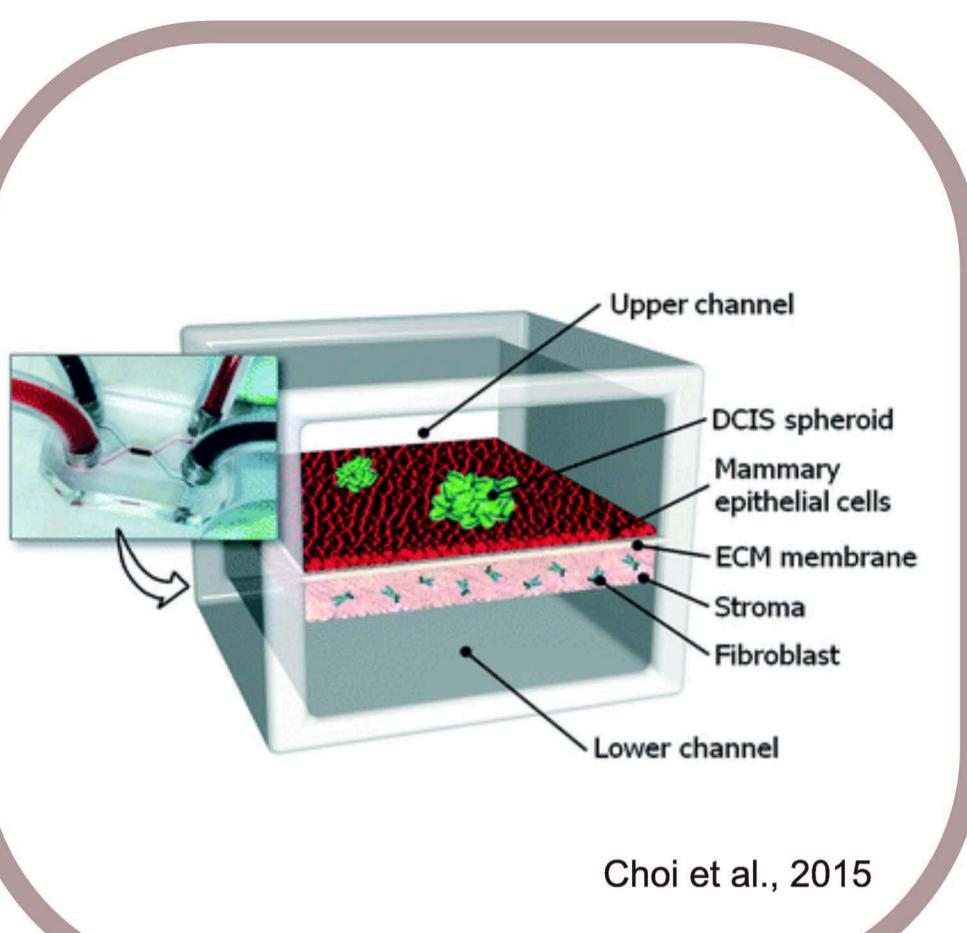
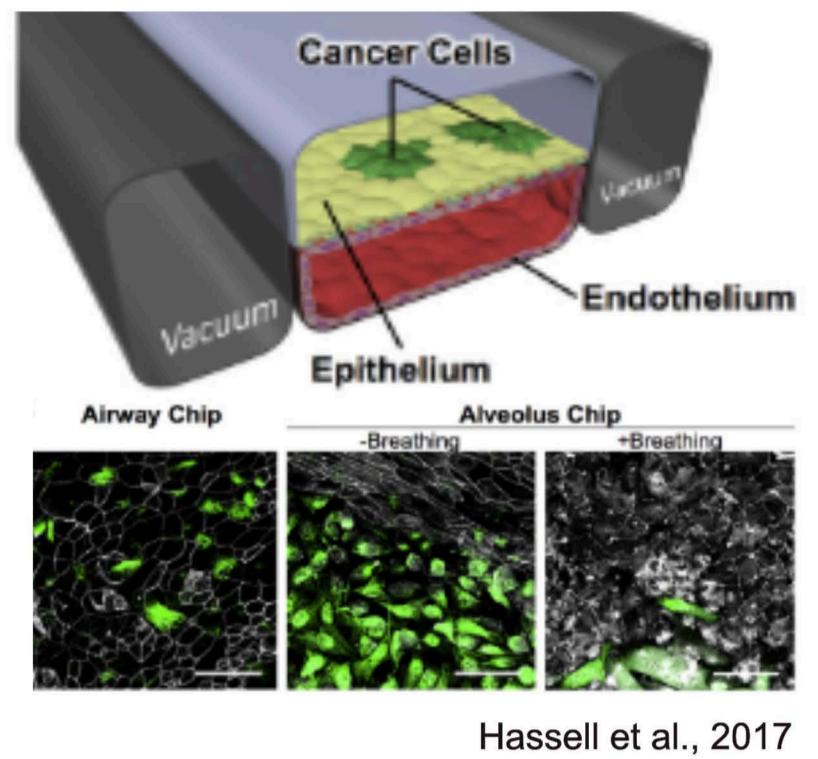
The Colorectal Cancer Microenvironment







