Metabolic network construction reveals probiotic-specific colonization resistance in a synthetic

small intestinal community

Jack Jansma [1], Anastasia Chrysovalantou Chatziioannou [1*], Kitty Castricum [2], Saskia van Hemert [2],

Sahar El Aidy [1#]

Affiliations

[1] Host-microbe Interactions, Groningen Biomolecular Sciences and Biotechnology Institute (GBB),

University of Groningen, Groningen, The Netherlands.

[2] Winclove Probiotics, Amsterdam, the Netherlands

*Present address: Department of Chemistry, Laboratory of Analytical Biochemistry, University of

Crete, Heraklion, Greece

[#] Corresponding author

Sahar El Aidy. Host-microbe Interactions, Groningen Biomolecular Sciences and Biotechnology

Institute (GBB), University of Groningen, Nijenborgh 7, 9747 AG Groningen, The Netherlands. P:

+31(0)503632201.

E: sahar.elaidy@rug.nl

Abstract

The gut microbiota plays a crucial role in maintaining overall health and probiotics have emerged as a

promising microbiota-targeted therapy for improving human health. However, the molecular

mechanisms of probiotics action in general and the targeting of small intestinal microbiota by

probiotics are not well understood. To address this, we constructed a synthetic community of three

species, which resembles the upper small intestinal microbiota. Our results indicate that probiotic

supplementation can directly affect the metabolism of the community, resulting in colonization

resistance in a probiotic specific manner. Supplementation with *Streptococcus thermophilus* led to

increased lactate production and a decrease in pH, while *Lactobacillus casei* supplementation

increased the resistance to perturbations and nutrient utilization without affecting lactate

production or pH. Additionally, when combined with kynurenine, *Lactobacillus casei* enhanced the

kynurenine pathway metabolism resulting in elevated kynurenic acid levels and possibly indirect

colonization resistance. Overall, our study reveals how selecting probiotics with distinct functional

capacities can unlock the full potential of microbiota-targeted therapies.

Keywords

Small intestine, microbiota, probiotics, dynamic metabolic networks, kynurenic acid

Introduction

In recent decades, there has been a growing interest in investigating the gut-microbiota and its significant roles in maintaining human health [1, 2]. The gut-microbiota is composed primarily of *Bacteroidaceae, Prevotellaceae, Rikenellaceae, Lachnospiraceae* and *Ruminococcaceae*, which reside mainly in the colon [3]. In contrast to the colon, the small intestinal microbiota is highly dynamic and location-specific. For example, *Streptococcaceae* and *Veilonellaceae* are abundant in the duodenum [4], while *Streptococcaceae, Carnobacteriaceae* and *Actinomycetaceae* prevail in the jejunum [5] and *Lactobacillaceae, Erysipelottrichaceae,* and *Enterobacteriaceae* are dominant in the ileum [6]. The small intestinal microbiota plays a crucial role in regulating the metabolic, endocrine, and immune functions of their hosts [7–9]. Consequently, altering the equilibrium between the host and the microbiota can negatively affect the health of the host [10, 11]. Thus, to re-establish this equilibrium and enhance the health of the host, multiple microbiota-targeted therapies, such as probiotics, are employed [12].

Probiotics are defined as live microorganisms that, when administered in adequate amounts, confer a health benefit on the host [13]. The development of probiotic therapies aiming at targeting the small intestinal microbiota poses a significant challenge due to the limited access to the human small intestine. Currently, investigations into the effect of probiotics on the small intestine are carried out either directly on animal models [9, 14] or indirectly through small clinical trials that lack detailed molecular mechanisms [15]. This lack of understanding hampers the development of novel probiotic therapies. In recent years, it has been demonstrated that the microbiota retrieved from stoma effluent exhibits similar compositional and behavioral characteristics with the healthy ileal microbiota [16]. These findings suggest that the stoma effluent-derived microbiota may serve as a promising source for probiotic research [17, 18]. However, the utilization of stoma effluent is limited for the study of the upper small intestine. To address this limitation, there is a need for construction of synthetic bacterial communities to bridge this research gap [19]. Synthetic bacterial communities

are bacterial mixtures produced through co-culturing of selected bacteria in controlled environments

[20]. These communities can comprise a range from two [21] to over one hundred species [22].

Synthetic communities can be subjected to perturbation through manipulation of growth parameters

or the introduction of new bacterial species, and can also be supplemented with specific compounds

such as prebiotics [23]. Analysis of the community's molecular response allows a deeper

understanding of the underlying molecular mechanisms [24], optimization of metabolites' production

[25] and assessment of emergent properties such as stability or resistance to perturbations [26, 27].

This methodology has the potential to guide the development of novel microbiota-targeted

therapies.

In humans, the kynurenine pathway is the primary route for the metabolism of free tryptophan, and

it plays a critical role in the generation of energy in the form of nicotinamide adenine dinucleotide

(NAD^+), immune system regulation [28] and the production of the neuroactive compounds quinolinic

and kynurenic acid (KYNA) [29]. Increased tryptophan metabolism via the kynurenine pathway in the

gut is associated with obesity and alterations in the microbiota [29]. G-protein coupled receptor 35

(GPR35), which is activated by KYNA, is abundantly expressed in the intestine, where it plays a role in

immune regulation and epithelial barrier function [30]. The concentration of KYNA in the gut lumen

increases from the proximal to the distal part of the intestine, indicating production by the gut

microbiota [31, 32]. Several gut bacteria can produce compounds from the kynurenine pathway [32,

33] and probiotics supplementation have been shown to reduce intestinal inflammation and improve

intestinal barrier function accompanied by reduced kynurenine but increased KYNA levels [34, 35].

