

Metabolomics

High resolution characterization of human metabolism and biochemical status on habitual diet and ketogenic diet

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Abstract

Introduction and background

Ketogenic diet (KD) consisting of mostly fat, adequate protein and very low carbohydrate consumption induces metabolic modifications where the main fuel switches from glucose to fat degradation. The diet is used in the treatment of several inborn errors of metabolism and as treatment for refractory epilepsy. The diet has also become popular as a rapid weight reduction method. High intake of fat, particularly saturated fat, can lead to high levels of total and LDL cholesterol concentrations known to be associated with development of atherosclerosis. This raises a concern that KD increases the risk of cardiovascular disease. However, the biochemical effects of KD are insufficiently described, and previous studies have been conducted primarily using overweight participants.

Objectives

This study aimed to identify metabolic changes in normal weight subjects as a result of following a ketogenic diet for three weeks. In addition, comparison between the ketogenic and habitual diet metabolomes of subjects with low and high relative LDL-C change after KD intervention was conducted.

Subjects and methods

In 2011-2012, a six-weeks, randomized controlled cross-over trial investigating the effect of three weeks on ketogenic diet (<20 g carbohydrates) was conducted in healthy normal weight subjects (N=30), mainly students and employees at the Department of Nutrition research at University of Oslo. As a continuation, comprehensive global metabolomics analyses of plasma with high-performance liquid chromatography coupled to mass spectrometry (LC-MS) was performed in this thesis to identify and quantify as many metabolites as possible in the biological samples.

Results

An average of 1085 and 419 features were detected in positive and negative ionization mode, respectively. Differential analysis using volcano plot showed that 48 features in positive ionization mode and 32 compounds in negative ionization mode were detected in significantly higher levels in samples taken after KD versus habitual diet with a log₂ fold change >1. Significantly higher levels of ketone bodies, lipids, acylcarnitines, trihomomethionine, acetylglycine, GABA and hydroxyvaleric were detected. Hydroxyvaleric acid had a greater increase in the group with the highest relative LDL-C increase compared to the lowest relative LDL-C increase group. Furthermore, citric acid, malic acid and alpha-ketoglutaric acid as intermediated in the citric acid cycle were significantly upregulated after KD. Levels of the amino acids threonine, asparagine, tryptophan, proline, b-alanine, leucine, arginine, methionine, glutamine and alanine were significantly decreased, while valine was elevated 43 % after 3 weeks on KD. Univariate correlation matrix showed a significant positive correlation ($p < 0.03$) between the amount of amino acid histidine and change in LDL-C with a correlation coefficient of 0.44 (CI 0.06, 0.71). Lower plasma levels of stachydrine, methyl indole 3-acetate, caffeine and N-acetylvaline was observed after KD.

Conclusion

Applying global metabolomics in a randomized controlled study has the potential to detect differences in the metabolome after dietary intervention, and further increase our knowledge of the biochemical effects and consequences of a KD. In particular, we found alterations in a diverse set of compounds. However, further investigations are needed to understand the mechanisms behind these changes and to determine long-term effects of following a KD.

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Abbreviations

ApoB	apolipoprotein B
ApoA1	apolipoprotein A1
CVD	cardiovascular disease
CM	chylomicrone
Da	dalton
ESI	electrospray ionization
GLUT1	glucose transporter protein type 1
HDL	high-density lipoprotein
HDL-C	HDL-cholesterol
IDL	intermediate-density lipoprotein
IEM	inborn error of metabolism
KD	ketogenic diet
LDL	low-density lipoprotein
LDL-C	LDL cholesterol
LDL-R	LDL receptor
LC-MS	liquid chromatography-mass spectrometry
MeOH	methanol
MS	mass spectrometry
<i>m/z</i>	mass-to-charge ratio

MCT	medium chain triglyceride
OUS	Oslo university hospital
PCA	principal component analysis
Rt	retention time
TC	total cholesterol
TG	triglycerides
V1	visit 1
V2	visit 2
V3	visit 3
VLDL	very-low-density lipoprotein

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1 Introduction

1.1 Ketogenic Diet

Ketogenic diet as treatment for disease

Ketogenic diet (KD) is a collective name for a diet consisting of mostly fat, adequate protein and very low carbohydrate consumption. The diet is called “ketogenic” because of the elevated production of ketone bodies. The diet was used as a treatment for epilepsy in the 1920s (1-3). Furthermore, KD is used in the treatment for several inborn errors of metabolism (IEMs) (4). IEMs are a large group of rare monogenetic diseases, usually caused by deficiency or defects of enzymes involved in anabolic or catabolic biochemical pathways, thereby affecting the levels of metabolites in cells, tissues and body fluids (5). Among them are glucose transporter protein type 1 (GLUT1) deficiency syndrome and pyruvate dehydrogenase deficiency, where KD is the primary treatment available (4). The biological mechanisms by which KD works for epilepsy are still unknown (6), except for GLUT1 and pyruvate dehydrogenase deficiency where KD contributes to an alternative energy source (7).

Low carbohydrate revolution

Later on the diet has been popular in the hope of achieving better health or a rapid weight loss (8, 9). The prevalence of obesity has been increasing worldwide over the past several decades in both children and adults (10). Obesity is associated with metabolic disorders like type-2-diabetes, cardiovascular disease (CVD), hypertension, cancers, osteoarthritis and the development of metabolic syndrome. Metabolic syndrome includes large waist circumference, dyslipidemia, elevated fasting glucose levels and elevated blood pressure (10). In 1972, Robert C. Atkin published his book *Dr. Atkin`s Diet revolution*. Atkin stated that a low intake of carbohydrates, mainly sugars and starches would be beneficial for health and for achieving better weight control. Atkin`s diet has been “One of the most popular fad diets in the United States” (11), and thus starting a low-carbohydrate revolution. High-fat foods such as vegetable oils, full-fat dairy products, margarine, mayonnaise, avocado and nuts represent a considerable proportion of the diet (12, 13). Moderate amounts of meat, fish, eggs and cheese are necessary to secure an adequate intake of protein, while bread, potato, pasta and rice are not compatible with a KD due to the high levels of carbohydrate. A moderate amount of other

carbohydrate sources such as vegetables, fruits and berries is accepted as they are important nutrient sources contributing with essential vitamins and minerals (14).

1.1.1 Variants of ketogenic diet

There are different variants of KD. How to manage the diet due to different levels of carbohydrate intake is the main difference that distinguish the diets. **Figure 1** illustrate the percentage distribution of the daily calorie intake of each macronutrient in a classic and modified KD compared to Nordic nutrition recommendations. Classic, modified and medium-chain triglyceride KD are most commonly used in Norway (13).

- **Classic ketogenic diet**

The classic KD is based on a ratio of grams of fat to grams of carbohydrate and protein combined (1). All meals have the same ketogenic ratio and are calculated with a 0.1 gram accuracy. A 2:1 ketogenic ratio equates to a meal that contains twice as many grams of fat as protein and carbohydrate combined. This signifies that the higher ketogenic ratio, the more fat in the diet.

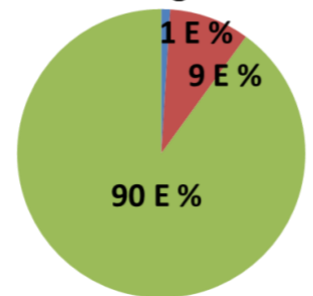
- **Modified ketogenic diet**

The modified KD limits carbohydrates to 10-20 g daily which corresponds to a slice of whole grain bread (15), but no limitations on protein intake and nearly 75 % of the daily calorie intake from fat. This KD is also known as the modified Atkins diet, as it is a modification of the diet that Robert C. Atkin established in 1972.