Tumor Microenvironment



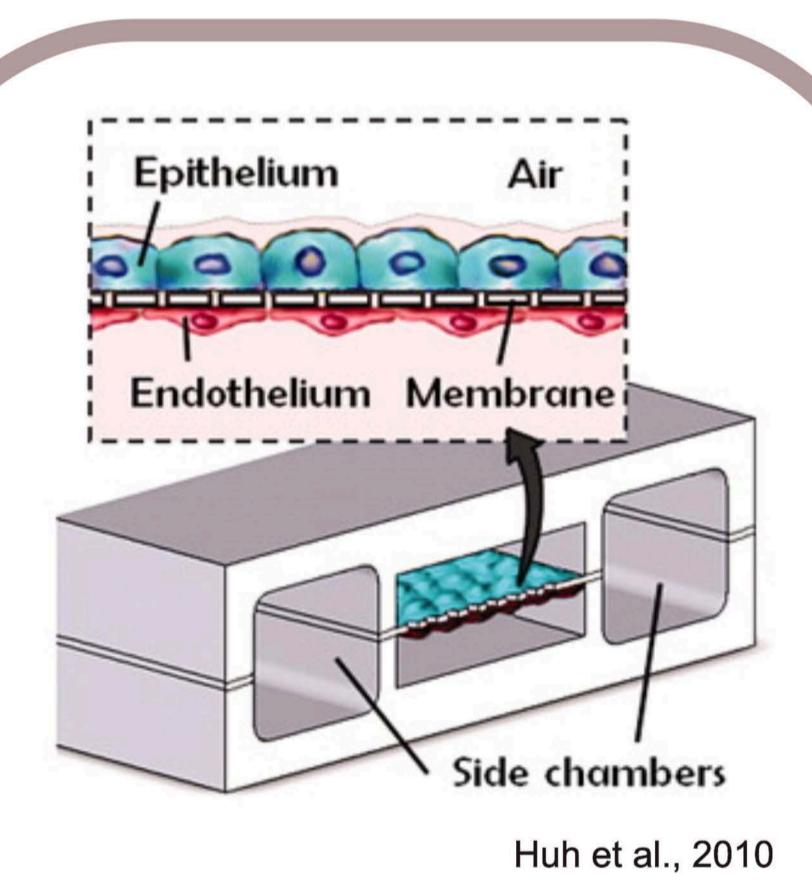
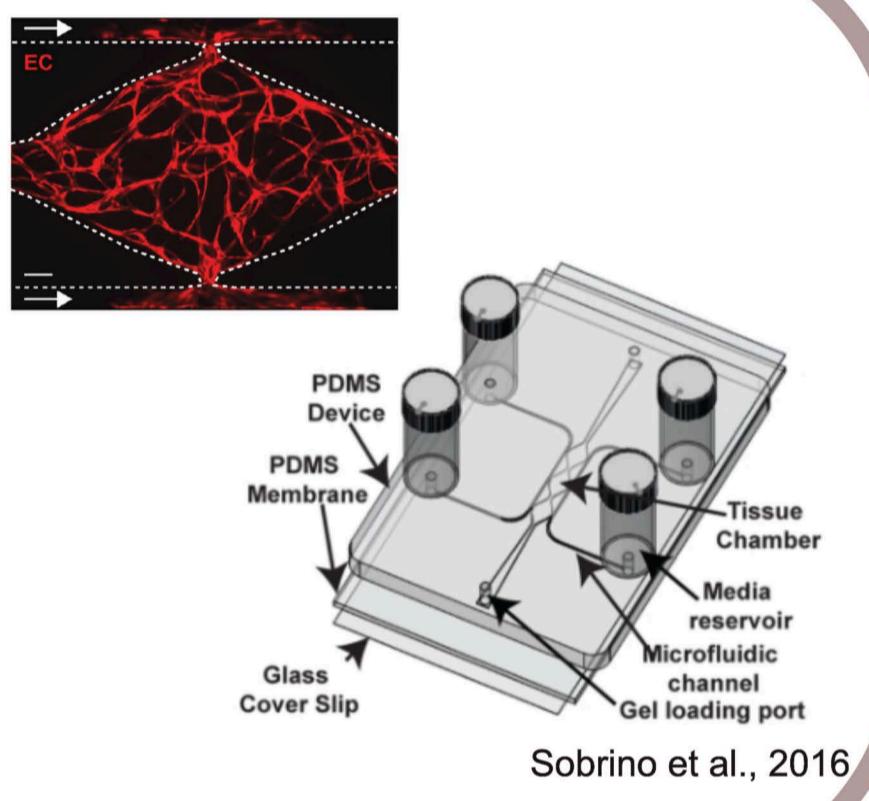
Microbe/Microbiome

