Competition for shared resources increases the impact of propagule size during

coalescence of gut microbial communities

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Abstract

The outcomes of species introductions depend on both propagule size and resource

competition, but the effects of these factors are inconsistent across community contexts.

Here, we investigate how propagule size and resource competition interact to shape

colonization during community coalescence by mixing pairs of *in vitro* gut microbial

communities at ratios that vary over seven orders of magnitude. Each resulting co-

culture contained species whose relative abundance depended on inoculation dose,

causing the composition of some co-cultures to vary substantially across mixture ratios,

even after >30 generations of growth. Using a consumer-resource model, we show that

dose-dependent colonization can arise when niche overlap is high and species compete

near-neutrally for shared resources. This model successfully predicts the outcomes of

mixtures of strain isolates, in which propagule size has larger, longer-lasting effects in

more diverse communities. Our study establishes a framework for predicting the

context-dependent impacts of propagule size, providing principles for the design of

microbiome therapeutics.

Introduction

Colonization of an ecosystem by new species plays a major role in shaping community composition and function, but the outcome of species introductions is notoriously difficult to predict[1–7]. To colonize, a species must disperse from its original population and overcome several ecological barriers, including the stochastic effects of ecological drift at small population sizes[8,9] and competition with resident species for a niche in the community[10–14]. Increases in the size and frequency at which propagules of dispersing species are introduced can help to overcome these barriers[15–19]. However, the effects of propagule size and frequency, known together as propagule pressure, are inconsistent across different contexts[5,20–22]. Large propagules can fail to colonize, and small propagules can successfully colonize, when other ecological features such as competition affect species introductions more strongly than propagule pressure[23–28]. Disentangling the influence of these factors is crucial for predicting the outcomes of species introductions.

Understanding the effect of propagule pressure on colonization is especially important in microbial communities, from which species can disperse in propagules varying in size across many orders of magnitude[29–31]. Dispersal between microbial communities can involve dozens or even hundreds of species[32,33], resulting in community coalescence events such as river confluences, leaves falling onto soil, and fecal-oral transmission of gut microbiota[13,30,34–37]. Determining the impacts of propagule size is crucial to predicting the effects of these large-scale introductions.

Understanding the ecological determinants of colonization in the human gut is

especially important because of their relevance to the development of successful

microbiome therapeutics[38,39]. Studies assessing colonization following fecal microbiota

transplants (FMTs) or other probiotic introductions into the gut have shown that both

inoculation dose and competition can independently affect colonization outcomes[32,40–44].

Systematic quantification of the relative importance of propagule size and competition is

more tractable using laboratory model microbial communities, which can be stable,

diverse, and reproducible[45–48]. Stool-derived *in vitro* communities can recapitulate

microbiota composition and *in vivo* responses to antibiotic perturbation[49] or iron

supplementation[50], making these communities a powerful model system for

investigating microbial ecology.

Here, we investigate the effects of propagule size on colonization using controlled

mixtures of human stool-derived *in vitro* communities at ratios varying across seven

orders of magnitude. While some species reached the same relative abundance

regardless of mixture ratio, the relative abundance of other species varied with

inoculation dose, even after >30 generations of growth. We utilize a consumer-resource

model to show that dose dependence can arise and persist when species have high

niche overlap and compete near-neutrally for shared resources, an effect that can be

enhanced by the presence of additional species. Using mixtures of strain isolates, we

experimentally confirmed the model prediction that the magnitude and duration of dose

dependence increases in more diverse communities. This work shows how the effects

of propagule size can be amplified by resource competition in complex communities and

have lasting effects on the outcomes of community coalescence.

Results

Diversity and composition of co-culture communities vary by inoculation dose

but deviate from neutral predictions

We performed coalescence experiments *in vitro* using a set of diverse, stable

communities of gut microbes. Following a previously established protocol[49], we derived

these communities from stool samples of eight healthy human subjects and passaged

them 15 times with a 1:200 dilution in fresh modified Brain Heart Infusion + mucin

(mBHI+mucin) medium every 48 h (**Fig. 1a, Methods**). To assess community

composition, we performed 16S rRNA gene amplicon sequencing on multiple passages

from each community and tracked the relative abundances of amplicon sequence

variants (ASVs). Community diversity decreased in the first few passages after initial

laboratory inoculation (**Fig. 1b, Extended Data Fig. 1a,b**), likely because some taxa in

the inoculum were nonviable or were outcompeted during *in vitro* passaging.

Nonetheless, diversity stabilized after ~3 passages, and most communities thereafter

contained ASVs that had made up >50% of the relative abundance of the stool inoculum

(**Extended Data Fig. 1c**). Similarly, family-level relative abundances quickly stabilized

(**Fig. 1c**, Pearson's tests, left: $n=25$, $\rho=0.25$, $P=0.23$; right: $n=22$, $\rho=0.93$, $P<10^{-9}$,

Extended Data Fig. 1c,d), with the remaining families after 15 passages collectively

representing >75% of the relative abundance of the corresponding stool inoculum in

most cases (**Extended Data Fig. 1c**). Thus, these stool-derived *in vitro* communities

are diverse and stable models of stool inocula that can be used to study ecological

interactions among gut microbes.

To systematically assess the effects of propagule size on colonization, we mixed eight

pairs of *in vitro* "parent" communities at seven mixture ratios ranging from 1000:1 to

1:1000 (**Fig. 1d, Extended Data Fig. 2, Table S1**). We co-cultured mixtures in triplicate

for five passages to allow composition to re-equilibrate, and we passaged two replicates

of each parent community alongside to allow for comparison. After passaging, we used

16S sequencing to assess the effects of inoculation dose on community composition

and the colonization outcomes of individual ASVs.

Ecological neutral theory predicts that if all species are competitively neutral, then

propagule size is the only determinant of colonization outcomes[12,28,51]. This theory

provides a useful null model for measuring the impact of propagule size on community

composition and the colonization of individual species during community coalescence.

We compared this model to an alternative hypothesis that non-neutral resource

competition determines colonization success, meaning that co-culture compositions

should be largely independent of inoculation dose[5,12,52]. To assess the similarity of our

data to the neutral model, we computationally generated a set of theoretical mixture

communities with compositions based on the inoculation dose of each species

(**Extended Data Fig. 3a**), which also represented the initial state of each experimental

co-culture before passaging. A large proportion of ASVs in each mixture were shared

between parent communities at similar levels (**Extended Data Fig. 3b,c**), and were

therefore predicted to maintain their abundance across mixture ratios; all remaining

ASVs were predicted to vary in relative abundance with inoculation dose (**Extended

Data Fig. 3a**).

