

ORIGINAL ARTICLE

Gut microbiota is associated with dietary intake and metabolic markers in healthy individuals

Line Gaundal¹, Mari C.W. Myhrstad¹, Ida Rud², Terje Gjøvaag³, Marte G. Byfuglien⁴, Kjetil Retterstøl^{5,6}, Kirsten B. Holven^{5,6}, Stine M. Ulven⁵ and Vibeke H. Telle-Hansen^{1*}

¹Department of Nursing and Health Promotion, Faculty of Health Sciences, Oslo Metropolitan University, Oslo, Norway; ²Nofima AS (Norwegian Institute of Food, Fisheries and Aquaculture Research), Ås, Norway; ³Department of Occupational Therapy, Prosthetics and Orthotics, Oslo Metropolitan University, Oslo, Norway; ⁴Mills AS, Oslo, Norway; ⁵Department of Nutrition, Institute of Basic Medical Sciences, Faculty of Medicine, University of Oslo, Blindern, Oslo, Norway; ⁶The Norwegian National Advisory Unit on Familial Hypercholesterolemia, Department of Endocrinology, Morbid Obesity and Preventive Medicine, Oslo University Hospital, Rikshospitalet, Oslo, Norway

Popular scientific summary

- Identifying gut bacteria associated with both diet and metabolic risk markers may be a potential strategy in dietary disease prevention.
- We explored the relationship between a panel of gut bacteria, diet, and metabolic and anthropometric markers in healthy adults.
- *Bacteroides stercoris* was associated with higher intake of healthy foods and lower diastolic blood pressure.
- Further studies are needed to address whether these findings are causally linked and whether targeting these bacteria can prevent metabolic diseases.

Abstract

Background: Metabolic diseases have been related to gut microbiota, and new knowledge indicates that diet impacts host metabolism through the gut microbiota. Identifying specific gut bacteria associated with both diet and metabolic risk markers may be a potential strategy for future dietary disease prevention. However, studies investigating the association between the gut microbiota, diet, and metabolic markers in healthy individuals are scarce.

Objective: We explored the relationship between a panel of gut bacteria, dietary intake, and metabolic and anthropometric markers in healthy adults.

Design: Forty-nine volunteers were included in this cross-sectional study. Measures of glucose, serum triglyceride, total cholesterol, hemoglobin A1c (HbA1c), blood pressure (BP), and body mass index (BMI) were collected after an overnight fast, in addition to fecal samples for gut microbiota analyzes using a targeted approach with a panel of 48 bacterial DNA probes and assessment of dietary intake by a Food Frequency Questionnaire (FFQ). Correlations between gut bacteria, dietary intake, and metabolic and anthropometric markers were assessed by Pearson's correlation. Gut bacteria varying according to dietary intake and metabolic markers were assessed by a linear regression model and adjusted for age, sex, and BMI.

Results: Of the 48 gut bacteria measured, 24 and 16 bacteria correlated significantly with dietary intake and metabolic and/or anthropometric markers, respectively. Gut bacteria including *Alistipes*, *Lactobacillus* spp., and *Bacteroides stercoris* differed according to the intake of the food components, fiber, sodium, saturated fatty acids, and dietary indices, and metabolic markers (BP and total cholesterol) after adjustments. Notably, *Bacteroides stercoris* correlated positively with the intake of fiber, grain products, and vegetables, and higher *Bacteroides stercoris* abundance was associated with higher adherence to Healthy Nordic Food Index (HNFI) and lower diastolic BP after adjustment.

Conclusion: Our findings highlight the relationship between the gut microbiota, diet, and metabolic markers in healthy individuals. Further investigations are needed to address whether these findings are causally linked and whether targeting these gut bacteria can prevent metabolic diseases.

Keywords: gut microbiota; metabolic markers; diet; healthy; humans; dietary fiber; vegetables; dietary fat; blood pressure; cholesterol

To access the supplementary material, please visit the article landing page

Received: 3 February 2022; Revised: 18 March 2022; Accepted: 4 May 2022; Published: 23 June 2022

New knowledge has pointed out gut microbiota as a mediator of dietary impact on host metabolism (1). The gut microbiota possesses important functions for the host, such as fermentation of dietary fibers and extraction of nutrients, synthesis of certain vitamins, improvement of gut integrity and the protection against pathogens, and the regulation of host immune system and host signaling pathways (2). It can therefore be regarded as a metabolically active organ complementing the host metabolism through its traits (3, 4). In healthy individuals, the gut microbiota comprises a wide range of bacteria belonging to the two major phyla: Firmicutes and Bacteroidetes (5–7). Changes in the gut microbiota composition may disrupt normal functions, and a wide spectrum of diseases, including obesity, type 2 diabetes (T2D), and cardiovascular diseases (CVDs) are associated with gut microbiota dysbiosis (4, 5, 8, 9). Dysbiosis is often referred to as general changes in the gut microbiota composition (10). For example, individuals with lower microbiota gene content are characterized by higher insulin resistance, adiposity, and dyslipidemia than those with higher bacterial richness (11). Furthermore, a promising role of the gut microbiota affecting human metabolism has been shown after microbiota transplant, in which fecal transplantation from lean donors to recipients with metabolic syndrome showed increased gut microbiota diversity and butyrate-producing bacteria, together with improvements in insulin sensitivity (12). The gut microbiota can therefore be an attractive future target for the prevention of metabolic diseases (4).

Both genetic and environmental factors shape the gut microbiota, in which the latter seem to dominate (13–15). Diet is one of the most important factors influencing the gut microbiota composition and function (16–18). The supply of dietary compounds such as non-digestible carbohydrates, protein, and fat strongly impacts the formation of microbiota-derived metabolites (19). A large body of evidence suggest that the beneficial effect of dietary fiber is mediated by the microbial formation of short chain fatty acids (SCFAs), mainly acetate, propionate, and butyrate (20). SCFAs have been shown to regulate metabolic processes, including glucose and lipid metabolism, and are implicated in the regulation of blood pressure (BP) (4, 9, 18, 19, 21). Furthermore, the capability of dietary fats to alter host metabolism through the gut microbiota

composition has also been shown (22, 23). Diet-derived free fatty acids may have antimicrobial effects or be used as substrates for the formation of microbiota-derived metabolites, which are involved in inflammatory processes and the regulation of glucose and lipid homeostasis (22, 23). The type of fat also seems to have distinct effects on host metabolism through the gut microbiota. Studies in mice show that dietary fat quality affects inflammation and insulin sensitivity through effects via the gut microbiota composition (24, 25).

Even though studies have shown associations between diet and gut microbiota (26–31), studies investigating metabolic markers associated with both gut microbiota and diet in healthy adults are scarce (32–35). Identifying specific gut bacteria associated with both diet and metabolic markers may be valuable and could represent a potential strategy for the prevention of disease. The aim of this study was therefore to explore the relationship between a panel of gut bacteria, dietary intake, and metabolic and anthropometric markers in a cross-sectional study in healthy individuals.

Materials and methods

Subjects and study design

Healthy participants originally recruited to a randomized controlled trial (RCT) (36) were included in this exploratory cross-sectional study. The participants who met to a screening visit prior to the randomized trial performed at Oslo Metropolitan University (OsloMet) were included in this study. Healthy volunteers (aged 18–65 years) with body mass index (BMI) between 18.5 and 27.0 kg/m² were recruited from advertisement and from the student mass and employees at OsloMet between April 2018 and January 2019. Seventy-two volunteers were assessed for eligibility. The exclusion criteria were fasting blood glucose values ≥ 6.1 mmol/L, any food allergies, intolerances, chronic metabolic diseases (e.g. diabetes, CVD, and cancer), or intestinal diseases, including inflammatory bowel disease (IBD), celiac disease, and irritable bowel syndrome (IBS). Those treated with antibiotics the previous 3 months, who were blood donors the previous 2 months, who were pregnant or lactating, who had $\geq 5\%$ weight change the previous 3 months, who used any hormonal treatment (except from oral contraception) and tobacco,

or who had a high alcohol consume (>40 g/day) were excluded. All participants signed a written informed consent form prior to participation. The study was approved by the Regional Committees for Medical and Health Research Ethics (2018/104) and conducted according to the Declaration of Helsinki guidelines. The study was registered in clinicaltrials.gov (NCT03658681).

Blood sampling and clinical assessment

Blood sampling and clinical assessment were performed after an overnight fast (≥ 12 h). Fasting glucose was measured in finger prick capillary blood samples using a HemoCue Glucose 201 Analyser and Micro cuvettes (HemoCue, USA). Fasting triglyceride and total cholesterol were measured in serum and sent to a routine laboratory (Fürst Medical Laboratory) within 24 h. EDTA (Ethylene-Diamine-Tetra-Acetic acid) tubes with whole blood were kept in room temperature for ≤ 24 h before analyzation of hemoglobin A1c (HbA1c) (Fürst Medical Laboratory). BP was measured twice with an automated BP monitor (ri-champion[®], Riester, Germany) after sitting still for at least 15 min. Participants were asked to relax during the measurements. The mean value from both measurements was used in the statistical analyses. Body weight and composition were measured after an overnight fast (≥ 12 h) using Bioimpedance analyzers (BC-418 Segmental Body Composition Analyser and InBody 720). One kg was subtracted compensating for clothing. Height was measured using a wall-mounted stadiometer (Acumed).