Host Cancer Cells

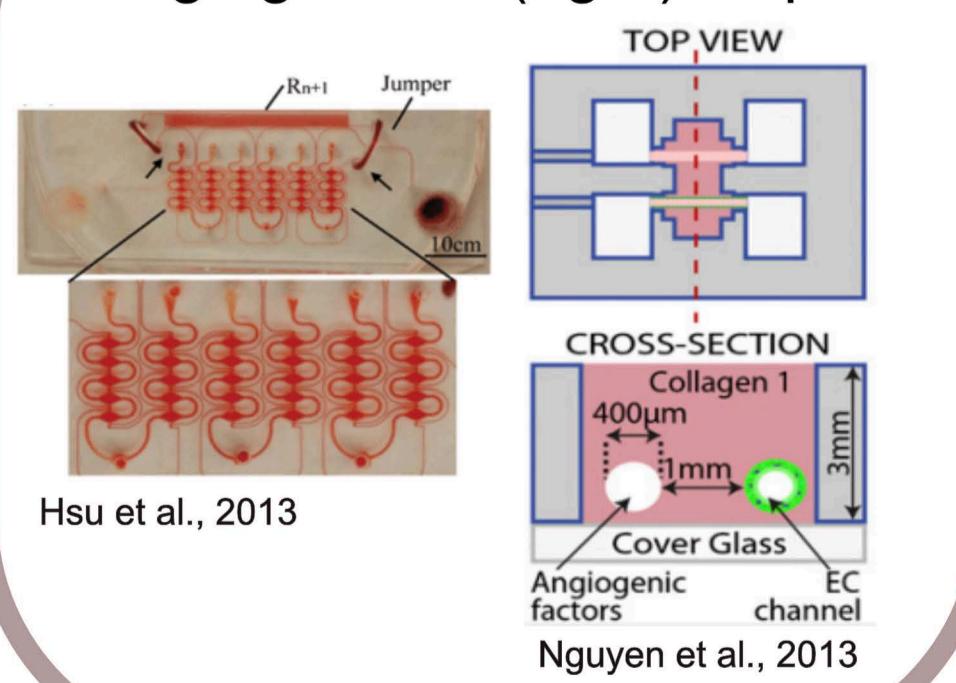
Vascularization

Immune Cells

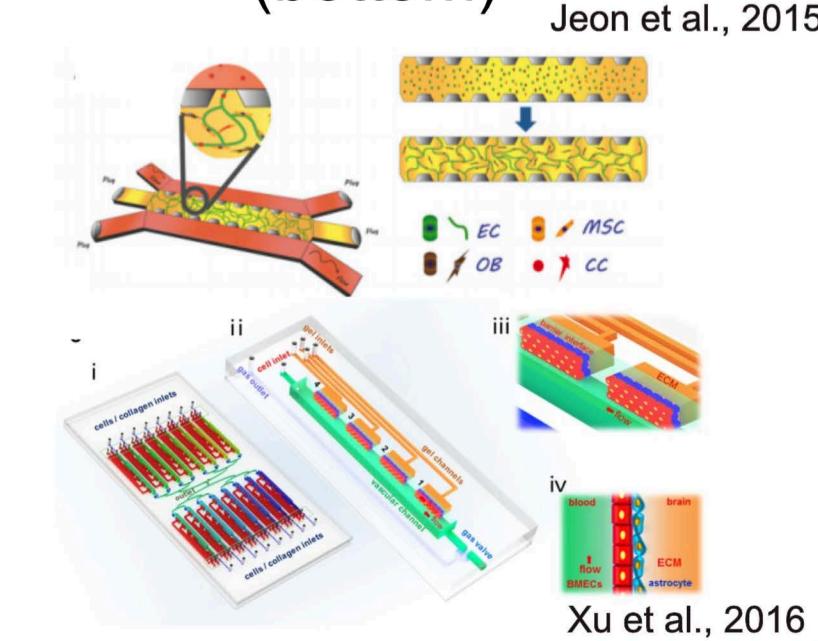
Tumor-Microbiome-on-a-chip



