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Losing weight for a better health: role for the gut microbiota

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Abstract

In recent years, there have been several reviews on gut microbiota, obesity and cardiometabolism summarizing interventions that may impact the gut microbiota and have beneficial effects on the host (some examples include [1–3]). In this review we discuss how the gut microbiota changes with weight loss (WL) interventions in relation to clinical and dietary parameters. We also evaluate available evidence on the heterogeneity of response to these interventions. Two important questions were generated in this regard: 1) *Can response to an intervention be predicted?* 2) *Could pre-intervention modifications to the gut microbiota optimize WL and metabolic improvement?* Finally, we have delineated some recommendations for future research, such as the importance of assessment of diet and other environmental exposures in WL intervention studies, and the need to shift to more integrative approaches of data analysis.

WEIGHT LOSS INTERVENTIONS, HEALTH OUTCOMES AND THE ROLE FOR GUT MICROBIOTA

**Effect of calorie restriction on gut microbiota – can we predict host responses based on pre-intervention health status and microbiota composition?**

Several studies in animal models and humans have addressed the impact of WL through calorie restriction (CR) on microbiota composition and its association with clinical outcomes. Some of these studies have analyzed whether certain phenotypes before WL may impact or predict the effect of the intervention on health outcomes.
**Rodent models**

Studies in rodent models have shed light on the role that gut microbiota may be playing in obesity. It has been demonstrated in rodents that an obese phenotype can be transmitted via the microbiota. Gut microbiota, depending on its composition and function, may be involved in several mechanisms leading to fat mass gain and eventually obesity. Among the mechanism the role of energy harvest from food (shown to be more efficient in certain bacterial groups) has been proposed. Germ free mice are resistant to diet-induced obesity,[6,7] but gain weight upon transfer of gut microbiota from conventionally raised mice or ob/ob mice, potentially through increased capacity for energy harvest.[8] Gut microbiota may also impact host metabolism in the development of rodent obesity through the induction of hepatic lipogenesis, and suppression of Fiaf in the gut epithelia, leading to upregulation of LPL activity and increased fat storage.[6] There is also a direct interaction between the gut microbiota, the gut-associated immune system, and adipose tissue through metabolic endotoxemia.[9–11] Therefore, other effects such as the regulation of lipogenesis and gluconeogenesis, gut hormone secretion and induction of inflammatory response have also been demonstrated in rodents.[5] In addition, rodent models have been used to investigate the relationship between genetics and gut microbiota,[12] and these studies have shown that different genetic backgrounds can lead to very diverse host-environment interactions.

Gut microbiota changes due to CR can be significant and depend on the type of intervention. For example, duration of CR can impact both gut microbiota composition and health outcomes. Zhang et al. showed in mice that lifelong CR led to large and
consistent changes in gut microbiota composition.[13] In this study, there was lower midlife serum LPS binding protein (LBP, a surrogate of metabolic endotoxemia) in mice fed a low fat and calorie diet, as opposed to other dietary compositions. Phyla that inversely correlated with LBP were positively correlated with lifespan, emphasizing on the important of low-grade inflammation in this context.

**In humans**

Divergence in human gut microbiota composition is associated to multiple factors. Microbiota enterotypes have been defined in different populations around the world. Differentiation into these enterotypes cannot be explained by individual factors such as age or degree of corpulence, geographical location, or by dietary modifications of short duration.[14] Instead, long-term dietary habits and certain clinical characteristics seem to be stronger determinants for these compositional differences.[15]

Obese and non-obese subjects have a different gut microbial profile.[16–20] Ley et al. showed that obese subjects have lower *Bacteroidetes* to *Firmicutes* ratio than lean subjects.[8] However, these findings have not been consistent in the literature.[21] Another study showed greater abundance in the *Firmicutes* group *Eubacterium rectale / Clostridium coccoides* in obese women with metabolic syndrome versus obese women with no metabolic complications and non-obese women.[19] There was a correlation between this bacterial group and certain clinical outcomes such as visceral adiposity. These findings suggest a different energy harvesting potential, consistent with the capacity of *Firmicutes* species to degrade non-digestible polysaccharides, although this remain to be proven.
An important aspect of gut microbial composition in relation to host health is microbial richness, referring to diversity in the gut ecosystem. Microbial richness is overall higher in lean vs. obese subjects, and this correlates with a healthier metabolic profile.[16,22] However even in subjects with different corpulence (lean vs. obese), metagenomic sequencing has revealed that different patterns of low or high diversity exist. When considering abundance of individual species, higher abundance of certain species such as *Faecalibacterium prausnitzii* (*F. prausnitzii*)[16,23,24] and *Akkermansia muciniphila* (*A. muciniphila*)[25,26] have been repeatedly associated with a healthier status.