Habitual dietary intake and dietary patterns

Forty-eight of the 49 participants included in this study completed a validated food frequency questionnaire (FFQ), reporting their habitual dietary intake the previous 12 months (37). The FFQ consisted of 270 food items and included questions about frequency of intake (from several times a day to never) and portions size based on household units (slices, glasses, cups, pieces, spoons and teaspoons) (37). Food groups included in the FFQ consisted of categories such as grain products; bread and vegetables; fruits and berries; nuts, olives, and seeds; meat, blood, and offal; fish and shellfish; cheese; and butter, margarine, and oil. Information about the consumption of these food groups and intake of specific foods within these categories (for example, intake of apples within the food group fruits and berries) was used to calculate adherence to dietary indices and to assess correlation with the gut microbiota and metabolic and anthropometric markers.

The dietary indices, the Healthy Nordic Food Index (HNFI) and the Healthy Diet Score (HDS), were used to assess correlation between adherence to the indices and the gut microbiota. The HNFI includes six food groups (fish, cabbage, apple and pears, root vegetables, rye bread,

and oatmeal) (38) (Supplementary Table 1). Rye bread was replaced with whole grain bread because information on rye bread specifically was not included in the FFQ, similar to previous studies (39, 40). Thus, in this study, the six food groups included in the modified HNFI were fish (cod, pollack, salmon, trout, herring, mackerel, and shellfish), cabbage (cabbage, cauliflower, broccoli, Chinese cabbage, and Brussels sprouts), apple and pears, root vegetables (carrot and rutabaga), oatmeal, and whole grain bread ($>50\%$ whole meal flour). One point was given for intake equal to or above the sex-specific median intake for each food group, and zero points were given for intake below the median. The maximum total score was six points. The total score was summarized for each participant, and a high score (4–6 points) indicates high adherence to the HNFI (38–40).

The HDS reflects adherence to the Norwegian dietary recommendations (41). In this study, the food groups included in the HDS were whole grain, vegetables, fruits and berries, low fat milk, fish, beans and lentils, vegetable oils (monounsaturated fatty acids (MUFAs) and polyunsaturated fatty acids (PUFAs)), red/processed meats, salt, sugar, and saturated fatty acids (SFAs). Each participant was given scores (0, 5, and 10 points) for the intake of food groups according to the Norwegian dietary recommendation and summed into an HDS (Supplementary Table 2). A high score (65–120 points) indicates high adherence to the recommendation, with a maximum score of 120 points.

Findings from a population-based study in Norway show that intakes of dietary fiber, SFA, salt, and sodium are not according to the recommendations, where the intake of fiber is less than recommended and intakes of SFA, sodium, and salt are higher than recommended (42). We therefore assessed whether the intake of these dietary components according to the recommendations was associated with the gut microbiota. Participants were divided into groups based on the average recommendations of fiber set by NNR (Nordic Nutrition Recommendations) (43) ($</\geq 30$ g/day, $n = 19/29$). In addition, participants were divided into groups based on recommended intake of SFA ($\leq/ > 10$ E%/day, $n = 9/39$), salt ($\leq/ > 5.0$ g/day, $n = 13/35$), and sodium ($\leq/ > 2.3$ g/day, $n = 23/25$). Participants were also divided into groups based on high or low adherence to the indices HNFI (0–3/4–6 points, $n = 30/18$) and HDS (0–60/65–120 points, $n = 10/38$).

Fecal collection and gut microbiota analyses

The participants were provided with a fecal sample collection kit for home use (GA-map[™] Dysbiosis Test, Genetic Analysis AS, Oslo, Norway). Participants were instructed to sample the stool from three different places and place it in the included tubes. Samples were kept in room temperature for maximum 3 days according to the manufacturers protocol. The GA-map[™] Dysbiosis Test

has shown stability in room temperature up to 5 days (44). All samples were stored at -80°C at OsloMet before they were collectively sent to Genetic Analysis AS (GA) for microbiota analyses after this study was completed.

The GA-map™ Dysbiosis Test is a commercially available genome-based test using fecal samples for analyses of gut bacteria associated with dysbiosis, described in detail elsewhere (44). In brief, the test was developed to identify and characterize bacterial groups that were able to distinguish patients with IBS and IBD from healthy controls. The test comprises 48 DNA probes targeting ≥ 300 bacteria on different taxonomic levels. Probes targeting seven variable regions (V3–V9) of the 16S rRNA gene were used to characterize and identify bacteria present, thus allowing mapping of the intestinal microbiota profile for a selected set of bacteria. Human fecal sample homogenization, and mechanical and enzymatic bacterial cell disruption were utilized to isolate and bind total bacterial genomic DNA to magnetic beads. The hypervariable regions V3–V9 of the 16S rRNA were further amplified by polymerase chain reaction. Single nucleotide extension and hybridization to a complementary DNA strand coupled to beads determined bacterial DNA labeling. To assess the abundance of bacteria, the strength of fluorescent signal (probe intensity) was detected and measured by Luminex 200 (Luminex Corporation). Probes are listed in Supplementary Table 3.

Statistical analysis

The current cross-sectional study is part of a randomized controlled dietary crossover study, with the primary aim to investigate the effect of fat quality on glycemic regulation, as described previously (36). Thus, this study is an exploratory study investigating the relationship between a panel of gut bacteria, dietary intake, and metabolic and anthropometric markers in healthy individuals. Gut bacteria were \log_2 -transformed before analysis. Correlations between gut bacteria, dietary variables, and metabolic and anthropometric markers were assessed with Pearson's correlation. Only bacteria showing statistically significant correlations with correlation coefficients ≥ 0.3 with either dietary intake or anthropometric and/or metabolic markers are presented. Differences in daily intake of dietary nutrients and dietary index score between males and females were assessed by independent sample *t*-test. Participants were divided into groups based on whether the intake was in accordance with the dietary recommendations or not. Differences between \log_2 -transformed gut bacteria based on whether the dietary intake (fiber, sodium, salt, and SFA [E%], and adherence to HNFI and HDS) and metabolic markers (systolic and diastolic BPs and total cholesterol) were below or above the recommended level were thereafter assessed by a linear regression model, hereafter called unadjusted model. Further adjustments

for age, sex, and BMI were performed in adjusted models. $P < 0.05$ was regarded as statistically significant. Statistical analyses were performed using IBM SPSS statistic (version 25), and figures were designed using GraphPad Prism 8 for Windows (version 8.0.0.).

Results

Seventy-two volunteers were assessed for eligibility, and 49 participants (12 males and 37 females) were included in this cross-sectional study. The mean age was 35.6 years (standard deviation [SD] 13.1 years), and measures of BMI, HbA1c, fasting glucose, triglyceride, total cholesterol, and diastolic and systolic BP were within the normal range (Table 1).

Data on dietary intake were collected by an FFQ reflecting dietary intake the past 12 months (Table 2). Mean daily intake of protein, total fat, MUFA, PUFA, and carbohydrates was 16.0 E%, 36.8 E%, 14.4 E%, 6.7 E%, and 41.4 E%, respectively. The intake of protein, total fat, MUFA, PUFA, and carbohydrates was within the recommended levels, whereas the intake of SFA was higher than recommended (12.5 E%) (45). The intake of fiber, 41.7 and 39.7 g/day for males and females, respectively, was higher than the recommended minimum daily intake, that is, 35 and 25 g/day, respectively (Table 2). Significant gender differences between dietary intake were found only for the intake of starch in g ($P = 0.005$) and E% ($P = 0.014$). For the HNFI and HDS indices, 18 (37.5%) and 38 (79.2%) participants had high adherence to the indices, respectively.

The abundance of the 48 gut bacteria probes included in this study represents bacteria belonging to the phyla Actinobacteria, Bacteroidetes, Firmicutes, Proteobacteria, Tenericutes, and Verrucomicrobia (Supplementary Table 3). Twenty-four bacteria significantly correlated (correlation coefficient ≥ 0.3) to one or more dietary

Table 1. Characteristics of the participants

Anthropometric and biochemical variables	Mean	SD
Female, <i>n</i> (%)	37 (75.5)	
Age (years)	35.6	13.1
BMI (kg/m ²)	22.8	2.2
HbA1c (%)	5.2	0.3
HbA1c (mmol/mol)	33.9	3.4
Glucose (mmol/L)	5.1	0.4
Triglyceride (mmol/L)	0.9	0.4
Total cholesterol (mmol/L)	4.8	0.9
Systolic blood pressure (mmHg) ¹	122.4	15.3
Diastolic blood pressure (mmHg) ¹	71.7	10.7

Variables are measured fasted.