The neutral model predicts that diversity (defined here as the effective number of species inferred from the Shannon diversity index) should be maximal at the 1:1 mixture ratio. Even though many ASVs were shared between parent communities, the predicted maximum diversity of theoretical co-cultures at the 1:1 mixture ratio was higher than the diversity of either parent community, accounting for the detection limit of 16S sequencing (**Fig. 1e,f**, Wilcoxon rank-sum one-sided test, $P=6.1\times10^{-5}$, **Extended Data Fig. 4, Methods**). Contrary to the neutral prediction, we found that co-cultures consistently maintained diversities similar to or lower than one of the parent communities, despite some variation over mixture ratios (**Fig. 1e,f, Extended Data Fig. 4**). This finding suggests that the diversity of the parent communities is already nearly maximized such that most niches are occupied by resident species, and that non-neutral competition between residents and new species makes it difficult for all species to coexist during coalescence.

Although propagule size was not the sole determinant of community diversity, diversity sometimes varied over mixture ratios, suggesting that it was impacted by inoculation dose. To further investigate the effects of propagule size on community composition, we compared each experimental co-culture to its corresponding theoretical co-culture using the Jensen-Shannon divergence (JSD, **Methods**). Experimental and theoretical co-cultures were less similar than replicate communities were to each other (**Fig. 1g**, Wilcoxon rank-sum two-sided test, $P<10^{-15}$), further supporting the importance of non-neutral competition during colonization. To assess the relative importance of propagule

size and non-neutral competition across mixtures, we used JSD to quantify the

difference between each co-culture and its corresponding parent communities.

Co-cultures consistently differed from the corresponding theoretical expectation at every

mixture ratio (**Fig. 1h**). In some mixtures, co-culture composition was similar across

mixture ratios (B1/C1, **Fig. 1h**), indicating that propagule size had a weak effect.

However, in other mixtures, composition varied substantially with inoculation dose,

despite deviating from neutral predictions (D1/D2, **Fig. 1h**). These findings highlight the

varied importance of propagule size and competition across mixtures.

Dose-dependent colonizers are present in every community mixture

To investigate why co-culture composition exhibits a stronger dependence on propagule

size in some mixtures than in others, we examined colonization outcomes at the ASV

level. We compared the relative abundance of each ASV after five passages to the

neutral prediction determined solely by its inoculation dose in the initial mixture

(**Extended Data Fig. 5**). Of the ASVs that had distinct relative abundances in the two

parent communities, some maintained a consistent relative abundance in co-cultures

across mixture ratios, suggesting a minimal effect of propagule size on their colonization

(**Fig. 2a, Extended Data Fig. 6a,b**). We classified these dose-independent ASVs as

strong or weak colonizers depending on whether their relative abundance was greater

or less than the neutral prediction, respectively.

In contrast, many of the remaining ASVs exhibited dose-dependent colonization, with

relative abundances that varied systematically across mixture ratios (**Fig. 2b,c,**

Extended Data Fig. 6c). Dose-dependent colonizers were present in every community

mixture, often in higher numbers than strong or weak colonizers, demonstrating that

propagule size can influence the colonization of a sizeable fraction of species in each

community during coalescence. Intriguingly, most strong, weak, and dose-dependent

colonizers exhibited distinct colonization patterns in different co-cultures and between

closely related taxa (**Fig. 2d, Methods**), suggesting that these behaviors were

influenced by community context rather than species identity or phylogeny.

Dose-dependent colonizers accounted for ~10-20% of the relative abundance of each

co-culture (**Fig. 2e**) except the one whose composition showed the least dose

dependence (B1/C1, **Fig. 1h**). A large fraction of the total relative abundance of each

co-culture was accounted for by resident ASVs that were at similar relative abundances

in the two parent communities and remained consistently at this relative abundance

across mixture ratios (**Fig. 2f, Extended Data Fig. 7a**). Another large fraction of each

community consisted of noisy ASVs whose relative abundance varied inconsistently

across inoculation doses (**Fig. 2g, Extended Data Fig. 7b**). Notably, many noisy ASVs

exhibited large changes in relative abundance across mixture ratios that were

consistent among the three replicates of each mixture (**Extended Data Fig. 7b**),

suggesting that changes in the surrounding community may have driven these

fluctuations. Overall, our results show that propagule size influences the colonization of

a distinct set of species in every co-culture, even though the colonization of other

species is primarily determined by non-neutral competition.

Transient dose dependence arises from high niche overlap in a consumer-

resource model

We next sought to identify the ecological forces that give rise to dose dependence

during community mixtures. In the D1/D2 co-cultures, we observed three dose-

dependent *Lactococcus* and *Enterococcus* colonizers from the D1 parent community.

The relative abundances of these ASVs were roughly anticorrelated with those of

another dose-dependent *Lactococcus* from the D2 community (**Fig. 3a**). This

observation led us to hypothesize that dose dependence could arise from competition

between phylogenetically related taxa that occupy overlapping niches.

To understand whether competition for shared resources could explain this dose-

dependent colonization, we used a standard consumer-resource (CR) model to

investigate the conditions under which dose dependence could arise[52] (**Methods**). We

first studied a system in which two species compete for a set of three common

resources: μ_1, μ_2, and μ_{12} (**Fig. 3b**). Two of these resources, μ_1 and μ_2, are consumed

exclusively by species 1 and 2, respectively, and the third resource, μ_{12}, is consumed

by both species at equal rates ($R_{1,12} = R_{2,12}$). We investigated how the niche overlap of

species 1 with species 2, defined as $\gamma_{12} = \frac{\mu_{12}}{\mu_1 + \mu_{12}}$, would affect the colonization behavior

of these two species. We varied γ_{12} from 0.025 to 1 while maintaining the same amount

of unique resources for each species $\mu_1 = \mu_2$. We initialized the abundances of the two

species at ratios from 1000:1 to 1:1000, as in our community coalescence experiments,

and we simulated the CR model to test how species abundances would change over

five passages.

In this simple system, the dose dependence of both species increased with the amount

of overlap between their niches (**Figure 3c**). For example, when niche overlap was low,

the two species rapidly reached consistent relative abundances across mixture ratios

($\gamma_{12} = 0.75$, **Figure 3c**). By contrast, when niche overlap was high, the relative

abundance of both species showed dose dependence, varying by ~20-fold between the

1000:1 and 1:1000 mixture ratios after three passages ($\gamma_{12} = 0.975$, **Figure 3c**),

consistent with the magnitude of dose dependence that we often observed during

community coalescence (**Fig. 2b, Fig. 3a**). Similar levels of dose dependence were

present even when the consumption rate of the shared resource was varied by ~10%

(**Extended Data Fig. 8a**). These results show that dose dependence can arise when

the growth of two species is dominated by their consumption of shared resources.