However, the precise effects of kynurenine pathway metabolism within the microbiota remain

unknown.

In the present study, a dynamic correlation-based network approach was utilized in

conjunction with multivariate analysis of experimental data obtained from chip-based digital

polymerase chain reaction (c-dPCR), proton-nuclear magnetic resonance (^{1}H-NMR) and high-

performance liquid chromatography-ultraviolet (HPLC-UV). The aim was to examine the impact of

introducing probiotic strains on the general metabolic activity and the kynurenine pathway activity of

a simplified community consisting of three intestinal isolates, which represent the small intestinal

microbiota.

Material and methods

Bacteria and growth media

Bacteria were selected based on their abundance in the upper small intestine, ability to metabolically

affect the kynurenine pathway or probiotic activity in the small intestine. *Escherichia coli* DSM11250

was obtained from the German collection of microorganisms and cell cultures; *Pseudomonas*

fluorescens MFY63 was obtained from the laboratory of microbiology signals and microenvironment

[36]; *Streptococcus salivarius* HSISS4 was obtained from the laboratory of host-microbe interactomics

[37]; *Streptococcus thermophilus* W69 and *Lactobacillus casei* W56 were obtained from Winclove

probiotics. All individual bacterial and community growth experiments were performed using

enriched beef broth (EBB) [38].

Bacterial kynurenine pathway metabolism

To investigate kynurenine pathway metabolism by the bacterial strains, 3 mL EBB was inoculated

from a glycerol stock stored at -80 °C and grown overnight at 37 °C. *E. coli* and *P. fluorescens* were

grown shaking at 220 RPM, *S. salivarius*, *S. thermophilus* and *L. casei* were grown without shaking.

Overnight grown pre-culture were diluted 1:100 in fresh EBB supplemented with 500 µM tryptophan

and kynurenine or kynurenic acid. At the start (0h) and at termination of the experiment (24h) 250 µL

culture was transferred to a new tube and immediately 1 mL cold (-20 °C) methanol was added. The

samples were stored at -20 °C until further use.

Growth experiments, community construction and sampling in a mini-bioreactor

All experiments were conducted in a MiniBio 250 mL bioreactor
(Applikon Biotechnology, The Netherlands) with a working volume of 250 mL. Each experiment was
performed in 200 mL EBB + 500 µM tryptophan and, if indicated 100 µM kynurenine. To prevent
foam accumulation 0.1 % (v/v) silicon antifoam emulsion (Carl Roth, Germany) was added. The
temperature, dissolved oxygen and pH were monitored online. The adaptive my-Control system
(Applikon Biotechnology, The Netherlands) was used to control process parameters. Setpoints of 50%
of the atmospheric oxygen and 37 °C were applied for dissolved oxygen and the temperature,
respectively [39, 40]. The culture was continuously stirred which was controlled with the adaptive
my-Control system. The stirrer speed increased linearly with decreasing oxygen below the setpoint
up to a maximum of 1200 RPM. The stirrer speed decreased linearly with increasing oxygen above
the setpoint down to a minimum of 220 RPM.

To determine kynurenine pathway metabolism of the individually grown bacteria in the bioreactor, a
single colony of *P. fluorescens* or 1 mL overnight grown precultures of *E. coli*, *S. salivarius*, *S.*
thermophilus or *L. casei* was added to 200 mL EBB supplemented with 500 µM tryptophan and, if
indicated, 100 µM kynurenine. At the start (0h) and at termination of the experiment (24h) 250 µL
culture was transferred to a new tube and immediately 1 mL cold (-20 °C) methanol was added. The
samples were stored at -20 °C until further use.

Community growth was obtained via inoculation of a single colony of *P. fluorescens* MFY63 in 200 mL
EBB supplemented with 500 µM tryptophan and, if indicated, 100 µM kynurenine. After 23 hours of
growth, 1 mL of combined overnight grown pre-cultures of *E. coli* (1:10000), *S. Salivarius* (1:100) and
L. casei (1:100) or *S. thermophilus* (1:100) were added. At 0, 3, 5, 7, and 24 hours after inoculation
with the overnight grown pre-cultures a 5 mL sample was obtained. Immediately 250 µL was added
to 400 µL NMR buffer (200 mM Na_2HPO_4, 44 mM NaH_2PO_4, 1 mM TSP, 3 mM NaN_3 and 20% (v/v)
D_2O), centrifuged at 4 °C, 21130 rcf for 20 min and 550 µL was transferred to a 5 mm NMR tube. For
biomass determination, 1.5 mL sample was centrifuged at room temperature for 1 min at 21130 rcf.
The supernatant was removed and the wet pellet was weighed. Lastly, 250 µL sample was

transferred to a new tube and immediately 1 mL cold (-20 °C) methanol was added. The samples

were stored at -20 °C until further use.