¹Expressed as mean values based on two measurements.

Table 2. Daily intake of nutrients and dietary index score assessed by FFQ

Dietary variables	Total (n = 48)		Male (n = 12)		Female (n = 36)	
	Mean g or score ¹ (SD)	Mean E% (SD)	Mean g or score ¹ (SD)	Mean E% (SD)	Mean g or score ¹ (SD)	Mean E% (SD)
kj	11301.6 (5576.6)		11964.0 (4364.9)		11080.9 (5964.2)	
kcal	2701.2 (1332.8)		2859.5 (1043.2)		2648.4 (1425.5)	
Protein	104.3 (41.7)	16.0 (2.5)	108.7 (36.7)	15.5 (2.3)	102.8 (43.7)	16.2 (2.6)
Alcohol	10.6 (10.0)	2.8 (2.5)	13.2 (12.7)	3.1 (6.1)	9.7 (9.0)	2.8 (2.5)
Total fat	112.6 (67.8)	36.8 (7.3)	113.1 (48.7)	35.3 (2.7)	112.5 (73.7)	37.3 (7.7)
SFA	37.8 (22.7)	12.5 (3.4)	39.9 (24.8)	12.0 (3.8)	37.2 (22.2)	12.6 (3.4)
MUFA	44.4 (29.1)	14.4 (3.3)	42.7 (17.1)	13.5 (2.9)	45.0 (32.3)	14.8 (3.4)
PUFA	20.8 (15.1)	6.7 (2.0)	20.5 (7.6)	6.7 (2.0)	20.9 (16.9)	6.7 (2.0)
Trans-fat	1.0 (0.7)	0.3 (0.2)	1.2 (1.1)	0.4 (0.2)	0.9 (0.6)	0.3 (0.1)
Cholesterol (mg)	310.3 (142.2)		317.3 (174.2)		307.9 (132.9)	
Carbohydrates	278.4 (147.5)	41.4 (7.2)	306.8 (114.1)	43.1 (5.4)	268.9 (157.3)	40.8 (7.7)
Fiber	40.2 (25.8)	2.9 (0.6)	41.7 (18.4)	2.9 (0.7)	39.7 (28.0)	2.9 (0.6)
Starch*	133.0 (61.3)	20.3 (5.3)	170.2 (70.8)	23.9 (5.3)	120.6 (53.3)	19.0 (4.8)
Mono- and disaccharides	126.8 (92.0)	18.3 (6.1)	119.9 (51.6)	16.8 (5.0)	129.1 (102.5)	18.9 (6.4)
Sugar	37.0 (34.1)	5.5 (3.2)	35.1 (22.2)	5.1 (2.7)	37.7 (37.5)	5.6 (3.3)
Salt	6.8 (2.8)		7.0 (3.0)		6.7 (2.8)	
HNFI	3.1 (1.3)		3.2 (1.2)		3.0 (1.3)	
HDS	73.8 (13.9)		70.0 (13.1)		75.0 (14.0)	

¹Mean score of the adapted version of Healthy Nordic Food Index (HNFI) and Healthy Diet Score (HDS). n = 48.

*Significant difference in dietary intakes between males and females, assessed by independent sample t-test, P < 0.05.

variables as assessed with Pearson's correlation. These bacteria and their correlations to dietary variables are illustrated in Fig. 1. Eleven bacteria were related to one or more macronutrients, 12 bacteria were related to one or more micronutrients, and 16 bacteria were related to one or more food groups or dietary indices (Fig. 1).

As shown in Fig. 1, the bacteria correlating with the most dietary variables were *Bacteroides stercoris*, *Clostridia*, and *Streptococcus* spp., showing correlations with 12, 10, and eight dietary variables, respectively. The strongest positive correlation was between *Streptococcus* spp. and the intake of bread with <50% whole flour ($r: 0.655$, $P < 0.001$), whereas the strongest negative correlation was between *Bacteroides* spp. & *Prevotella* spp. and the intake of sugar (E%) ($r: -0.488$, $P < 0.001$). The intake of fiber and fiber-rich foods such as grain products and vegetables, and micronutrients commonly found in these food groups (tocopherol, thiamine, niacin, vitamin B6, folate, and phosphorus) correlated positively with *Bacteroides stercoris* (Fig. 1).

Correlations between gut bacteria and metabolic and anthropometric markers were thereafter investigated. Sixteen gut bacteria showed a statistically significant correlation with metabolic and/or anthropometric markers (correlation coefficients ≥ 0.3), as outlined in Fig. 2. Of these, two bacteria from the Bacteroidetes phylum (*Alistipes onderdonkii* and *Parabacteroides* spp.) and

nine bacteria from the Firmicutes phylum (*Lachnospiraceae*, *Bacilli*, *Ruminococcus albus* & *R. bromii*, *Lactobacillus* spp., *Eubacterium bifforme*, *Eubacterium rectale*, *Streptococcus salivarius* spp. *thermophilus* & *S. sanguinis*, and *Dialister invisus* & *Megasphaera micronuciformis*) correlated significantly to metabolic and/or anthropometric markers. In addition, Actinobacteria and *Bifidobacterium* spp. belonging to the Actinobacteria phylum, and *Akkermansia muciniphila* and *Enterobacteriaceae* from the phyla Verrucomicrobia and Proteobacteria, respectively, correlated to metabolic and/or anthropometric markers. The strongest positive correlation was between *Alistipes onderdonkii* and body fat (%) ($r: 0.428$, $P = 0.002$), and the strongest negative correlation was between *Lactobacillus* spp. and diastolic BP ($r: -0.411$, $P = 0.003$). Furthermore, BP correlated with several gut bacteria showing a positive correlation with Actinobacteria and *Bifidobacterium* spp. and a negative correlation with *Parabacteroides* spp., *Bacilli*, *Eubacterium bifforme*, and *Lactobacillus* spp. The abundance of *Lachnospiraceae*, *Bacilli*, *Alistipes onderdonkii*, *Dialister invisus*, and *Megasphaera micronuciformis* were significantly correlated to blood lipids (total cholesterol and/or triglyceride) (Fig. 2).

Based on the correlation analysis shown in Figs. 1 and 2, significant correlations between gut bacteria and both dietary variables and metabolic and anthropometric

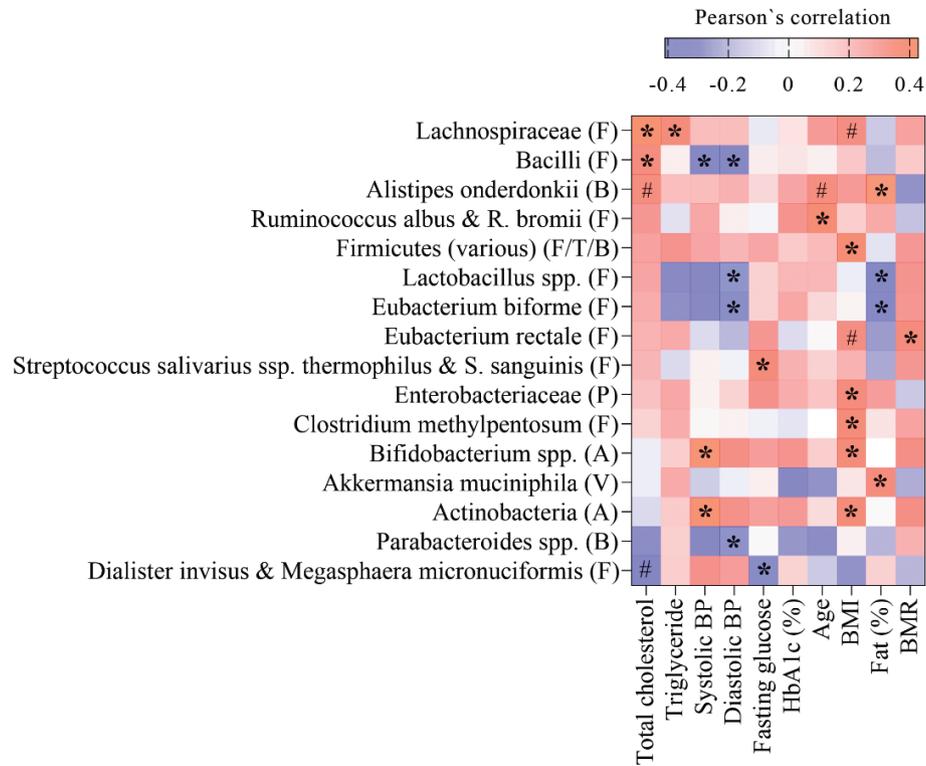


Fig. 2. Heat map of Pearson's coefficient between gut bacterial taxa and metabolic and anthropometric markers. The bacterial taxa are sorted from negative (blue) to positive (red) correlation toward total cholesterol levels, assessed by Pearson's correlation. Significant correlations are marked by * (correlation coefficient ≥ 0.3) or # (correlation coefficient < 0.3). Phylum is indicated within parentheses; A, Actinobacteria; B, Bacteroidetes; F, Firmicutes; P, Proteobacteria; T, Tenericutes; V, Verrucomicrobia; BMI, body mass index; BMR, basal metabolic rate; BP, blood pressure; HbA1c, hemoglobin A1c.