Although these species showed dose dependence when niche overlap was high, the

magnitude of this dose dependence declined over time (**Fig. 3c, Extended Data Fig.**

8b). For example, when $\gamma_{12} = 0.975$, the dose dependence of the two species declined

from ~20-fold to ~10-fold from passage three to passage five (**Fig. 3c**). This declining

dose dependence was consistent with decreases in dose dependence over passages

that we observed among some ASVs during community coalescence (**Extended Data**

Fig. 8c). This finding demonstrates that high niche overlap can slow a community's

convergence to equilibrium, creating transient, neutral-like colonization dynamics.

Strain isolates show less dose dependence in pairwise mixtures than during

community coalescence

Based on the predictions of our CR model, we hypothesized that the *Lactococcus* and

Enterococcus species that showed dose-dependent colonization in the D1/D2 co-

cultures had high levels of niche overlap that would lead them to show dose

dependence in pairwise mixtures as well. To test this prediction, we isolated

representatives of these ASVs (**Methods**). We performed Sanger sequencing of the

16S region and whole-genome sequencing (**Methods**) to identify the strains from the

D1 community as *Lactococcus garvieae*, *Enterococcus faecalis*, and *Enterococcus*

casseliflavus, and the strain from the D2 community as *Lactococcus lactis* (**Fig. 4a**).

These strains exhibited similar maximum growth rates in monoculture, although small

differences in growth dynamics suggested the potential for distinct profiles of resource

utilization (**Fig. 4a**).

We performed a set of pairwise mixtures of *L. lactis* with the strains isolated from

community D1 at ratios from 1000:1 to 1:1000 (**Fig. 4b-d**). Most strains exhibited some

degree of dose-dependent colonization after three passages, suggesting that they had

some niche overlap (**Fig. 4b-d**). However, after five passages, each strain reached a

relative abundance that was consistent across mixture ratios (**Fig. 4b-d**). This lack of

dose dependence contrasted with their behavior in the D1/D2 co-cultures, in which dose

dependence was maintained even after five passages (**Fig. 3a**). This discrepancy

shows that pairwise strain mixtures are unlikely to fully capture the magnitude and

duration of dose dependence that we observe in more complex communities.

Dose dependence is enhanced in mixtures with more diverse communities

We used our CR model to investigate how dose dependence would be affected by the

addition of a third species. We introduced a third species into the model that consumes

a unique resource that was initialized each passage at the same level as the other

resources, such that $\mu_3 = \mu_2$ (**Fig. 5a**). Species 3 also competes with species 1 for a

fraction β of resource μ_1, which it consumes at relative rate r compared to species 1. To

investigate how this new species affects colonization, we varied β from 0 to 0.9 (**Fig.**

5b). As before, we initialized the abundances of species 1 and 2 at a range of ratios

from 1000:1 to 1:1000, and we set the abundance of species 3 equal to that of species

2. We simulated the CR model to investigate how species abundances changed over

five passages.

In this three-species model, species 1 showed higher dose dependence as its niche

overlap with species 3 increased (**Fig. 5b**). When species 3 had minimal niche overlap

with species 1 ($\beta = 0.1$, **Fig. 5b**) or was a weak competitor ($r \ll 1$, **Extended Data Fig.**

8d), species 1 showed similar levels of dose dependence as when species 3 was

absent (**Fig. 3c**), as expected. However, when niche overlap was high ($\beta > 0.5$) and

species 3 was a strong competitor ($r \gg 1$), species 3 effectively outcompeted species 1

for the use of their shared resource (**Fig. 5b, Extended Data Fig. 8d**). Under these

conditions, the dose dependence of species 1 increased substantially due to its

increased reliance on the resource shared with species 2 (**Fig. 5b**, **Extended Data Fig.**

8d). At the same time, the dose dependence of species 2 was minimally affected by the

amount of niche overlap between species 1 and 3, showing that the addition of new

species does not necessarily alter the colonization behavior of all community members

(**Fig. 5b, Extended Data Fig. 8d**). Together, these results predict that initial mixture

ratios should have larger, more lasting effects on community composition in more

diverse communities, in which competition among multiple species limits the unique

resources available for any species.

To test these predictions, we introduced the *L. lactis* strain into an even mixture of *E.*

faecalis and *E. casseliflavus* at ratios from 1000:1 to 1:1000. Consistent with our

prediction, after three passages, the degree of dose-dependence exhibited by *L. lactis*

increased to ~200-fold (**Fig. 5c**) from ~60-fold in pairwise co-cultures (**Fig. 4b,c**), while

neither *E. faecalis* nor *E. casselifavus* were classified as dose-dependent colonizers

(**Fig. 5c**). However, after five passages, none of the three strains exhibited dose-

dependent colonization (**Fig. 5c**). These results support our model's prediction that the

addition of a new species can increase the magnitude of dose dependence, even if this

dose dependence still declines more rapidly than during community coalescence.

To test the prediction that introducing strains into a more diverse community would

increase the magnitude and duration of dose dependence by increasing niche overlap,

we performed a set of strain-community mixtures in which we mixed each of the two

Enterococcus and *Lactococcus* strains into the D1 or D2 communities at ratios from

1000:1 to 1:1000 (**Fig. 5d,e, Extended Data Fig. 9**). In striking contrast to the pairwise

and three-strain mixtures, a subset of strains in each strain-community mixture showed

dose dependence after five passages (**Fig. 5d,e, Extended Data Fig. 9**). For instance,

E. casseliflavus and *L. lactis* were no longer dose-dependent by the fifth passage in

pairwise co-cultures (**Fig. 4c**). However, when *L. lactis* was introduced into the D1

community, both strains exhibited dose dependence after five passages, varying by

~10-fold in relative abundance across all mixture ratios (**Fig. 5d**). Similarly, when *L.*

garvieae was introduced into the D2 community, it showed complementary patterns of

dose dependence with *L. lactis*, even though *E. casseliflavus* remained at consistent

relative abundances across mixture ratios (**Fig. 5e, Extended Data Fig. 9c**). Together,

these results demonstrate that dose dependence can increase with community

diversity. These findings are consistent with the model prediction that the presence of

additional species increases niche overlap, amplifying the effect of inoculation dose on

colonization.

