A standardized gnotobiotic mouse model harboring a minimal 15-member

mouse gut microbiota recapitulates SOPF phenotype

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Abstract

Mus musculus is the classic mammalian model for biomedical research. Despite global efforts

in standardizing breeding and experimental procedures, the undefined nature and inter-

individual diversity of laboratory mouse microbiota remains a limitation. In an attempt to

standardize preclinical studies, we have developed a simplified mouse microbiota composed

of 15 strains from 7 of the 20 most prevalent bacterial families representative of the fecal

microbiota found in specific opportunistic- and pathogen-free (SOPF) C57BL/6J animals and

derived a standardized gnotobiotic mouse model called GM15. GM15 recapitulates extensively

the functionalities found in C57BL/6J SOPF microbiota metagenome and GM15 animals are

phenotypically similar to SOPF. They even perform better in a model of post-weaning

malnutrition. The GM15 model ensures an increased reproducibility and robustness of

preclinical studies by limiting the confounding effect of microbiota composition fluctuation and

offers new possibilities for research focusing on how the microbiota shapes host physiology in

health and diseases.

Background

The intestinal microbiota is a complex and dynamic ecosystem largely composed of bacteria whose activity profoundly influences our health and diseases [1]. Advances in sequencing and analytical technologies coupled with improved computing tools have revolutionized the field of host-microbiota interaction [2]. These developments have provided an increased depth and accuracy in the study of intestinal microbial assemblages and activity for correlative studies with human health or disease traits. Despite these sophisticated descriptions of host-microbiome interaction phenomena, the underlying causal mechanisms remain largely elusive [3].

The use of model organisms plays a decisive role in the challenge to move from correlation to causal links in the host-microbiome field as they have long enabled researchers to identify the shared biological functions among living organisms, and facilitated the discovery of conserved molecular mechanisms governing the fundamental principles of biology [4]. Thanks to its genetic and physiological similarities to humans, in addition to its rapid and prolific breeding, the mouse has been a classic mammalian model of choice for the past decades for biomedical research and the host-microbiome field is no exception [5]. While the use of defined genetic backgrounds, as well as the absence of specific pathogens [6] is now common practice in mouse studies, an important confounding factor is the variability in the composition of the intestinal microbiota among experimental animals and animal facilities [7,8]. This parameter is under the influence of multiple elements such as genetics, diets, biological rhythms and breeding conditions [9]. As a consequence, to restrain microbial diversity, efforts have been made to tailor protocols for microbiota-related mice studies and to standardize mouse microbiota composition [9, 10, 11, 12, 13].

The mouse gut microbiota richness is usually estimated at more than 300 bacterial genera [5] and common inhabitants of the mouse intestine belong to seven bacterial phyla with *Firmicutes*, *Bacteroidetes* and *Proteobacteria* being the most abundant ones [12, 14, 15]. The first attempt in standardizing the mouse microbial environment (initially to study immunocompromised mouse models) arose 15 years ago with the wide implementation of the

Specific Pathogen-Free (SPF) hygienic status of mouse husbandries [6] by re-deriving mouse

strains by two-cell-stage embryo transfer to SPF recipients and subsequent post-natal

inoculation with a cocktail of bacteria devoid of pathogens to homogenize microbial

colonization among a given animal facility. SPF, and then SOPF inbred lines (lacking Specific

Opportunist Pathogens, such as *Staphylococcus aureus* or *Pseudomonas aeruginosa*) now

represents the common standard health status for experimental mouse breeding. However,

despite the global efforts in standardizing the SPF procedures, the undefined nature and

important inter-individual diversity of the SPF microbiota remains a limitation in host-

microbiome studies, since the scientific community still lacks a common SPF standard cocktail

and rather use a facility-specific cocktail of bacteria [9]. Indeed, the microbiota fluctuates a lot

with diet and environment, so it is strictly impossible to have the exact same microbiota of SPF

mice in two different facilities.

Microbial cultivation and gnotobiology offer attractive strategies to standardize the

microbiota of mouse models. Germ-Free (GF) animals (i.e. animals devoid of any living micro-

organisms) are the originators of gnotobiotic animals (i.e. animals with a controlled microbiota)

obtained by colonization with pure culture or cocktails of bacterial strains [16]. Recent efforts

have been put into isolating, cultivating and archiving isolated culture of the dominant members

of the mouse microbiota [17]. Gnotobiotic animals can be kept in isolators for several

generations and offer the possibility of a strict control of their microbial status. Gnotobiotic

models offering different degree of microbial complexity have been developed in the past

ranging from mono-colonization (monoxenic animals) to high diversity microbiota models such

as conventionalized ex-GF animals using a donor microbiota [9, 12, 18]. Two models have

emerged for breeding and long-term experimental purpose: the Altered Schaedler Flora (ASF)

model and the recent oligo-Mouse-Microbiota$_{12}$ (oligoMM$_{12}$) model [19, 20]. These models offer

an enlarged microbial potential as compared to monoxenic mice, while keeping the model

simple and experimentally tractable as compared to conventionalized animals (SPF or SOPF).

The ASF was developed in the late 70s by adding bacterial strains which better

represented the microbiota of conventional mice to the initial Schaedler flora, a minimal

microbial consortium that protected ex-GF mice from opportunistic pathogen colonization

during breeding [21]. The ASF is composed of 8 defined bacteria, which are stable over mouse

generations. Mice colonized with ASF display a certain degree of immune normalization when

compared with SPF mice [19]. Limitations of the ASF model are that the strains are not publicly

available of the strains and that they are not all representative of the dominant members of the

mouse microbiota [9, 12]. In addition, ASF mice differ substantially from SPF mice with respect

to microbiota-associated traits beyond immune parameters and resistance to opportunistic

pathogen colonization, probably owing to the limited phylogenic diversity and metabolic

capabilities of the ASF consortium [9, 12, 22]. Recently, the oligoMM$_{12}$ model was developed

[20]. It is a minimal microbiota gnotobiotic model composed of 12 defined cultivable mouse

commensal bacteria from the miBC collection representing members of the major bacterial

phyla of the mouse gut [17, 23, 24]. The community is transmissible and stable over

consecutive mouse generations and unlike ASF offers colonization resistance against

Salmonella enterica serovar Typhimurium albeit not to the degree of a conventional complex

microbiota [20]. However, how the oligoMM$_{12}$ model performs relative to the SPF for

microbiota-associated traits beyond *S.* Typhimurium colonization resistance remains elusive.

In an attempt to standardize preclinical studies in the host microbiome field, we have

developed a simplified mouse microbiota that is representative of SOPF microbiota at the

functional level and derived a standardized gnotobiotic mouse model called GM15, which

phenotypically mimics SOPF mice under standard diet conditions. We show that, under

conditions of chronic physiological stress such as post-weaning malnutrition on a low-protein

diet, a dietary condition triggering stunting, GM15 microbiota seems superior to an SOPF

microbiota to buffer the deleterious effect of a depleted diet on mouse juvenile growth.

Results and Discussion

***In silico* identification of the main bacterial families of the C57BL/6J SOPF fecal**

microbiota

To define a minimal microbiota containing representative and prevalent bacteria from the gut

of C57BL/6J SOPF mice, we analyzed the composition of fecal pellets from 4 C57BL/6J SOPF

mice (two females and two males) by whole-genome sequencing. An average of 13.4 million

of paired-end reads was obtained per sample with a length of 300 bp. The metagenomics data

sets generated were classified using the Centrifuge software [25] and compared to the RefSeq

complete genome database [26], and 20 dominant families consistently present in all mice

were identified (Fig. 1a). The profiling of metagenomic sequencing data pointed out a

comparable distribution of bacterial families among the four tested C57BL/6J SOPF mice.

Moreover, bacterial species identification was possible for genome sequences with good

phylogenetic resolution and already referenced in taxonomy databases. Interestingly, among

the identified species, *Bacteroides acidifaciens, Clostridium cocleatum, Lactobacillus*

johnsonii, Lactobacillus murinus and *Lactobacillus reuteri* were previously identified as mouse-

enriched and dominant intestinal bacteria [17] and *Lactobacillus murinus*, *Parabacteroides*

goldsteinii, Clostridium ASF 356 and *Clostridium* ASF 502 were already part of the ASF model

[19].

Isolation and taxonomic characterization of GM15 bacterial strains

We established four different strategies in order to isolate and culture a maximal number of

representative strains of the 20 dominant bacteria families identified in our metagenomic

sequencing analysis (Fig. 1b). First, we isolated the most prevalent strains from fecal pellets

of C57BL/6J SOPF mice using non-selective agar media. Then, we used antibiotic selection

to isolate resistant strains. We also used rumen enrichment to isolate strains from caecal

content. Finally, we used fecal pellets of ASF mice to isolate additional strains. We obtained a

collection of approximately 400 cultivable bacterial isolates. All isolates were pre-screened by

MALDI-TOF MS for dereplication prior to the first taxonomic identification by 16S rRNA gene

Sanger sequencing. We selected eleven strains covering seven of the most representative and

prevalent families of the intestinal microbiota of C57Bl/6J SOPF mice and obtained four

additional strains from the DSMZ miBC collection to establish the GM15 consortium that covers

most of the dominant bacteria families found in C57Bl/6J SOPF animals (Fig.1a,b). In

summary, GM15 is composed of two strains of Bacteroidaceae, one strain of Tannerellaceae,

six strains of Lachnospiraceae, three strains of Lactobacillaceae, one strain of

Erysipelotrichaceae, one strain of Ruminococcaceae, and one strain of Enterobacteriaceae

(Additional file 1: Figure S1). The genomes of the fifteen strains were sequenced and the draft

genomes sequences aligned against the NCBI database [27], which allowed the identification

of 14 strains at the species level and 1 strain at the family level.