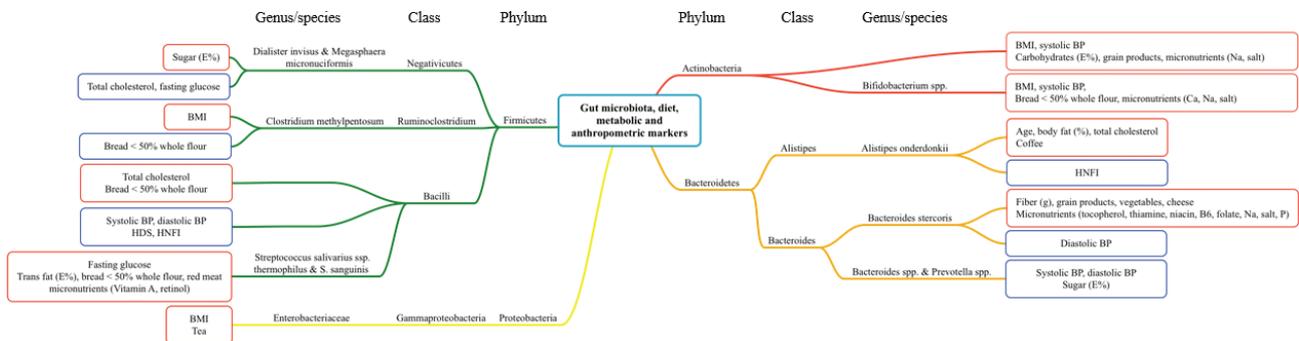


Fig. 3. Correlation map between gut bacteria, dietary intake, and metabolic and anthropometric markers. The gut bacteria are grouped according to representative phyla, indicated by different colors (red: Actinobacteria; orange: Bacteroidetes; green: Firmicutes; yellow: Proteobacteria). Correlation between the gut bacteria with one or more dietary variables, metabolic and/or anthropometric markers is illustrated by red boxes (positive correlation) or blue boxes (negative correlation). BMI, body mass index; BMR, basal metabolic rate; BP, blood pressure; E%, percentage of total energy intake; HDS, healthy diet score; HNFI, Healthy Nordic food index.

HNFI (≥ 4 points) was significantly associated with higher abundance of *Bacteroides stercoris* and lower abundance of *Bacilli*, *Lactobacillus* spp., *Eubacterium bifforme*, and *Streptococcus salivarius* ssp. *thermophilus*, while a higher adherence to the HDS (≥ 65 points) was associated with lower abundance of *Lactobacillus* spp. 2 after adjusting for age, sex, and BMI (Table 3).

Differences in the abundance of gut bacteria between participants stratified according to their systolic BP ($< / \geq 120$ mmHg) ($n = 25/24$), diastolic BP ($< / \geq 80$ mmHg) ($n = 39/10$), and total cholesterol levels ($< / \geq 5.0$ mmol/L) ($n = 29/20$) were thereafter investigated. Gut bacteria showing a significant difference in abundance between groups in the linear regression analyses are shown in

Table 3. Gut bacteria associated with the intake of fiber, sodium, salt, SFA (E%), and adherence to HNFI and HDS indices¹

Gut bacteria	Unadjusted values		Adjusted values		P [‡]	P [§]
	B	95% CI	B	95% CI		
	Fiber intake (≥ 30 g/d) (n = 29)					
(B) <i>Alistipes</i>	-0.306	-1.697, -0.068	-0.301	-1.730, -0.008	0.034	0.048
	Sodium intake (> 2.3 g/d) (n = 25)					
(B) <i>Bacteroides fragilis</i>	-0.297	-2.672, -0.063	-0.250	-2.495, 0.198	0.040	0.093
(B) <i>Bacteroides stercoris</i>	0.367	0.325, 2.303	0.329	0.118, 2.241	0.010	0.030
	Salt intake (> 5.0 g/d) (n = 35)					
(P) <i>Proteobacteria</i>	0.314	0.087, 1.595	0.288	-0.044, 1.585	0.030	0.063
	SFA intake (≥ 10 E%/d) (n = 20)					
(A) <i>Actinomycetales</i>	-0.385	-1.928, -0.325	-0.404	-2.079, -0.283	0.007	0.011
(F) <i>Firmicutes</i>	-0.300	-0.766, -0.022	-0.319	-0.839, 0.001	0.038	0.050
(F) <i>Ruminococcus gnavus</i>	0.204	-0.264, 1.520	0.238	0.050, 1.966	0.163	0.040
(F) <i>Phascolarctobacterium</i> sp.	0.246	-0.202, 2.620	0.331	0.056, 3.187	0.091	0.043
(T) <i>Mycoplasma hominis</i>	-0.221	-2.175, 0.289	-0.327	-2.716, -0.076	0.130	0.040
	High adherence to HNFI (≥ 4 points) (n = 18)					
(B) <i>Bacteroides stercoris</i>	0.333	0.198, 2.266	0.317	0.114, 2.229	0.021	0.031
(F) <i>Bacilli</i>	-0.338	-1.056, -0.182	-0.420	-1.123, -0.219	0.006	0.005
(F) <i>Lactobacillus</i> spp.	-0.260	-2.481, 0.122	-0.332	-2.721, -0.291	0.075	0.016
(F) <i>Eubacterium bifforme</i>	-0.276	-2.639, 0.041	-0.345	-2.905, -0.335	0.057	0.015
(F) <i>Streptococcus salivarius</i> spp. <i>thermophilus</i>	-0.294	-1.556, -0.029	-0.329	-1.673, -0.098	0.042	0.028
	High adherence to HDS (≥ 65 points) (n = 38)					
(F) <i>Lactobacillus</i> spp. 2	-0.383	-1.680, -0.278	-0.337	-1.618, -0.107	0.007	0.026

¹Gut bacteria values were log₂-transformed before analysis.

[‡]P for unadjusted values assessed by a linear regression model.

[§]P for values adjusted for age, sex, and BMI, assessed by a linear regression model.

Phyla are indicated within parentheses; (A), Actinobacteria; (B), Bacteroidetes; (F), Firmicutes; (P), Proteobacteria; (T), Tenericutes

The level of significance was set at $P < 0.05$ and are indicated in bold italic.

Table 4. A higher systolic BP (≥120 mmHg) was associated with lower abundance of *Lactobacillus* spp. after adjusting for age, sex, and BMI (Table 4). A higher diastolic BP (≥80 mmHg) was associated with lower abundance of *Bacteroides stercoris*, *Bacteroides* spp., *Bacilli*, *Eubacterium bifforme*, *Eubacterium rectale*, *Lactobacillus* spp., and *Streptococcus* spp. 2, and higher abundance of *Dialister invisus* and *Megasphaera micronuciformis*, after adjusting for age, sex, and BMI. A higher total cholesterol level (≥5.0 mmol/L) was associated with higher abundance of *Ruminococcus albus* and *Ruminococcus bromii*, but this association was no longer significant after adjusting for age, sex, and BMI (Table 4).

Discussion

In this study, we explored the relationship between a panel of gut bacteria commonly found in the human gut, dietary intake, and metabolic and anthropometric markers in healthy adults. Of the 48 gut bacteria analyzed in this study, 24 bacteria were shown to correlate with dietary intake and 16 bacteria correlated with metabolic and/or anthropometric markers. Several of the gut bacteria

correlated with both dietary intake and metabolic and/or anthropometric markers. Furthermore, we show that specific gut bacteria differed in relation to the intake of food components such as fiber, sodium, SFA, and healthy food indices, and between participants stratified according to BP and total cholesterol.