Discussion

In this study, we quantified the effect of propagule size on hundreds of simultaneous colonization events during community coalescence. We found that propagule size has persistent effects on colonization for a subset of species in complex communities, even after >30 generations of competition—long enough to effect significant ecological changes in a community regardless of ultimate colonization success[53,54]. Using a consumer-resource model, we showed that the magnitude and duration of dose dependence increase when niche overlap is high, as would be expected in diverse communities. In accordance with model predictions, we observed stronger dose dependence during community coalescence and strain-community mixtures than in two- and three-strain mixtures. Our model and results suggest that neutral-like colonization dynamics can arise through resource competition alone, without a need to invoke more complex phenomena like resource fluctuations[55], higher-order species interactions[56,57], or environmental modification[7,11].

The framework we propose suggests several principles for the design of ecological interventions like microbiome therapeutics[38] and agricultural inoculants[58]. First, our model predicts that introduced species will show higher levels of dose dependence in diverse communities. This prediction suggests that experiments with small, model communities may systematically underestimate the effects of propagule size in more diverse, natural communities, emphasizing the importance of conducting large-scale field experiments and clinical studies to test ecological interventions.

Our work identifies niche overlap as the main reason propagule size has stronger

impacts on colonization in more diverse communities. While niche overlap is

challenging to measure, it is often higher among closely related species, which tend to

consume more similar resources[59]. We might therefore predict higher levels of dose

dependence when introduced species are taxonomically related to existing residents, as

our *Lactococcus* and *Enterococcus* example illustrates. A more precise method for

predicting niche overlap in microbial communities is metabolomics, which can be a

powerful tool for quantifying resource utilization[46,60,61]. Future studies that combine

controlled colonization experiments with metabolomics of microbial isolates will shed

further light on the ecological factors that determine colonization success.

Materials and Methods

Data availability

Sequencing data will be made publicly available from the NCBI SRA upon manuscript publication.

Code availability

The computer code that performs the analysis is available at

https://github.com/DoranG1/dose-dependent-colonization.

Collection of stool samples from human subjects

Stool samples were collected as a part of a study[62] approved by the Stanford University Institutional Review Board under Protocol 54715. Informed written consent was obtained from all participants. Samples were provided by healthy, non-pregnant adults living with at least one roommate or housemate. At least two participants were recruited to provide samples from each household, one of whom took a five-day course of the antibiotic ciprofloxacin during sampling. Subjects collected stool samples in sterile plastic tubes and froze them immediately at -20°C inside multiple layers of insulating protection. Samples were transported to the laboratory on dry ice within six weeks of collection and were stored at -80°C. For this study, we derived *in vitro* communities from the last pre-antibiotic stool sample collected by eight study subjects from four household pairs. These eight subjects lived in the San Francisco Bay Area and ranged from 25-30 years old.

Derivation of top-down *in vitro* communities

To obtain stool fragments for inoculation of *in vitro* communities, we used a Basque

Engineering CXT 353 Benchtop Frozen Aliquotter to drill cores from the frozen stool

samples of eight study subjects: A1, A2, B1, B2, C1, C2, D1, and D2 (**Table S1**). Cores

were kept frozen by immersing the drill area in liquid nitrogen and were stored at -80 °C

after drilling.

We generated *in vitro* communities following methods established in our previous

studies[49,50]. All steps of community inoculation and passaging were performed in an

anaerobic chamber (Coy Instruments), using sterilized pipette tips. First, a 50-mg

fragment of each stool core was resuspended in 500 µL of filter-sterilized PBS. Next, 20

µL of the stool resuspension were added to 180 µL of Brain Heart Infusion (BHI, BD

Biosciences 237200) supplemented with 0.2 mg/mL L-tryptophan, 1 mg/mL L-arginine,

0.5 mg/mL L-cysteine, 5 µg/mL vitamin K, 1 µg/mL haemin, and 5 g/mL mucin (referred

to here as mBHI+mucin) in a flat-bottom 96-well plate (Greiner Bio-One 655161). Each

sample was inoculated as three replicates in separate wells. All media were autoclaved

before inoculation to ensure sterility.

After inoculation, communities were grown at 37 °C in the anaerobic chamber and

passaged every 48 h by transferring 1 µL of saturated culture into a new 96-well plate

containing 199 µL of fresh mBHI+mucin. We passaged each community 15 times and

stored each plate of saturated cultures at -80 °C after passaging by mixing 100 μL of

saturated culture with 100 μL of 50% filter-sterilized glycerol.

Whole community mixtures

We thawed glycerol stocks of one replicate of each community from passage 15 and

reinoculated these cultures in an anaerobic chamber (Coy Instruments) by adding 3 μL

of glycerol stock to 197 μL of mBHI+mucin. These communities were grown at 37 °C for

48 h.

To create community mixtures, we first generated a dilution series of each of the

unmixed parent communities. Fresh mBHI+mucin was added to each saturated culture

to create communities with concentrations 1/10, 1/100, and 1/1000 of the parent

community. We generated each dilution in triplicate from the single replicate of each

parent community, resulting in 24 communities for each dilution factor from the eight

parent communities, and 72 dilution communities in total across all three dilution factors.

To generate each pairwise mixture, we added 1 μL of each dilution community to 1 μL

of another parent community and 198 μL of fresh mBHI+mucin. We mixed communities

in eight combinations: A1/A2, B1/B2, C1/C2, D1/D2, A1/B1, B1/C1, C1/D1, and D1/A1.

Four pairs of communities (A1/A2, B1/B2, C1/C2, D1/D2) were derived from subjects

from the same household, while the other four mixtures (A1/B1, B1/C1, C1/D1, D1/A1)

involved communities from different households. Each pair of communities was mixed

at seven ratios: 1000:1, 100:1, 10:1, 1:1, 1:10, 1:100, and 1:1000. Each mixture was

passaged in triplicate. This protocol resulted in 168 co-culture communities (8 mixtures

× 7 ratios × 3 replicates), which were passaged in two flat-bottom 96 well plates. We

also inoculated 2 μL of each undiluted parent community in 198 μL of mBHI+mucin in

duplicate, comprising 16 unmixed control parent communities. The four corners of each

96-well plate were filled with 200 μL of mBHI+mucin as blank controls to monitor

contamination.

Both 96-well plates containing mixture co-cultures and unmixed control communities

were grown at 37 °C and passaged every 48 h five times by adding 1 μL of saturated

culture to 199 μL of mBHI+mucin. Saturated cultures from each passage were stored at

-80 °C.

16S rRNA gene sequencing

DNA was extracted from 50 μL of each community from each mixture using the Qiagen

DNeasy Ultraclean 96 Microbial Kit (Qiagen 10196-4) according to the manufacturer's

instructions. Extracted DNA was stored at -20 °C.

We performed 16S rRNA gene amplicon sequencing targeted to the V4 region.