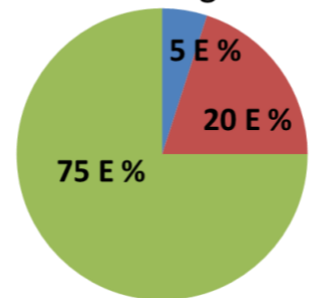
- **Medium-chain triglyceride ketogenic diet**

The medium-chain triglyceride KD is similar to the classic diet except that 30- 60 % of the daily calorie intake from fat is specified to be medium chain triglyceride (MCT) fat (1). MCT fat provides more ketones per calorie of energy than long chain triglycerides (LCT) (1).

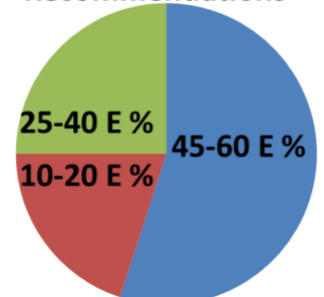
Classic ketogenic diet



Modified ketogenic diet



Nordic Nutrition Recommendations



■ Carbohydrate

■ Protein

■ Fat

Figure 1 Pie chart that illustrates the percentage amounts of macronutrients in a classic and modified ketogenic diet compared to Nordic Nutrition Recommendations.

Changing the diet from a recommended Nordic diet (16) where 45-60 % of the energy requirements comes from carbohydrates to a KD low on carbohydrate results in physiological

changes. The KD induces metabolic modifications associated with a state of starvation where the main fuel switches from glucose to fat degradation. This induces “physiological ketosis”.

1.1.2 Physiology of ketosis

In a normal physiological state the beta cells of the pancreatic islets secrete the peptide hormone insulin in response to a carbohydrate rich meal. Insulin is an important regulator of the metabolism of carbohydrates, fat and protein. The glucose level in the blood after a ketogenic meal is low, thus the secretion of insulin is reduced. This results in reduced absorption of glucose from the blood into the liver. The lower level of glucose in the liver reduces the production of and hence the availability of oxaloacetate in the liver. Oxaloacetate is a product of glucose and amino acid degradation and reacts with acetyl CoA to produce citrate for ATP generation in the citric acid cycle. Increased release of fatty acids from the adipose tissue and subsequent transport into the liver where the fatty acids are metabolized leads to production of Acetyl CoA. Carnitine is essential in transport of long-chain fatty acids into mitochondria for oxidation (1).

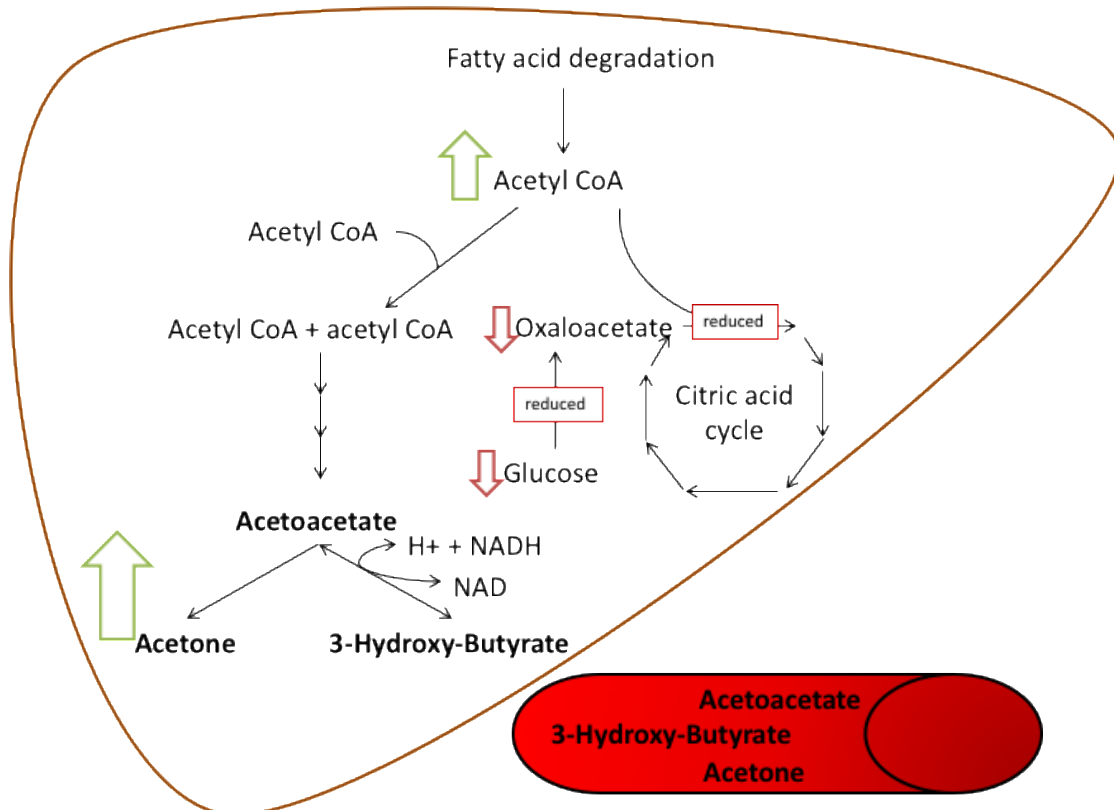


Figure 2 ketogenesis simplified production of ketone bodies in the liver in response to ketogenic diet

Lower levels of oxaloacetate combined with elevated levels of Acetyl CoA that exceed the amount needed for citrate synthesis lead to shunting of Acetyl CoA to production of ketone bodies, ketogenesis, resulting in a metabolic state of ketosis (17). This is illustrated in **figure 2**. Glucose levels remains stable due to synthesis from glucogenic amino acids and glycerol liberated via lysis of triglycerides (1, 18, 19).

Ketone bodies

Ketone bodies are water soluble molecules that contain hydrogen, carbon and oxygen.

Acetoacetate, beta-hydroxybutyrate and acetone are three well known endogenous ketone bodies used to produce adenosine triphosphate (ATP) as a result of metabolic degradation.

Ketones can be measured in urine and blood (1). Circulating levels of ketone bodies vary across the populations due to variations in basal metabolic rate, hepatic glycogen stores and mobilization of amino acids from muscle proteins (20).

Ketolysis is the metabolic degradation of ketone bodies illustrated in **figure 3**. The ketone bodies are exported to the circulation system for uptake by the mitochondria of extrahepatic organs, where they are converted to acetyl coA and enters the citric acid cycle to generate ATP.