The abundance of *Bacteroides stercoris* was positively associated with higher adherence to the HNFI index and negatively associated with a higher diastolic BP after adjusting for age, sex, and BMI. Moreover, *Bacteroides stercoris* correlated positively with the intake of healthy foods and food components, including fiber, grain products, and vegetables. Dietary carbohydrates provide important substrates for microbial metabolism in the human gut. Many members of the *Bacteroides* genus are enriched with genes encoding carbohydrate-active enzymes and are involved in the breakdown of complex carbohydrates from the diet (46–48). As such, *Bacteroides* members are the predominant organisms involved in carbohydrate metabolism and are considered generalists due to their capacity to switch between host and diet-derived energy sources (19). Furthermore, *Bacteroides*

Table 4. Gut bacteria associated with metabolic markers¹

Gut bacteria	Unadjusted values		Adjusted values		P [#]	P [#]
	B	95% CI	B	95% CI		
Systolic BP (≥ 120 mmHg) (n = 24)						
(A) <i>Actinobacteria</i>	0.285	0.014, 2.028	0.238	-0.196, 1.903	0.047	0.108
(A) <i>Bifidobacterium</i>	0.305	0.097, 2.201	0.251	-0.152, 2.045	0.033	0.090
(F) <i>Lactobacillus</i> spp.	-0.229	-2.269, 0.251	-0.309	-2.583, -0.140	0.114	0.030
Diastolic BP (≥ 80 mmHg) (n = 10)						
(B) <i>Bacteroides stercoris</i>	-0.300	-2.570, -0.087	-0.320	-2.718, -0.123	0.036	0.033
(B) <i>Bacteroides</i> spp. and <i>Prevotella</i> spp.	-0.288	-1.259, -0.015	-0.225	-1.163, 0.166	0.045	0.138
(B) <i>Bacteroides</i> spp.	-0.279	-2.348, 0.010	-0.335	-2.635, -0.166	0.052	0.027
(F) <i>Bacilli</i>	-0.350	-1.286, -0.155	-0.387	-1.404, -0.189	0.014	0.011
(F) <i>Dialister invisus</i> and <i>Megasphaera micronuciformis</i>	0.330	0.363, 4.138	0.384	0.644, 4.584	0.020	0.010
(F) <i>Eubacterium bifforme</i>	-0.345	-3.487, -0.389	-0.313	-3.361, -0.157	0.015	0.032
(F) <i>Eubacterium rectale</i>	-0.325	-2.661, -0.209	-0.357	-2.814, -0.343	0.023	0.013
(F) <i>Lactobacillus</i> spp.	-0.378	-3.553, -0.580	-0.348	-3.400, -0.407	0.007	0.014
(F) <i>Streptococcus</i> spp. 2	-0.294	-1.217, -0.029	-0.305	-1.284, -0.007	0.040	0.048
Total cholesterol (≥ 5.0 mmol/L) (n = 20)						
(F) <i>Ruminococcus albus</i> and <i>R. bromii</i>	0.374	0.433, 2.739	0.245	-0.579, 2.659	0.008	0.202

¹Gut bacteria values were log-transformed before analysis.

#P for unadjusted values assessed by a linear regression model.

#P for values adjusted for age, sex, and BMI, assessed by a linear regression model.

Phyla are indicated within parentheses; A, Actinobacteria; B, Bacteroidetes; F, Firmicutes.

The level of significance was set at $P < 0.05$ and are indicated in bold italic.

stercoris has been identified as part of a common set of microbial species referred to as a common bacterial core, which are largely shared between individuals (7). Interestingly, *Bacteroides stercoris* has been postulated as a keystone species of the human gut microbiome influencing the microbial community structure including the growth of butyrate-producing bacteria (49).

Higher abundance of *Bacteroides stercoris* in healthy individuals compared with patients with IBD has been shown (50), suggesting a potential beneficial impact of the presence of *Bacteroides stercoris* (51). Taken together, these findings indicate a potential protective role of *Bacteroides stercoris* on human health. However, these studies cannot establish causality and the role of *Bacteroides stercoris* for health and disease needs to be further elucidated in intervention studies.

A higher intake of fiber was associated with lower abundance of *Alistipes* after adjusting for age, sex, and BMI in this study. Interestingly, diets low in fiber and high in fats have shown to increase the abundance of *Alistipes* (52, 53). Whether these differences are related to metabolic regulation by diet needs further investigation, as the impact of *Alistipes* has been reported as both protective and detrimental on CVD (54). In addition to *Alistipes*, members of the Proteobacteria phylum are reported to increase with the presence of fat, indicating that these members utilize dietary fats for growth (52). Here,

we show that Proteobacteria abundance correlated positively with the intake of dietary cholesterol and cheese. A high-cholesterol diet was recently shown to increase Proteobacteria abundance in zebrafish (55), and high-fat feeding resulted in higher abundance of Proteobacteria in mice (56–58). Furthermore, Proteobacteria has been reported to increase in low-grade inflammation, a common feature of metabolic diseases (59). *Enterobacteriaceae*, a member of the Proteobacteria phylum, correlated positively with BMI in this study. In line with our results, *Enterobacteriaceae* has been associated with obesity (60), while weight-loss has been shown to reduce the abundance of *Enterobacteriaceae* (61–63). Taken together, these findings may indicate that dietary components such as fiber and fat modulate the gut microbiota composition, which in turn may impact host metabolic regulation.

In addition to *Bacteroides stercoris*, we show that a lower abundance of bacteria belonging to the Firmicutes phylum, specifically *Bacilli*, *Eubacterium bifforme*, *Eubacterium rectale*, *Streptococcus* spp. 2, and *Lactobacillus* spp., was associated with higher diastolic BP after adjusting for age, sex, and BMI. Moreover, a higher systolic BP was also associated with lower abundance of *Lactobacillus* spp. after the same adjustments. The relationship between *Lactobacillus* abundance and BP has also been reported by others. Palmu and colleagues recently demonstrated a strong negative correlation

between certain *Lactobacillus* species and BP, and sodium intake in a Finish cohort including 6,953 participants (64). In a study by Wilck et al., a high salt diet was shown to increase BP and reduce the abundance of *Lactobacillus* species in mice (65). In this study, the administration of *Lactobacillus* species to mice resulted in a reduced systolic BP and normalization of diastolic BP. The authors further tested this effect in a pilot study in healthy males receiving 6 g sodium for 14 days, which led to a higher BP and a loss of *Lactobacillus* species (65). The relationship between *Lactobacillus* and BP was also evaluated in a meta-analysis of RCTs including 702 individuals (66). The effect of probiotic fermented milk, commonly adding *Lactobacillus* species for the fermentation purpose, reduced systolic and diastolic BP with 3.10 and 1.09 mmHg, respectively, compared with placebo, and the effect on systolic BP was suggested to be even more effective in hypertensive individuals (66). Another meta-analysis of RCTs with 543 participants also showed an effect of probiotic consumption on BP (67). The intake of probiotics, including probiotics from dairy, reduced systolic BP with 3.56 mmHg and diastolic BP with 2.38 mmHg compared with control groups. Interestingly, this effect was found to be similar to the effect of salt reduction and resistance training (67). Furthermore, pre-hypertensive and hypertensive individuals displayed a reduced gut microbiota diversity and richness compared with healthy controls (68). Furthermore, transplantation of fecal bacteria from the hypertensive donors to germ-free mice increased BP and reduced microbiota diversity in hypertensive mice compared with control mice (68). These findings may, therefore, indicate that the gut microbiota, in particular the abundance of *Lactobacillus*, may be involved in the regulation of BP.

A positive correlation between *Lachnospiraceae* and fasting total cholesterol and triglyceride levels was found in this study. The gut microbiota has been reported to play an important role in the variation of blood lipid levels in humans (69). A positive correlation between cholesterol levels and *Lachnospiraceae* has been shown previously (70), while others show an inverse relationship (34, 71, 72). *Lachnospiraceae* is a heterogenic family comprised of taxa reported as both potentially beneficial and harmful (73, 74). For example, differences in *Lachnospiraceae* members between healthy controls and patients with T2D and metabolic syndrome have been shown (75, 76), and different members have been related to either higher or lower lifetime CVD risk (77). The heterogeneity of *Lachnospiraceae* emphasizes these discrepancies, and investigating gut microbes in lower taxonomic levels may give more insight to the potential role of *Lachnospiraceae* in host lipid metabolism.

In this study, we demonstrate associations between a panel of gut bacteria, dietary intake, and metabolic and

anthropometric markers in healthy adults. This study is explorative by nature, and the results obtained may be used to further generate new hypothesis. Hence, we did not control for multiple testing. In this study, the sample size, the inclusion of relatively young adults (mean age 35.6 years), and mostly women (75.5%) in a Norwegian population residing in the Oslo area represent limitations. In addition, dietary data were collected by a FFQ and showed a high fiber intake (41.7 g/day for males and 39.7 g/day for females). In comparison, findings from the Norwegian Women and cancer cohort, also using FFQ data, reported fiber intakes of 20 g/day (78). Hence, our findings may not be generalized to other populations in different geographical regions as the gut microbiota has been shown to differ among age, gender, and geographical regions (79–81). The findings in this study may, however, highlight the relationship between a high fiber intake and the gut microbiota.