Extracted DNA was amplified using primers modified from the Earth Microbiome Project

spanning 515F–806R, with gene-specific sequences GTGYCAGCMGCCGCGGTAA for

the forward primer and GGACTACNVGGGTWTCTAAT for the reverse primer[63,64]. The

16S region was amplified using 27 cycles of PCR with an annealing temperature of 50

°C, and successful amplification of the 16S region was confirmed using gel

electrophoresis. Sequencing primers were attached using 10 cycles of PCR with an

annealing temperature of 54 °C. Excess primer was removed using a 0.8X Ampure

bead cleanup. Libraries were pooled to contain equal volume from each sample and

sequenced using an Illumina NovaSeq SP at the DNA Services Lab, Roy J. Carver

Biotechnology Center, University of Illinois at Urbana-Champaign.

Raw reads were annotated and demultiplexed using UMI-tools[65], and primer and

adapter sequences were trimmed using cutadapt[66]. To remove the blank media

controls, samples with <100 reads were excluded. DADA2[67] was used to filter and

truncate reads and to assign amplicon sequence variant (ASV) taxonomy based on the

SILVA (release 138) reference database[68], and to obtain a phylogeny of all ASVs.

Our final community coalescence dataset contained 91,504,441 reads with a median of

201,425 reads (standard deviation 69,902) per non-blank sample. Two samples in our

dataset (A1 replicate 1 from passage 3, and D1/A1 100:1 replicate 1 from passage 5)

received zero sequencing coverage, so these were excluded prior to subsequent

analyses. 1715 unique ASVs were detected across all samples, and 198 of these were

present at ≥10^{-3} relative abundance in at least one sample, which we estimated as the

limit of detection. Although subjects from the same households often shared a higher

proportion of gut microbial strains compared to subjects from different households[62], we

observed no differences in the proportion of ASVs shared in mixtures of communities

from the same versus different households (**Extended Data Fig. 3**). We used the

relative abundance of each ASV to calculate Jensen-Shannon divergence.

In addition to our community coalescence dataset, we obtained a second dataset

representing the *L. garvieae*/*L. lactis* co-cultures after five passages and the D2/*L. lactis*

co-cultures after five passages. This dataset contained 48,347,562 reads with a median

of 683,158 reads (standard deviation 370,504) per non-blank sample. We also obtained

a third dataset representing the *E. faecalis*/*L. lactis*, *E. casseliflavus*/*L. lactis*, three-

strain, and all strain-community co-cultures after three and five passages. This third

dataset contained 9,429,310 reads with a median of 12,927 reads (standard deviation

12,375) per non-blank sample. We excluded one replicate each from the *L. garvieae*/*L.*

lactis, D1/*L. lactis, L. garvieae*/D2, and *E. casseliflavus*/D2 (passage 5) mixtures based

on poor sequencing coverage at one or more mixture ratios in those replicates.

Estimation of theoretical community compositions

To assess the effect of propagule size in our community mixtures, we computationally

generated a set of theoretical community mixtures according to the predictions of

ecological neutral theory[51]. The expected abundance of each ASV across mixture ratios

was calculated based on a weighted average of the relative abundance in the parent

communities: $\frac{x_1}{A} + \frac{x_2}{B}$, where x_1 and x_2 represent the relative abundance of the ASV in

parent community 1 and 2, respectively, and A and B represent the dilution factor of

parent community 1 and 2, respectively, at the given mixture ratio. For example, the

predicted abundance of an ASV in a 1:100 mixture of A1 and A2 in which the

inoculation dose of A2 is 100 times greater than that of A1 is $\frac{x_{A1}}{100} + x_{A2}$. We normalized

these weighted averages to calculate a theoretical relative abundance for each ASV in a

mixture, and we ignored ASVs with predicted relative abundance $<10^{-3}$, our sequencing

limit of detection. We generated two theoretical communities for each pair of parent

communities based on the two replicates of each parent community in our experiments.

We generated separate sets of theoretical communities from passages 3 and 5 of the

parent communities to account for changes to parent community composition that may

have occurred during passaging.

Categorization of ASVs by colonization behavior

We developed a set of summary statistics for each ASV based on its average change in

relative abundance across mixture ratios and the deviation from predicted neutral

relative abundances, which enabled us to systematically categorize each ASV into one

of six distinct patterns: low abundance, noisy, resident, dose-dependent, strong

colonizer, or weak colonizer (**Extended Data Fig. 5**).

For each ASV in each mixture, we calculated four measures. The average $\log_{10}$ fold

change in relative abundance across mixture ratios (net dose difference) and average

absolute value of $\log_{10}$ fold change in relative abundance across mixture ratios

(magnitude dose difference) were used to quantify each ASV's level of dose

dependence. The average $\log_{10}$ fold difference between experimental and theoretical

relative abundances across mixture ratios (net theoretical difference) and average

absolute value of $\log_{10}$ fold difference between experimental and theoretical relative

abundances across mixture ratios (magnitude theoretical difference) were used to

quantify each ASV's deviation from neutral predictions and distinguish between

resident, strong colonizing, and weak colonizing ASVs.

Net dose difference was calculated as $\frac{\sum(x_i - x_{i-1})}{n}$, where x_i represents the $\log_{10}$(relative

abundance) of an ASV at a single mixture ratio (e.g., 100:1), x_{i-1} represents the

$\log_{10}$(relative abundance) at the preceding mixture ratio (e.g., 1000:1), and $n = 7$

represents the total number of mixture ratios. Magnitude dose difference was calculated

as $\frac{\sum|x_i - x_{i-1}|}{n}$. Both net dose difference and magnitude dose difference were first

calculated individually for each replicate of the ASV and then averaged across

replicates.

Net theoretical difference was calculated as $\frac{\sum(\bar{x}_a - \bar{x}_t)}{n_{\text{ratio}}}$, where $\bar{x}_a$ and $\bar{x}_t$ represent the

average experimental and theoretical, respectively, $\log_{10}$(relative abundance) across

replicates at a single mixture ratio. Magnitude theoretical difference was calculated as

$\frac{\sum|\bar{x}_a - \bar{x}_t|}{n_{\text{ratio}}}$. For ASVs that were not detected in the co-cultures at some mixture ratios

and/or parent communities, we set the relative abundance to 10^{-3}.

To categorize ASVs, we used five binary decisions based on our summary statistics and

other measures (**Extended Data Fig. 5**). First, ASVs were classified as "low

abundance" if their relative abundance was $<10^{-2.58}$, a threshold slightly above our limit

of detection, in both parent communities. Next, ASVs were classified as "noisy" if the

ratio of magnitude dose difference to the absolute value of net dose difference was

>2.25, indicating large, non-monotonic changes in relative abundance over mixture

ratios. To prevent ASVs with extremely low change over mixture ratios (absolute value

net dose difference <0.1) from artificially inflating the number of noisy ASVs, we

replaced these low values with 0.1 as a pseudocount.