In CR studies there have been some consistent shifts in microbial composition. Interestingly, it appears that certain characteristics in the gut, together with diet, associate with individual response to CR and lifestyle interventions. Such baseline differences and varied outcomes have been identified in the MICRO-Obes study, where a population of 49 overweight and obese individuals has been thoroughly studied in terms of gut microbiota composition, clinical parameters, and dietary intake. It was first shown that these individuals could be clustered by their response profile to 6 weeks of CR followed by a 6 week weight stabilization period. There were baseline differences in clinical parameters and microbiota among the three WL response clusters. Namely, *Lactobacillus/Leuconostoc/Pediococcus* group, was most abundant at baseline in the cluster of worst responders to CR and WS. However, the response to the intervention could be better predicted by baseline insulin sensitivity and inflammatory parameters illustrating the fact that we need deeper insight into the predictive potential of gut microbiota in dietary intervention.[27]
More recently, it was shown in both the MICRO-Obes and MetaHIT studies that individuals can be stratified by their microbial richness, and those with higher richness (about 60-80%) tend to have a healthier metabolic status [22] and dietary intake.[28] MICRO-Obes subjects that had higher baseline microbial richness tended to respond better to the dietary intervention in terms of blood lipids, insulin sensitivity and low-grade inflammation.

Finally, as it will be described in more detail in the following section, higher baseline *A. muciniphila* was associated with a healthier metabolic profile in the same study.[26] Individuals with a higher baseline abundance of this species had better outcomes from the intervention, namely a greater reduction in waist circumference, blood lipids, and increase in insulin sensitivity. Individuals with higher *A. muciniphila* in the context of higher microbial richness were also the most metabolically healthy throughout the intervention, illustrating the importance to take into account the overall gut microbial ecosystem, rather than focusing solely on one species.

The functional capacity of the gut microbiota in CR can be studied through modelisation of metagenomic information and through direct measure of metabolites in fluids (metabolomics). In a randomized cross-over study comparing a 4-week high protein/low carbohydrate diet to a high protein/medium carbohydrate regime in obese men, a reduction in abundance of *Roseburia* spp. and *E. rectale*, as well as fecal butyrate, correlated with lower carbohydrate intake.[29] Total fecal short chain fatty acids (SCFA), acetate, propionate, isovalerate and valerate increased with higher carbohydrate intake. On the other hand, the high protein/low carbohydrate diet was characterized by a potentially deleterious fecal metabolite profile, high in branched chain
fatty acids, phenylacetic acid and \textit{N}-nitroso compounds.\cite{30} Similarly, another study in obese adults found lower fecal SCFA production in an 8-week low carbohydrate/high fat regime. This was accompanied by an exacerbation of bowel habits and a decrease in \textit{Bifidobacterium}.\cite{31}

CR interventions in obese adolescents have also demonstrated changes in microbial composition.\cite{32,33} Interestingly, baseline microbial composition differences were found between good (>4 kg WL) and bad (<2 kg WL) responders to CR, and changes in certain bacterial groups were associated with WL or improvement in clinical outcomes (\textbf{Table 1}).

Given the intricate relationship between the gut microbiota and host, a key question is \textit{whether modification of gut microbiota before interventions through diet and/or prebiotic treatment (defined later in this review) has the potential to optimize WL and metabolic improvement}. Studying baseline differences between responders and non-responders is key to answer this question (\textbf{Figure 1}).

In conclusion, baseline profiles in microbiota and metabolic status, together with dietary macronutrient intake, may play a role in outcomes from CR interventions. More detail is needed on the role of micronutrients. An interaction between diet and microbiota has been identified in the development of obesity in human-to-mouse microbial transplantation studies.\cite{34,35} This evidence shows the importance of analyzing diet in CR interventions. For the most part, intervention periods have lasted most commonly 1 to 3 months, with a few exceptions going up to 6 months. Longer follow up periods should be included in future studies.

While these studies have adequately phenotyped the changes in gut microbiota composition with dietary interventions, it is difficult to go beyond strong correlations and
elucidate mechanisms from these results. As shall be discussed in the last section, data integration approaches allow the simultaneous analysis of environment, gut microbiota and host, which may lead to the identification of mechanistic links and therapeutic targets.

**Effects of prebiotic and probiotic on host metabolism: putative links with gut microbes**

Numerous studies have demonstrated that manipulating the gut microbiota with dietary intervention (i.e., prebiotics and probiotics) may affect host metabolism (i.e., glucose, lipid and energy metabolism) (Figure 2). In this section, we briefly discuss examples showing the impact of such intervention in preclinical models as well as recent evidence suggesting that dietary interventions using pre and probiotics may also be linked with gut microbes in humans.

Twenty years ago, Gibson and Roberfroid have developed the prebiotic concept, recently revised as "A non digestible compound that, through its metabolization by microorganisms in the gut, modulates composition and/or activity of the gut microbiota, thus conferring a beneficial physiological effect on the host".[36] Over the last decades, this concept has led to the investigation of key questions such as how changes in the gut microbiota induced by prebiotics but also specific bacteria contribute to regulate energy intake, fat mass development and glucose/lipid metabolism? We will first discuss data obtained in rodents and in the second part the effectiveness of such interventions on human health.
Animal models