Using a targeted method to detect only 48 bacteria may be criticized, as other bacteria than those analyzed may have an impact on metabolic regulation. However, the method provides a rapid, high-throughput analysis of human fecal samples allowing researchers to assess fecal microbiota composition without advanced expertise. Here, we explored a panel of bacteria commonly found in the human gut, shown to differ between healthy controls and patients with IBD and IBS, characterized as dysbiotic (44). Although no clear consensus about the definition of dysbiosis in the literature currently exist, dysbiosis often refers to as general changes in the gut microbiota composition, an imbalance in composition, or as changes in specific taxa in that composition (10). Here, we further investigate the associations of the selected set of gut bacteria related to dysbiosis with diet and metabolic markers in healthy adults. Identifying potentially harmful or beneficial bacteria by targeted methods may, in the future, be useful to predict disease risk and/or be used as a therapeutic potential for treating disorders. However, the lack of measures of microbial metabolites such as SCFAs limits our possibility to investigate the functionality of the gut microbiota and its relationship with dietary intake and metabolic and anthropometric markers. A complete characterization of the gut microbiota and measures of microbial metabolites are encouraged in future studies. Our findings may provide useful information about the association between a panel of gut bacteria, dietary intake, and metabolic and anthropometric markers in healthy adults. The cross-sectional design does not allow us to establish causality, and we encourage more and larger randomized controlled studies to further investigate this relationship.

Conclusion

In this study, we demonstrate associations between a panel of gut bacteria, dietary intake, and metabolic and

anthropometric markers in healthy individuals. Of special interest was the abundance of *Bacteroides stercoris*, which correlated positively with the intake of fiber. Further adjustments for age, sex, and BMI showed that *Bacteroides stercoris* was positively associated with higher adherence to the HNFI index and negatively associated with higher diastolic BP. These findings may indicate a role for the gut microbiota in metabolic regulation through diet, or that the gut microbiota may reflect a healthy versus unhealthy lifestyle. Whether we can modify the gut microbiota by diet and consequently impact metabolic status remains to be elucidated.

Data described in the manuscript will be made available upon request pending application and approval.

Conflict of interest and funding

Funding for this research was provided by OsloMet, University of Oslo, Genetic Analysis, and Mills AS. Mills AS had no role in the analysis of the data. L.G., I.R., and T.G. report nothing to declare. M.G.B. is employed at Mills AS. She does not own any stocks in the company. K.R. reports research grants and/or personal fees from Akcea, Amgen, Sanofi, and Sunnovion, and none of which are related to the content of this manuscript. K.B.H. reports grants from Tine SA, Mills AS, Olympic Seafood, Amgen, Sanofi, and Kaneka, and personal fees from Amgen, Sanofi, and Pronova, outside the submitted work. S.M.U. has received research grants from Tine DA, Mills AS, and Olympic Seafood, and none of which are related to the content of this manuscript. M.C.W.M. is involved in projects with research grants from Tine SA and Olympic Seafood and has received research fund from Mills AS, and none of which are related to the content of this manuscript. V.H.T.-H. reports research grants from Mesterbakereen, Det Glutenfrie Verksted, and Norsk cøliakiforening, and none of which are related to the content of this manuscript. She has been employed at Mills AS. She does not own any stocks in the company, and the work performed in this paper was done after she left the company.

Acknowledgments

We are grateful to all volunteers who participated in this study, and we thank the bioengineer Ellen Rael for her contribution in the conduction of this study.

Authors' contributions

The author contributions were as follows: L.G., M.C.W.M., I.R., T.G., K.R., K.B.H., S.M.U., and V.H.T.H. designed the research; L.G., M.C.W.M., and V.H.T.H. conducted the research; L.G., M.C.W.M., and V.H.T.H. analyzed data and performed statistical analysis; L.G., M.C.W.M., and V.H.T.H. wrote the manuscript; L.G., M.C.W.M.,

I.R., T.G., M.G.B., K.R., K.B.H., S.M.U., and V.H.T.H. provided critical editorial comments; L.G., M.C.W.M., and V.H.T.H. had primary responsibility for final content; and all authors read and approved the final manuscript.

References

1. Sonnenburg JL, Backhed F. Diet-microbiota interactions as moderators of human metabolism. *Nature* 2016; 535: 56–64. doi: 10.1038/nature18846
2. Thursby E, Juge N. Introduction to the human gut microbiota. *Biochem J* 2017; 474: 1823–36. doi: 10.1042/BCJ20160510
3. O'Hara AM, Shanahan F. The gut flora as a forgotten organ. *EMBO Rep* 2006; 7: 688–93. doi: 10.1038/sj.embor.7400731
4. Koh A, Bäckhed F. From association to causality: the role of the gut microbiota and its functional products on host metabolism. *Mol Cell* 2020; 78: 584–96. doi: 10.1016/j.molcel.2020.03.005
5. Arora T, Backhed F. The gut microbiota and metabolic disease: current understanding and future perspectives. *J Intern Med* 2016; 280: 339–49. doi: 10.1111/joim.12508
6. Eckburg PB, Bik EM, Bernstein CN, Purdom E, Dethlefsen L, Sargent M, et al. Diversity of the human intestinal microbial flora. *Science (New York, N.Y.)* 2005; 308: 1635–8. doi: 10.1126/science.1110591
7. Qin J, Li R, Raes J, Arumugam M, Burgdorf KS, Manichanh C, et al. A human gut microbial gene catalogue established by metagenomic sequencing. *Nature* 2010; 464: 59–65. doi: 10.1038/nature08821
8. Hansen TH, Gøbel RJ, Hansen T, Pedersen O. The gut microbiome in cardio-metabolic health. *Genome Med* 2015; 7: 33. doi: 10.1186/s13073-015-0157-z
9. Hur KY, Lee M-S. Gut microbiota and metabolic disorders. *Diabetes Metab J* 2015; 39: 198–203. doi: 10.4093/dmj.2015.39.3.198
10. Hooks KB, O'Malley MA, Davies JE. Dysbiosis and its discontents. *mBio* 2017; 8: e01492-17. doi: 10.1128/mBio.01492-17
11. Le Chatelier E, Nielsen T, Qin J, Prifti E, Hildebrand F, Falony G, et al. Richness of human gut microbiome correlates with metabolic markers. *Nature* 2013; 500: 541–6. doi: 10.1038/nature12506
12. Vrieze A, Van Nood E, Holleman F, Salojarvi J, Kootte RS, Bartelsman JF, et al. Transfer of intestinal microbiota from lean donors increases insulin sensitivity in individuals with metabolic syndrome. *Gastroenterology* 2012; 143: 913–6.e7. doi: 10.1053/j.gastro.2012.06.031
13. Rothschild D, Weissbrod O, Barkan E, Kurilshikov A, Korem T, Zeevi D, et al. Environment dominates over host genetics in shaping human gut microbiota. *Nature* 2018; 555: 210–5. doi: 10.1038/nature25973
14. Yatsunenkov T, Rey FE, Manary MJ, Trehan I, Dominguez-Bello MG, Contreras M, et al. Human gut microbiome viewed across age and geography. *Nature* 2012; 486: 222–7. doi: 10.1038/nature11053
15. Goodrich JK, Waters JL, Poole AC, Sutter JL, Koren O, Blekhman R, et al. Human genetics shape the gut microbiome. *Cell* 2014; 159: 789–99. doi: 10.1016/j.cell.2014.09.053
16. Graf D, Di Cagno R, Fåk F, Flint HJ, Nyman M, Saarela M, et al. Contribution of diet to the composition of the human gut microbiota. *Microb Ecol Health Dis* 2015; 26: 26164. doi: 10.3402/mehd.v26.26164
17. Wu GD, Chen J, Hoffmann C, Bittinger K, Chen YY, Keilbaugh SA, et al. Linking long-term dietary patterns with gut microbial