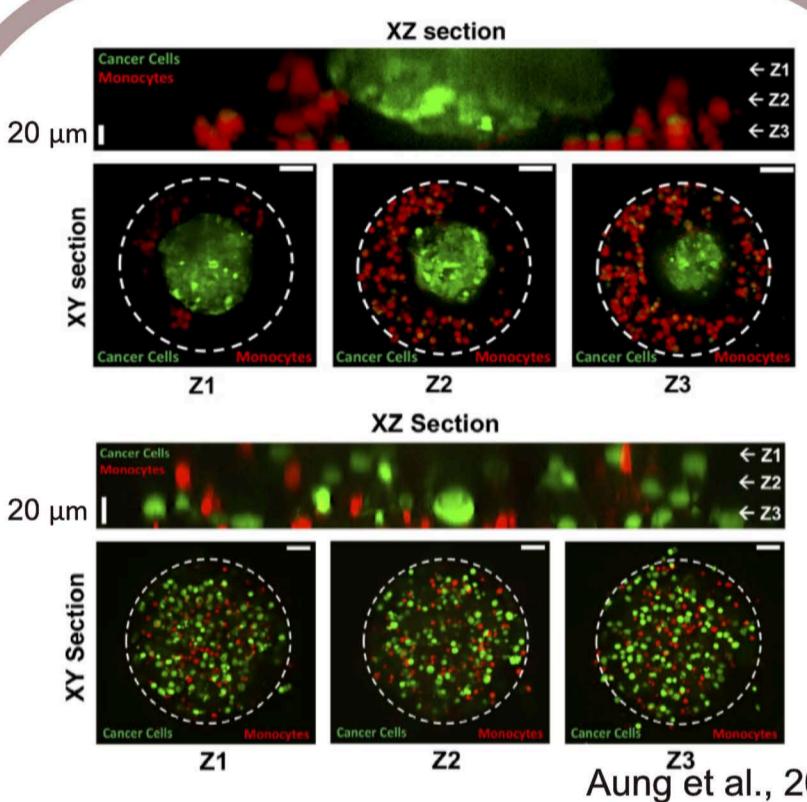
Vasculogenesis (left) and Angiogenesis (right) chips



Metastasis-on-a-chip: Breast (top) and Brain (bottom)