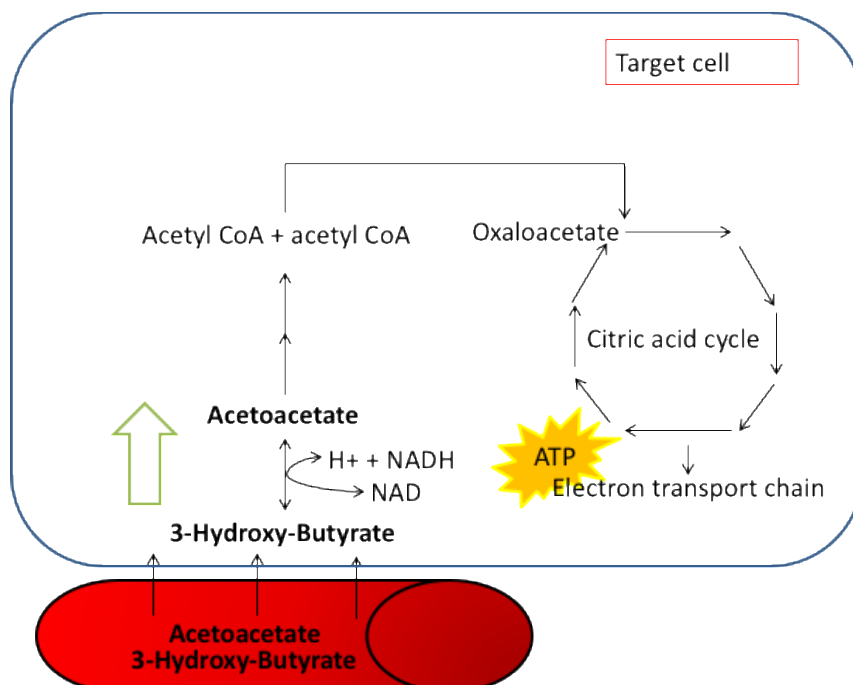


Figure 3 ketolysis simplified presentation of metabolic degradation of ketone bodies in a target cell generating ATP

So, two hallmarks of a KD is the reduction in blood sugar levels and the elevation of ketone levels in the blood (17). However, these changes can cause undesirable consequences.

1.1.3 Side effects of the ketogenic diet

Constipation, vomiting, diarrhea, headache, fatigue and weight loss are common side effects reported on a classic and modified KD (21). Furthermore, there is an increased risk for nutritional deficiencies since the diet limits food groups such as fruit and starchy vegetables. Thus, multi-vitamin supplementation is recommended, especially among those who are not aware of their own nutrient intake. Hypoglycemic episodes can occur during the first weeks before the metabolism adapts to ketone bodies and fat as primary sources of energy (1). The ketosis during the first couple of weeks on the diet lowers the pH in the blood and may cause a mild metabolic acidosis (12). A severe metabolic acidosis, ketoacidosis, can occur in persons with diabetes mellitus and is a life-threatening condition.

Additionally, elevated total cholesterol and LDL cholesterol have been reported as adverse effects of the KD (21-25).

1.1.4 Raised lipid levels in a ketogenic diet

Raised lipids levels are common in people on KD (21-24). Elevation of total cholesterol, lipoproteins, total apolipoprotein B (apoB) and apolipoprotein A1 (apoA-1) was reported among children on KD for six months, but there was no description regarding type of fat intake in the report (22). Studies on adults on KD report similar finding (26, 27).

Cholesterol and lipoprotein

Cholesterol is an essential component of cell membranes and is a precursor of bile acids and steroid hormones (10). Cholesterol is effectively transported in the blood stream packed within lipoproteins that delivers lipids to cells and tissue. Lipoproteins consist primarily of cholesteryl esters and triglycerides in their core and phospholipid, free cholesterol and apolipoproteins (apo) on the surface (28). Lipoproteins can be classified based on their densities. In order of increasing density; Chylomicrones (CM), very low-density lipoprotein (VLDL), intermediate density lipoproteins (IDL), low-density lipoproteins (LDL) and high-

density lipoproteins (HDL). ApoB-containing lipoproteins (VLDL, IDL and LDL) in plasma are the main transporters for both exogenous and endogenous cholesterol from the liver to peripheral tissues (10, 29). ApoB100 is the main structural apolipoprotein of LDL with a single molecule of apoB-100 per LDL particle. The HDL contains apoA-1 as its major apolipoprotein and transport the cholesterol from peripheral tissues to the liver, termed reverse cholesterol transport (10). Triglycerides are formed from exogenous fatty acids and endogenously either produced by the liver or released from adipose tissue (10). As a rule of thumb, the following values have been classified as optimal; A total cholesterol < 5 mmol/L, LDL <3mmol/L, triglycerides <2 mmol/l and HDL >1 mmol/l for men and <1.3 for women (30).

Elevated total cholesterol and LDL cholesterol (LDL-C) are known risk factors for developing cardiovascular disease. This raises a concern that a KD with a high intake of fat increases the risk of atherosclerosis.

1.1.5 Cardiovascular disease and atherosclerosis risk

Atherosclerosis is a complex and degenerative disease of the vascular endothelium where plaques inside the arterial walls reduce the internal vessel diameter and inhibit the blood supply to vital organs, in particular when plaques rupture and cause platelet aggregation and coagulation dramatically reducing or blocking blood flow, causing chronic and acute cardiovascular disease (CVD)(31). CVD is a class of diseases affecting the heart and circulatory system and is the major cause of death and morbidity in the Norwegian population (32). Important metabolic risk factors for developing CVD includes high blood pressure, high fasting blood glucose and high cholesterol, all of which are modifiable risk factors (33, 34). In addition, an unhealthy diet is one of the modifiable risk factors that causes the most deaths in Norway (32). Furthermore, age, gender and genetics contribute in addition as non-modifiable risk factors.

High blood cholesterol, especially LDL-C is a causal factor in the pathophysiology of atherosclerosis and is associated with increased risk of CVD (10, 29). Lipids, primarily cholesterol-rich apoB containing lipoproteins and cholesteryl esters, and cells of the immune system are major contributors in the atherogenic process. In short, lipid accumulation in intima causes endothelial dysfunction as one of the first steps in the development of

atherosclerosis. This recruits monocytes from the bloodstream that migrate into the intima of the arterial wall where they differentiate into macrophages that ingest oxidized LDL and further progress to form foam cells (31, 35).

1.1.6 Role of fat quality for development of atherosclerosis

Total fat includes saturated fat, monounsaturated fat, omega-3 polyunsaturated fat and omega-6 polyunsaturated fat (10). National dietary guidelines in Norway focus on a diet and an eating pattern that prevents chronic disease and promotes health (36). The guidelines recommend a diet with 25-40 E % total fat, <10 E% saturated fat, 10-20 E % monounsaturated fat, 5-10 E% polyunsaturated fat (36). High intake of fat, particularly saturated fat in the diet, is associated with development of health problems. Large amounts of saturated fat in the diet can lead to high levels of total and LDL cholesterol concentrations associated with the development of atherosclerosis (29). The plasma cholesterol raising effects of saturated fat, particularly myristic (C14:0) and palmitic (16:0) acids, have been well established (10). However, the evidence for how a diet rich in saturated fat can cause atherosclerotic vascular disease is lacking despite several studies that have attempted to address this question. Replacing saturated fat with polyunsaturated fatty acids improves the lipid profile and reduces the cardiovascular disease risk (36-38). Thus, replacing foods from animal sources rich in saturated fat with foods high in mono- and polyunsaturated fat e.g. vegetable oils are preferable.

The distribution between saturated and unsaturated fat in a KD can be of high importance, especially regarding changes in the lipid profile. There is a lack of knowledge of the long-term physiological changes of adherence to KD. Previous studies have been conducted primarily using overweight participants (39, 40). Losing weight will change and improve the metabolic profile (10). But what is the effect of KD on normal weight individuals?

1.1.7 Effect of Ketogenic diet on LDL cholesterol and gene expression in normal-weight, young adults: A randomized controlled study

The low carbohydrate high fat (LCHF) study was conducted by Retterstøl et al in 2011/2012. A total of 30 normal-weight, young adults mainly students and employees at the Department

of Nutrition research at the University of Oslo completed the study. The participants were randomly allocated to a LCHF diet or habitual diet for three weeks. The group assigned to LCHF diet had a plasma LDL-C increase from 2.2 ± 0.4 mmol/l at baseline to 3.1 ± 0.8 mmol/l after intervention. The group continuing their habitual diet had no difference in plasma LDL-C levels (2.5 ± 0.8 mmol/l) $p < 0,001$ after the three weeks on the continued habitual diet.