More than a decade ago, Cani et al. described that the three different prebiotics (i.e. inulin-type fructans, which varied according to their degree of polymerization (i.e., number of fructose moieties), differentially affected gut peptides secretion. They found that the administration of prebiotic compounds profoundly changes the gut microbiota composition and metabolic function contributing to the upregulation of two gut peptides involved in reduced food intake, namely Glucagon-like peptide-1 (GLP-1) and PYY, and a decreased plasma levels of the orexigenic peptide, ghrelin.[37,38] By using culture and non-culture dependent tools it has been shown that the three prebiotics used were able to change the gut microbiota in favor of *Bifidobacterium* spp. The abundance of *Bifidobacterium* spp. was inversely associated with body weight, fat mass as well as metabolic endotoxemia and inflammation.[39] More recently, thanks to metagenomics tools, novel results have clearly shown that the modulation of the gut microbiota was more complex than a simple change in *Bifidobacterium* spp., indeed, dozens of taxa were changed upon prebiotic treatment in obese and diabetic rodents.[40] Among the taxa increased by the treatment, *Akkermansia muciniphila* was increased by about 100 fold.[40] Interestingly, the abundance of this bacteria was positively associated with a lower fat mass, an improved glucose tolerance and gut barrier function as well as with the number of intestinal L cells secreting GLP-1 and PYY.[40] Since this discovery, several studies have shown that the administration of *Akkermansia muciniphila* in obese and diabetic rodents reduces fat mass gain, insulin resistance, metabolic endotoxemia and low grade inflammation,[12,41,42] thereby showing that this bacteria may play a crucial role. Although the overall mechanisms are not fully elucidated, this bacterium reinforced the gut barrier function and contribute to regulate energy homeostasis.[41]
Thus, taken together, a variety of rodent model studies indicate that prebiotics may elicit beneficial impacts in metabolic disorders associated with obesity and diabetes. Moreover, several studies indicate that some of these effects may be obtained with specific bacteria often misinterpreted as probiotic. Notably, the term probiotic is often misused (see the International Scientific Association for Probiotics and Prebiotics published a consensus statement clarifying the scope of and appropriate use for the term ‘probiotic’ (for a review, see [43]).

Besides this important opinion, various strains of *Lactobacillus* and *Bifidobacterium* have demonstrated beneficial effects, most of the time by maintaining glucose homeostasis and decreasing inflammation and hepatic steatosis. Importantly, some of these strains also affect body weight and fat mass development, whereas others do not (for comprehensive reviews on this topic).[44,45]

In summary, abundant literature have reported the impact of specific *Lactobacillus* or *Bifidobacterium* strains on obesity and associated disorders in rodents, however strains are not equally potent in terms of body weight and fat mass loss or improvement of glucose/lipid metabolism and inflammatory markers.

The following examples illustrate the concept that strains are not equipotent. *Lactobacillus gasseri* BNR17 reduces body weight and fat mass in overweight rats,[46] whereas in diet-induced obese mice, *Lactobacillus plantarum* 14 reduces the mean adipocyte size and *Lactobacillus paracasei* F19 induces a reduction of total fat mass and plasma triglycerides.[47] Conversely, *Lactobacillus acidophilus* NCDC supplementation did not affect body fat mass and/or hepatic steatosis and muscle fat in obese mice.[48] *Lactobacillus casei* Shirota reduces insulin resistance and metabolic endotoxemia, without affecting fat mass and body weight in diet-induced obese mice.[49] Finally,
Lactobacillus plantarum WCFS1 did not change body weight, fat mass or inflammation in diet-induced obese mice.\[41\] These examples clearly illustrate that although they are all Lactobacillus, specific strains are efficient on metabolic parameters whereas other not.

Similar to the Lactobacillus spp. examples, specific strains of Bifidobacterium have been shown to metabolic disorders in obese and diabetic models.\[44\] For example, a recent study has shown that Bifidobacterium pseudocatenulatum CECT 7765 reduces body weight gain, fat mass, plasma glucose and inflammation in diet-induced obese mice.\[50\] In a similar model, Bifidobacterium longum supplementation has been found to reduce body weight gain, fat mass, insulin resistance, systolic blood pressure, and metabolic endotoxemia.\[51\] Another study demonstrated that supplementation with Bifidobacterium animalis subsp lactis 420 reduced inflammation and improved insulin in obese and diabetic mice.\[52\] Again, these selected examples also illustrate that Bifidobacterium strains may affect metabolism, not always by inducing a body weight loss but most likely by improving intestinal barrier.

In humans

A limited number of studies have evaluated whether effects observed in rodents can similarly be achieved in humans. Among these studies, the impact of fermentable carbohydrates (including prebiotics) feeding on enteroendocrine hormones such as GLP-1, PYY and ghrelin, the reduced plasma glucose and inflammatory tone has been generally replicated in both healthy or obese humans,\[53–55\] however, the impact on fat
mass and body weight remain limited.[56] Interestingly, in these studies the gut microbiota composition was not studied, except in Dewulf et al. 2013, who shows that specific bacteria are positively and negatively correlated with fat mass, metabolic endotoxemia and glucose/lipid markers.[56]