***In silico* functional metagenomics analysis of the GM15 strains**

To gain insights into the functionalities encoded in the individual genomes of the GM15

members, the coding sequences of the 15 strains were converted into their respective protein

sequences, which were annotated for clustering into KO (KEGG Orthology) groups. By

merging the 15 assembled individual genomes, we found that the GM15 metagenome

possesses 3890 non-redundant KO groups. Besides, all GM15 strains possessed 3 to 64

unique functions, although *E. coli* Mt1B1 exhibited a vast repertoire of unique KO groups (10_3),

indicating that *E. coli* Mt1B1 is responsible for the GM15's functional metagenomic profile at

33%, while the 14 other strains contribute all together at 8% and the 59% remaining are

associated to non-unique KO groups (Fig. 1c). It is noteworthy that KEGG module analysis is

biased towards gene sets, pathways and functional groups of well-characterized bacteria such

as *E. coli*, which is by far the most studied bacterial species to date.

Next, we highlighted the *in silico* functionalities of the 15 selected strains associated

with known enzymatic activities in the gut [28] (Fig. 1d). Again *E. coli* Mt1B1 is a major

contributor, but each functionality is also covered by other strains at equivalent or lower levels.

As expected, the enzymatic activities in the gut are correlated with the phylogenetic

membership of the strains. For example, Lactobacilli, which are a major part of the lactic acid

bacteria group, the principal contributor to lactate dehydrogenase [29], clustered together.

Additionally, Lachnospiraceae clustered with Ruminococcaeae and Erysipelotrichaceae,

which include mainly bacteria with sporulation capabilities [30, 31, 32]. Finally, Bacteroides

clustered with Parabacteroides, whose species are predominant in the colonic mucus barrier

and promote mucinase activity [33], and are significant producers of succinate, a major

metabolic by-product [34, 35]. Thus, different strains of a bacteria family and of other closely

related bacteria families are capable of the same enzymatic activities in the gut. This can be

essential for the generation of simplified non-specific gnotobiotic models.

Then, we determined the functional coverage of the GM15 metagenome (i.e. the sum

of the genomes of the 15 strains) relative to the KEGG modules of C57BL/6J SOPF mouse

microbiota found in our initial metagenomic analysis (Fig.1a). In addition, the KEGG modules

from the OligoMM$_{12}$ and ASF microbiota were included for comparative analysis, as these

consortia were previously used to generate gnotobiotic mice with a stable and defined mouse-

derived microbiota [23, 36]. The presence and completeness of KEGG modules was

determined for each metagenome and used for hierarchical clustering (Fig. 1e). One cluster

contained highly conserved modules in all mouse models (Fig. 1e, cluster 4; Additional file 3:

Table S2), including 140 pathways. We also identified clusters of modules that were not

represented among the ASF and OligoMM$_{12}$ consortia, but specifically common to GM15 and

SOPF (Fig. 1e, clusters 9 and 6; Additional file 3: Table S2), comprising a total of 155 pathways

indicating that qualitatively the GM15 metagenome covers functionalities found in SOPF

microbiota that were lacking in ASF and OligoMM$_{12}$ models. Quantitatively, the defined

consortia of GM15, OligoMM$_{12}$ and ASF covered respectively 72, 54 and 48% of the KEGG

modules of the C57BL/6J SOPF microbiome suggesting a superior functional potential of the

GM15 community as compared to OligoMM$_{12}$ and ASF models. Thus, taken collectively our *in*

silico analysis suggests that the GM15 community carries a significant potential for enzymatic

activities in the gut and recapitulates widely the functionalities found in C57BL/6J SOPF murine

metagenome.

Monitoring and stability assessment of GM15 gut microbiota

To explore *in vivo* the functional potential of the GM15 community, we next generated GM15

isobiotic animals. First, we investigated whether the GM15 consortium can stably colonize the

mouse intestine over several generations. To this end, we developed a strain-specific

quantitative polymerase chain reaction (qPCR) microfluidic assay, which allows simultaneous

absolute quantification of the 15 strains, along with the global bacterial load in a given biological

sample (Additional file 2: Table S1). Fecal samples from GF and SOPF mice were used as

negative and positive controls. Only *C. clostridioforme* YL32, which was obtained from the

DSMZ miBC collection, was not detected in our SOPF mice. Co-monitoring of specific and total

bacteria aimed to detect any bacterial load imbalance caused by a contamination in gnotobiotic

isolators.

Eight-week-old GF mice (five breeding pairs; GM15 founders, F0; Additional file 4:

Figure S2) were inoculated by oral gavage with the frozen mixtures of the 15 strains and bred

in sterilized positive pressure isolators up to the F4 filial generation. All strains except

Lachnospiraceae bacterium COE1, *Lachnospiraceae* sp. MD329, and Lachnospiraceae sp.

MD308 were above the detection limit of our qPCR microfluidic assay in the fecal samples of

most individual mice (GM15 founders and progenies from the 4 consecutive generations;

Additional file 5: Figure S3a; Fig. 2a). *Anaerotruncus colihominis* JM4-15, *Clostridium* ASF

356, *Clostridium* ASF 502, and *Clostridium cocleatum* I50 were occasionally below detection

limit. The detection of strains was reproducible between fecal and caecal samples from

individual mice of the second filial generation (Additional file 5: Figure S3b). These results

indicate a stable colonization and effective vertical transmission of at least 12 strains out of the

15 inoculated. Based on the results of our qPCR assay, we consider that the 3 remaining

strains: *Lachnospiraceae* bacterium COE1, *Lachnospiraceae* sp. MD329, and

Lachnospiraceae sp. MD308, either did not efficiently colonize the animals, live in the caecum

or the colon below the detection limit of our qPCR assay, or predominantly live in other

gastrointestinal niches than those sampled.

Next, we evaluated the effect of aging on the GM15 community by following individual

mice of the first filial generation between 2 and 12 months of age (Additional file 4: Figure S2).

Overall, no changes in the qualitative and quantitative composition of the GM15 consortium

were detected (Fig. 2b). Then, we asked if the composition of the GM15 community was

modulated by substituting the breeding diet by an alternative maintenance diet. This alternative

diet is quasi isocaloric but its nutritional composition differs by 1.3 and 1.6-fold fewer proteins

and lipids respectively, and 1.2-fold more carbohydrates than the breeding diet. An alternative

diet was administrated to 8-week-old GM15 mice for 4 weeks. We collected fecal samples

before diet change, after 2 and 4 weeks, and again after 4 weeks back to the breeding diet

(Additional file 4: Figure S2). We did not detect any significant changes in the GM15

composition under these conditions (Fig. 2c). Additionally, successful fecal microbiota

transplantation (FMT) from GM15 founders to GF mice was confirmed by strain-specific qPCR

of the ex-GF mice' feces collected at week 1, 2 and 3 post-transplantation (Additional file 4:

Figure S2a; Fig. 2d).

When all the data are analyzed together (Fig. 2e), we notice a limited fluctuation

(maximum 1Log10-fold change) of each member of the GM15 gut microbiota's load among

the conditions tested (Fig. 2e). Therefore, we conclude that the GM15 community is stable

upon adult colonization, among filial generations, during aging, upon mild dietary fluctuations

and can be transmitted efficiently by FMT.

The GM15 microbial community recapitulates SOPF macroscopic phenotype

GF and published gnotobiotic mice display anatomical alterations compared to SOPF mice,

such as enlarged caecum, along with physiological and metabolic differences [20, 37]. To

phenotypically assess the gnotobiotic GM15 model, we designed a comparative study to

evaluate the steady state macroscopic, immune, metabolic and endocrine phenotypes of GF,

SOPF and GM15 mice. In addition, all phenotyping was achieved across two generations, F1

and F2, to strengthen data analysis.

Initially, we evaluated the reproduction performance of the GM15 model by recording

the period from mating to offspring delivery (Fig. 3a), the number of pups per litter (Fig. 3b),

and the perinatal mortality (Fig. 3c). GM15 mice behaved like SOPF mice with the exception

of one less progeny per mean litter. Then, we quantified the food intake relative to body weight

after weaning at 4 weeks of age and observed no significant difference between the three

groups despite a marked increased variation among GF animals which is not detected in GM15

and SOPF animals (Fig. 3d). Next, we studied post-natal growth parameters. Male and female

GM15 animals gained weight (Fig. 3e) and size (Fig. 3f) like SOPF mice, although the growth

curves from the two sexes differ. We then studied internal organ size. As expected, the

characteristic caecum enlargement seen in GF animals was reduced in GM15 mice (Fig. 3g).

The weights of GM15 and SOPF brain, liver and spleen were equivalent and larger than those

of GF mice (Fig. 3g). Bone size was also identical in GM15 and SOPF mice and larger than

that of GF mice (Fig. 3h). Taken together, these results confirm that the gut microbiota is

associated with somatic tissue growth [38, 39], and that the GM15 simplified microbiota is

sufficient to largely recapitulate the breeding and growth performance of SOPF mice by

compensating the physiological limitations of GF mice.