In the LCHF study specific assays were used to identify and quantify metabolites of interest coupled to cholesterol pathways, lipid profile and peripheral blood mononuclear cells gene expression. But what about everything else? What about all the possible information related to the hundreds and thousands of metabolites that we do not detect with specific assays?

1.2 Metabolomics

Metabolomics is the systematic study of low molecular weight molecules (metabolites), typically less than 1500 Da, in a biological sample (41). Carbohydrates, amino acids, nucleotides and lipids are substrates, intermediates and end products of metabolism that can be investigated using metabolomics to complement genomics, transcriptomics, proteomics, and other omics studies as illustrated in **figure 4 (42)**. Metabolomics provides information closer to the phenotype. The phenotype refers to the physical and biochemical characteristics of an organism that is determined by its genotype and the influence of a myriad of environmental factors over time directly upon the organism and on the expression of its genes (43). The metabolome can be defined as the entire collection of metabolites in a biological sample (44). Metabolomics can be divided into a targeted or an untargeted, global approach (45). Specific assays are used in targeted metabolomics to identify and quantify specific metabolites of interest in a biochemical pathway. A global approach aims to investigate as many metabolites as possible in a biological sample(45).

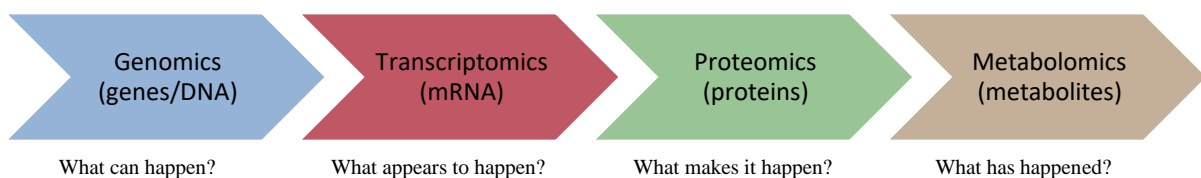


Figure 4 The omics cascade. Metabolome is downstream of the genome, thus metabolomics provides information closer to the phenotype. The phenotype is the physical and biochemical characteristics of an organism that is the results of the combined effect of genome, lifestyle and environmental factors. The figure was adapted from (46).

1.2.1 Targeted analyses

Targeted analyses are specific assays used to identify and quantify selected metabolites or metabolite classes (47). These metabolites are substrates, intermediates, products or other substances involved in known biochemical pathways, usually endogenous, but many are exogenous. Targeted analyses are used in standard clinical chemistry for diagnostics of IEMs and other conditions (48). It is possible to analyze one or several compounds based on a hypothesis (47). However, the metabolism is complex and it is preferred to analyze as many compounds as possible if no clear hypothesis can be defined. Global metabolomics make this feasible.

1.2.2 Global metabolomics

Global metabolomics is an untargeted approach to identify and quantify as many metabolites as possible in a biological sample, thus being able to generate new hypotheses based on the findings (44, 45, 47). Untargeted assays using liquid chromatography-mass spectrometry (LC-MS) provide semi-quantitative data, where chromatographic peak areas are reported instead of absolute concentrations (49). Global metabolomics can be used to examine differences in the metabolic profile between a control and a treatment group (47). One single analytical method can not detect the entire collection of metabolites in a biological sample due to different properties of the metabolites, thus different strategies and methods are necessary to focus on different metabolite classes and to increase the number of metabolites detected.

1.2.3 LC-MS used in global metabolomics

The analytical platform used for this project was LC-MS with an electrospray ionization (ESI) source illustrated in **figure 5**. This is one of the most used analytical tools in global metabolomics (47, 50). The analytical platform and extraction method used will influence which metabolites that can be detected and measured.

A mass spectrometer (MS) is an analytical instrument applied to detect positively or negatively charged gas phase metabolites based on their mass-to-charge ratios (m/z) (51). Thousands of peaks referred to as metabolite features can usually be detected in a biological sample (47). Each feature consists of a measured mass-to-charge ratio (m/z) and a unique

retention time that is used to perform metabolomic profiling (45). Metabolites can be separated in time prior to mass spectrometry analysis to reduce the chance of matrix effects, e.g. ion suppression. Liquid chromatography (LC) is a technique to separate the compounds in a solution based on equilibrium between a solid stationary phase and a liquid mobile phase. A pump generates high pressure, allowing the liquid mobile phase to transport the sample compounds through a closed column containing the stationary phase (47). Each compound has its own unique properties and interacts differently with the solid and mobile phases chosen. The time from the sample is injected until the compound elutes from the LC column is referred to as the retention time of the compound (52).



Figure 5 picture of the instrument used in untargeted metabolomics

The electrospray ionization (ESI) source is used to transform ions in a liquid phase to gas phase using electrical energy, before entering the MS (47, 53). Compounds in solution are differently charged (positively or negatively). The voltage applied determines whether the positively or negatively charged metabolites enter the MS, referred to as positive and negative ionization mode, respectively (47).

1.2.4 Untargeted metabolomics workflow

A common workflow to perform untargeted metabolomics using LC-MS starts with the experimental design and consist of collecting relevant samples, sample preparation, data acquisition, data processing and statistical analyses of the data generated to investigate trends and identify metabolites and metabolic pathways involved (45). This can generate hypotheses that may be validated with further quantitative measurements. **Figure 6** illustrates the untargeted metabolomics workflow.

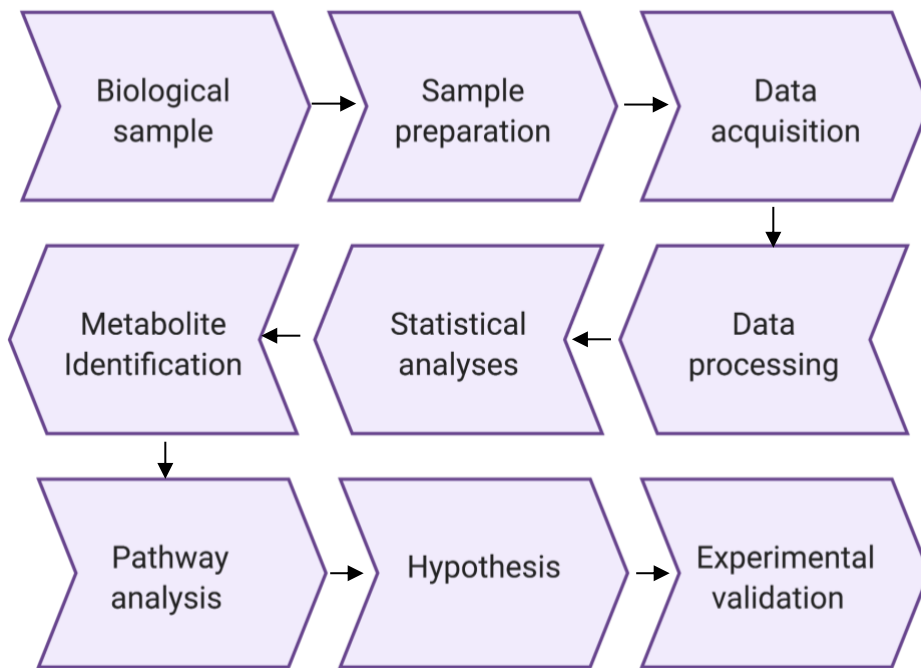


Figure 6 Untargeted metabolomics workflow for hypothesis generation

Experimental design

A well-established experimental design is essential to obtain objective and valid results. Furthermore, it is important to evaluate the number of individuals, samples and matched controls to be included in the study (47). Standardization and control of variables influencing the metabolome must be addressed, e.g. sex, age, ethnicity, diet, physical activity, medication and diseases, which all contribute to biological variation. The design of the LCHF study is shown in **figure 7**. The LCHF had a cross over design, meaning that both groups performed the diet intervention but at different times.