- enterotypes. *Science* (New York, N.Y.) 2011; 334: 105–8. doi: 10.1126/science.1208344
18. Valdes AM, Walter J, Segal E, Spector TD. Role of the gut microbiota in nutrition and health. *BMJ* (Clinical research ed.) 2018; 361: k2179. doi: 10.1136/bmj.k2179
 19. Flint HJ, Duncan SH, Scott KP, Louis P. Links between diet, gut microbiota composition and gut metabolism. *Proc Nutr Soc* 2015; 74: 13–22. doi: 10.1017/S0029665114001463
 20. Koh A, De Vadder F, Kovatcheva-Datchary P, Bäckhed F. From dietary fiber to host physiology: short-chain fatty acids as key bacterial metabolites. *Cell* 2016; 165: 1332–45. doi: 10.1016/j.cell.2016.05.041
 21. Miyamoto J, Kasubuchi M, Nakajima A, Irie J, Itoh H, Kimura I. The role of short-chain fatty acid on blood pressure regulation. *Curr Opin Nephrol Hypertens* 2016; 25: 379–83. doi: 10.1097/mnh.0000000000000246
 22. Schoeler M, Caesar R. Dietary lipids, gut microbiota and lipid metabolism. *Rev Endocr Metab Disord* 2019; 20: 461–72. doi: 10.1007/s11154-019-09512-0
 23. Candido FG, Valente FX, Grzeskowiak LM, Moreira APB, Rocha D, Alfenas RCG. Impact of dietary fat on gut microbiota and low-grade systemic inflammation: mechanisms and clinical implications on obesity. *Int J Food Sci Nutr* 2017, 69, 1–19. doi: 10.1080/09637486.2017.1343286
 24. Caesar R, Tremaroli V, Kovatcheva-Datchary P, Cani PD, Bäckhed F. Crosstalk between gut microbiota and dietary lipids aggravates WAT inflammation through TLR signaling. *Cell Metab* 2015; 22: 658–68. doi: 10.1016/j.cmet.2015.07.026
 25. Cao W, Liu F, Li RW, Chin Y, Wang Y, Xue C, et al. Docosahexaenoic acid-rich fish oil prevented insulin resistance by modulating gut microbiome and promoting colonic peptide YY expression in diet-induced obesity mice. *Food Sci Hum Wellness* 2022; 11: 177–88. doi: 10.1016/j.fshw.2021.07.018
 26. Falony G, Joossens M, Vieira-Silva S, Wang J, Darzi Y, Faust K, et al. Population-level analysis of gut microbiome variation. *Science* (New York, N.Y.) 2016; 352: 560–4. doi: 10.1126/science.aad3503
 27. Zhernakova A, Kurilshikov A, Bonder MJ, Tigchelaar EF, Schirmer M, Vatanen T, et al. Population-based metagenomics analysis reveals markers for gut microbiome composition and diversity. *Science* (New York, N.Y.) 2016; 352: 565–9. doi: 10.1126/science.aad3369
 28. Partula V, Mondot S, Torres MJ, Kesse-Guyot E, Deschasaux M, Assmann K, et al. Associations between usual diet and gut microbiota composition: results from the Milieu Intérieur cross-sectional study. *Am J Clin Nutr* 2019; 109: 1472–83. doi: 10.1093/ajcn/nqz029
 29. Noh H, Jang H-H, Kim G, Zouiouich S, Cho S-Y, Kim H-J, et al. Taxonomic composition and diversity of the gut microbiota in relation to habitual dietary intake in Korean adults. *Nutrients* 2021; 13: 366. doi: 10.3390/nu13020366
 30. Trefflich I, Jabakhanji A, Menzel J, Blaut M, Michalsen A, Lampen A, et al. Is a vegan or a vegetarian diet associated with the microbiota composition in the gut? Results of a new cross-sectional study and systematic review. *Crit Rev Food Sci Nutr* 2020; 60: 2990–3004. doi: 10.1080/10408398.2019.1676697
 31. De Filippis F, Pellegrini N, Vannini L, Jeffery IB, La Stora A, Laghi L, et al. High-level adherence to a Mediterranean diet beneficially impacts the gut microbiota and associated metabolome. *Gut* 2016; 65: 1812–21. doi: 10.1136/gutjnl-2015-309957
 32. Asnicar F, Berry SE, Valdes AM, Nguyen LH, Piccinno G, Drew DA, et al. Microbiome connections with host metabolism and habitual diet from 1,098 deeply phenotyped individuals. *Nat Med* 2021; 27(2): 321–332. doi: 10.1038/s41591-020-01183-8
 33. Ma W, Nguyen LH, Song M, Wang DD, Franzosa EA, Cao Y, et al. Dietary fiber intake, the gut microbiome, and chronic systemic inflammation in a cohort of adult men. *Genome Med* 2021; 13: 102. doi: 10.1186/s13073-021-00921-y
 34. Companys J, Gosalbes MJ, Pla-Pagà L, Calderón-Pérez L, Llauroadó E, Pedret A, et al. Gut microbiota profile and its association with clinical variables and dietary intake in overweight/obese and lean subjects: a cross-sectional study. *Nutrients* 2021; 13: 2032. doi: 10.3390/nu13062032
 35. Li Y, Wang DD, Satija A, Ivey KL, Li J, Wilkinson JE, et al. Plant-based diet index and metabolic risk in men: exploring the role of the gut microbiome. *J Nutr* 2021; 151: 2780–9. doi: 10.1093/jn/nxab175
 36. Gaundal L, Myhrstad MCW, Leder L, Byfuglien MG, Gjøvaag T, Rud I, et al. Beneficial effect on serum cholesterol levels, but not glycaemic regulation, after replacing SFA with PUFA for 3 d: a randomised crossover trial. *Br J Nutr* 2020; 125, 1–11. doi: 10.1017/S0007114520003402
 37. Carlsen MH, Lillegaard ITL, Karlsen A, Blomhoff R, Drevon CA, Andersen LF. Evaluation of energy and dietary intake estimates from a food frequency questionnaire using independent energy expenditure measurement and weighed food records. *Nutr J* 2010; 9: 37. doi: 10.1186/1475-2891-9-37
 38. Olsen A, Egeberg R, Halkjær J, Christensen J, Overvad K, Tjønneland A. Healthy aspects of the nordic diet are related to lower total mortality. *J Nutr* 2011; 141: 639–44. doi: 10.3945/jn.110.131375
 39. Roswall N, Sandin S, Löf M, Skeie G, Olsen A, Adami H-O, et al. Adherence to the healthy Nordic food index and total and cause-specific mortality among Swedish women. *Eur J Epidemiol* 2015; 30: 509–17. doi: 10.1007/s10654-015-0021-x
 40. Puaschitz NG, Assmus J, Strand E, Karlsson T, Vinknes KJ, Lysne V, et al. Adherence to the Healthy Nordic Food Index and the incidence of acute myocardial infarction and mortality among patients with stable angina pectoris. *J Hum Nutr Dietet* 2019; 32: 86–97. doi: 10.1111/jhn.12592
 41. Garnweidner-Holme L, Torheim LE, Henriksen L, Borgen I, Holmelid S, Lukasse M. Adherence to the Norwegian dietary recommendations in a multi-ethnic pregnant population prior to being diagnosed with gestational diabetes mellitus. *Food Sci Nutr* 2020; 8: 3031–40. doi: 10.1002/fsn3.1248
 42. Totland THM, Kjerpeseth B, Lundberg-Hallén N, Helland-Kigen KM, Lund-Blix NA, Myhre JB, et al. Norkost 3 En landsomfattende kostholdsundersøkelse blant menn og kvinner i Norge i alderen 18-70 år, 2010-11. Oslo. Helsedirektoratet; 2012.
 43. NNR. Nordic nutrition recommendations 2012: integrating nutrition and physical activity; 9289326700. Copenhagen: Nordic Council of Ministers; 2014.
 44. Casén C, Vebo HC, Sekelja M, Hegge FT, Karlsson MK, Ciemniejewska E, et al. Deviations in human gut microbiota: a novel diagnostic test for determining dysbiosis in patients with IBS or IBD. *Aliment Pharmacol Ther* 2015; 42: 71–83. doi: 10.1111/apt.13236
 45. Nasjonalt råd for ernæring. Kostråd for å fremme folkehelsen og forebygge kroniske sykdommer: Metodologi og vitenskapelig kunnskapsgrunnlag. Oslo, Helsedirektoratet, 2011.
 46. Hollister EB, Gao C, Versalovic J. Compositional and functional features of the gastrointestinal microbiome and their effects on human health. *Gastroenterology* 2014; 146: 1449–58. doi: 10.1053/j.gastro.2014.01.052