The GM15 simplified microbiota partially restores SOPF immune phenotype

It is now well established that host-specific bacteria consortia influence intestinal and systemic

immune maturation [40, 41]. We thus profiled the basal immune parameters of GM15 animals

and compared them to SOPF and GF animals at 7-8 weeks of age. We observed that the

GM15 community induces the production of IgA in the gut and in serum at levels equivalent to

those detected in SOPF mice (Fig. 4a). Notably, the highest levels of circulating IgA were not

correlated with the highest levels of fecal IgA, and intragroup variability was not related to gut

microbiota composition, which was homogeneous between individuals. Additionally, IgG2b

serum levels were equivalent in GM15 and SOPF mice compared to GF mice, IgG3 levels

were increased in GM15 mice compared to GF and SOPF mice, and IgM levels were slightly

increased in GM15 mice compared to SOPF (Fig. 4a), which was consistent with published

comparative analyses carried out with SPF mice even though IgG3 and IgM levels were lower

in our SOPF mice [42, 43]. On the other part, no difference of circulating IgG1 was observed

between GF, GM15 and SOPF, and all IgG2a levels were below levels of detection (data not

shown). Surprisingly, GM15 mice presented elevated levels of IgE in serum compared to GF

and SOPF mice (Fig. 4a). High IgE levels may result from parasitic infections, or

immunodeficiencies and long-lived IgE producing plasma cells generated by food antigens in

mice with low-diversity microbiota during early life [42, 44, 45]. Independent tests following

FELASA guidelines (CR EU RADS, France) rejected the infection hypothesis of the GM15

mice, which were negative for ectoparasites and endoparasites, respiratory and intestinal

specific pathogenic bacteria and infectious agents, as well as viruses. It is also known that

Clostridia-abundant juvenile mice have lower IgE levels [46]. Of note, our GF mice do not show

increased IgE levels as SOPF, and GM15 mice harbor a similar load of Clostridia as SOPF

despite elevated levels of IgE. Therefore, we speculate that specific members of the GM15

consortium are inducers of the observed high IgE titer.

Next, we assayed the circulating levels of key intestinal cytokines IL-17a and IL-22,

whose production is induced by commensal bacterial via TLR2 activation in the gut mucosa

[47], and linked to local inflammation [48, 49] or associated with anti-microbial responses and

control of bacterial colonization [50, 51]. GM15 microbiota induced release of both cytokines

compared to GF mice (Fig. 4b) and in a similar manner to SOPF mice for IL-22. To note, serum

IL-17a concentrations in our SOPF mice detected by Luminex were slightly lower than those

published for SOPF animals in previous studies using Luminex detection, but do not explain

the difference observed between GM15 and SOPF mice levels [52]. We also measured IL-1b,

IL-10, IL-12p70, and IL-17f, which are cytokines involved in autoimmune and inflammatory

bowel diseases, but we did not detect any difference between the three groups of mice.

In parallel, we assessed viable leukocytes by measuring CD45+ cell counts and

analyzed T cells, B cells, NK cells and monocytes or dendritic cells (DC) populations in whole

blood and several lymphoid organs (spleen, thymus, Peyer's patches (PP) and mesenteric

lymph nodes (MLN)). It has been previously shown that isolated lymphoid structures such as

PP and associated cellularity were strongly increased with microbiota diversity whereas the

total cell numbers in MLN were comparable between GF and SPF mice [53]. As described in

literature, the number of PP collected where higher in SOPF mice compared to GF and partially

restored in GM15 mice (data not shown), and no or minor difference were observed between

the 3 groups for CD45+ cells count that showed slight increased levels in SOPF compared to

GF mice with apparent intermediary levels in GM15 mice in spleen, thymus and PP, and no

difference in MLN (Fig. 4c).]. In 2018, Kennedy *et al.* published an overview of the literature

describing the main immune cell populations modifications observed in GF mice in different

organs [54]. In our study, very few immune cell populations seemed to be impacted by flora

composition. No difference between the 3 groups was observed for MLN DC, circulating and

splenic CD4+ and CD8+ T cells and splenic B cells as described in literature (data not shown).

Minor differences were observed for PP CD4+ T cells and splenic DC between GM15 mice

and SOPF but not with GF mice whereas both populations were described to be decreased in

GF mice (Additional file 6: Fig. S4b/e). Only PP NK cells showed similar and increased levels

in GM15 and GF mice compared to SOPF mice (Additional file 6: Fig. S4e). Interestingly, the

great majority of populations impacted showed similar levels in GM15 mice and SOPF

compared to GF mice (Additional file 6: Fig. S4a-e). Circulating monocytes, MLN CD8+ T cells

and splenic NK cells were increased in GM15 and SOPF compared to GF as described in

literature for circulating monocytes and MLN CD8+ T cells and though splenic NK cells were

not described to be modified in GF mice. Thymic CD4+ T cells, circulating and MLN B cells

were decreased in GM15 and SOPF mice compared to GF though circulating B cells were

described to be decreased in GF mice.

Taken collectively, our results indicate that the GM15 community is sufficient to dampen

the characteristic immunodeficiencies of GF animals and restore an immunophenotype close

to that detected in SOPF animals with a peculiar signature of the GM15 model with high IL17a

and IgE levels in sera. To note, no correlation was observed between those two markers in

our study although positive or negative correlations have been both described in literature [55,

56, 57, 58]. Furthermore, as for IgE, IL-17a levels modulations have been described to be

associated to particular flora compositions [42, 59, 60, 61, 62, 63, 64, 65, 66, 67]. Such an

observation paves the way to further investigation to decipher which and how members of the

GM15 community trigger these immunophenotypes.

Low-complexity GM15 microbiota shares more metabolic traits with SOPF than GF mice

The gut microbiota influences multiple host metabolic pathways by providing metabolites to its

host and also shapes inter-organ communication within the body by influencing the production

and activity of endocrine signals [68, 69]. One-dimensional proton nuclear magnetic resonance

spectrometry ($_1$H NMR) was previously applied to investigate how the gut microbiota impacts

host metabolism using mouse models [70, 71], or human cohorts [72]. Using this technology,

we analyzed the metabolic profile of plasma samples from GF, GM15 and SOPF mice and

were able to quantify a total of 51 polar metabolites and 5 non-polar metabolites whose levels

were not significantly affected by sex (Additional file 7: Table S3; Additional file 8: Table S4).

A principal component analysis based on quantified polar metabolites showed sequential

alignment on first principal component (8.8% of total variance) according to microbiota

complexity of GF, GM15 and SOPF samples, independently of F1 and F2 generations (Fig.

5a). The metabolic composition of GF plasma resulted only from the host metabolic activity

and represented a basal phenotype. At the other end, the SOPF mice harboring a very diverse

gut microbiota, exhibited a larger panel of metabolic activities based on host-bacterial and

bacterial-bacterial interactions. In-between, the GM15 low-complexity community contributed

in a lower extent to the metabolic phenotype. We performed the equivalent statistical analysis

using binned NMR spectra instead of quantified metabolites and as expected we obtained the

same results (Additional file 9: Figure S5). Next, we performed a discriminant analysis across

all samples to highlight the specific metabolic signatures of each group. Eight metabolites

emphasized significant variation, although it is noteworthy that the calculated distance placed

GM15 closer to SOPF than to GF (Fig. 5b). As previously reported, GF mice had higher plasma

levels of glycine [73] and reduced plasma acetate concentration [74]. On the contrary, SOPF

mice harboring a diverse microbiota had higher plasma levels of acetate, acetoacetate and

dimethyl sulfone [75]. As for GM15 mice, less methanol and more citrate, ethanol and pi-

methylhistidine were detected. Interestingly, methanol may occur as a result of fermentation

by gut bacteria and can stimulate citric acid fermentation [76, 77]. Thus, the simplified gut

microbiota of GM15 mice may produce less methanol and/or microbially-produced methanol

may be used to form citrate. On the other part, ethanol is produced by some intestinal bacteria,

such as *E. coli* and lactic acid bacteria, which are well represented in the GM15 community

[78]. These specific metabolic signatures may be used in the future as a panel of biomarkers

to identify the GM15 model. Besides, the analysis of the plasma non-polar metabolites

indicated that GM15 colonization was sufficient to reduce free cholesterol and

phosphatidylcholine as observed in SOPF mice (Fig. 5c). The three additional non-polar

metabolites detected were equivalent in all mice (Additional file 8: Table S4).

Finally, we investigated the circulating levels of key metabolic hormones and growth

factors. To allow blood sample mutualization for different analyses, mice were not fasted.

Ghrelin was below limit of detection for all groups, and glucagon, insulin and leptin did not

show any difference between GM15 and GF mice, whereas levels detected in SOPF mice

were similar to those previously published under basal conditions [79, 80, 81] (data not shown).

However, as compared to levels detected in GF animals the GM15 gut microbiota was able to

restore the levels of circulating insulin growth factor 1 (IGF-1) as seen in SOPF animals. IGF-

1 is an essential growth factor promoting systemic and tissue growth [82, 83] and this

observation correlates well with the improved macroscopic growth of GM15 mice compared to

GF animals (Fig. 5d, Fig. 3e-h). Corticosterone levels, which are high in GF and low in SOPF

animals were also normalized by GM15 bacterial colonization [84] (Fig. 5e), indicating that

GM15 community seems as efficient as a complex SOPF microbiota at utilizing host

metabolites and promoting steroidogenesis and growth factor production.

Collectively, our results reveal that GM15 animals stand out from GF mice and

recapitulate many of the SOPF metabolic features even though some differences exist.

GM15 mice are less sensitive to diet-induced stunting than SOPF mice

Previous work by our lab and others [85, 86, 38] has shown that the gut microbiota influences

pathogenesis associated to chronic undernutrition, particularly diet-induced stunting.

Consequently, we sought to study our newly-established gnotobiotic mouse model under a

severe nutritional stress, induced by a macronutrient-depleted diet (containing 4% protein and

2% lipids) expected to trigger stunting. We thus fed male mice a breeding diet (BD) or a

depleted diet (DD) from post-natal day 21 (i.e. the day of weaning) until post-natal day 56.