UHPLC-UV

Cell debris was removed from the sample stored at -20 °C in methanol by centrifugation at 4 °C,

21130 rcf for 10 min. The supernatant was transferred to a new tube and the methanol fraction was

evaporated using a Savant speed-vacuum dryer (SPD131, Fisher Scientific, Landsmeer, The

Netherlands) at 60 °C for 75 min. The dried material was reconstituted to 1 mL with 0.7% perchloric

acid. Samples were filtered using a 0.2 µM RC membranes (Phenomenex, Utrecht, The Netherlands)

and injected into the UHPLC system (Dionex UltiMate 3000 autosampler; Dionex UltiMate 3000 LPG-

3400SD pump, Thermo Fisher Scientific, Waltham, Massachusetts, USA). All samples were analyzed

on a C18 column (Kinetex 5 µm, C18 100 Å, 250 × 4.6 mm, Phenomenex, Utrecht, the Netherlands)

using a gradient of water/methanol with 0.1% formic acid (0 to 10 min, 95% to 80% H_2O; 10 to 20

min, 80% to 5% H_2O; 20 to 23 min, 5% H_2O; 23 to 31 min, 95% H_2O), with a flowrate of 1 mL/min.

The column was kept at a temperature of 35 °C. UV-detection at 260 nm was performed with an

UV6000LP Detector (Dionex Ultimate 3000 variable wavelength detector, Thermo Fisher Scientific,

Waltham, Massachusetts, USA) Data recording and analysis were performed using Chromeleon

software (version 6.8 SR13).

DNA extraction

DNA extraction was performed as previously described [41] from the cell pellets obtained by

centrifuging 1.5 mL culture for 1 min at 21130 rcf. The supernatant was removed, the pellets were

resuspended in 750 mL lysis buffer (500 mM NaCl, 50 mM Tris-HCl (pH 8.0), 50 mM EDTA, 4 % SDS)

and transferred to a 2 mL screw cap tube containing 0.5 g, 0.1 mm zirconia beads and 4, 3 mm glass

beads. The cells were disrupted with a mini-bead beater (3X 1 min, with 1 min intervals on ice) and

incubated at 95 °C for 15 min. Cell debris was removed by centrifugation at 4 °C, 21130 rcf for 20

min. 500 µL of the supernatant was transferred to a clean tube. 200 µL 10 M ammonium acetate was

added, and the samples were incubated for 10 min on ice. The samples were centrifuged at 4 °C,

21130 rcf for 10 min. 600 µL was transferred to a clean tube and 600 µL isopropanol was added. The

samples were incubated overnight at -20 °C and afterwards centrifuged at 4 °C, 21130 rcf for 15 min

to pellet the nucleic acids. The supernatant was removed and to wash the nucleic acid pellet 700 µL

70 % ethanol was added, the samples were mixed thoroughly and centrifuged at room temperature,

21130 rcf for 5 min. The pellets were completely dried by leaving the tubes open for 60 min at room

temperature and were dissolved in 100 µL MilliQ.

Digital chip-based PCR analysis

Chip-based digital polymerase chain reaction (c-dPCR) was performed with the QuantStudio™ 3D

Digital PCR System (User Guide, Catalog Number A29154, Thermo Scientific). Each c-dPCR reaction

was prepared in a final volume of 15,5 µL containing 5 µL QuantStudio™ 3D Digital PCR Master Mix

v2 (Applied Biosystems), 1 µM end concentration forward and reverse primer, 0,25 µM end

concentration probe **(Supplementary table 1, Eurogentec)** and nuclease free water. 1 µL extracted

genomic DNA was added to the reaction mix. The DNA concentration was estimated by using the

cycle threshold values from the qPCR run. For no-template control 1 µL nuclease free water was

added instead of genomic DNA. The positive control was extracted genomic DNA from pure cultures.

14,5 µL was loaded into the QuantStudio™ 3D Digital 20K Chip v2 using the QuantStudio™ 3D Digital

PCR Chip Loader. The amplification reactions were performed in the QuantStudio™ 3D Digital PCR

System. The PCR conditions were: 10 minutes at 96 °C for DNA polymerase activation, followed by 40

two-step cycles of 2 minutes at 60 °C, 30 seconds at 98 °C and a final extension for 2 minutes at 60 °C

and an infinite 10°C hold until the chips were read. After DNA amplification the chips were

transferred to the QuantStudio™ 3D Digital PCR Chip reader for imaging. The end-point fluorescence

data were collected and analyzed using the QuantStudio™ 3D AnalysisSuite cloud software (version

3.1.6-PRC-build18). Droplets were considered positive when the fluorescence signal was above the

threshold. The threshold was set to the positive control DNA and no-template control. The number

of positive and negative reactions were counted and using Poisson statistics to measure absolute

copies per µL.