If ASVs were not classified as low abundance or noisy, they were classified as "dose-

dependent colonizers" if the absolute value of the net dose difference was >0.09,

meaning that they exhibited large monotonic changes in relative abundance over

mixture ratios. Of the remaining ASVs, if the magnitude theoretical difference was >0.3

and the absolute value of the net theoretical difference was ≥0.2, indicating a strong

deviation from neutral predictions, they were classified as "strong colonizers" or "weak

colonizers" based on whether the sign of net theoretical difference was positive or

negative, respectively. All other ASVs, which showed little dose dependence or

deviation from neutral predictions, were classified as "resident". All numeric cutoffs

between categories were chosen to minimize the number of ASVs classified as low

abundance, noisy, resident, or low abundance, while ensuring that dose-dependent,

strong, and weak colonizers could be readily distinguished by inspection.

Consumer-resource model

We implemented a standard consumer-resource model[52]:

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$$\frac{dX_i}{dt} = X_i \sum_{\mu=1}^{M} R_{i,\mu} Y_\mu$$

$$\frac{dY_\mu}{dt} = -Y_\mu \sum_{i=1}^{N} R_{i,\mu} X_i$$

where X_i denotes the absolute abundance of species i, Y_μ the amount of resource μ,

and $R_{i,\mu}$ the consumption rate of resource μ by species i. Species abundances were

simulated in Matlab (Mathworks, Inc.) using the ode45 solver until all resources were

depleted. After each such passage, the abundances were diluted 1:200, and resource

concentrations were refreshed. We initialized the abundances of the two species at a

range of ratios from 1000:1 to 1:1000, as in our community coalescence experiments,

and we simulated five passages for each community. We defined dose dependence as

the ratio of a species' relative abundances at starting ratios of 1000:1 and 1:1000.

We first investigated a simple system in which two species, denoted 1 and 2, compete

to consume use a set of three common resources: μ_1, μ_2, and μ_{12} (**Fig. 3b**). Two of

these resources, μ_1 and μ_2, are consumed exclusively by species 1 and 2 respectively,

and the third resource, μ_{12}, is consumed by both species at equal rates ($R_{1,12} = R_{2,12} = $

1). To investigate how the niche overlap of species 1 with species 2, defined as $\gamma_{12} = $

$\frac{\mu_{12}}{\mu_1 + \mu_{12}}$, would affect the colonization behavior of these two species (**Fig. 3c**), we varied

γ_{12} from 0.025 to 1 while maintaining the same amount of unique resources for each

species $\mu_1 = \mu_2 = 1$. To test how varying the rates of shared resource consumption

affects colonization behavior, we varied $R_{1,12}$ from 0.9 to 1.1 while maintaining a

constant $R_{2,12} = 1$ (**Extended Data Fig. 8a**).

We investigated how dose dependence would be affected by the addition of a third

species. We introduced a third species into the model that consumes a unique

resource, $\mu_3 = \mu_2 = 1$. Species 3 also competes with species 1 for a fraction β of

resource μ_1, which it consumes at relative rate $r = \frac{R_{3,13}}{R_{1,13}}$ compared to species 1. To

investigate how the presence of this new species affects colonization, we varied β from

0 to 0.9 while maintaining a constant $r = 10$ (**Fig. 5b**). As before, we initialized the

abundances of species 1 and 2 at a range of ratios from 1000:1 to 1:1000, and we set

the abundance of species 3 equal to that of species 2. To investigate how the

consumption rate of this shared resource affects colonization behavior, we also tested

the effect of varying r from 10^{-2} to 10^{2} while maintaining a constant $\beta = 0.5$ (**Extended**

Data Fig. 8d).

Strain isolation

Lactococcus garvieae, *Lactococcus lactis*, *Enterococcus faecalis*, and *Enterococcus*

casseliflavus strains representing dose-dependent ASVs were isolated by sorting single

cells from communities D1 and D2 into mBHI-filled wells of a 96-well plate using a

fluorescence-activated cell sorter (FACS)[69]. To do so, communities were transported

from the anaerobic chamber to the FACS facility in an anaerobic container (BD

Biosciences) to ensure minimal oxygen exposure. We confirmed the purity of each

strain isolate by streaking multiple times on BHI agar plates supplemented with 5%

horse blood and submitted individual colonies for Sanger and whole-genome

sequencing. Colonies verified as pure were grown for 48 h in 200 μL of mBHI+mucin in

a 96-well plate, and 75 µL of each saturated culture were mixed with 75 µL of 50%

glycerol and stocked in plastic crimp vials (ThermoFisher C4011-11) for future use.

To confirm isolate purity, we performed Sanger sequencing targeted to the V4 region of

the 16S rRNA gene. We amplified the 16S gene with 35 cycles of PCR with an

annealing temperature of 55 °C using the previously described primers. The resulting

PCR products were sent to Elim Biopharmaceuticals for sequencing.

We determined the species identity of strain isolates using whole-genome sequencing.

DNA was extracted from isolate cultures using the DNeasy Ultraclean 96 Microbial Kit

(Qiagen 10196-4) according to the manufacturer's instructions. Libraries were prepared

for whole-genome sequencing using the Nextera DNA Flex Library Prep Kit (Illumina

20018705) according to the manufacturer's instructions. The DNA concentration of each

library was quantified using the Quant-iT High-Sensitivity dsDNA Assay Kit (Invitrogen

Q33130) and a Synergy HT Microplate Reader (Biotek Instruments), and libraries were

pooled at equal DNA input. Pooled libraries were sequenced at the DNA Services Lab,

Roy J. Carver Biotechnology Center, University of Illinois at Urbana-Champaign.

Co-cultures of strain isolates and strain-community mixtures

We performed a set of four mixtures of two or three strains (*L. garvieae*/*L. lactis*, *E.*

faecalis/*L. lactis*, *E. casseliflavus*/*L. lactis*, and *E. faecalis*/*E. casseliflavus*/*L. lactis*) and

five strain-community mixtures (D1/*L. lactis*, *L. garvieae*/D2, *E. faecalis*/D2, *E.*

casseliflavus/D2, *E. faecalis*/*E. casseliflavus*/D2) using the glycerol stocks of our four

strain isolates and of the D1 and D2 parent communities. We performed mixtures at

ratios from 1000:1 to 1:1000 in triplicate as described above for whole community

mixtures. For mixtures that included both *E. faecalis* and *E. casseliflavus*, 0.5 μL of each

strain culture were used. Mixtures were passaged in mBHI+mucin for 48 h at 37 °C five

times.