As shown in Fig. 6a-b and similarly to our previous observation (Fig. 3e-f) GM15 and

SOPF mice grew well on the BD as they show similar body weight and size gains. Next, we

confirmed that the DD triggered an almost full stunting of both the juvenile GM15 and SOPF

mice, characterized by the flattering of their weight and size curves (Fig. 6a-b). However, the

GM15 mice performed slightly better in terms of growth as both their body weight and size

were significantly less impacted than SOPF animals by the DD. While the caecum of GM15

was enlarged (Fig. 6c, Additional file 10: Fig. S6a), this variation could not account for the total

weight difference between GM15 and SOPF animals. We thus compared the sizes and weights

of nine other organs (Fig. 6d-f, Additional file 10: Fig. S6b-g) in order to account for these

variations and using two-way ANOVA, we confirmed that diet was the main driver of the growth

phenotype. However, we found that the increase in body size observed in GM15 mice on the

DD could be correlated to a significant increase in the size of the tibia (Fig. 6d). Although we

did not find any other significant difference in other single organ size or weight between GM15

and SOPF animals on DD, we observed a clear tendency of an increase in GM15 compared

to SOPF animals for most parameters tested (Fig. 6e-f, Additional file 10: Fig. S6b-g). We

tested this tendency by integrating all the phenotypical parameters in a Principal Component

Analysis (PCA). We first established which parameters were correlated to the phenotype

(Additional file 10: Fig. S6h), thus excluding brown adipose tissue from the analysis as a non-

correlating parameter. The PCA revealed that, under BD, the GM15 phenotype was part of the

spectrum of the SOPF phenotype (Fig. 6g). However, under DD, there is a clear shift between

the GM15 and SOPF phenotypical space (Fig 6g, dotted lines). Our results thus indicate that

under nutritional stress, GM15 microbiota buffers diet-induced stunting slightly more effectively

than an SOPF microbiota. As IGF1 is an important driver of the diet and microbiota mediated

growth promotion [38], we assessed IGF1 levels in animals and while the levels massively

drop in the DD conditions, we did not detect any difference between GM15 and SOPF mice at

day 56 on the DD (Fig. 6h). However, we cannot exclude that slight variation in size between

GM15 and SOPF animals on DD results from a differential secretion of IGF1 before day 56,

when the growth rate of the animal is maximal (Fig. 6a-b).

Taken together, our data show that animals bearing the simplified GM15 microbial

community perform similarly than animal bearing a complex SOPF microbiota under non

stressful conditions, however the GM15 microbiota seem partially protective against the

deleterious effects of chronic undernutrition, as compared to SOPF mice.

Conclusions and perspectives

Here we describe GM15, a simplified and controlled murine gut microbiota and its related

GM15 C57BL/6J gnotobiotic mouse model. The GM15 community is composed of pure

cultures of 15 strains from 7 of the 20 most prevalent bacteria family present in the fecal

microbiota of SOPF C57BL/6J mice. GM15 carries significant potential for enzymatic activities

in the gut and recapitulates extensively the functionalities found in C57BL/6J SOPF gut

microbiome. *In vivo* GM15 is stable upon adult colonization for up to 12 months, during natural

transmission among 4 filial generations, upon mild dietary fluctuations and can be transmitted

efficiently by FMT. GM15 compensates the steady state breeding, growth, immune, endocrinal

and metabolic limitations of GF mice and recapitulates many of the SOPF phenotypical

features. Taken together our results establish that GM15 is a novel controlled preclinical model

phenotypically similar to SOPF with the potential to ensure an increased reproducibility and

robustness of preclinical studies by limiting the confounding effect of microbiota composition

fluctuation. Importantly, the reduced microbial complexity of the GM15 community, the

tractability of its members and the control offered to the experimenter on its composition allow

easy quantification of gut microbiota dynamics. Indeed, upon manipulation of the GM15

community composition, the correlation of such dynamics with fluctuating host traits allows the

establishment of causal relationships between specific microbiota members and host traits.

GM15 model offers exciting perspectives for improvement. Indeed, under stressful nutritional

environment, the simplified GM15 microbiota performs slightly better than a complex SOPF

community in terms of physiological host response. We have previously identified a

Lactobacillus strain that is capable of buffering the deleterious effects of such challenge in

monocolonized mice [38]. Interestingly, out of its 15 strains, GM15 microbiota contains 3

Lactobacillus strains. Further genetic manipulations coupled to gnotobiotic studies focusing on

modifying the Lactobacilli components of GM15 will pave the way to understanding how this

minimal bacterial community buffers the host response to chronic undernutrition.

As all experimental models GM15 has its limitations. First, phenotypically, despite being

macroscopically similar, GM15 and SOPF animals differ in specific immune and metabolic

signatures. This is not particularly surprising given the marked reduced microbial diversity that

the GM15 model carries compare to SOPF microbiota (15 strains vs hundreds of species in

SOPF animals). These specific immune and metabolic signatures of the GM15 models pave

the way to further studies aiming at defining if the presence or absence of specific community

members triggers these phenotypes. Another limitation worth mentioning here is that the GM15

model is a standardized model, and such standardization, essential for establishing robust

causal relationships between a microbiota configuration and a host trait, may decrease the

translational potential of the observation.

In addition, microbiota-mediated resistance to *Salmonella* infection may be tested in

GM15 animals and compared to ASF and OligoMM$_{12}$ animals as recently done by Stecher and

colleagues [20]. Second, the GM15 model, by its low complexity, offers the possibility to use it

as a template for establishing further complex or complete consortia, e.g. by complementing it

with representative strains of the missing prevalent bacteria family found in SOPF microbiota

such as Deferribacteraceae, Oscillospiraceae, Clostridiaceae and Eubacteriaceae. Despite

these limitations and room for improvement, our study establishes that the GM15 model offers

new possibilities for preclinical research focusing on host-microbes and microbe-microbe

interactions, and how the microbiota shapes the environmental impact on health and diseases

or drug efficacy.

Methods

Bacterial strains isolation and identification

Fresh caecal contents and fecal pellets of C57BL/6J SOPF mice (Charles River Lab., France) were resuspended (1/10 wt/vol) in reduced broth media for direct dilution plating on agar plates and growth at 37°C under anaerobic atmosphere (90% N_2, 5% H_2, 5% CO_2). *Lactobacillus johnsonii* NCC 533 was isolated on MRS agar. *Lactobacillus murinus* 313 and *Parabacteroides goldsteinii* WAL 12034 were isolated on Colombia CNA agar with 5% sheep blood. *Bacteroides acidifaciens* A40 and *Lachnospiraceae* sp. MD308 were isolated on GAM agar. *Bacteroides caecimuris* I48 and *Lactobacillus reuteri* I49 were isolated on GAM agar supplemented respectively with 32 µg/mL vancomycin and 32 µg/mL erythromycin. *Lachnospiraceae* bacterium COE1 and *Lachnospiraceae* sp. MD329 were isolated on M2GSC agar (modified Hobson, containing (per 100 mL) 1 g of casitone, 0.25 g of yeast extract, 0.4 g of NaHCO3, 0.2 g of glucose, 0.2 g of cellobiose, 0.2 g of soluble starch, 30 mL of clarified rumen fluid, 0.1 g of cysteine, 0.045 g of K2HPO4, 0.045 g of KH2PO4, 0.09 g of (NH4)2SO4, 0.09 g of NaCl, 0.009 g of MgSO4.7H2O, 0.009 g of CaCl2, 0.1 mg of resazurin, and 1.5 g agar). *Lachnospiraceae* bacterium COE1, *Lachnospiraceae* sp. MD329, and *Lachnospiraceae* sp. MD308 were isolated from caecal contents, the rest from fecal pellets. Fecal pellets of ASF mice (Taconic, USA) were cryopreserved at -80°C, and then resuspended in reduced broth media for direct FMT in GF mice. Fresh caecal content and fecal pellets were resuspended in reduced broth media for direct dilution plating on agar plates and growth at 37°C under anaerobic atmosphere (90% N_2, 5% H_2, 5% CO_2). *Clostridium* ASF356 and *Clostridium* ASF502 were isolated on M2GSC agar, respectively from caecal content and fecal pellets. For identification of isolates, colonies were pre-screened for dereplication by MALDI-TOF MS (Vitek MS, Biomérieux) according to the manufacturer's instructions and database enrichment using RUO mode. Then, gDNA was extracted from pure cultures and analyzed by 16S rRNA gene sequencing (GATC Biotech). Following Edgar's recommendation [87], a full-length 16S rRNA sequence similarity ≥99% using either NCBI blast [27], Ribosomal Database Project [88], or EzTaxon [89] programs allowed the identification of 13 isolates at the species level, and

isolates MD329 and MD308 could only be assigned to the *Lachnospiraceae* family. A more

precise annotation could be given for the 2 isolates *Clostridium* ASF356 and *Clostridium*

ASF502 obtained from the defined ASF microbial consortium. *Anaerotruncus colihominis* JM4-

15 (DSM-28734), *Clostridium clostridioforme* YL32 (DSM-26114), *Clostridium cocleatum* I50

(DSM-1551), *Escherichia coli* Mt1B1 (DSM-28618) were obtained from DSMZ.

Culture conditions

Freshly grown cultures of individual bacterial strains were supplemented with 20% glycerol

prior cryopreservation at -80°C. Each culture was systematically validated for culture purity

and identity by MALDI-TOF. Culture media and material were introduced in the anaerobic

chamber at least 2 days before use. Anaerobic bacterial strains were grown in GAM, except

Clostridium ASF502, *Lachnospiraceae* sp. MD308, *Lachnospiraceae bacterium* COE1 and

Lachnospiraceae sp. MD329 in M2GSC, and *A. colihominis* JM4-15 in Bifidobacterium

medium. For mouse colonization and absolute quantification of bacteria, a single colony of

each of the 15 bacterial strains was grown and amplified at 37°C. Each bacterial pellet was

resuspended in medium, 1 mL was cryopreserved with 10% glycerol, 1 mL was centrifuged

and bacterial pellet was stored at -20°C for gDNA extraction, and the rest was used for

numeration by dilution plating on agar plates. A frozen mixture of the GM15 bacterial

community containing the 15 individual strains at equivalent concentration (6.67E+06 CFU)

was prepared to enable easy inoculation.