A study using synbiotic approaches that is a supplementation with prebiotics and probiotic (inulin-type fructans and *Bifidobacterium longum*) has shown in 66 overweight patients with non-alcoholic steatohepatitis a reduced steatosis, metabolic endotoxemia, insulin resistance, and inflammation.[57] Excluding these studies using prebiotic supplementation, only few studies have reported a beneficial impact of probiotics on obesity and type 2 diabetes in humans, with again a certain strain specificity (for review[58]). More recently, similar to the results obtained in rodents, it has been shown that important variations of *Akkermansia muciniphila* quantity may be observed in the intestine of obese/overweight subjects. Although, no one knows with precision the level of *Akkermansia muciniphila* required to detect beneficial/healthy versus pathological situation, as discussed earlier in this review, Dao et al. have recently demonstrated in human that below a given fecal amount of *Akkermansia muciniphila* obese/overweight subjects were less disposed to respond to the beneficial effect of a caloric restriction diet in terms of improved cardiometabolic risk factors (i.e., plasma cholesterol, inflammation, insulin resistance and glycemia).[26]

**Bariatric surgery induces substantial shifts in gut microbiota composition**

Gut microbiota changes have been thoroughly assessed in bariatric interventions both in animal models and humans. In general, bariatric surgery leads to a dramatic
improvement of pre-surgical obesity co-morbidities, with some differences observed between the types of bariatric interventions. The gastric band, for example, leads to a more attenuated WL than sleeve, although they are both considered restrictive procedures. Roux-en-Y gastric bypass (RYGB) leads to the most important changes in health outcomes, potentially due to a change in the gut architecture and gut hormonal secretion, together with extensive WL (Table 2). This particular intervention causes greater improvements in type 2 diabetes and other obesity co-morbidities.[59] The effect of bariatric surgery on health has been extensively reviewed in previous publications. [60–63]

**Rodent models**

Studies in mice that have compared different bariatric surgery procedures with non-operated or SHAM operated mice have allowed the definition of surgery-specific changes in gut microbiota. Liou et al. compared mice that had undergone RYGB, non-operated controls weight matched to the RYGB group, and Sham operated mice fed a HFD *ad libitum.*[64] Gut microbial composition from Sham and weight-matched groups was different from that in the RYGB group. Of interest, among other phylogenetic changes, there was an increase in abundance of *Verrucomicrobia* (genus *Akkermansia*) and *Gammaproteobacteria* (genus *Escherichia*) with RYGB, which correlated with improved metabolic outcomes. Gut microbiota transfers (i.e. transfer of postsurgery caecal content) to germ-free mice led to weight improvement. This study showed that microbial changes in RYGB are due to gastrointestinal reconfiguration and not just to WL, changes in diet or intestinal transection. The RYGB group had the highest fecal energy output.
Vertical sleeve gastrectomy (VSG) is becoming popular practice in bariatric interventions. It was previously believed to be a purely restrictive procedure, but there is now evidence suggesting that several aspects of digestion, bile acid metabolism and gastrointestinal hormonal secretory profile are modified. To this point, it was recently published that circulating bile acids are altered in mice undergoing VSG, which was correlated with shifts in gut microbial composition.[65] Furthermore, knockout of the bile acid receptor FXR reduced WL and clinical improvement.

A recent study by Tremaroli et al. compared phenotypes in mice receiving fecal transfer from morbidly obese women, or women that had undergone either RYGB or vertical banded gastroplasty.[66] One unique feature of this study was that microbiota composition was studied long term, with fecal samples obtained 9 years after surgery, when the women were weight-stable. Changes in microbiota were not only maintained over time, they were also surgery-specific but independent of BMI. Even though the phenotype was transmitted from the two surgical groups to the mice, there were some functional and compositional differences in microbiota, such as higher *Proteobacteria* in the RYGB group, and lower abundance of *E. rectale* and *Roseburia intestinalis* in the sleeve group compared to the obese group. The fecal and circulating metabolite profiles were different between groups. This study provides compelling evidence of the role of microbiota in long term weight maintenance of bariatric patients.

*In humans*

The potential role of microbiota in human health improvement stemming from bariatric surgery has been recently summarized.[67,68] As in mouse studies, the composition of gut microbiota in humans is extensively changed with bariatric surgery
(Table 2). For example, Furet et al. showed important changes in microbiota measured with 16S qPCR, after bypass. This included an increase in *F. prausnitzii*, which was inversely associated with inflammation regardless of diet.[17] Later, Kong et al. published more detailed gut microbiota information on this group obtained with 16S pyrosequencing.[69] This analysis showed that microbial richness increased after RYGB, and that approximately half of the correlations seen between diet and gut microbiota could be explained by dietary intake.

Damms-Machado et al., compared the effect of a very low calorie diet (VLCD) to VSG over 6 months, with 3 patients per group. They saw a reduction in *Firmicutes* to *Bacteroidetes* ratio, less butyrate fermentation, and more NEFA and bile acid secretion in the VSG group.[70] The authors argue that the decrease in proportion of *Firmicutes* would account for the decrease capacity to ferment SCFA, leading to less calorie extraction from diet and therefore greater benefit from the intervention. It is difficult to link this to clinical outcomes because the VSG group was heavier at baseline than the VLCD group.

Other bariatric interventions have included a small number of subjects.[71–73] Their design has been either cross-sectional, or with short-term follow-up (Table 2). Some changes in gut microbiota have been consistent, such as a decrease in *Firmicutes* after surgery, increase *Proteobacteria* and a tendency towards an increase in *Verrucomicrobia* (*Akkermansia*).