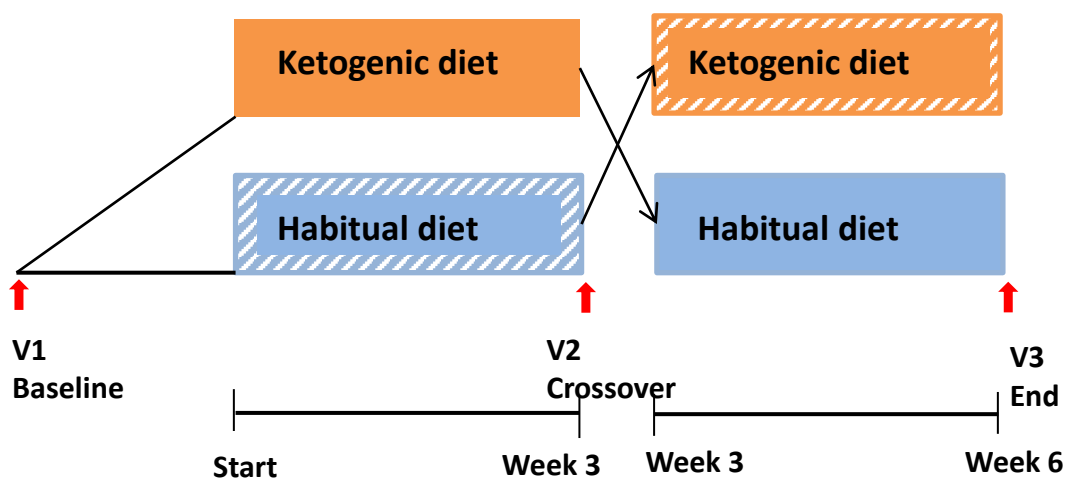


Figure 7 The LCHF study design with three visits (V1-V3) when samples were obtained.

Biological material

The composition of metabolites differs in the different sample materials. Plasma, serum, tissue, urine, tear fluid, spinal fluid, whole blood and dried blood spots are different biological materials that can be used in metabolomics (54-58). Choice of material depends on the metabolites of interest and availability of samples.

Sample preparation

Intervention and control samples must be treated equally prior to analysis. Information about the preanalytical factors e.g. sampling, sample handling, preparation, transport and storage is important (45, 47). Sample preparation in metabolomics is usually non-specific to ensure coverage of as many metabolites as possible in the biological sample .

Data acquisition

In LC-MS global metabolomics a process blank sample and quality control samples are usually included in addition to the biological samples of interest (49). A process blank sample, e.g. water, is prepared using the same procedure as for the biological samples, and analyzed to remove the background signal which do not reflect the biological effects being studied (49). Quality control samples are used to correct for changes in instrument performance over time (49). A pooled quality control (PQC) sample is often used in global metabolomics studies. A small volume of each biological sample is mixed together to generate a PQC sample, followed by aliquotation to a set of identical PQC samples (49). These PQC samples should be analyzed at designated time points through the analytical sequence (47).

Full scan and tandem mass spectrometry data acquired

Data acquisition by LC-MS can be done in full MS and/or MS/MS scan mode (59). In full MS scan mode, intact molecular ions in a given m/z range are detected (51). In MS/MS scan mode, molecular (precursor) ions are fragmented to produce product ions(51). The fragmentation pattern of a compound can be used for identification purposes by determination of the molecular structure. Data acquired using full MS (molecular mass and isotope pattern) and MS/MS (fragmentation spectra) scan mode can be compared to experimental and established MS libraries and databases for compound identification(51, 60).

Data preprocessing

Mass and retention time shifts are corrected for using the PQC samples (61). The ionization process may produce ions (fragments and adducts) referred to as features that are not necessarily unique metabolites (45). Features are ideally grouped together to the single metabolite they originate from based on retention times using comprehensive software (47, 61).

Data analysis

Several bioinformatics statistical tools can be used to determine whether there are differences in the metabolome of the KD compared to the habitual diet samples, and identify significantly altered metabolites (45, 47). Univariate statistical methods are used to analyze metabolomics features independently, while multivariate methods are used to analyze all the metabolomics features, revealing possible patterns (47). Principal component analysis (PCA) is a method that summarizes the complex metabolomics data to visualize the difference between two or more groups (62). PCA is not a statistical test but an ordination analysis to simplify the complex data and aims to uncover factors that account for the major patterns across the original variables and explain most of the variability in the data (63). PCA forms a new coordinate system for separation of samples, and the largest variance of the dataset is illustrated by the first principal component (64). In untargeted metabolomics one point in a PCA plot represents all the features detected in a specific sample.

Differential analysis using volcano plot detect specific compounds differing in amount between two groups based on defined sample ratios (fold change) and calculated p-values. Volcano plot shows which compounds that are significantly up and down-regulated (65). **Figure 8** shows a box and whisker plot and how the data is distributed. The height of the rectangle represents the peak areas in the interquartile range. The first quartile represents 25 % of the distribution, the third quartile represents 75 % of the distribution. In small number of data points, the whiskers end at the highest and lowest data points (65).

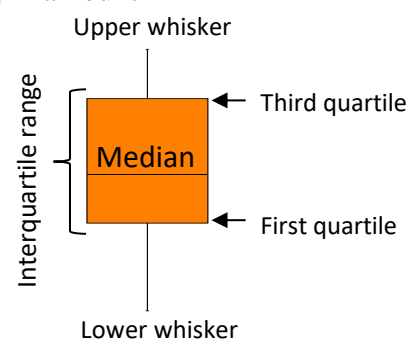


Figure 8 Box and whisker plot distribution of data

Metabolite identification

Metabolite identification is partially performed manually, thus being a time consuming step (45). To identify a feature of interest, metabolite databases, e.g. the Human Metabolome Database (<http://www.hmdb.ca/>) are used by searching the accurate mass of the specific compound (45). Chempider searches databases of MS1 scans by using the mass or predicted formulas and mzCloud database for matching and finding similar fragmentation spectra (65). The metabolite identification is a comprehensive process, where a considerable number of the features detected from biological samples cannot be identified without extensive follow up investigations (60).

1.2.5 Metabolomics in nutrition research

There has been an increased interest in using metabolomics for nutrition research (66). The complexity of different diets and how they affect our health has for a long time been of interest among researchers. Untargeted metabolomics has been used to identify biomarkers associated with food intake (67, 68) and as a screening tool for revealing patterns of different dietary interventions (69). However, the associations between complex diets and their desired results as well as adverse outcomes are difficult to address. Untargeted metabolomics can contribute to a wider understanding of how diets effect the health and identify individual metabolites and complex profiles that change in response to the dietary intervention (66).

As earlier described, Retterstøl et al investigated the effect of following a KD for three weeks with targeted analyses to identify and quantify metabolites of interest coupled to cholesterol pathways, lipid profile and peripheral blood mononuclear cells gene expression (70). A global metabolomics analysis approach was not performed at that time.

2 Aim of thesis

Studies investigating the impact of KD in human health are sparse. The aim of this thesis was to apply an untargeted metabolomics profiling approach on plasma samples from healthy, young adults on a habitual and ketogenic diet, respectively. In addition to analyzing samples from the first three weeks of the six week cross over intervention study published so far (70), this thesis will also study the effect after cross over from habitual to ketogenic diet.