WGS

DNA samples from the 15 bacterial cultures were prepared for whole-genome sequencing,

using the Nextera XT DNA library preparation kit (Illumina, Cat.Nos. FC-XXX, California, USA)

according to manufacturer's recommendations. The resulting libraries were checked for their

quality using the High-sensitivity DNA chip using the Agilent 2100 Bioanalyzer (Waldbroon,

Germany) and quantified using the QuantiFluor One dsDNA kit (Promega). Paired-end

(2x300bp) sequencing was performed on a MiSeq sequencer using the "MiSeq v3 kit (600

cycles)" (Illumina, California, USA).

De novo **genome assembly**

After a quality control with FastQC [90], the paired-end reads were assembled *de novo* using

the "A5-miseq" assembly pipeline [91], comprising the following steps: adapter trimming,

quality trimming and filtering, error correction, contigging and scaffolding. The 15 *de novo*

assemblies resulted in draft genomes composed of few scaffolds (from 30 to 268) with high

N50 values (from 13 099 to 943 892). Genomes were then ordered using Mauve [92] and

annotated with PGAP of the NCBI database. Default parameters were used for all software tools.

Taxonomic annotation

WGS generated data have been trimmed and quality controlled by AfterQC software [93]. A k-

mer counting strategy with the Centrifuge software [25] have been privileged to infer taxonomy,

and reads were confronted to the RefSeq complete genome database [26], with bacteria,

archaea and viruses domains and the mouse representative genome (taxid 10090), in order

to estimate the amount of host DNA contamination and remove it from sequenced data.

Genomic functional analysis

Genes were predicted and turned into protein sequences using Prodigal [94]. Marker genes

were searched using the HMM3 package [95]. Predicted proteins sequences of genomes were

blasted against the KEGG microbial database [96] which contains 13 millions of proteins

sequences and trimmed with following parameters: best-hit with an expected value threshold

< 10_{-5}. The matrix obtained was consolidated into KEGG orthologs count (KO, which

represents a set of genes the have sequence similarities and do the same function), into KEGG

modules (which represent short enzymatic pathways, involving few proteins and doing a

targeted function), and into KEGG pathways (i.e. large metabolic pathways). KO were

analyzed for their presence absence among genomes. The modules were analyzed for their

completion (4 levels: Full, lack 1 enzymes, lack 2 enzymes, absent), and only modules with a

score with 3 or 4 were presented and integrated for their KO count. The KEGG pathways were

analyzed for their number of related KO count affiliated to them. A list of functions of interest

have been designed and their presence among genomes have been analyzed in details

(Additional file 3: Table S2). Because pathways and functions of interest did not have the same

number of KO of interest and a different distribution among the genomes, functional data were

normalized among each function in order to obtain values that can be comparable. For each

function/pathway, the number of different KO was normalized by the total number of KO

retrieved. Data were then log-transformed +1. Clusterization of both functions and communities

was performed using Euclidian distance and ward's method, and a kmeans clustering was

performed in order to define the community clusters.

Identification of specific regions for primers design

NUCmer, a part of the MUMmer package [97], was used to perform pairwise alignment of the

15 genomes. Based on these alignments, PanCake [98] was used with default parameters to

identify specific regions of each genome. Specific regions with a length of 200 bp were

extracted, meeting the following criteria: GC content between 48 and 52%, distance to a border

of a scaffold higher than 300bp, unique in the draft genome. The specificity of each 200bp

region was double-checked with BLAST [27] on the 15-genome database and on the NCBI *nr*

database. The design of primers on the specific regions was performed by Fluidigm. The

primer specificity was checked with BLAST.

Animal experiment

All mice were bred according standardized procedures in the gnotobiology unit of BIOASTER

at the ANSES animal facility (Lyon, France), housed in sterilized positive pressure isolators

(Noroit) under a 12h light/dark cycle at 22±2°C and 50±30% of humidity, and fed ad libitum

with irradiated R03-40 diet (3395 kcal/kg, 25.2% kcal proteins, 61.3% kcal carbohydrates,

13.5% kcal lipids; Safe) and autoclaved water. Irradiated corn-cob granules (Safe) were used

as bedding. Sterile enrichment was provided in all cages and was constituted by cotton rolls

as nesting material, poplar bricks and a polycarbonate red mouse igloo (Safe). Nesting material

and poplar bricks were renewed every two weeks. All breeders were mated by trio (2 females

and 1 male) between 8 weeks and 6 months of age, and all mice were weaned at 4 weeks

after birth. C57BL/6J GF mice were obtained by aseptic hysterectomy of a C57BL/6J SOPF

female (Charles River Lab.), and neonates were fostered on C3H GF mothers (CDTA). Axenic

status was assessed weekly by gram staining and cultures of fecal suspension on solid and

liquid media. GM15 founders were 8-week-old C57BL/6J GF mice colonized by oral

inoculations of 215 µL of the GM15 bacterial community, twice at 48h interval. GM15

microbiota composition was analyzed by qPCR microfluidic assay from feces collected at 6-

week-old. Alternative diet R04-40 (3339 kcal/kg, 19.3% kcal proteins, 72.4% kcal

carbohydrates, 8.4% kcal lipids; Safe) was given at 8-week-old GM15 mice for 4 weeks. FMT

was done by inoculating orally a suspension of fresh fecal pellets to 7-week-old C57BL/6J GF

mice, twice at 48h interval. For undernutrition experiments, GM15 and SOPF mice were bred

and randomly assigned to be given either the above R03-40 diet or a custom-made low-protein

diet (3500 kcal/kg, 4.7% kcal proteins, 90.1% kcal carbohydrates, 5.3% kcal lipids, Envigo) for

5 weeks after weaning. Mice were killed by cervical dislocation and biocollections were

performed aseptically. All animal procedures were approved by ANSES/ENVA/UPEC ethic

committee   (APAFIS#4529-2016022616404045v3;   APAFIS#785-2015042819315178v2;

APAFIS#18918-2019020118003843v3) and were conducted in accordance with the National

and European legislation on protection of animals used for scientific purposes.

gDNA extraction from caecal contents and fecal pellets

Caecal and fecal gDNA were extracted using the DNeasy® PowerLyzer® PowerSoil® Kit

(Qiagen) following the manufacturer's instructions with modifications. Samples (approximately

0.1g) were heat-treated at 65°C for 10 min, and 95°C for 10 min, before a double bead beating

at 30 Hz for 5 min. 50µL of DNA were obtained with two consecutive elutions.

Quantitative PCR microfluidic assay

In order to quantify specific and global bacteria load per g of caecal or fecal samples (wet

weight), qPCR microfluidic assay was performed using respectively specific primers for GM15

and "universal" primers amplifying the genes encoding 16S rRNA from most bacteria groups

[99]. Amplicons generated using these primers range between 60 and 99 base pairs. qPCR

microfluidic assays were conducted in 48.48 Dynamic Array™ IFCs for EvaGreen Fast Gene

Expression on a Biomark HD (Fluidigm) according to the manufacturer's instructions, with

cDNA diluted 100-fold, preamplified with pooled primers, and diluted again 100-fold. Each IFC

included triplicate reactions per DNA sample, standards and negative control. Standards were

generated by serial dilution of a gDNA extract from pure bacterial cultures of known

concentration. The efficiency of each qPCR reaction was calculated based on the slope of

standard curves and within the range of 78-107%. An equivolume mixture of standards was

used to normalize data between runs. Due to the different individual detection limits of the

qPCR assay for each primer, the detection limit for GM15 ranged between $2.73 \times 10_2$ and 6.57

$\times 10_5$ CFU/g (Additional file 2: Table S1).

Sample preparation for immunophenotyping

For flow cytometry analyses whole blood was collected on EDTA tube. Spleen, thymus, MLN

and PP were collected in RPMI (Gibco). Single-cell suspensions were achieved using a 100

μm cell-stainer (Becton Dickinson) and a 5 mL syringe plunger in RPMI supplemented with 2%

heat-inactivated fetal bovine serum (Sigma) and 100 μg/mL DNASE1 (Roche). Cells were then

spun at 400xg for 5 min at room temperature. Medium was discarded and cells were washed

using 5 mL of supplemented medium. For whole blood, spleen and thymus samples, red blood

cells were lysed by resuspension in 1 mL PharmLyse 1X (Becton Dickinson) for 10 min. Cells

were then spun at 400xg for 5 min at room temperature. Lysing solution was discarded and

cells were washed using 2 mL of PBS (Gibco). Cells were pelleted a second time and

resuspended in PBS supplemented with 2% heat-inactivated fetal bovine serum (Sigma).

Numeration and viability were determined using Propidium Iodide marker exclusion and

MACSQUANT Flow cytometer (Miltenyi). Cells were then resuspended to a working

concentration of 10_6 cells / tube for organs and 100 µL / tube for whole blood, and analyzed by

flow cytometry.

Whole blood was collected on dry Eppendorf tube for sera analysis. Sera were obtained by

centrifugation 2000xg for 15 min at 4°C and stored at -20°C before Luminex and ELISA

analyses. Feces were collected in Eppendorf low-binding tubes and stored at -80°C before

Luminex analysis. Feces supernatant was obtained by disrupting 100 mg feces in 1mL PBS-

Protease Inhibitor 1X (Sigma) using Lysing Matrix E Tube (MP Biomedicals) and Fast Prep

homogenizer (MP Biomedicals). Samples were spun at 8000xg for 15 min at 4°C and

supernatants were collected for IgA Luminex analysis.