Most importantly, very few bariatric intervention studies assessing microbiota have included dietary information and food intake behavior or other kinds of environmental exposures. Our group has recently reported that dietary quality in bariatric patients is poor, particularly protein intake.[74] In addition to change in food intake after
bariatric surgery, these subjects also receive protein supplementary that could impact on gut microbiota. Therefore, it will also be important to focus on dietary quality of bariatric patients before and after surgery to optimize response and increase the likelihood of a shift to a healthier gut microbiota.

Interpretation of microbial changes with human bariatric interventions need to be made with caution and with a thorough knowledge of the clinical background of the patients, as morbidly obese populations are usually taking multiple medications. The effect of polypharmacy, including metformin and other diabetes treatments, on the gut microbiota and its relation to health is only now being elucidated.[75,76]

INTEGRATION OF KNOWLEDGE AND POTENTIAL FOR FUTURE

Throughout this review we have discussed the interactions between three main elements: the host, the gut microbiota, and the environment. The advancement of available technologies for the assessment of gut microbiota is key in the work presented here. The field is shifting from targeted measurement of specific bacterial groups to a gut microbiota ecology approach. This is complementary to the thorough analysis of particular species of interest. With these advances in technology, microbiota will be more thoroughly characterized and quantified. This will include RNAseq and more detailed functional annotations. Other relevant measures include the gut environment, architecture and ecosystem, in conjunction with functional characteristics of the gut microbiota as a metabolic organ through the use of metabolomics.

From a clinical point of view, extensive phenotyping of populations is mandatory to identify subgroups that may be responding differently to an intervention. Indeed, even if a population seems uniform in terms of BMI, there is non-negligible heterogeneity in
body composition, which in turn would be associated with different profiles of metabolic health, as explained by Ahima and Lazar.[77] Clinical parameters, pathologies and other traits of the host must be studied in detail to identify subgroups that may respond differently to interventions.

Regarding the environment, there is a wide array of exposures influencing host and gut microbiota that are very difficult to measure. Diet is the factor with the greatest potential to influence the gut microbiota and, although it is often assessed, it is very difficult to measure it reliably. Dietary intake and habits should be routinely taken into consideration in the kinds of interventions we have covered in this review. At the same time, there are many other environmental factors that could be influencing microbiota, including drug intake, pollution and physical activity.

The gut microbiota is at the interphase between environment and host. It is important to study profiles from these three elements in parallel using data integration and systems biology approaches.[78,79] This would allow a more profound understanding of the factors that may be influencing, or may be influenced, by gut microbiota,[80] as well as differentiation of individual subpopulations that may undergo different responses after a WL intervention (Figure 1).

**Ecosystem modelisation: a first step toward truly personalized nutrition?**

An example of a potential approach for personalized improvement in metabolic status can be seen in the recently published work by Shoaie et al.[81] Given the complexity of the intestinal bacterial ecosystem characterized by microbe-microbe interactions, and interactions between microbes, the environment and the biology of the
host, informatics and mathematics experts have used novel approaches to model these interactions. These modelisation approaches aim at better understanding at the individual level the interactions between the microbiota ecosystem and dietary intake, and to infer the potential impact on metabolic health (Figure 3). As such, knowledge of the individual composition in gut microbiota lead to the identification of metabolites produced in excess or otherwise deficient and to propose appropriate individualized diet to correct a potential imbalance. Although this approach may seem a bit theoretical, a first step has been taken with the modeling of amino acid exchanges between different bacterial groups. Dietary protein and amino acids are, in fact, important substrates for colonic fermentation, where they serve as a nitrogen source for the microbiota. A model called CASINO (Community And Systems-level INteractive Optimization) was applied to analyze these exchanges in people with enriched or depleted microbiota of the MICRO-Obes study. CASINO was actually able to predict differences in production of SCFA and amino acids (such as phenylalanine and branched chain amino acids) between subjects. Fecal and blood metabolomics analysis allowed validation of the relevance of this theoretical model. Actually subjects with lower microbial richness had a greater elevation of amino acids such as phenylalanine and branched chain amino acids (valine, leucine, isoleucine). Blood elevation of some of these amino acids has been linked to insulin resistance and also identified as risk factor for type 2 diabetes (e.g. phenylalanine). The dietary intervention led to a significant decrease of these metabolites together with increased gut microbiota richness. CASINO also modeled which specific bacterial groups contributed significantly to the production of these “deleterious” metabolites. Finally, by comparing subjects with low or high gut microbiota richness during the
intervention, the model proposed what specific dietary changes (i.e. food categories) individuals with low richness potentially should consume to improve their metabolism.

CONCLUSIONS

Several studies described a positive impact of CR, bariatric surgery and dietary interventions such as prebiotic and probiotic supplementation on diet-induced metabolic disorders in rodents and in humans. Additional studies are warranted to suggest the use of one or another strain as therapeutic tool in the current clinical practice. It is worth noting that evidence suggests that body weight loss is not a prerequisite to observe beneficial impact upon health. This implies that changes in gut microbiota may contribute to the improvement of metabolic disorders via complex mechanisms that can be indirectly related to energy homeostasis.