Specific objectives of this thesis were:

- Compare the metabolomics profile before and after a three weeks long KD intervention. Is there a separation between the metabolomes of habitual diet and KD?
- Investigate which metabolites that significantly change in concentration between the metabolomes of habitual and KD
- How does a KD affect the amino acid profiles among the participants?
- How does a KD affect metabolites involved in energy metabolism?
- Compare the metabolomic profiles between those with the lowest and highest relative LDL-C increase in response to the KD
- Investigate which metabolites that are significantly changed in concentration between those with lowest and highest relative LDL-C increase

3 Materials and methods

3.1 Low carbohydrate high fat diet study

3.1.1 Participants and design of the LCHF diet study

Participants were mainly students and employees at the Department of Nutrition research at the University of Oslo in September 2011 to December 2012 that were invited to participate in a ketogenic (low carbohydrate high fat) diet study after information meetings. Inclusion and exclusion criteria have been described elsewhere in detail (70). **Figure 9** illustrates the six weeks randomized controlled parallel crossover intervention study.

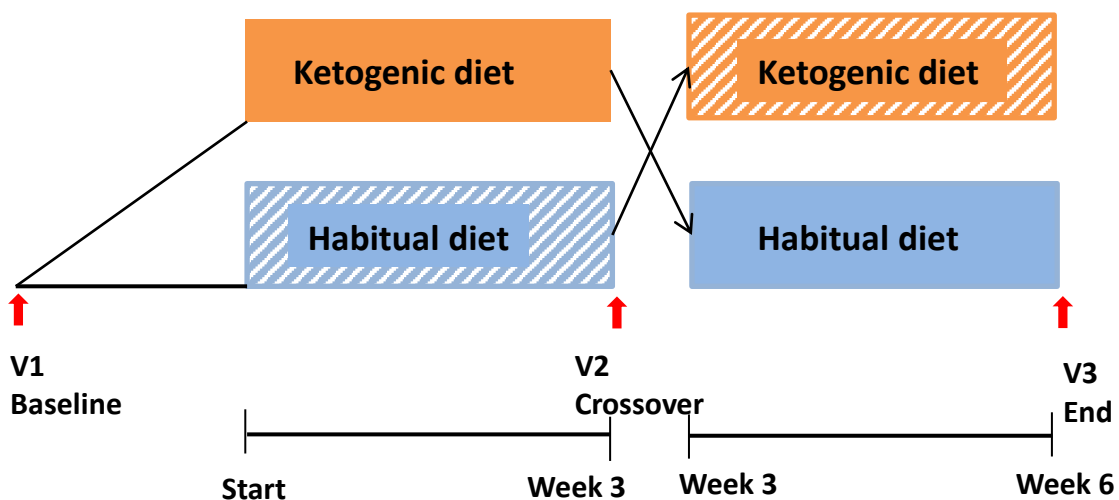


Figure 9 Design of the six weeks randomized controlled parallel crossover intervention study. Participants were allocated to either a three weeks ketogenic diet or to continue their habitual diet for three weeks, followed by crossover and another three weeks on the other type of diet. Data were collected at baseline (V1), crossover (V2) and end of intervention (V3)

The participants were divided randomly into two groups. One group, called the LK (low carbohydrate) group, started with a KD and ended the dietary intervention after 3 weeks. The LK group then resumed their habitual diet for the next three 3 weeks. The other group continued with their habitual diet for the first three weeks, followed by the KD for the last 3 weeks. The dietary specifications were as reported in Retterstøl et al (70). Specifically, the participants were advised to follow the guidelines in Dr. Atkins' new Diet Revolution (71).

Carbohydrate intake was limited to 20 g per day or were not to exceed 5 E% for the calorie intake. They had no restrictions on type of fat during the dietary intervention.

3.1.2 Dietary assessment

21 of the participants performed a four days weighed food record assessment prior study start to evaluate their habitual diet and 19 of the participants performed a three days weighed food record during the LCHF diet period (70). KostBeregningsSystem Version 7.1, University of Oslo, Norway and “MAT på DATA” version 5.2 were used for energy and nutrients calculations (70). This was to check for compliance among the participants on the KD.

3.2 Materials for metabolomics and LCHF diet

3.2.1 Sample material and storage

All the sample materials were from the LCHF diet study (70). Anthropometric measurements, blood pressure and fasting blood samples were drawn at baseline, after three weeks (crossover) and after six weeks (end of study). Plasma was obtained using EDTA tubes (Becton Dickinson) and serum was collected from silica gel tubes (Becton Dickinson) as described in detail elsewhere (70). Plasma and serum were stored at -80 °C until further analysis.

The frozen samples were placed on blocks of dry ice from the biobank at the Department of Nutrition Research at the University of Oslo and shipped to Department of Medical Biochemistry, Oslo University Hospital, Rikshospitalet, Norway for global metabolomics analysis.

The samples were numerated from p1, p2...p30, where p1 to p15 represented one group (LK group) and p16 to p30 constituted another group (KL group).

3.2.2 Preparations for LC-ESI-MS analyses (chemicals)

Solvents

Methanol (MeOH) $\geq 99.9\%$, formic acid (FA) 98-100 % and ammonium acetate $\geq 99.0\%$ purity were obtained from Merck (Dramstadt, Germany). Type 1 water was used from a

Millipore Milli-Q purification system with Q-guard and Quantum purification cartridge to reach a resistivity of 18,2 M Ω .cm at 25°C (Merck) (72).

Mobile phases

Mobile phases were prepared prior to each global metabolomics analysis.

Mobile phases A and B used for positive ionization method consisted of water with 0.1 % formic acid (FA) and MeOH with 0.1 % FA, respectively. Mobile phases A and B used for negative ionization method consisted of water with 10 mM ammonium acetate and MeOH with 10mM ammonium acetate, respectively. Mobile phases were stored at room temperature after preparation.

Extraction solution

The extraction solution consisted of cold MeOH stored at 4 °C.

Calibration solutions

Pierce LTQ Velos ESI positive Ion Calibration solution and Pierce ESI Negative Ion Calibration solution were obtained from Thermo Fisher Scientific (Waltham, MA, USA) stored at -18 °C.

3.3 Sample preparation

Protein precipitation with organic solvent was performed. The sample preparation method was conducted for both plasma and serum samples. The sample preparations were distributed over four and three days for serum and plasma, respectively (See **appendix**, section 7.1.1) However, only plasma is described further due to limited time and that the plasma not the serum is the entity circulating in the blood vessels. Plasma constitutes the liquid portion of blood when blood cells have been removed. For the majority of analyses the measured concentrations will be similar in plasma and serum. The plasma samples were thawed for 50 minutes at room temperature, followed by vortexing for 5 seconds (Genie2, Scientific Industries, Bohemia, NY, USA) and then centrifuged at 3600 rpm at 20°C for 10 minutes (Megafuge 1.0 R, Heraeus instruments).

30 μ l blood plasma was transferred to filter paper cards from Whatman 903 Protein Saver Cards from GE Healthcare Life Science (Chicago, IL, USA), and stored at -80°C for possible

future analysis.

90 μl of cold methanol was added to 30 μl blood plasma in a 1.5 ml Eppendorf tube, and then mixed on a table vortex for 5 seconds. The samples were centrifuged at 14800 rpm (14.8 $\times g$ rpm) at 4°C for 10 minutes with Heraeus Fresco 21 centrifuge from Thermo scientific. 90 μl of the clear supernatant were transferred to an HPLC vial with insert and cap, all from Matriks AS, Oslo, Norway. **Figure 10** shows the sample preparation method (my own photos)



Figure 10 Sample preparation (private photos)

To obtain a pooled quality control (PQC) to be used throughout the whole run (analysis), a representative subset consisted of 30 μl of each sample from day1 were collected as shown in **figure 11**. See **appendix section 7.1.2** for detailed description of the preparations of the different pooled quality controls used in the study.