^{1}H NMR spectroscopy and data processing

All ^{1}H-NMR spectra were recorded using a Bruker 600 MHz AVANCE II spectrometer equipped with a

5 mm triple resonance inverse cryoprobe and a z-gradient system. The temperature of the samples

was maintained at 25 °C during measurement. Prior to data acquisition, tuning and matching of the

probe head followed by shimming and proton pulse calibration were performed automatically for

each sample. One-dimensional (1D) ^{1}H-NMR spectra were recorded using the first increment of a

NOESY pulse sequence with presaturation ($\gamma B_1 = 50$ Hz) for water suppression during a relaxation

delay of 4 s and a mixing time of 10 ms. 256 scans of 65,536 points covering 13,658 Hz were recorded

and zero filled to 65,536 complex points prior to Fourier transformation, an exponential window

function was applied with a line-broadening factor of 1.0 Hz. The spectra were phase and baseline

corrected and referenced to the internal standard (TSP; δ 0.0 ppm), using the MestReNova software

(v.12.0.0-20080, Mesterlab Research). With the same software, spectral alignment was performed

selecting manually areas and applying a linear filling method of the missing values, after the

exclusion of the water signal and the surrounding empty area (4.50-5.30 ppm). Spectral binning

followed from -0.50 to 9.00 ppm with an equal size binning step of 0.005 ppm. Noise removal was

performed by averaging each integrated bin separately and removing the bins with an average below

100. Before ordination analysis bins belonging to TSP (-0.50 – 0.72 ppm), ethanol (1.170 – 1.220 ppm

and 3.640 – 3.690 ppm) and a pH sensitive area (7.850 – 8.190 ppm) were removed. The annotation

of the bins was performed with the Chenomx Profiler software (Chenomx NMR Suite 8.6 and

Chenomx 600 MHz, version 11) and the HMDB database 5.0 (http://www.hmdb.ca).

Statistical and network analysis

All analysis of variance (ANOVA) were performed using GraphPad Prism 7.0. When a comparison

resulted in a statistically significant difference ($P < 0.05$), multiple comparisons testing was

performed by controlling the false discovery rate (FDR) according to the Benjamini-Hochberg method

($\alpha < 0.05$). The R package MixOmics was used for ordination and multivariate statistical analysis of

the ^{1}H-NMR spectra [42]. The dynamic profile comparison using Kendall's τ correlation with the BH (α

< 0.05) was performed using the R package psych. In CytoScape 3.9.1 the plugin CoNet [43] was used

for network construction. The network properties were obtained using the network analyzer tool. To

analyze the PLS-DA separation, the accuracy, R2, Q2 and *P*-value of a permutation test performed

with the separation distance statistic and 1000 permutations are calculated using MetaboAnalyst 5.0

[44].

Results

Kynurenine pathway metabolism and pH changes in a three species synthetic community

To investigate the metabolic network changes resulting from probiotic supplementation, it is critical

to carefully select bacterial members for the community. These members should be isolated from

the normal small intestinal microbiota and capable of metabolizing key proximal small intestinal

metabolites such as, acetate and lactate [7]. Additionally, the kynurenine pathway should be affected

by the community. In this study, *Streptococcus salivarius* and two bacteria capable of affecting the

kynurenine pathway, namely *Pseudomonas fluorescens* and *Escherichia coli* [33, 45–47], were

selected. Two commonly used probiotic species, which were previously shown to be active in the

small intestine, *Streptococcus thermophilus* W69 [48, 49] and *Lactobacillus casei* W56 [50], were

added separately to distinguish between general and strain-specific effects of probiotic

supplementation. Before constructing the community, each species was separately grown in flasks

with 500 μM tryptophan and kynurenine or KYNA to investigate the ability of the various community

strains to metabolize kynurenine pathway metabolites. After 24 hours of incubation, kynurenine

pathway metabolism was observed for *P. fluorescens* and *E. coli*, but not for *S. salivarius*, *S.*

thermophilus and *L. casei* **(Supplementary figure 1)**. To confirm the kynurenine pathway metabolism

in conditions resembling the upper small intestine, each strain is separately grown in a

MiniBio 250 mL bioreactor (Applikon Biotechnology, The Netherlands) at 37 °C with controlled pO2

(50% of the atmospheric air) as previously described [39, 40] with and without 100 µM kynurenine.

After 24 hours of incubation, the results revealed that *E. coli* effectively metabolized tryptophan and

converted kynurenine into KYNA. Conversely, *P. fluorescens* produced only a small amount of

kynurenine from tryptophan. In contrast, *S. salivarius*, *S. thermophilus*, and *L. casei* did not

metabolize either tryptophan or kynurenine **(Supplementary figure 2A)**. These findings confirm the

differential metabolic capabilities of bacterial strains in relation to kynurenine pathway metabolism.

Concurrently, the pH dynamics of each strain cultured in the presence or absence of kynurenine were

assessed. *P. fluorescens* did not exhibit any noticeably pH changes during the course of the

experiment, whereas *E. coli*, *S. salivarius* and *S. thermophilus* demonstrated a decrease in pH after

3.5h of growth. *L. casei* reduced the pH after 12h of growth. Among all bacteria, only *E. coli*

displayed an increase in pH after 6h. Comparable pH profiles were observed for all bacteria, except

for *E. coli* when incubated with 100 µM kynurenine. Co-culturing *E. coli* with kynurenine resulted in

an earlier and more pronounced decline in pH **(Supplementary figure 2B)**.

The synthetic community was then grown in the bioreactor under the same conditions stated above.