Analysis of pairwise mixture composition using Sanger sequencing data

To analyze the composition of co-cultures from the *L. garvieae*/*L. lactis* mixture after

three passages, we performed Sanger sequencing of each co-culture as described

above and used the CASEU R package[70] to quantify the relative abundance of both

strains at each mixture ratio.

Quantification of strain isolate growth

Growth curves were obtained using an Epoch 2 Microplate Spectrophotometer (Biotek

Instruments). Strains were grown overnight and then diluted 1:10, 1:100, or 1:1000 into

fresh BHI+mucin in 96-well plates. We measured the optical density at 600 nm (OD_{600})

of 10 replicates of each strain over 48 h of anaerobic growth at 37 °C with continuous

orbital shaking. Instantaneous growth rate was quantified as $d \ln(OD_{600})/dt$ using the

gcplyr R package[71].

Author contributions

D.A.G., K.S.X., D.A.R., and K.C.H. designed the research; D.A.G., K.S.X., A.B.P., and

R.R.J. performed the research. D.A.G., K.S.X., J.G.L., J.C.C.V., D.A.P., B.H.G., D.A.R.,

and K.C.H. analyzed the data; D.A.G., K.S.X., D.A.R., and K.C.H. wrote the paper; and

all authors reviewed it before submission.

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Figures

Figure 1: Diversity and composition of co-cultures vary by inoculation dose and deviate from neutral predictions

a) *In vitro* communities were inoculated in triplicate from stool samples collected from eight healthy human subjects and passaged 15 times to reach stability.

b) The diversity of each community over time, quantified as the effective number of species ($e^{H'}$) calculated from the Shannon diversity index (H'), initially decreased and then plateaued by passage 3.

c) Correlation of family-level relative abundance across passages 0, 3, and 7 (Pearson's tests, left: n=25, ρ=0.25, P=0.23; right: n=22, ρ=0.93, P<10^{-9}). Points are colored by family. The dashed line indicates a perfect correlation in relative abundance between passages.

d) Experimental design. Eight pairs of *in vitro* communities were mixed in triplicate at mixture ratios ranging from 1000:1 to 1:1000. The resulting co-cultures were passaged five times.

e) Diversity of co-culture communities was similar to or lower than the diversity of parent communities across mixture ratios. Gray lines represent both replicates of predicted neutral mixtures for an example mixture based on the diversity of replicate parent communities (1:0 and 0:1, shown in green), and orange lines represent three replicates of experimental community mixtures. Diversity was quantified by the effective number of species ($e^{H'}$) calculated from the Shannon diversity index (H').

f) Distribution of diversity across parent and co-culture communities. The effective number of species at each mixture ratio more closely resembled the diversity of the parent communities than the higher-diversity theoretical prediction (Wilcoxon rank-

sum one-sided tests, parent ($n=32$) vs. theoretical ($n=16$) co-cultures: $P=6.1\times10^{-5}$;

theoretical vs. experimental 1:1 ($n=24$) co-cultures: $P=6.6\times10^{-7}$; theoretical vs. other

($n=48$ at each mixture ratio) co-cultures: $P<10^{-4}$).

g) Comparisons of co-culture communities to theoretical communities. Violin plots

represent the distribution of Jensen-Shannon divergence (JSD) between all pairs of

parent communities (including those that were not mixed), theoretical and

experimental mixtures at the 1:1 mixture ratio, and inoculation replicates of the 1:1

mixture ratio (significance calculated using Wilcoxon rank-sum one-sided tests,

parents ($n=56$) vs. replicates ($n=174$): $P<10^{-15}$; parents vs. theoretical-experimental

($n=334$): $P=1.8\times10^{-7}$; theoretical-experimental vs. replicates: $P<10^{-15}$).

h) Variation of community composition across mixtures and mixture ratios. Each plot

shows the Jensen-Shannon divergence (JSD) of all co-culture and theoretical

communities relative to both parent communities (brown and gold) for a single

mixture. Grey points along the dotted line represent theoretical mixtures, and colored

points represent data from experimental co-culture communities. Points show

averages and error bars show full range of values across inoculation replicates.

Figure 2: Colonization behaviors of individual ASVs influence community composition.

a) Representative examples of strong and weak colonizers. Each panel shows an individual ASV from one mixture. Gray lines represent both replicates of predicted neutral relative abundances for each ASV, and orange lines represent three replicates of experimental relative abundances.

b) Dose-dependent colonizers exhibit large changes in relative abundance across mixture ratios. Lines are colored as in (**a**).

c) The number of ASVs that display each colonization pattern in each set of mixtures.

Only strong, weak, and dose-dependent colonizers are shown. ASVs were counted

as present if they were detected in the 1:1 mixture at the fifth passage.

d) Colonization behavior is not associated with ASV phylogeny. The heatmap shows

the number of times each ASV exhibited each colonization pattern in any mixture.

Only ASVs exhibiting dose-dependent (DD), strong (S), or weak (W) colonization in

any mixture are shown.

e) Total relative abundance of each type of colonizer in each set of mixtures. Only the

1:1 mixture is shown. Black lines indicate separate ASVs. Relative abundances are

averaged across the three inoculation replicates.

f) A representative example of a resident ASV that was present at similar relative

abundances in both parent communities. Lines are colored as in (**a**).

g) A representative example of a noisy ASV whose relative abundance showed large

fluctuations between adjacent mixture ratios. Lines are colored as in (**a**).

Figure 3: A consumer-resource model predicts transient dose dependence.

a) ASVs from the *Enterococcus* and *Lactococcus* genera showed dose dependence

    during community coalescence. Origin of strain isolates from the D1/D2 co-cultures.

    Enterococcus faecalis, *Enterococcus casseliflavus*, and *Lactococcus lactis* were

    isolated from the D1 community, and *Lactococcus lactis* was isolated from the D2

    community. Gray lines represent both replicates of predicted neutral relative

    abundances for each ASV after five passages, and colored lines represent three

    replicates of experimental relative abundances after five passages.

b) Schematic of a simple consumer-resource model in which two species compete near

    neutrally (equal consumption rates $R_{1,12} = R_{2,12} = 1$) for resource μ_{12}, and resources

    μ_1 and μ_2 are unique to species 1 and 2, respectively (**Methods**).

c) The dose dependence of both species in the consumer-resource model increases as

    the niche overlap of species 1 with species 2 (γ_{12}) increases. Dose dependence also

    decreases from passage 3 (dotted line) to passage 5 (solid line). Colored points

    highlight three values of γ_{12} for which the relative abundances over mixture ratios of

749   each species are shown at right. Dose dependence is defined as the ratio of relative

750   abundances after passaging starting from a 1000:1 versus 1:1000 mixture ratio.