Pooled quality control preparation

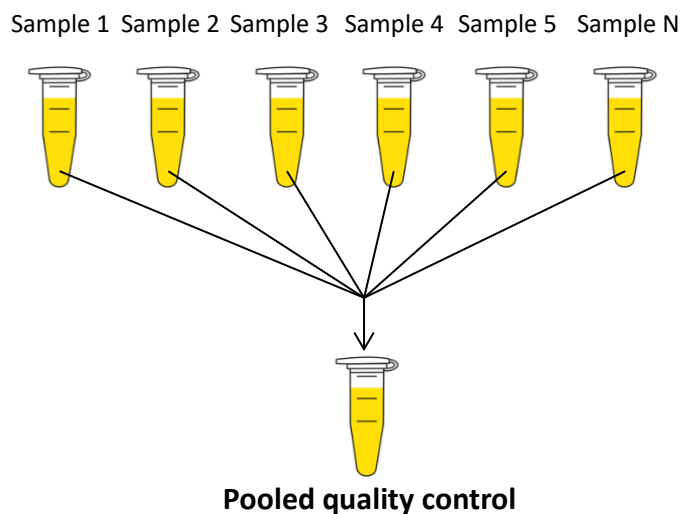


Figure 11 Preparation of pooled quality control that contains the same metabolites as all the samples to control for possible analytical retention time drift and peak area drift throughout the sequence.

See **appendix 7.1.3** for the sample sequence used for global metabolomics analysis, including PQC samples injected at regular intervals throughout the analysis.

3.4 Liquid chromatography – electrospray ionization– mass spectrometry settings

Metabolites were measured with LC-ESI-MS instrumentation. The LC-ESI-MS used was Dinoex Ultimate 3000 UHPLC system coupled to a Q Exactive Orbitrap MS (Thermo Scientific). The analytical column used was Pursuit XRs (250 x 2.0 mm, particle size 3 μ m purchased from Agilent Technologies (Santa Clara, CA, USA). The LC-ESI-MS settings used are shown in **tables 1-4** and **figure 12**.

Table 1 Liquid chromatography settings used for all the analyses

Parameter	Setting
Mobile Phase A	Positive ionization: Water with 0.1 % formic acid Negative ionization: Water with 10 mM ammonium acetate
Mobile Phase B	Positive ionization: Methanol with 0.1% formic acid Negative ionization: Methanol with 10 mM ammonium acetate
Gradient	See Table 2 and Figure 12
Injection Volume	2 μ l
Column temperature	30 $^{\circ}$ C
Flow rate	300 μ l/min
Analysis time	32.5 minutes
Re-equilibration time	10 minutes

Table 2 Flow gradient

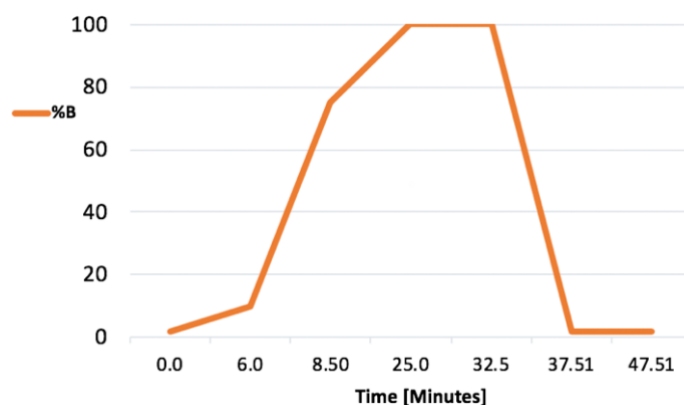
Mobil phase A: Water with 0.1% formic acid (positive ionization)

Water with 10 mM ammonium acetate (negative ionization)

Mobil phase B: Methanol with 0.1 % formic acid (positive ionization)

Methanol with 10 mM ammonium acetate(negative ionization)

Time	Flow	%A	%B
0	0.3	98	2
6	0.3	90	10
8.5	0.3	25	75
25	0.3	0	100
32.5	0.3	0	100
37.51	0.3	98	2
47.51	0.3	98	2

**Figure 12** flow gradient

Mobil phase A: Water with 0.1% formic acid

(positive ionization)

Water with 10 mM ammonium acetate

(negative ionization)

Mobil phase B: Methanol with 0.1 % formic acid

(positive ionization)

Methanol with 10 mM ammonium acetate

(negative ionization)

Table 3 Mass spectrometric settings

Parameter	Setting
Scan ranges (<i>m/z</i>)	70-1050
Fragmentation	Yes
Resolution	Full MS: 70 000 MS/MS: 17 500
Polarity	Positive and negative
Automatic gain control target value	Full MS: 1.00E+06 ion counts MS/MS: 5.00E+05
Maximum injection time	Full MS: 250 ms MS/MS: 100 ms
Analysis time	32.5 minutes
Re-equilibration time	10 minutes
MS/MS parameter	Setting
Scan type	Data dependent acquisition, top 5
Dynamic exclusion time	10 s
Intensity threshold	5.00E+04
Stepped normalized collision energy	20, 50, 80

Table 4 Electrospray settings

Electrospray parameter	Setting
Sheath gas (N2) flow rate	40 a.u.
Auxiliary gas (N2) flow rate	10 a.u.
Sweep gas (N2) flow rate	2 a.u.
Capillary temperature	250 °C
S-lens RF level	50.0
Auxiliary gas heater temperature	300 °C
Electrospray voltage	3.5 kV
Electrospray needle position	C

3.4.1 Ethics

All of the participants signed a written consent form, agreeing to the donation of blood, anthropometric measures and terms of the study. The study was approved by Regional Committee for Medical and Health Research Ethics (REC) (protocol approval 2018/2019 – 2011/1365-26). There were no conflicts of interest declared in this thesis.

3.5 Data analysis

Xcalibur (Version 4.0) was used to make the global metabolomics analysis sequence and to control the liquid chromatography. Tune (version 2.5 Build 2042) was used to calibrate and to control the mass spectrometric parameters. Compound discover software (version 2.1 SP1) from Thermo Scientific was used for processing raw data files from the global metabolomics analyses using LC-ESI-MS, perform statistics and use the processed data to search in online databases (Chemical formula in ChemSpider and fragmentation spectra (MS/MS) in mzCloud). **Figure 13** shows the workflow tree used for preprocessing the data in compound discover. The workflow template for the pre-processing used in this project is “Metabolomics/untargeted Metabolomics with Statistics Detect Unknowns with ID using Online Databases” created by compound discoverer.

Statistical tools used for differential analysis were mainly principal component analysis (PCA), volcano plot and box plot. Metabolic peak intensities were log transformed and scaled in CD to normalize skewed distribution.