P. fluorescens was initially cultured in the bioreactor for a period of 23h. Thereafter, *S. salivarius*, *E.*

coli and one of the probiotic species were added to the bioreactor to create a bacterial community.

Furthermore, we supplemented each community with 100 µM kynurenine to simulate increased

activity of the small intestinal kynurenine pathway. The small intestine is known to undergo rapid

changes in response to food ingestion, necessitating the need for the bacterial community to adapt

[16, 51]. Moreover, substantial metabolism was observed in individually grown bacteria after 3.5h of

growth, as indicated by the pH profile **(Supplementary figure 2B).** Therefore, we collected samples

from the community at 0, 3, 5, 7 and 24h after community growth and analyzed the biomass,

absolute species distribution using c-dPCR, metabolic environment using ^{1}H-NMR, and kynurenine

metabolism using HPLC-UV **(Figure 1A)**. Our results showed that both kynurenine (between 3 and 7h)

and KYNA (between 7 and 24h) were produced in the control and probiotic-supplemented

communities, with kynurenine addition increasing KYNA production in all communities

(Supplementary figure 3). However, we observed that the addition of kynurenine together with *L.*

casei increased the production of kynurenine after 5 and 7h of growth, subsequently leading to

increased production of KYNA after 7 and 24h of growth compared to the control and *S.*

thermophilus supplemented communities **(Supplementary figure 3, figure 1B)**. Moreover, the pH

levels declined sharply after 3h of community growth, followed by a linear increase and a plateau

after 10h of growth **(Figure 1C)**, and the addition of kynurenine did not affect the lowest pH and the

time it was reached (ordinary two-way ANOVA; $F(1,12) = 1.78$, $P = 0.21$ and $F(1,12) = 0.76$, $P = 0.4$

respectively) **(Table 1)**. However, probiotic supplementation altered both the lowest pH and the time

it was reached (ordinary two-way ANOVA; $F(2,12) = 6.85$, $P = 0.01$ and $F(2,12) = 4.0$, $P = 0.047$)

(Table 1), mostly attributed to differences between *L. casei* and *S. thermophilus* supplementation

(Figure 1C, Supplementary figure 4), indicating a probiotic specific effect.

Overall, the addition of kynurenine together with *L. casei* increased the production of kynurenine and

subsequently KYNA compared to the control and *S. thermophilus* supplemented communities.

Probiotic supplementation altered both the lowest pH and the time it was reached, mostly attributed

to differences between *L. casei* and *S. thermophilus* supplementation, indicating a probiotic-specific

effect. These findings suggest that probiotics, particularly *L. casei*, may influence the production of

kynurenine pathway metabolites in the small intestine, which could have important implications for

human health.

Effect of kynurenine and probiotic supplementation on biomass and cell counts in a simplified

bacterial community

Bacteria can gain advantages in a community through utilizing the energy produced by converting

compounds present in the environment, inhibiting the growth of competitors, or stimulating other

bacteria to secrete beneficial metabolites, thereby affecting the structure and composition of the

community [52]. Therefore, we hypothesized that the supplementation of a probiotic species or

kynurenine, which was only metabolized by *P. fluorescens* and *E. coli* (**Supplementary figure 2A**),

may alter the composition of the simplified community. The total biomass and the community

composition were compared (**Figure 1A**) to investigate the effects of kynurenine and probiotic

supplementation on community growth. Kynurenine addition did not affect the biomass, while

probiotic addition altered total biomass production after 7h of growth (**Supplementary data 1).**

Specifically, the addition of *L. casei* resulted in an increase in biomass in the absence of kynurenine,

while the introduction of *S. thermophilus* in the presence of kynurenine led to a reduction in the total

biomass (**Supplementary figure 5**).

Next, the total cell count and cell count of specific species in the community were monitored using c-

dPCR. Kynurenine supplementation did not affect the total cell count or the cell count of any

individual species (**Supplementary data 2).** Furthermore, probiotic supplementation did not alter the

cell count of *P. fluorescens* (**Figure 2, Supplementary data 2**), but probiotic supplementation altered

the total cell count and the cell count of *E. coli* (after 3 and 5h of growth, although multiple

comparison testing did not identify specific alterations), and of *S. salivarius* (after 5 and 7h of growth)

(**Figure 2, supplementary data 2**), which coincides with the decrease in pH (**Figure 1C**) and with the

growth of *S. thermophilus* (**Figure 2).** In contrast, the cell counts of *L. casei* continuously increased

over time (**Figure 2**). Cell counts of the probiotic species were not affected by kynurenine

supplementation at any sampling point (**Figure 2**). Specifically, *L. casei* addition inhibited the growth

of *S. salivarius* by altering the community metabolism. These results suggest that the

supplementation of certain compounds, exemplified here by kynurenine, or probiotics, or a

combination of both, can alter the growth and composition of bacterial communities, which could

have implications for the design of probiotic therapies.

Metabolic network analysis of probiotic supplementation and kynurenine metabolism in the synthetic

bacterial community.

In order to elucidate the metabolic changes that are associated with alterations in pH, kynurenine

metabolism and cell count, we employed ^{1}H-NMR to analyze the supernatant **(Figure 1A).** Following

spectral binning and processing, a total of 943 bins were subjected to principal component analysis

(PCA). Our results revealed time-dependent clustering of the samples **(Figure 3A)**, whereby the most

significant contributors to the observed separation were lactate (bin 366, 368 and 369), acetate (bin

485 and 486), glucose (bin 868) and betaine (bin 880).