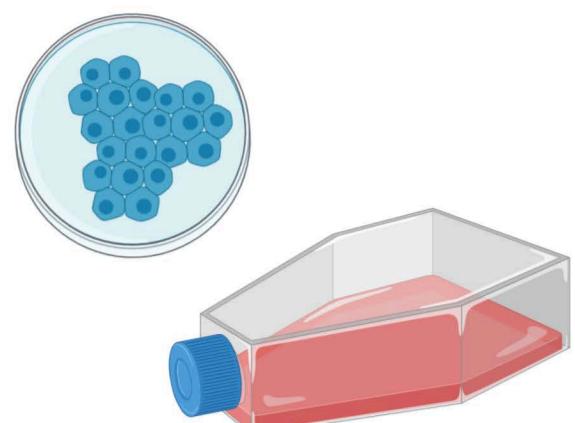
Immune interactions



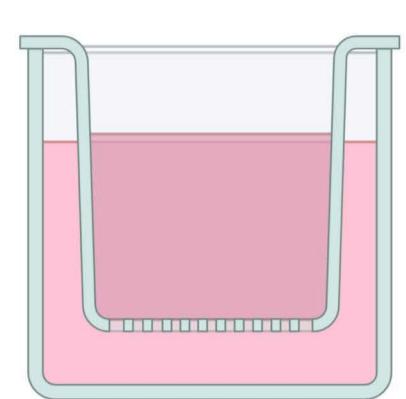
Reproducibility Controlability

In vivo physiological relevance Complexity

2D



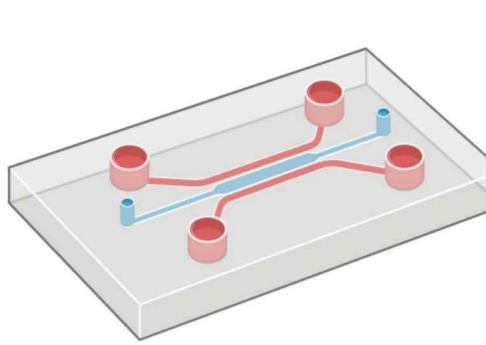
2.5D



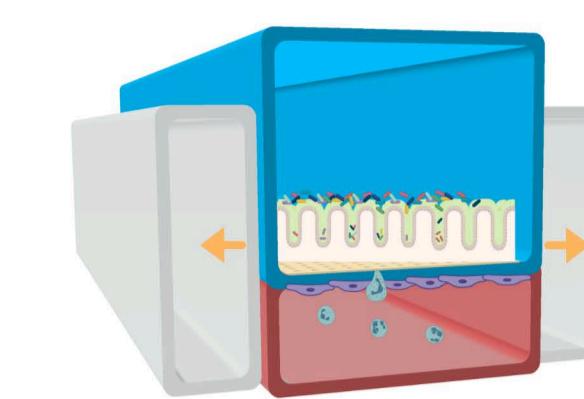
3D



Microfluidic
Models



Tissue-
Engineered
Models



Animal
Models



Examples

Petri dishes
Culture flasks
Falcon tubes
96-well plates

Transwells

Organoids
Hydrogels
Spheroids
Rotating Wall Vessel

Organ-on-chip
Physiome-on-chip
OrganoPlate

HuMiX
Gut-chip models
Tumor-chip models

Mouse
Rabbit
Monkey

Strengths

Low cost

Rapid construction

Complex 3D microarchitecture

Soluble factor signaling & metabolite sampling

Reproducible soluble factor gradients

Multiple cell types & Differentiated Cells

Oxygen control & suitability for hypoxia studies

Microscale flow control and shear forces

Mechanical stimulation

Multi-organ effects

Study of immune interactions

Cell migration and metastasis

Cell-cell/Cell-ECM interactions

Angiogenesis studies

Community modeling

Relative Strength
Low ————— High

Host
Model

Monitor cytokine secretions

Air-liquid interface
3D differentiation
Mucus secretion
Multiple cell types

Realistic cell proliferation and differentiation
Cell-cell communication

Better imitate cellular natural environment
Cell-cell interactions

Better micropatterning of stimuli
High cellular biofunctionality

Complex tissues
Native immune system

Microbe
Model

Growth
Adherence
Cytotoxicity

Growth
Biofilm formation

Microbial community modeling

Mechanistic studies

Multi-species modeling

Whole microbiome compositional alterations

Utility in Tumor-
Microbiome Studies

Adapt for high-throughput screening

Development of biomaterials/bioprinting technologies

Methods to improve cell differentiation

Visualization techniques/Live microscopy/Holographic imaging

Metabolite analysis techniques - Advancements in Mass Spectrometry

Hypoxia generation/Oxygen sensing

Controlling microbial over-proliferation

Long-term co-culture maintenance

Incorporation of a vascular interface/Vascularization/Angiogenesis

Incorporation of immune cells

Better flow/perfusion control

Measurement of barrier permeability

Sensor development

Quantitative analytical tools

Advancements Needed