Flow Cytometry

10_6 cells or totality of cells for some of PP samples were stained for surface markers.

Leukocytes were stained using anti-CD45 Viogreen (Miltenyi, clone30F11) and were checked

for viability using Zombie Green Fixable Viability (Biolegend). T cells were stained using anti-

CD3 BV421 (BD, clone 145-2C11), anti-CD4 PE (Miltenyi, clone REA604) and anti-CD8a PE

Vio615 (Miltenyi, clone REA601). B cells were stained using anti-CD45 RB220 PE Vio770

(Miltenyi, clone RA3-6B2). NK cells were stained using anti-CD335 APC (NKp46) (Miltenyi,

clone 29A1.4.9). Monocytes were stained using anti-CD11b APC Vio770 (Miltenyi, clone

REA592) in spleen, thymus and whole blood samples. Dendritic cells were stained using anti-

CD11c APC Vio770 (Miltenyi, clone REA754) in spleen, PP and MLN samples. Cells were

analyzed using a MACSQuant Ten Flow cytometer (Miltenyi) and raw data were analyzed

using FlowJo software (Tree Star, Becton Dickinson). For CD45+ cell count comparison in

PP, results were normalized to 10_6 cells using number of cells stained for each mouse when

less than 10_6 viable cells were isolated from PP. Data normalization was not possible for whole

blood sample, thus CD45+ cell count comparison has not been performed for this

compartment. For frequency results, data are represented as a percentage of CD45+ cells for

all organs.

Metabolic Panel, Ig and Cytokines Luminex

Serum concentrations of Metabolic Panel (Ghrelin, Glucagon, Insulin and Leptin) were
determined using the Mouse Metabolic Magnetic Bead Panel Milliplex MAP kit (Millipore).
Samples were not diluted, and assay was performed according to the manufacturer's
instructions. Serum concentrations of Ig Panel (IgA, IgG1, IgG2a, IgG2b; IgG3, IgM, IgE) and
IgA in feces supernatant were determined using the Mouse Immunoglobulin Isotyping
Magnetic Bead Panel Milliplex MAP kit (Millipore) and Mouse IgE Single Plex Magnetic Bead
Milliplex MAP kit (Millipore). Samples were diluted 1:12500 (IgA, IgG1, IgG2a, IgG2b; IgG3,
IgM), 1:100 (IgE), and 1:100 (IgA in feces supernatant). Assays were performed according to
the manufacturer's instructions. Serum concentrations of Cytokines Panel (IL-1b, IL-10,
IL12p70, IL-17a, IL-17f, IL-22) were determined using the Mouse Th17 Magnetic Bead Panel
Milliplex MAP kit (Millipore). Samples were no diluted and assay was performed according to
the manufacturer' instructions. Samples were assayed in monoplicate.

Corticosterone and IGF-1 ELISAs

Serum concentrations of Corticosterone were determined using the Corticosterone ELISA kit
(Abnova). Samples were diluted 1:50, and assay was performed according to the
manufacturer' instructions. Serum concentrations of IGF-1 were determined using the
Mouse/Rat IGF-1 Quantikine ELISA kit (R&D Systems). Samples were diluted 1:500, and
assay was performed according to the manufacturer' instructions. Samples were assayed in
monoplicate.

Sample preparation for metabophenotyping

It is known that the quantification of some plasma metabolites, such as tryptophan and
tyrosine, is biased since they bind to albumin, a highly abundant protein in plasma [100]. In
addition, protein precipitation methods with organic solvents can induce loss of volatile
metabolites and overlay of residual broad resonances of lipids with some polar metabolites

[101]. Thus, the polar and non-polar metabolites extraction from the same plasma sample were

prepared with some in-house modifications to previously described sequential approaches

[102, 103]. Briefly, we deproteinized the plasma samples by acidified ultrafiltration in order to

increase desorption yields of aromatic amino acids and then, quantify the polar metabolites.

We also extracted the filter residue of the same sample to analyze the non-polar metabolites.

Polar metabolites preparation

Frozen mice plasma samples are thawed in thermoshaker Eppendorf (10 min, 10°C and 1000

rpm). The entire amount of plasma about 180 µL is filtered using 0.2 µm centrifugal tube from

VWR (5 min, 10°C, 10000xg). Next, the lipoproteins removal was performed by mixing 150 µL

of filtered plasma with 50 µL milli-Q water and 10 µL deuterated formic acid (2.1% in milli-Q

water) on a clean 10 kDa cut-off ultracentrifugation tube (VWR) using thermoshaker Eppendorf

(10 min, 10°C, 750 rpm) and then centrifuged (30 min, 10°C, 10000xg, soft ramp). Vortexed

10 kDa-filtred plasma samples (135 µL) were transferred in 0.5 mL 96-wellplate Agilent and

mixed with phosphate buffer solution (45 µL, 0.6 M, pH=7.4) containing internal standard DSS-

d4 (1.54 mM), pH-reference standard DFTMP (4 mM) and D_2O (40%) on thermoshaker

Eppendorf (1.5 min, 10°C, 650 rpm). Finally, the resulting sample solutions (155 µL) were

transferred in 3 mm SampleJet NMR tubes. The DSS-d4 concentration was calibrated by $_1H$

NMR using sodium succinate dibasic hexahydrate standard solution. This protocol was

systematically applied to prepare blank and quality control samples. The concentration of the

formic acid was optimized to allow desorption of the metabolites like tryptophan from plasma

albumin. Deuterated formic acid instead of protonated form was used to decrease the

exogenous NMR signal and allow the quantification of the endogenous formiate. All

ultracentrifugation filters are previously recovered from tubes and washed separately from

residual glycerol in a 250 mL Duran bottle, plunged in milli-Q water and then sonicated in a

bath for 10 min. This washing procedure is repeated five times. After the last step of the

washing the ultracentrifugation filters could be stored in water for at least 3 months. Just before

use traces of milli-Q water from ultracentrifugation filters are dispersed outside with nitrogen

stream. We observed that this procedure of washing removes better the residual glycerol than

supplier's protocol and gives higher yields of ultrafiltred plasma.

Non-polar metabolites extraction after ultrafiltration

The lipoproteins on the 10 kDa filters was further diluted with 150 µL phosphate buffer solution

(1 M, pH=7.4) and mixed on thermoshaker Eppendorf (10 min, 10°C, 750 rpm). The

lipoproteins samples were then transferred in clean Eppendorf tubes and extracted with 400

µL methanol-dichlormethan (1:2 v/v). Samples were centrifuged (5 min, 10°C, 10000xg) for

better phase separation. The dichlormethan layer was transferred in clean Eppendorf tube and

the aqueous phase was extracted again with dichlormethan. The pulled organic phase was

evaporated under nitrogen stream. The dry lipidic residue was dissolved with 200 µL

deuterated chloroform containing 0.03% TMS internal standard and 155 µL of resulting solution

was transferred in 3 mm SampleJet NMR tube. The TMS concentration was calibrated by NMR

using 1,3,5-tritertbutylbenzen standard solution and it was found to be 0.435 mM.

Metabolites analysis and quantification

The 1D ^{1}H NMR spectra were acquired at 298 K with 600 MHz Ascend (Avance III HD)

spectrometer from Bruker equipped with a 5 mm QCI cryoprobe. All samples were stored at

6°C in the SampleJet autosampler. Polar metabolites were analyzed using *noesygppr1d* pulse

sequence. For each spectrum, 128 scans were collected into 32k data points within 14 ppm

spectroscopic width and a recycling delay of 4 s. The mixing time was calibrated to 50 ms and

the acquisition time was 3.9 s. The non-polar metabolites NMR spectra were acquired using

zg30 pulse sequence. The spectra were recording using 256 scans, into 32k data point and a

spectroscopic width of 20 ppm. The relaxation delay was 4 s. The FIDs were zero-filled to 64k

points and Fourier transformed using 0.3 Hz exponential line-broadening function. All spectra

were aligned to DSS-d4 and TMS respectively internal standard. The concentrations of the

polar and non-polar metabolites were quantified using Chenomx NMR suite 8.1. The Chenomx

software was applied also for spectra binning of 10_{-3} ppm width for each bin. The triacyl glycerol

(TAG), phosphatidylcholine (PC), lysophosphatidylcholine (LysoPC), sphingomyelin (SM), free

cholesterol (FC) and cholesterol ester (EC) quantification an in-house lipid database within

Chenomx was built with authentic lipid standards. The lipids database spectra were recorded

using the same parameters as described above for non-polar metabolites.

Metabolomics data analysis

Discriminant analyses were performed using the partial least square (PLS-DA) algorithm to

extract metabolomics signatures [104, 105]. A variable selection algorithm based on Elastic-

Net was used to improve model performance by selecting most significant metabolic

signatures that explain the groups (GF, GM15, SOPF). The statistical performances of the

regression models were assessed using the balanced error rate with and without cross-

validation (E2 and CV-E2) and permutation tests. Permutation tests consisted in building the

null distribution of the balanced error rate E2 by randomly permuting observations. Regression

models were thus challenged by testing if the cross-validation error rate CV-E2 is significantly

different from the null distribution with a p-value < 0.05. Metabolites involved in the cross-

validated signatures were ranked by order of importance in the PLS-DA model using their VIP

(variable importance in projection) scores. The biplot allows projecting onto the two first

components, the samples and the metabolites that significantly discriminate each sample

groups. Metabolites that are positively correlated (or positively contribute) to a sample group

will point to the direction of this group. They will point to the reverse direction for a negative

correlation.