To determine whether the dynamic metabolic profiles of the community were altered by probiotic or

kynurenine supplementation, we conducted partial least squares – discriminant analysis (PLS-DA)

with cross validation (CV) and a permutation test was performed with the 943 bins as input. Our

analysis revealed no significant differentiation between communities with and without kynurenine

supplementation, nor between control and *S. thermophilus*-supplemented communities

(Supplementary figure 6A and 6B). However, we observed significant separation in communities

supplemented with *L. casei,* with lactate (bin 923), glucose (bin 851 and 853) and threonine (bin 820)

contributing most to the separation from the control **(Figure 3B)**. Additionally, glucose (bin 848, 851

and 853 – 856), glycerol (bin 815) and cadaverine (bin 707) were the major contributors to the

differentiation between *L. casei* and *S. thermophilus* supplemented communities **(Figure 3C)**.

To demonstrate the metabolic contributions responsible for the variance obtained from the entire

spectral dataset, the chemical library of Chenomx profiler software and the online human

metabolome database version 5.0 were utilized. We identified 30 metabolites, and subjected the

area under the curve (AUC) of a representative peak for each metabolite to ordination analysis. The

results of this approach were similar to the ordination analyses conducted with the bins. PCA

revealed temporal clustering of the samples and PLS-DA revealed significant separation of the *L. casei*

supplemented community from the control and *S. thermophilus* supplemented communities

(Supplementary figure 7A, B and C). Therefore, we postulated that the AUC values of the identified

set of metabolites may serve as a surrogate for the full spectra dataset and could be employed for

subsequent analysis.

Given that the metabolic network of a community is highly interconnected, a substantial alteration in

certain metabolites can lead to multiple smaller changes in related metabolites that might not be

detectable by ordination analysis **(Figure 3A, B and C)**. Similarly, minor effects of kynurenine

supplementation on metabolism may go undetected via ordination analysis. To investigate smaller

metabolic changes induced by kynurenine or probiotic addition to the synthetic community, we

compared dynamic metabolic profiles between experiments, which revealed similarities

(Supplementary figure 8) and differences in the dynamics of saccharolytic and proteolytic

community metabolism **(Figure 3D, Supplementary data 3)**. Of note, we observed increased lactate

production in the *S. thermophilus* supplemented community, which could potentially account for the

observed pH variation **(Figure 1C)**.

To explore the coordinate behavior of metabolites and network topology features across the tested

conditions [53], we constructed dynamic correlation-based networks using the Cytoscape plugin

CoNet [43], whereby the nodes represent the metabolites and the edges represent correlations

between the metabolic profiles per condition. Without kynurenine, *S. thermophilus* supplementation

decreased the number of edges and the clustering coefficient, while the metabolite with the highest

number of edges was aspartate (7), compared to cadaverine (12) in the control network **(Figure 3E,**

table 2). *L. casei* supplementation had a more drastic effect on the network, increasing the number

of edges, clustering coefficient and network density substantially **(Figure 3E, table 2)**. Kynurenine

supplementation reduced the number of edges, including indole, present in the control community

network, and reduced the number of edges of cadaverine from 12 to 1. Consequently, cadaverine

and indole no longer connected two major hubs of connected metabolites compared to control

community. The clustering coefficient and network density of the control community supplemented

with kynurenine was 1, due to the connections between all the metabolites with all the other

metabolites in each hub **(Figure 3E, table 2)**. Kynurenine supplementation to the *S. thermophilus* and

L. casei supplemented communities increased the number of edges, clustering coefficient and

network density, with indole and cadaverine being the metabolites with the most edges, similar to

the control network without kynurenine supplementation **(Figure 3E, table 2).** The node degree and

betweenness centrality of individual metabolites were also analyzed, showing probiotic-specific

effects on the community metabolism as well as an effect of kynurenine supplementation on

cadaverine metabolism **(Supplementary data 4)**. Together the network analysis revealed probiotic-

specific effects on the community metabolism and an effect of kynurenine supplementation on

clustering coefficient, network density and node parameters of the metabolic networks, mainly due

to alterations in cadaverine utilization.

Discussion

The present study investigated the impact of probiotic supplementation on metabolism of the small

intestinal microbiota. To achieve this, we constructed a synthetic community consisting of three gut

isolates that mimicked the microbial community of the upper small intestine. Specifically, we

examined the metabolic impact of probiotic supplementation on both general metabolism and the

kynurenine pathway, which has been implicated in the modulation of immune responses and

neuroinflammation. Our findings shed light on the probiotic-specific mechanisms of action that drive

colonization resistance in the small intestine and suggest that the kynurenine pathway may be a

promising target for future probiotic-based therapies aimed at promoting intestinal health.