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REFERENCES


### TABLES

#### Table 1. Overview of CR studies reporting changes in microbiome composition and/or function, along with clinical outcomes and/or dietary intake.

<table>
<thead>
<tr>
<th>First author</th>
<th>Study design</th>
<th>Population</th>
<th>Method</th>
<th>Changes in gut microbiome</th>
<th>Clinical or dietary outcome associated with gut microbial changes</th>
<th>MICROBIOTA VS.:</th>
</tr>
</thead>
<tbody>
<tr>
<td>Ley et al. (2006)</td>
<td>CR intervention: two diets (low carb or low fat) for 1 yr.</td>
<td>12 obese adults</td>
<td>16S rRNA sequencing</td>
<td>↑ Bacteroidetes:Firmicutes</td>
<td>Increase in Bacteroidetes abundance correlated with %WL.</td>
<td>CLINICAL METABOLITES</td>
</tr>
<tr>
<td>Duncan et al. (2007)</td>
<td>Randomized cross-over study: two 4-wk diets high in protein with low or medium carb, with a 3-day high carb maintenance diet before each regime.</td>
<td>19 obese men, no co-morbidities</td>
<td>FISH targeting 16S rRNA of 10 dominant bacterial groups and total bacteria</td>
<td>Yes</td>
<td>↑ total bacteria in maintenance diet. ↓ carb intake correlated with ↓ Roseburia spp/E. rectale group and Bifidobacteria. Abundance of other groups did not change. •When compared to medium or low carb diets, SCFA content was higher in the maintenance diet. •Butyrate production was positively correlated with carb intake.</td>
<td>DIET</td>
</tr>
<tr>
<td>Santacruz et al. (2009)</td>
<td>CR and exercise intervention for 10 wks.</td>
<td>36 overweight /obese adolescent s. There were low (WL&lt;2kg) and high (WL&gt;4kg) responders</td>
<td>16S rRNA qPCR of 11 bacterial groups and total bacteria</td>
<td>Yes</td>
<td>Different E. coli, B. longum and B. adolescentis between low and high responders before and after the intervention. Greater change in bacterial group abundance for high responders. •Bacteroides and Lactobacillus groups were positively correlated, and E. coli inversely correlated with WL. •Complex carb intake was negatively correlated with B. fragilis. •There was no difference in dietary intake between groups.</td>
<td></td>
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<tr>
<td>Nadal et al. (2009)</td>
<td>CR and exercise intervention for 10 wks.</td>
<td>39 overweight /obese adolescent s. There were low (WL&lt;2.5kg ) and high (WL&gt;4kg) responders</td>
<td>FISH targeting 16S rRNA of 11 dominant bacterial groups and total bacteria</td>
<td>Yes</td>
<td>In high responders: ↓ C.histolyticum, E. rectale/C. coccoides, and C. lituseburensis; ↑ Bacteroides/Prevotell a. No changes were seen for low</td>
<td></td>
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<tr>
<td>Study</td>
<td>CR intervention</td>
<td>Responders</td>
<td>Correlated with</td>
<td>Weight Loss</td>
<td>Fecal Output and SCFA</td>
<td>Microbiota Composition</td>
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<td>Brinkworth et al. (2009)</td>
<td>CR intervention: low-carb/high-fat vs. high-carb/low-fat diet for 8 wks.</td>
<td>91 responders</td>
<td>Bifidobacteria in low-carb/high-fat diet group at 8 wks.</td>
<td>• Low carb/high fat group had more WL than high carb/low fat group.</td>
<td>• Fecal output and concentration of fecal butyrate, acetate, propionate, and total SCFA were lower in the low carb/high fat diet group.</td>
<td>Bangladesh</td>
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<td>Wu et al. (2011)</td>
<td>Cross-sectional study (COMBO study). Subgroup was part of a 10-day controlled feeding study (CAFE study) (randomized to high-fat/low-fiber or low-fat/high-fiber).</td>
<td>16S rRNA sequencing, with a subset of shotgun metagenomics. Rectal biopsy for CAFE.</td>
<td>Two main enterotypes identified: Bacteroides and Prevotella.</td>
<td>• Association between nutrients and enterotypes was seen for the FFQ data (habitual diet) and not for 24h recalls.</td>
<td>● Taxa that correlated with fat-derived nutrients and with calories were also correlated with BMI.</td>
<td>● Associations between nutrients and enterotypes were seen for the FFQ data (habitual diet) and not for 24h recalls.</td>
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<td>Walker et al. (2011)</td>
<td>CR intervention: 1-wk maintenance diet followed by 3-wk diet high in resistant starch or non-starch polysaccharides using a cross-over design. Finally, 3-wk high-protein CR diet was consumed.</td>
<td>14 responders</td>
<td>Microbiota composition changed rapidly with diet, but inter-individual differences &gt;&gt; within-subject changes.</td>
<td>• Total SCFA were lower in high protein/low carb diet.</td>
<td>• Both diets increased fecal branched-chain fatty acids, phenylacetic acid and N-nitroso compounds, and decreased butyrate and</td>
<td>Microbiota composition changed rapidly with diet, but inter-individual differences &gt;&gt; within-subject changes.</td>
</tr>
<tr>
<td>Russell et al. (2011)</td>
<td>Cross-over study: 1-wk maintenance diet followed by 4-wk high</td>
<td>17 responders</td>
<td>Roseburia/E. rectale group and Bacteroides spp. in high protein/low carb diet.</td>
<td>• Total SCFA were lower in high protein/low carb diet.</td>
<td></td>
<td>FISH for detection of dominant bacterial groups and Roseburia/E. rectale group and Bacteroides spp. in high protein/low carb diet.</td>
</tr>
<tr>
<td>Study</td>
<td>CR Intervention</td>
<td>Total Bacteria</td>
<td>Fiber-Derived Antioxidants</td>
<td></td>
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<tr>
<td><strong>Kong et al. (2013)</strong></td>
<td>6-wk CR followed by 6wk WS. Diet was high in protein and fiber, with low glycemic index. Subjects were clustered into 3 groups according to their WL response.</td>
<td>16S rRNA qPCR of 7 dominant bacterial groups</td>
<td>( \uparrow ) baseline Lactobacillus/Leuconostoc/Pediococcus group in non-responders.</td>
<td>( \text{Lactobacillus/Leuconostoc/Pediococcus} ) group abundance correlated to weight regain during the WS period, and with consumption of starch in the diet. ( \text{Weight regain profiles could not be fully explained by differences in clinical parameters or dietary intake.} )</td>
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<td><strong>Cotillard et al. (2013)</strong></td>
<td>6-wk CR followed by 6wk WS. Diet was high in protein and fiber, with low glycemic index.</td>
<td>Shotgun metagenomic sequencing</td>
<td>( \uparrow ) microbial richness in subjects with low baseline gene richness. 26 out of 39 gene clusters varied significantly with time; ( \downarrow ) E. rectale and Bifidobacterium spp. ( \downarrow ) several gene clusters during WS.</td>
<td>( \text{Higher baseline microbial richness correlated with improved metabolic status after intervention.} ) ( \text{Change in abundance of the 26 gene clusters were correlated with changes in diet, clinical outcomes, or both.} )</td>
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<td><strong>Dao et al. (2015)</strong></td>
<td>6-wk CR followed by 6wk WS. Diet was high in protein and fiber, with low glycemic index.</td>
<td>Shotgun metagenomic sequencing and 16S rRNA qPCR of A. muciniphila</td>
<td>( \downarrow ) A. muciniphila in subjects with highest baseline abundance, but it remained 100 times more abundant than in subjects with low baseline abundance. There was a core to 26 MGS associated with A. muciniphila abundance at least one point during the intervention.</td>
<td>( \text{Higher baseline } A. \text{ muciniphila abundance was associated with a metabolically healthier status and with better outcomes from the dietary intervention.} ) ( \text{The most metabolically healthy subgroup was characterized by higher } A. \text{ muciniphila abundance and microbial richness.} ) ( \text{Baseline correlation between } A. \text{ muciniphila and serum acetate, which decreased after the intervention.} ) ( \text{No correlation between } A. \text{ muciniphila and diet, including diet quality index.} )</td>
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Carb, carbohydrate; CR, calorie restriction; sAT, subcutaneous adipose tissue; SCFA, short chain fatty acids; wk(s), week(s); WL, weight loss; WS, weight stabilization; yr, year
Table 2. Effect of bariatric intervention on gut microbiota composition in humans.