Statistics

Reproductive performance and body growth were analyzed respectively by 1-way and 2-way

ANOVA. Phenotyping data impacted by age, filial generation or sex, were analyzed by the F-

test for multiple linear regressions (R v3.4.2), otherwise by 1-way ANOVA and Tukey's multiple

comparison parametric test or Kruskal-Wallis and Dunn's multiple comparison non-parametric

test after D'Agostino et Pearson test for data set normality assessment (GraphPad Prism v7

and v8). PCA for phenotyping analysis was performed using R v3.4.2 and the ade4 package

[106].

Additional files

Additional file 1: Figure S1. Morphological features of the GM15 strains. (TIFF 2312 kb)

Additional file 2: Table S1. GM15 strains-specific primers. (XLSX 11 kb)

Additional file 3: Table S2. KEGG clusters. (XLSX 142 kb)

Additional file 4: Figure S2. *In vivo* experimental design. (TIFF 591 kb)

Additional file 5: Figure S3. Assessment of gut microbiota stability of GM15 founders, and of

reproducibility between fecal and caecal samples of individual mice. (TIFF 799 kb)

Additional file 6: Figure S4. Immune cell populations profiling. (TIFF 471 kb)

Additional file 7: Table S3. Polar metabolites. (XLSX 18 kb)

Additional file 8: Table S4. Non-polar metabolites. (XLSX 16 kb)

Additional file 9: Figure S5. PCA based on binned NMR spectra. (TIFF 1374 kb)

Additional file 10: Figure S6. Measurements of several organs after nutritional challenge.

(TIFF 1011 kb)

Abbreviations

ASF: Altered Schaedler flora; BD: Breeding diet; bp: Base pair; CFU: colony forming units;

CNA: Colistin and naladixic acid; DD: Depleted diet; FMT: Fecal microbiota transplantation;

GAM: Gifu anaerobic medium; gDNA: Genomic deoxyribonucleic acid; GF: Germ-free; IFC:

Integrated fluidic circuit; KO: KEGG orthology; MLN: mesenteric lymph nodes; MRS: Man

Rogosa Sharpe medium; OTU: Operational taxonomic unit; PCA: Principal coordinate

analysis; PP: Peyer's patches; NMR: Nuclear magnetic resonance; qPCR: Quantitative

polymerase chain reaction; SEM: Standard error of the mean; SOPF: Specific opportunistic

pathogen-free; WGS: Whole genome sequencing

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Availability of data and materials

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VIRD00000000, VIRC00000000, VIRB00000000, VIRA00000000, VIQZ00000000,

VIQY00000000, VIQX00000000, VIQW00000000, VIQV00000000, VIQU00000000,

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Author's contributions

MD, FL and AT conceived the project, analyzed, interpreted and integrated all the data. MD

and FL wrote the manuscript. MD prepared figures. LB and JY achieved *in silico* design. MD

performed bacteria isolation and identification, gDNA extraction, developed the qPCR

microfluidic assay and analyzed macroscopic phenotyping data. AS carried out WGS, CE

identified specific regions for primers design, and AM performed taxonomic annotation and

functional bioinformatic analyses. PB, HD and MD performed animal experiments. CC, AD, JC

and SP performed the immunophenotyping, analyzed and interpreted data. AB and DAO

performed the metabophenotyping, analyzed and interpreted data. MS helped design the diet

induced stunting protocol. FDV performed, analyzed and contributed to the text and figures

related to the stunting model experiments. ALB provided technical support during the stunting

model experiments. All authors read and approved the final manuscript.

Ethics approval

Mouse experiments were performed as approved under the MESR and ANSES/ENVA/UPEC

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Consent for publication

Not applicable.

Competing interests

The authors declare that they have no competing interests.

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Legends

Fig. 1. Selection, isolation and functional analysis of GM15 strains. a Composition of

representative and prevalent bacterial families identified from fecal samples of 4 C57BL/6J

mice of our colony. **b** 11 bacterial strains were isolated from fecal or caecal samples of either

C57BL/6J or ASF mice using various culture methods. 4 additional strains were obtained from

the DSMZ collection. **c** Unique KO (KEGG Orthology) groups are those present in only one

GM15 member. The unique contribution of each GM15 member is relatively evenly distributed

(<2%), except *E. coli* whose unique functions contribute for 33%, although non-unique KO

groups are primarily responsible for the GM15's metagenomic profile with 59% of contribution.

d Heat map of hierarchical clustering of enzymatic activities in the gut for the 15 strains of the

GM15 model. In quorum sensing (QS) function, only effector proteins were screened (because

receptors suffer of similarities with nonspecific QS receptors). **e** Heat map of hierarchical

clustering of KEGG modules distribution in the metagenomes of GM15, SOPF, ASF and Oligo-

MM12 models. Clusters of KEGG modules are highlighted with their respective size (number of

KEGG modules). The list of KEGG modules and clusters is shown in Additional file 3: Table

S2.

1181

Fig. 2. Stability assessments of the GM15 mice gut microbiota over filial generations, in

aging, under diet change and through FMT. SOPF groups show the distribution of each

GM15 strains in the complex gut microbiota of 8-week-old SOPF mice. The absolute

quantification of each strain was determined by specific qPCR microfluidic assay. To gain

clarity, each GM15 strain is associated to a number between 1 and 15. * Strains 4, 6, 7 and 14

were at the detection limit of the qPCR microfluidic assay, and thus were not detected in all

samples. ** Strains 8, 9 and 10 were below detection limit of the qPCR microfluidic assay.

Strain 5, obtained from the DSMZ collection, was not detected in our SOPF colony. **a** Radar

plot showing the GM15 strains distribution in feces of GF C57Bl/6J mice colonized with the

GM15 community (F0, n=10) and bred for consecutive generations (F1 to F4, n=19 to 8). **b**

Radar plot showing the overall stability of the GM15 community composition in feces collected

from 9 mice between 2 and 12 months of age (2 mice died at 12-month-old). **c** Radar plot

showing that an alternative diet, such as a maintenance diet, can be used for 4 weeks, and

then reversed to the breeding diet for 4 more weeks, without modifying the composition of the

gut microbiota of GM15 mice (n=18) compared to mice fed all along with the breeding diet

(n=9). **d** Radar plot showing the feasibility of GM15 fecal microbiota transplantation to GF mice

(n=9). **e** Box plots showing the low variability of the GM15 strains concentrations considering

all fecal samples of mice from generation F1 to F4 (n=47), from generation F1 at 12-month-old

(n=7), from generation F1 under diet change (n=18), and fecal samples collected 3 weeks post

fecal microbiota transplantation in GF mice. Whiskers represent min and max data.

1202

Fig. 3. Macroscopic phenotyping on mice from two consecutive filial generations (F1-

F2). a-c 1-way ANOVA analyses of the reproductive performance, where dots, lines and error

bars represent respectively individual litters (34 GF, 31 GM15 and 19 SOPF), means and SEM.

a GM15, like SOPF, delivered pups around 24 days after mating, against 33 days for GF mice.

b GM15 mean litter of 5 pups was in between GF (4 pups) and SOPF (6 pups). **c** Perinatal

mortality was similar for GM15 and SOPF (12 and 11%), and much lower than for GF mice

(45%). **d** Box plots showing no difference between GF, GM15 and SOPF mice in terms of feed

intake, although much variability was observed with GF mice. Whiskers represent min and max

data. **e-f** 2-way ANOVA analyses of body growth of mice bred with their mothers until week 4,

where lines and error bars represent respectively means and SEM. Body weight and size

curves of GM15 mice (9 females (F), 12 males (M)) were similar or equivalent to those of SOPF

mice (10 F, 10 M), whereas GF mice showed a significant delayed growth (7 F, 10 M). **g-h**

Organs weight or size impacted by age, filial generation or sex, were analyzed by the F-test

for multiple linear regressions, otherwise by 1-way ANOVA. Each animal is represented by a

dot at the age of the sacrifice. **g** The characteristic caecum enlargement of GF mice was

partially, but significantly, reduced in GM15 mice. The weights of the brain, liver and spleen

were equivalent in GM15 and SOPF mice, and significantly bigger than those of GF mice. **h**

Femur and tibia lengths were also equivalent in GM15 and SOPF mice, and significantly bigger

than those of GF mice. *P<0.05, **P<0.01, ***P<0.001, ****P<0.0001.

1222

Fig. 4. Immune phenotyping through serum and fecal immunoglobulin subtyping,

circulating cytokines analysis and immune cell populations analysis in different organs

a-c Dot plots where dots, lines and error bars represent respectively individual mice (15-17

GF, 20-21 GM15 and 20 SOPF), means and SEM. **a** Fecal IgA, serum IgA, IgG2b, IgG3, IgM

and IgE Luminex analysis. F-test for multiple linear regression analyses (fecal IgA) and Dunn's

multiple comparison showing restoration of fecal IgA, serum IgA and serum IgG2b levels in

GM15-colonized mice at the same level than SOPF, slightly increased IgM levels in GM15

mice compared to SOPF mice and elevated levels of IgG3 and IgE levels in GM15 mice

compared to SOPF and GF mice. **b** Circulating IL-17a and IL-22 levels Luminex analysis.

Dunn's multiple comparison showing elevated IL-17a levels in GM15 mice compared to SOPF

and GF mice and restoration of IL-22 levels in GM15 mice at the same level than SOPF. **c**

CD45+ count comparison in spleen, thymus, MLN and PP by flow cytometry. Dunn's multiple

comparison showing increased CD45+ cell count in spleen, thymus, and PP of SOPF mice

compared to GF mice, levels of GM15 mice are in between, and no difference between groups

for MLN. Tukey's multiple comparison showing increased PP in SOPF mice, then in GM15

mice, compared to GF mice. *P<0.05, **P<0.01, ***P<0.001, ****P<0.0001.