Probiotics are known to exert their effects through competitive exclusion often by reducing the pH of

the gut milieu through increased lactate production [54]. In our study, supplementation with *S.*

thermophilus led to increased lactate production, resulting in a reduction in pH, and accompanied by

a decrease in formate and acetate levels **(Figure 3D and 1C)**. However, this supplementation did not

result in significant changes to the composition of the community **(Figure 2)** or the dynamic

metabolic network **(Figure 3E)**. Conversely, when *L. casei* was supplemented, we observed a

reduction in the growth of *S. salivarius*, which was not accompanied by increased lactate production

(Figure 2, 3D). Other studies have also reported beneficial effects of *L. casei* supplementation

independent of acidification or lactate production [55]. These results suggest that *L. casei*

supplementation alters the metabolic environment, enhancing the resistance of the community to

perturbations and improving metabolic utilization **(Figure 3D and E).** This finding is consistent with

previous studies showing a positive association between habitual *L. casei* Shirota intake and long-

term compositional stability of the fecal microbiota [56].

The concept of direct colonization resistance refers to the ability of the resident microbiota to resist

the colonization and expansion of invading species [57]. However, decreased diversity can lead to

unoccupied ecological niches that can be occupied by invading species potentially harmful to host

health [58]. Probiotics have been proposed as a means to increase colonization resistance by filling

unoccupied niches before pathogenic species can colonize [59]. However, current probiotic

interventions do not take into account the functional capacity of probiotics to occupy specific niches

in the gut [60], which may explain the variable effectiveness of probiotics [61]. To address this,

synthetic communities are constructed [22, 62–65], but the lack of molecular insights has hampered

the development of probiotic therapies. The genetic capacity of *L. casei* in our synthetic community

of three species made it more functionally diverse **(Figure 3D)** compared to the three initial

community members compared to *S. thermophilus*, which resembles *S. salivarius*. Thus, *L. casei*

increased the resistance to perturbations and metabolic utilization **(Figure 3E, table 3),** by occupying

an empty ecological niche, whereas *S. thermophilus* had to compete for the ecological niche

occupied by *S. salivarius*. However, the ability to occupy a niche is determined by the functional

capability [66] and the environment [67]. Therefore, selecting strains with a wide range of functional

capacities and constructing a diverse multispecies probiotic formula could be a promising approach

for developing effective probiotic therapies [18, 68]

Indirect colonization resistance involves improving the gut epithelial barrier or enhancing the hosts

immune system [69]. Metabolites produced by the microbiota can interact with the host and

improve gut integrity. For example, stimulation of GPR35 by KYNA, was shown to enhance mucosal

repair and stimulate mucus secretion, which increases the integrity of the gut epithelial barrier [70,

71]. Thus, increasing microbial production of KYNA in the small intestine may indirectly increase

colonization resistance. Although no in vivo studies were conducted in our study to confirm

improved intestinal barrier integrity due to increased microbial KYNA production resulting from

probiotic addition, we show that in the *L. casei* supplemented community, *E. coli* had a reduced

impact on the metabolic environment since the node degree and betweenness centrality of the *E.*

coli produced metabolites cadaverine [72] and indole [73] were lower **(Figure 3E)**. In *kynurenine*

and S. thermophilus supplemented community, E. coli had an increased impact on the metabolic

environment **(Supplementary figure 1 and 2),** even though this effect was not translated into

increased cell counts for *E. coli* **(Figure 2)**.

Overall, the present study demonstrates that supplementation with different probiotics can lead to

distinct alterations in the metabolic environment of the microbiota community, with *L. casei*

enhancing resistance to perturbations and improving metabolic utilization. Our findings suggest that

selecting probiotic strains with a wide range of genetic capabilities, which complement the functional

capacity of the residing microbiota or constructing a diverse multispecies formula may be a

promising approach for developing effective probiotic therapies. Additionally, the present findings

suggest that increasing microbial production of metabolites such as KYNA in the small intestine may

indirectly improve intestinal barrier integrity and increase colonization resistance. Further in vivo

studies are necessary to confirm the effects of probiotic supplementation on intestinal barrier

integrity and to develop probiotic therapies that target specific ecological niches in the gut.

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Data availability: The raw ^{1}H-NMR data was deposited in the Metabolights public repository [74]

under the accession number: MTBLS7370

Tables

Table 1: properties of the pH profile of each community.

The mean ± SD of the lowest pH value and time the lowest pH value was reached for all communities

with and without kynurenine (kyn) supplementation.

Communities	Lowest pH value	Time of the lowest pH value (h)
Control	5.69 ± 0.09	4.7 ± 0.2
S. thermophilus	5.56 ± 0.08	3.8 ± 0.1
L. casei	5.77 ± 0.03	4.5 ± 0.7
Control + kyn	5.77 ± 0.17	4.0 ± 0.2
S. thermophilus + kyn	5.53 ± 0.11	3.9 ± 0.3
L. casei + kyn	6.05 ± 0.25	4.6 ± 0.4

479   **Table 2: Properties of the dynamic correlation-based metabolic networks for each community**. The

480   number of nodes, edges, clustering coefficient, network density and the nodes with the most edges

481   of the networks depicted in figure 3E for the communities with probiotic or kynurenine (kyn)

482   supplementation.