47. Jandhyala SM, Talukdar R, Subramanyam C, Vuyyuru H, Sasi-kala M, Nageshwar Reddy D. Role of the normal gut micro-biota. *World J Gastroenterol* 2015; 21: 8787–803. doi: 10.3748/wjg.v21.i29.8787
48. Sorbara MT, Pamer EG. Microbiome-based therapeu-tics. *Nat Rev Microbiol* 2022, 20, 365–380. doi: 10.1038/s41579-021-00667-9
49. Fisher CK, Mehta P. Identifying keystone species in the human gut microbiome from metagenomic timeseries using sparse linear regression. *PLoS One* 2014; 9: e102451. doi: 10.1371/journal.pone.0102451
50. Strömbeck A, Lason A, Strid H, Sundin J, Stotzer P-O, Simrén M, et al. Fecal microbiota composition is linked to the postoperative disease course in patients with Crohn's disease. *BMC Gastroenterol* 2020; 20: 130. doi: 10.1186/s12876-020-01281-4
51. Nomura K, Ishikawa D, Okahara K, Ito S, Haga K, Takahashi M, et al. Bacteroidetes species are correlated with disease activity in ulcerative colitis. *J Clin Med* 2021; 10: 1749. doi: 10.3390/jcm10081749
52. Agans R, Gordon A, Kramer DL, Perez-Burillo S, Ru-fián-Henares JA, Paliy O, et al. Dietary fatty acids sustain the growth of the human gut microbiota. *Appl Environ Microbiol* 2018; 84: e01525-18. doi: 10.1128/AEM.01525-18
53. David LA, Maurice CF, Carmody RN, Gootenberg DB, But-ton JE, Wolfe BE, et al. Diet rapidly and reproducibly alters the human gut microbiome. *Nature* 2014; 505: 559–63. doi: 10.1038/nature12820
54. Parker BJ, Wearsch PA, Veloo ACM, Rodriguez-Palacios A. The genus *alisticipes*: gut bacteria with emerging implications to inflammation, cancer, and mental health. *Front Immunol* 2020; 11: 906. doi: 10.3389/fimmu.2020.00906
55. Kong Y, Li Y, Dai Z-R, Qin M, Fan H-L, Hao J-G, et al. Gly-cosaminoglycan from *Ostrea rivularis* attenuates hyperlipid-emia and regulates gut microbiota in high-cholesterol diet-fed zebrafish. *Food Sci Nutr* 2021; 9: 5198–210. doi: 10.1002/fsn3.2492
56. Choi Y, Kwon Y, Kim D-K, Jeon J, Jang SC, Wang T, et al. Gut microbe-derived extracellular vesicles induce insulin resistance, thereby impairing glucose metabolism in skeletal muscle. *Sci Rep* 2015; 5: 15878. doi: 10.1038/srep15878
57. Jeong M-Y, Jang H-M, Kim D-H. High-fat diet causes psychi-atric disorders in mice by increasing Proteobacteria population. *Neurosci Lett* 2019; 698: 51–7. doi: 10.1016/j.neulet.2019.01.006
58. Hildebrandt MA, Hoffmann C, Sherrill-Mix SA, Keilbaugh SA, Hamady M, Chen YY, et al. High-fat diet determines the com-position of the murine gut microbiome independently of obe-sity. *Gastroenterology* 2009; 137: 1716–24.e1–2. doi: 10.1053/j.gastro.2009.08.042
59. Rizzatti G, Lopetuso LR, Gibiino G, Binda C, Gasbarrini A. Proteobacteria: a common factor in human diseases. *BioMed Res Int* 2017; 2017: 9351507. doi: 10.1155/2017/9351507
60. Peters BA, Shapiro JA, Church TR, Miller G, Trinh-Shevrin C, Yuen E, et al. A taxonomic signature of obesity in a large study of American adults. *Sci Rep* 2018; 8: 9749. doi: 10.1038/s41598-018-28126-1
61. Xiao S, Fei N, Pang X, Shen J, Wang L, Zhang B, et al. A gut microbiota-targeted dietary intervention for amelioration of chronic inflammation underlying metabolic syndrome. *FEMS Microbiol Ecol* 2014; 87: 357–67. doi: 10.1111/1574-6941.12228
62. Sotos M, Nadal I, Marti A, Martínez A, Martín-Ma-tillas M, Campoy C, et al. Gut microbes and obesity in adolescents. *Proceed Nutr Soc* 2008; 67: E20. doi: 10.1017/S0029665108006290
63. Fei N, Zhao L. An opportunistic pathogen isolated from the gut of an obese human causes obesity in germfree mice. *ISME J* 2013; 7: 880–4. doi: 10.1038/ismej.2012.153
64. Palmu J, Salosensaari A, Havulinna AS, Cheng S, Inouye M, Jain M, et al. Association between the gut microbiota and blood pressure in a population cohort of 6953 indi-viduals. *J Am Heart Assoc* 2020; 9: e016641. doi: 10.1161/JAHA.120.016641
65. Wilck N, Matus MG, Kearney SM, Olesen SW, Forslund K, Bartolomeaus H, et al. Salt-responsive gut commensal mod-ulates TH17 axis and disease. *Nature* 2017; 551: 585–9. doi: 10.1038/nature24628
66. Dong J-Y, Szeto IMY, Makinen K, Gao Q, Wang J, Qin L-Q, et al. Effect of probiotic fermented milk on blood pressure: a meta-analysis of randomised controlled trials. *Br J Nutr* 2013; 110: 1188–94. doi: 10.1017/S0007114513001712
67. Khalesi S, Sun J, Buys N, Jayasinghe R. Effect of probiotics on blood pressure: a systematic review and meta-analysis of ran-domized, controlled trials. *Hypertension* 2014; 64: 897–903. doi: 10.1161/hypertensionaha.114.03469
68. Li J, Zhao F, Wang Y, Chen J, Tao J, Tian G, et al. Gut micro-biota dysbiosis contributes to the development of hypertension. *Microbiome* 2017; 5: 14. doi: 10.1186/s40168-016-0222-x
69. Fu J, Bonder MJ, Cenit MC, Tigchelaar EF, Maatman A, De-kens JA, et al. The gut microbiome contributes to a substantial proportion of the variation in blood lipids. *Circ Res* 2015; 117: 817–24. doi: 10.1161/circresaha.115.306807
70. Koren O, Spor A, Felin J, Fåk F, Stombaugh J, Tremaroli V, et al. Human oral, gut, and plaque microbiota in patients with ath-erosclerosis. *Proc Natl Acad Sci* 2011; 108: 4592–8. doi: 10.1073/pnas.1011383107
71. Liu Y, Song X, Zhou H, Zhou X, Xia Y, Dong X, et al. Gut microbiome associates with lipid-lowering effect of rosu-vastatin in vivo. *Front Microbiol* 2018; 9: 530. doi: 10.3389/fmicb.2018.00530
72. Tindall AM, McLimans CJ, Petersen KS, Kris-Etherton PM, Lamendella R. Walnuts and vegetable oils containing oleic acid differentially affect the gut microbiota and associations with car-diovascular risk factors: follow-up of a randomized, controlled, feeding trial in adults at risk for cardiovascular disease. *J Nutr* 2020; 150: 806–17. doi: 10.1093/jn/nxz289
73. Vacca M, Celano G, Calabrese FM, Portincasa P, Gobetti M, De Angelis M. The controversial role of human gut lach-nospiraceae. *Microorganisms* 2020; 8: 573. doi: 10.3390/microorganisms8040573
74. Sorbara MT, Littmann ER, Fontana E, Moody TU, Kohout CE, Gjonbalaj M, et al. Functional and genomic variation be-tween human-derived isolates of lachnospiraceae reveals inter-and intra-species diversity. *Cell Host Microbe* 2020; 28: 134–46. e134. doi: 10.1016/j.chom.2020.05.005
75. Zhang X, Shen D, Fang Z, Jie Z, Qiu X, Zhang C, et al. Human gut microbiota changes reveal the progression of glucose in-tolerance. *PLoS One* 2013; 8: e71108. doi: 10.1371/journal.pone.0071108
76. Chávez-Carbajal A, Nirmalkar K, Pérez-Lizaur A, Hernán-dez-Quiroz F, Ramírez-Del-Alto S, García-Mena J, et al. Gut microbiota and predicted metabolic pathways in a sample of Mexican women affected by obesity and obesity plus met-abolic syndrome. *Int J Mol Sci* 2019; 20: 438. doi: 10.3390/ijms20020438

77. Kelly TN, Bazzano LA, Ajami NJ, He H, Zhao J, Petrosino JF, et al. Gut microbiome associates with lifetime cardiovascular disease risk profile among Bogalusa heart study participants. *Circ Res* 2016; 119: 956–64. doi: 10.1161/CIRCRESAHA.116.309219
78. Enget Jensen TM, Braaten T, Jacobsen BK, Barnung RB, Olsen A, Skeie G. Adherence to the Healthy Nordic Food Index in the Norwegian Women and Cancer (NOWAC) cohort. *Food Nutr Res* 2018; 62. doi: 10.29219/fnr.v62.1339
79. Huttenhower C, Gevers D, Knight R, Abubucker S, Badger JH, Chinwalla AT, et al. Structure, function and diversity of the healthy human microbiome. *Nature* 2012; 486: 207–14. doi: 10.1038/nature11234
80. Mueller S, Saunier K, Hanisch C, Norin E, Alm L, Midtvedt T, et al. Differences in fecal microbiota in different European study populations in relation to age, gender, and country: a cross-sectional study. *Appl Environ Microbiol* 2006; 72: 1027–33. doi: 10.1128/aem.72.2.1027-1033.2006
81. Haro C, Rangel-Zúñiga OA, Alcalá-Díaz JF, Gómez-Delgado F, Pérez-Martínez P, Delgado-Lista J, et al. Intestinal microbiota is influenced by gender and body mass index. *PLoS One* 2016; 11: e0154090. doi: 10.1371/journal.pone.0154090

***Vibeke H. Telle-Hansen**

Faculty of Health Sciences,
Oslo Metropolitan University,
Post box 4, St. Olavs plass, 0130 Oslo,
Norway.
Email: vtelle@oslomet.no