Figure 4: Transient dose-dependent colonization in pairwise strain mixtures.

a) Growth of the *Enterococcus* and *Lactococcus* strains in monoculture. Blank values were subtracted from the OD_{600} measurement of each strain, and growth curves were then $\log_{10}$ transformed. Per-minute growth rates were then calculated from these curves and smoothed over intervals of five minutes (**Methods**).

b-d) Species relative abundances in co-cultures of (b) *E. faecalis* and *L. lactis*, (c) *E. casseliflavus* and *L. lactis*, and (d) *L. garvieae* and *L. lactis*. Lines are colored as in **Fig. 3a**. Dose dependence decreased substantially from passage three (dotted lines) to passage five (solid lines).

Figure 5: Dose dependence increases in mixtures with more diverse communities.

a) Modification to the consumer-resource model in (**Fig. 3a**) in which a third species competes with species 1 for a fraction β of resource μ_1 with relative consumption rate r compared to species 1.

b) The dose dependence of species 1 after five passages increases with niche overlap between species 1 and 3 (β) at constant $r = 10$, while the dose dependence of species 2 is largely unaffected. Dose dependence is defined as in **Fig. 3c**.

770   c-e) Species relative abundances in co-cultures of (c) *E. faecalis, E. casseliflavus*, and

771   *L. lactis*, (d) the D1 community and *L. lactis*, and (e) *L. garvieae* and the D2

772   community. Lines are colored as in **Fig. 3a**. Dotted lines with open points show data

773   after three passages, and solid lines with closed points show passage five.

Extended Data Figure 1: Diversity and composition of stool-derived communities.

a) ASV richness of each community over passages. Only ASVs present in the

community at relative abundance $>10^{-3}$ were included in the richness calculation.

b) Family-level richness of each community over passages.

c) Proportion of the stool inoculum (passage 0) retained at each subsequent passage.

Each point shows the total ASV or family relative abundance of the stool inoculum

represented by taxa retained in communities at a given passage. Taxa were counted

as retained if they were detected at a relative abundance $>10^{-3}$.

d) Composition of one replicate of each community over passages. ASV relative

abundances are separated with solid lines and colored by family. Only the most

abundant families are shown in the key.

Extended Data Figure 2: Composition of co-culture communities.

ASV relative abundances are separated with solid lines and colored by bacterial family.

Only the most abundant families are shown in the key. The first replicate of the 100:1

mixture ratio of the D1/A1 mixture had low sequencing depth and was removed from

subsequent analyses.

Extended Data Figure 3: Theoretical community composition is dose-dependent, even though many ASVs are shared between parent communities.

a) The theoretical neutral community composition of an example mixture after five passages. 1:0 and 0:1 ratios represent unmixed parent communities. ASV relative abundances are separated with solid lines and colored by family.

b) Most ASVs in each co-culture were present in both parent communities. ASVs were counted as originating from one or both parent communities if they were present at relative abundance $>10^3$.

c) Relative abundances of ASVs at the 1:1 mixture ratio, colored by parent community of origin. Black lines indicate separate ASVs. Relative abundances are averaged across the three inoculation replicates.

Extended Data Figure 4: Diversity is limited in co-culture communities is robust

across multiple metrics.

a-b) Diversity of each co-culture community, quantified by (a) the effective number of

species ($e^{H'}$) calculated from the Shannon diversity index (H'), or (b) ASV richness,

assuming a limit of detection of 10^{-3} relative abundance. Lines are colored as in **Fig.**

1e.

c) Distribution of ASV richness across parent and co-culture communities.

Extended Data Figure 5: Classification scheme for ASV colonization behavior.

ASV totals for each category refer to the total number of observations of an ASV with

that behavior across all mixtures.

Extended Data Figure 6: Example strong, weak, and dose-dependent colonizers

from the C1/C2 mixture.

a-c) Relative abundance of (a) strong, (b) weak, and (c) dose-dependent colonizers

from the C1/C2 mixture. Each panel shows one ASV. Lines are colored as in **Fig.**

2a.

826 **Extended Data Figure 7: Resident and noisy ASVs from the C1/C2 mixture.**

827 a,b) Relative abundance of (a) resident and (b) noisy ASVs from the C1/C2 mixture.

828 Each panel shows one ASV. Lines are colored as in **Fig. 2a**.

Extended Data Figure 8: Predicted effects of niche overlap and resource consumption rates in a consumer-resource model.

a) Dose dependence after five passages in a two-species consumer-resource model is robust to ~10% variation in relative consumption rates $R_{1,12}$ and $R_{2,12}$ of a shared resource μ_{12}. Dose dependence is defined as in **Fig. 3c**.

b) Dose dependence declines over passages in a two-species consumer resource model (equal consumption rates $R_{1,12} = R_{2,12} = 1$), even when the niche overlap of species 1 and 2 (γ_{12}) is high.

c) The dose dependence of an ASV in the Tannerellaceae family declined substantially between passages three and five. Lines are colored as in **Fig. 2a**.

d) The dose dependence of species 1 after five passages increases with relative resource consumption rate of species 3 and 1 (r) at constant $\beta = 0.5$, while the dose dependence of species 2 is largely unaffected. Dose dependence is defined as in **Fig. 3c**.

Extended Data Figure 9: Colonization behavior in strain-community mixtures.

a) Species relative abundances in the *E. faecalis*/D2 co-cultures after five passages.

Only *L. lactis* is classified as dose-dependent.

b) Species relative abundances in the *E. casseliflavus*/D2 co-cultures after five

passages. *E. faecalis*, *E. casseliflavus*, and *L. lactis* are classified as dose-

dependent.

c) Species relative abundances in the *L. garvieae*/D2 co-cultures after three passages.

L. garvieae and *L. lactis* are classified as dose-dependent.

d) Species relative abundances in the *E. faecalis*/D2 co-cultures after three passages.

Only *L. lactis* is classified as dose-dependent.

e) Species relative abundances in the *E. casseliflavus*/D2 co-cultures after three

passages. *E. faecalis*, *E. casseliflavus*, and *L. lactis* are classified as dose-

dependent.

Lines are colored as in **Fig. 2a**.

Table S1. Sample origin of each *in vitro* community.

In vitro communities were derived from day 0 pre-antibiotic samples collected from

healthy adults[62].

Community	Subject	Household
A1	XBA	XB
A2	XBB	XB
B1	XCA	XC
B2	XCB	XC
C1	XDA	XD
C2	XDB	XD
D1	XFA	XF
D2	XFB	XF

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