1239

Fig. 5. Metabolic phenotyping of GM15 animals. **a** PCA score plot representing the

distribution of the sample in the two first principal components. b Biplot representing the

projection the samples in the PLS-DA score plot given by the two first components and the

projection of the contribution of significant metabolites. **c** F-test for multiple linear regressions

showing equivalent decreased levels of Free Cholesterol and Phosphatidylcholine in GM15

and SOPF mice compared to GF mice. **d-e** Circulating hormones Luminex analysis. Dot plots

where dots, lines and error bars represent respectively individual mice (16 GF, 21 GM15 and

19-20 SOPF), means and SEM. **d** Dunn's multiple comparison showing restoration of serum

IGF-1 titers in GM15-colonized mice at the same level than SOPF. **e** F-test for multiple linear

regressions showing equivalent and reduced serum corticosterone concentrations in GM15

and SOPF mice, compared to GF mice.

1251

Fig. 6. GM15 mice and SOPF respond differently under post-weaning chronic

undernutrition. Body weight (**a**) and body size (**b**) were measured every week from weaning

to post-natal day 56. **c.** Representative picture of caecum, subcutaneous fat and right femur

of mice at day 56. Tibia length (**d**), liver (**e**) and epididymal white adipose tissue (**f**) weight were

measured. **g.** PCA of tissue weight and size showed clustering of GM15 and SOPF under

control conditions but a distinct pattern under nutritional stress. Percentages on each axis

indicate the variance associated to each coordinate. **h**. Serum IGF1 at day 56. BD: breeding

diet; DD: depleted diet. P-values after two-way ANOVA were adjusted for Sidak's post-hoc test

for multiple comparisons. *P<0.05, **P<0.01, ***P<0.001.

1261

Additional file 1: Figure S1. Morphological features of the GM15 strains. Each bacterial

strain was grown individually from a single colony isolated on agar medium and amplified in

liquid culture to exponential growth phase. Bacteria were Gram-stained and imaged by light

microscopy (80-fold magnification, NanoZoomer S60, Hamamatsu). Members of

Bacteroidaceae, Tannerellaceae and Enterobacteriaceae stained Gram-negative, while

Lactobacillaceae, Erysipelotrichaceae and Ruminococcaceae stained Gram-positive.

Lachnospiraceae stained Gram-positive, except *Lachnospiraceae* sp. MD329 and

Lachnospiraceae bacterium COE1, which stained Gram-negative likely due to their cell wall

structure as already reported for other Clostridiales [107].

1271

1272 **Additional file 2: Table S1. GM15 strains-specific primers**. Sequences of primers designed

1273 in this study and detection limits by qPCR microfluidic assay.

1274

1275 **Additional file 3: Table S2. KEGG clusters**.

1276

Additional file 4: Figure S2. *In vivo* experimental design. First, 5 couples of 8 week-old

C57BL/6J GF mice were colonized by oral inoculations of the GM15 bacterial community, twice

at 48h interval (F0). GM15 microbiota stability was assessed by qPCR microfluidic assay from

feces collected after 1, 2 and 3 weeks. Then, in order to evaluate the reproducible transfer of

the GM15 microbiota by fecal microbiota transplantation, 7 week-old C57BL/6J GF mice were

colonized by oral inoculations of a suspension of fresh fecal pellets from GM15 mice (F0),

twice at 48h interval. Again, qPCR microfluidic assay from feces collected after 1, 2 and 3

weeks was carried out. Next, the GM15 mouse line was amplified to monitor the GM15

microbiota through four filial generations at 6 weeks of age (F1-F4), and allow the phenotyping

study from two consecutive generations (F1.2 and F2.1). Reproduction performance and

perinatal mortality were recorded, 4 week-old mice were randomly selected at weaning,

monitored weekly for body weight and size, and feed intake, until sacrifice at 8-9 weeks of age.

GF and SOPF mice were also studied as control groups. Besides, a comparative analysis was

done on the fecal microbiota of 8 week-old GM15 mice (F1.1) either fed with the breeding diet

or an alternative isocaloric diet given for 4 weeks. Finally, the fecal microbiota of 6 month-old

and 12 month-old control GM15 mice fed with the breeding diet was analyzed.

1293

Additional file 5: Figure S3. Assessment of gut microbiota stability in GM15 founders

(F0, n=10), and of reproducibility between fecal and caecal samples of individual GM15

mice (F2, n=11). SOPF group shows the distribution of each GM15 strains in the complex gut

microbiota of 8-week-old SOPF mice. The absolute quantification of each strain was

determined by specific qPCR microfluidic assay. To gain clarity, each GM15 strain is

associated to a number between 1 and 15. * Strains 4, 6, 7 and 14 were at the detection limit

of the qPCR microfluidic assay, and thus were not detected in all samples. ** Strains 8, 9 and

10 were below detection limit of the qPCR microfluidic assay. Strain 5, obtained from the DSMZ

collection, was not detected in our SOPF colony. **a** Radar plot showing the GM15 strains

distribution in feces of GF C57Bl/6J mice before and 1, 2 and 3 weeks after the oral colonization

with the GM15 community. **b** Radar plot showing the reproducible detection of the GM15

strains in feces and caecum collected from the same GM15 mice.

1306

Additional file 6: Figure S4. Immune markers. Immune cell populations profiling analysis

by flow cytometry in different organs a-e Dot plots where dots, lines and error bars

represent respectively individual mice (13-17 GF, 21 GM15 and 19-20 SOPF), means and

SEM. **a** Tukey's multiple comparison showing restoration of whole blood monocytes levels in

GM15 mice at the same levels than in SOPF mice and difference of B cells levels between the

3 groups. **b** Tukey's multiple comparison showing splenic dendritic levels slightly decreased

in GM15 mice compared to SOPF, and Dunn's multiple comparison showing restoration of NK

cells levels in GM15 mice at the same levels than in SOPF mice. **c** Tukey's multiple comparison

showing restoration of MLN CD8+ T cells and B cells levels in GM15 mice at the same levels

than in SOPF mice. **d** Tukey's multiple comparison thymic CD4+ T cells levels restored in

GM15 mice at the same levels than in SOPF. **e** Tukey's multiple comparison showing slightly

decreased PP CD4+ T cells in GM15 mice compared to SOPF and Dunn's multiple comparison

showing increased PP dendritic cells levels in GF mice compared to SOPF mice and increased

NK cells levels in GF and GM15 mice compared to SOPF.

1321

1322   **Additional file 7: Table S3. Polar metabolites concentration in plasma samples of GF,**

1323   **GM15 and SOPF mice**. The metabolites quantification was performed with the help of

1324   Chenomx NMR suite 8.3. The concentrations are reported as mean and standard deviation.

1325

1326   **Additional file 8: Table S4. Non-polar metabolites concentration in plasma samples of**

1327   **GF, GM15 and SOPF mice.** The metabolites quantification was performed with the help of

1328   Chenomx NMR suite 8.3. The non-polar database profiles were created with the help of

1329   Compound builder module using $_1$H NMR spectra of authentic lipid standards. The

1330   concentrations are reported as mean and standard deviation.

1331

Additional file 9: Figure S5. PCA screen plot based on $_1$H NMR plasma fingerprints.

Processed $_1$H NMR plasma spectra of polar metabolites were binned with AMIX v3.9.14

software from Bruker Biospin, using 0.04ppm width from 0.5 to 10ppm spectral window. The

residual water region from 4.68 to 4.88pp was excluded from analysis. All spectra were

normalized to the total spectral area and the data table was exported into SIMCA v13.0.3

software for statistical analysis. The PCA analysis was performed using UV-scaling of data

and the model was autofit using cross validation rules to determine the number of significant

components. The clouds of sample points GF, GM15 and SOPF are distributed according to

microbiota complexity from left to right on PC1 (18%) and top to down on PC2 (12%).

1341

Additional file 10: Figure S6. Measurements of several organs after nutritional

challenge. Measurements at day 56 of weights of caecum (**a**), visceral (**b**) and subcutaneous

(**c**) white adipose tissue, brown adipose tissue (**d**), kidneys (**e**) and soleus muscle (**f**) and length

of right femur (**g**). P-values after two-way ANOVA were adjusted for Sidak's post-hoc test for

multiple comparisons. **h.** Correlation matrix showing Pearson r values of all the measured

parameters. For PCA, brown adipose tissue was excluded, given its lack of correlation with

other parameters.

a

Filial generations
(calculated CFU/g of feces)

- F0
- F1
- F2
- F3
- F4
- SOPF

b

Aging
(calculated CFU/g of feces)

- 2 months
- 3 months
- 4 months
- 6 months
- 12 months
- SOPF

c

Alternative diet (AD) vs. breeding diet (BD)
(calculated CFU/g of feces)

- BD_8 weeks
- BD_10 weeks
- BD_12 weeks
- BD_16 weeks
- AD_8 weeks
- AD_10 weeks
- AD_12 weeks
- AD_16 weeks
- SOPF

d

Fecal Microbiota Transplantation
(calculated CFU/g of feces)

- GF receivers
- 1 week
- 2 weeks
- 3 weeks
- Donors (F0)
- SOPF

e

1. *Bacteroides acidifaciens* A40
2. *Bacteroides caecimuris* I48
3. *Parabacteroides goldsteinii* WAL 12034
4. *Clostridium cocleatum* I50 *
5. *Clostridium clostridioforme* YL32
6. *Clostridium* ASF356 *
7. *Clostridium* ASF502 *
8. *Lachnospiraceae* bacterium COE1 **
9. *Lachnospiraceae* sp. MD329 **
10. *Lachnospiraceae* sp. MD308 **
11. *Lactobacillus johnsonii* NCC 533
12. *Lactobacillus murinus* 313
13. *Lactobacillus reuteri* I49
14. *Anaerotruncus colihominis* JM4-15 *
15. *Escherichia coli* Mt1B1