<table>
<thead>
<tr>
<th>First author</th>
<th>Study design</th>
<th>Population</th>
<th>Method</th>
<th>Diet reported?</th>
<th>Changes in gut microbiome</th>
<th>Clinical or dietary outcome associated with gut microbial changes</th>
</tr>
</thead>
</table>
| Damms-Machado et al. (2015) | Comparison of LSG vs. VLCD, 6-mo follow up. | 10 morbidly obese women, but microbiome data available for 6 (3 VLCD, 3 LSG). | Shotgun metagenomic sequencing | No             | ↓ Bacteroidetes in VLCD group; ↓ Firmicutes in LSG group. In LSG, ↓ Firmicutes, Bacteroidetes, E. rectale, and F. prausnitzii and ↑ F. prausnitzii. ↑ Several Firmicutes species in VLCD. | ● The VLCD had an improvement in blood lipids while the LSG did not.  
● Metabolic capacity for butyrate fermentation was increased for the VLCD group after the intervention, but there was no difference in SCFA between groups.  
● Fecal excretion of NEFA and bile acids was increased after LSG. |
| Zhang et al. (2009)   | Cross-sectional comparison of lean, obese and RYGB patients. | 3 lean, 3 morbidly obese, 3 unrelated BS patients 8-15 mo after surgery. | 16S rRNA sequencing              | No             | ↓ Firmicutes, ↑ Gammaproteobacteria in RYGB.  
↑ Prevotellaceae (H2 producers) and Archaea (H2 consumers) in obese.  
Verrucomicrobia variable in normal weight group, undetectable in obese group and highest in RYGB. | None reported. |
| Furet et al. (2010)  | Bariatric intervention (RYGB) with 3 and 6-mo follow-up, and comparison to lean controls. | 13 lean and 30 morbidly obese adults (7 with T2D) | 16S rRNA qPCR of total bacteria and 7 select bacterial groups | Yes            | ↓ F. prausnitzii in obese diabetic patients.  
↓ Bacteroides/Prevotella group in both obese groups. After RYGB: ↑ Bacteroides/Prevotella and E. coli; ↓ Bifidobacterium and Lactobacillus/Leuconostoc/Pediococcus groups.  
F. prausnitzii ↑ at 3 mo and remained stable at 6 mo. | ● F. prausnitzii inversely related to inflammation independently of diet. There was an inverse correlation between Leptin decrease and E. coli increase after surgery.  
● Some of the associations were observed between gut microbiota, corpulence and energy intake. |
| Kong et al. (2013)    | Bariatric intervention (RYGB) with 3 and 6-mo follow-up. | 30 morbidly obese adults (7 with T2D) | 16S rRNA sequencing              | Yes            | ↑ microbial richness 3 mos post-surgery and then stabilized.  
Microbiome composition shifted throughout intervention.  
↑ Bacteroides, Escherichia, and Alistipes increased;↓ | ● There were more correlations between microbial genera and sAT gene expression 3 mos after RYGB.  
● Changes in the abundance of 14 discriminant genera were associated with changes in clinical parameters and sAT |
<table>
<thead>
<tr>
<th>Study</th>
<th>Intervention</th>
<th>Study Group</th>
<th>Methodology</th>
<th>Findings</th>
</tr>
</thead>
</table>
| Graessler et al. (2013) | Bariatric intervention (RYGB) with 3-mo follow-up | 6 morbidly obese adults (5 with T2D). Lean controls for microbiome analysis only. | Shotgun metagenomic sequencing           | Microbial composition shifted after BS, including changes in 22 microbial species. ↑ obese vs. lean differences after surgery. ↓ Firmicutes and Bacteroidetes; ↑ Proteobacteria and Verrucomicrobia. Some species level changes: ↓ F. prausnitzii and ↑ A. muciniphila.  
From PCA analysis, species from component 1 (characterized by Enterobacter cancerogenus) were correlated to BMI and CRP. Most correlations observed between CRP and bacterial species were BMI-dependent. There were 10 species associated with blood lipids and 2 with HbA1c and F. prausnitzii correlated with fasting glucose. |
| Ward et al. (2014)     | Bariatric intervention (RYGB) measuring effect of PPI use on gut microbiota before and 6 mo after RYGB. | 8 morbidly obese adults                                                     | 16S rRNA sequencing                      | ↑ Firmicutes ↓ Bacteroides pre-surgery in PPI users. ↑ Akkermansia abundance pre-surgery in PPI users and increased in both groups. PPI users tended to have less excess weight loss than non-users. Only weight loss reported. |