Communities	Nodes	Edges	Clustering coefficient	Network density	Node with the most edges (edges)
Control	24	54	0.857	0.368	Cadaverine (12)
S. thermophilus	23	40	0.794	0.374	Aspartate (7)
L. casei	17	94	0.935	0.886	Gluconolactone/gluconate, glycerol (14)
L. casei + kyn	24	103	0.892	0.654	Glucose, tryptophan, glycerol (16)
S. thermophilus + kyn	20	56	0.901	0.549	Indole, cadaverine (12)
Control + kyn	20	54	1	1	Tryptophan, phenylalanine, alanine, methionine, glutamate, proline, isoleucine, glycine, valine, tyrosine (9)

Figure legends

Figure 1: *Lactobacillus casei* supplementation increases kynurenine pathway metabolism in a

synthetic community

A) Overview of the construction of the community and sampling procedure. *P. fluorescens* was

inoculated 23h prior to inoculating other species with probiotics. PH and pO_2 (indicated with dashed

line) were continuously measured. Black arrows indicate sampling points (0, 3, 5, 7 and 24h) for

obtaining metabolomics, compositional and biomass data. All experiments were performed in

triplicates. B) Boxplots showing the area under the curve (AUC) for kynurenine and kynurenic acid as

measured from the HPLC-UV chromatograms normalized to the AUC of the initial tryptophan peak

obtained from -23h. Significant comparisons are depicted with letters (a-d): a: q < 0.05, b: q < 0.01, c:

q < 0.001 and d: q < 0.0001 as obtained with an ordinary two-way ANOVA and multiple comparisons

testing by controlling the false discovery rate according to the Benjamini-Hochberg procedure. C) pH

profile for control, *S. thermophilus* supplemented and *L. casei* supplemented communities

with/without kynurenine. Sampling points indicated by dashed lines. Boxplot indicates the lowest pH

value per condition. Significant differences are indicated with asterisks; *, *q* < 0.05, as obtained with

an ordinary one-way ANOVA and multiple comparison testing by controlling the false discovery rate

according to the Benjamini-Hochberg procedure.

Figure 2: Supplementation of *Lactobacillus casei* reduces the growth of *Streptococcus salivarius* in a

simplified community

Boxplots of total cell count and cell count per species of a simplified community (*E. coli*, *S. salivarius*

and *P. fluorescens*) determined after 3, 5, 7 and 24h of growth. Cell counts of the communities grown

with the supplemented probiotic species *L. casei* or *S. thermophilus* are depicted in different colors.

Control or probiotic supplemented communities grown with 100 µM kynurenine are shown with

stripes. Statistical differences between communities were determined by ordinary two way ANOVA

approach per timepoint, and significant differences (q < 0.05) after multiple comparisons testing by

controlling the false discovery rate according to the Benjamini-Hochberg procedure are indicated

with asterisks. Probiotic addition altered the cell count of *E. coli* after 3 and 5h of growth, and the cell

count of *S. salivarius* after 7h of growth (ordinary two-way ANOVA: $F_{(2,12)}$ = 4.43 and P = 0.036,

$F_{(2,12)}$ = 4.15, P = 0.043 and $F_{(2,12)}$ = 7.62, P = 0.007 respectively), but multiple comparisons testing

did not yield any significant differences. Cell counts of added probiotic species grown with and

without kynurenine per timepoint were compared by performing an unpaired t test.

Figure 3: Probiotic and kynurenine supplementation alter the metabolism of a simplified

community in a specific manner

A) PCA with binned ^{1}H-NMR spectra as input showing separation of metabolites with lactate (bin

366,368 and 369), acetate (bin 485 and 486), glucose (bin 868), and betaine (bin 880) being the most

distinguishing factors (indicated with arrows). The ellipses represent the 95 % confidence interval. B)

PLS-DA with binned ^{1}H-NMR spectra of control and *L. casei* supplemented communities, indicating

lactate (bin 923), glucose (bin 851 and 853), and threonine (bin 820) as contributing factors. C) PLS-

DA with binned ^{1}H-NMR spectra of *S. thermophilus* and *L. casei* supplemented communities

indicating glucose (bin 848, 851 and 853 – 856), glycerol, (bin 815) and cadaverine (bin 707) as

contributing factors. Statistical parameter accuracy, R2 and Q2 and *P*-values are included. The *P*-

value is the results of a permutation test performed with the separation distance statistic and 1000

permutations. All parameters were calculated using MetaboAnalyst 5.0. D) Dynamic profiles of the

significantly different saccharolytic and proteolytic metabolites with repeated measure two-way

ANOVA showing significant differences. The error bars indicate the standard deviation. The

timepoints where a significant result of multiple comparison testing by controlling the false discovery

rate according to the Benjamini-Hochberg procedure is obtained are indicated with an asterisk; *: q <

0.05, detailed results of the multiple comparison tests can be found in the supplementary data 3. E)

Dynamic correlation-based networks of all measured metabolites in the communities showing

positive and negative correlations via red solid and blue dashed edges, respectively. Nodes represent

metabolites. Edges represents correlations between the two connected nodes, obtained when 2/3

methods give a positive results; Kendall's -0.9 > τ > 0.9, Spearman's -0.9 > ρ > 0.9. Brown's

randomization method with 1000 iterations had a Benjamini-Hochberg corrected *P*-value < 0.05. The

calculation of the edges was performed using the Cytoscape plugin CoNet.

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