HbA1c, hemoglobin A1c; LSG, laparoscopic sleeve gastrectomy; PCA, principal component analysis; PPI, proton pump inhibitor; RYGB, Roux-en-Y gastric bypass; T2D, type 2 diabetes; VLCD, very low calorie diet; yr, year
FIGURE LEGENDS

Figure 1. Comparing responses to weight loss interventions through extensive phenotyping and data integration. There are phenotypic and behavioral traits that differentiate responders vs. non-responders to weight loss interventions. These differences can be compared 1) at baseline, between responders (status Y) and non-responders (status X) for prediction (yellow profile vs. orange profile), and 2) before vs. after the intervention (yellow profile vs. blue profile) to study mechanisms that may be involved in a good response to the intervention. Environment may refer to diet, exercise, behavior, and other environmental exposures. Omics may refer to genomics, epigenomics, transcriptomics, proteomics and metabolomics in different tissues.

Figure 2. Dietary intervention such as prebiotic supplementation as well as gastric surgery impact gut microbiota and host metabolism and thereby represent interesting approaches for the treatment of obesity and metabolic disorders. Obesity is associated with alterations in metabolism and energy homeostasis. Gastric bypass surgery is associated with changes in gut microbiota composition and metabolic functions and represents one of the more effective approaches to treat obesity and metabolic disorders. Dietary interventions targeting the gut microbiota, such as prebiotics, induce changes in gut microbiota composition that are associated with modification of the secretion of gut enteroendocrine hormones as well as with a reduction in metabolic inflammation, and glucose, lipid and energy homeostasis dysfunctions.
Figure 3. Modelisation of the gut ecosystem as a first step for personalized nutrition. Individuals with low and high gut microbial richness differ in certain clinical parameters, dietary intake and metabolite profile. The CASINO toolbox predicts, at the individual level, differences in metabolite production by gut bacteria and proposes changes in dietary intake for individuals with low gene richness to improve their gut microbiome metabolism. BCAAs, branched chain amino acids.
High Microbial Richness
Greater diversity of intestinal microbiota

Cardiometabolic risk
Bacterial genes

Low Microbial Richness
Dyslipidemia
Insulin resistance
Low grade inflammation
Weight gain
More BCAAs

Improved gut microbiome metabolism
Dietary intervention

CASINO Toolbox