IBM SPSS Statistics 24 for Windows software was used for performing data analysis regarding baseline description and blood sample parameters. Paired t-test was used on normally distributed data. Data are expressed with mean values and standard deviation. The nonparametric test, wilcoxon signed ranks, test was used if the data were skewed to compare two related samples. “R” software was used for performing correlations analysis, both Pearson and Spearman were used. Statistical significance levels were determined to $p < 0.05$

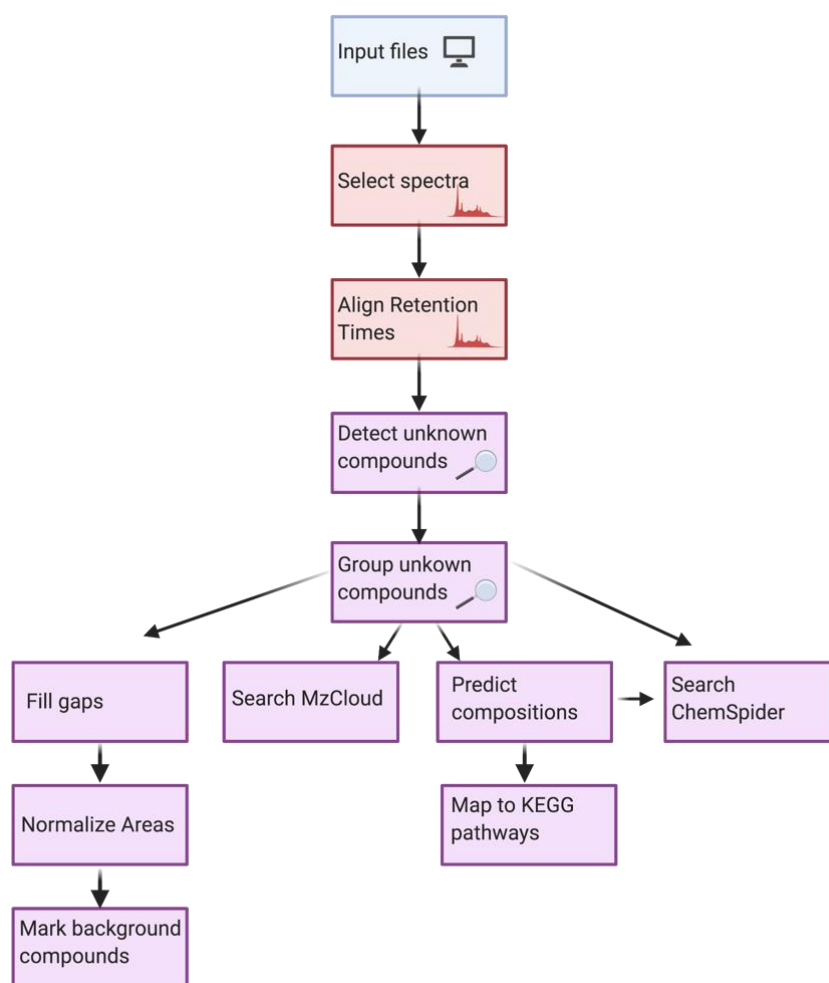


Figure 13 Workflow tree used for preprocessing the data in compound discoverer(CD). Figure modified from CD software.

4 Results

Metabolomic profiling was performed on plasma samples using high-performance liquid chromatography coupled to mass spectrometry. Analysis results were based on retention time (rt), mass to charge ratio (m/z), and peak area of each feature.

4.1 Participant characteristics

30 participants were included. Four participants were excluded after metabolomics analysis due to results and further investigations revealing non-compliance with respect to the diet, resulting in a total of 26 participants. See **appendix** section 7.2.1 for details regarding the exclusion. Baseline characteristics of the 26 participants and the changes in anthropometrics measurements and blood parameters following dietary intervention are shown in **table 5**. The mean BMI in the population corresponded to “normal weight”, however both men and women lost weight during the study period. The level of atherogenic lipoprotein particles increased on KD, with the exception of triglyceride and triglyceride-rich lipoprotein (TRL) cholesterol which remained unchanged. HDL-C increased significantly.

Table 5 Characteristics of participants in the project

	Baseline (n=26)	End-of-intervention (n=26)	P-value ¹
Female, n (%)	21(80.8)		1.000
Age (years)	25.1±4.7		1.000
Weight (kg)	63.2±7.9	61.6±7.9	<0.001
BMI (kg/m ²)	21.8±1.3	21.2±1.4	<0.001
Total cholesterol (mmol/l)	4.3±0.6	5.8±1.2	<0.001
HDL cholesterol (mmol/l)	1.7±0.4	2.0±0.4	<0.001
LDL cholesterol (mmol/l)	2.4±0.6	3.4±1.11	<0.001
Triglyceride (mmol/l)	0.8(0.4-1.5)	0.8(0.4-1.5)	0.380 ^a
Non-HDL-C (mmol/l)	2.7±0.7	3.8±1.2	<0.001
TRL cholesterol (mmol/l)	0.3±0.1	0.4±0.2	0.090
ApoA1 (g/L)	1.6±0.2	1.9±0.3	<0.001
ApoB (g/L)	0.7±0.2	1.0±0.3	<0.001
FFA (mmol/l)	0.4(0.1-0.9)	0.8(0.3-1.3)	<0.001 ^a

¹Data presented as n, number (%) or mean ± SD. Wilcoxon or paired t-test.

ApoB; apolipoprotein B; ApoA1; apolipoprotein A1; HDL, high density lipoprotein; LDL, low density lipoprotein; FFA, serum free fatty acid; non-HDL-C (TC minus HDL-C); TRL cholesterol (triglyceride-rich lipoprotein cholesterol; TC minus LDL-C minus HDL-C)

^aWilcoxon signed ranks test

Compliance to the ketogenic diet

Compliance to the KD among the participants is shown in **table 6**. The mean intake of reported carbohydrate was 17 ± 3 (9-20) g. The energy intake varied a lot between the participants, but there was no difference between the energy intake between habitual diet and following the KD. The reported calorie intake on the habitual diet ranged from 1223 to 4152 kcal and 1326 to 3013 kcal on the KD.

Table 6 Dietary intake on habitual diet and on ketogenic diet

	Habitual diet	Ketogenic diet	P-value ¹
Energy, kcal	2109±637 ^a (21) ^{b*}	2043±432 (19)	0.797
Carbohydrate, E%	46.4±4.5 (21)	3.2±0.6 (19)	<0.001*
Protein, E %	17.3±2.8 (21)	26.3±4.6 (19)	<0.001*
Fat, E%	30.3±4.8 (21)	69.5±5.2 (19)	<0.001*
Saturated fat, E%	10.9±1.8 (20)	29.5±7.5 (19)	<0.001 ₂
Monounsaturated fat, E%	10.6±2.5 (20)	23.9±3.7 (19)	<0.001 ₂
Polyunsaturated fat, E%	5.6±1.2 (20)	12.6±6.3 (19)	0.001 ₂
Fiber	2.8±0.7 (20)	2.9±1.0 (19)	0.430 ₂
Cholesterol, mg	266±76 (18)	1007±370 (18)	<0.001

^aMean ± SD

^b number of weighed dietary record for the specific nutrient included in calculations of mean

¹Paired t-test

*Missing two values on KD, 19 related samples included in paired t-test

₂Missing one value on KD, 19 related samples included in paired t-test

4.1.1 Metabolome at baseline shows random metabolite profiles

Visit 1 was used as baseline for both groups. For the group that continued with their usual diet from week 1 to week 3, visit 1(V1) and visit 2(V2) were compared to investigate whether the participants changed their habitual diet as an unintended response to participation in the study. A principal component analysis plot (PCA) was used to explore and visualize the variation between the samples. The PCA plot showed no substantial difference between the participants' metabolomes at these two visits in the habitual diet group (see **appendix** section 7.2.2), and thus the effect of study inclusion alone had not altered their behaviours in a way that markedly affected their plasma metabolomes.

To explore the potential baseline metabolome difference between the two groups, a PCA plot with all the baseline samples were performed. As presented in **Figure 14**, there were no obvious differences between the two groups at baseline according to the PCA plot.

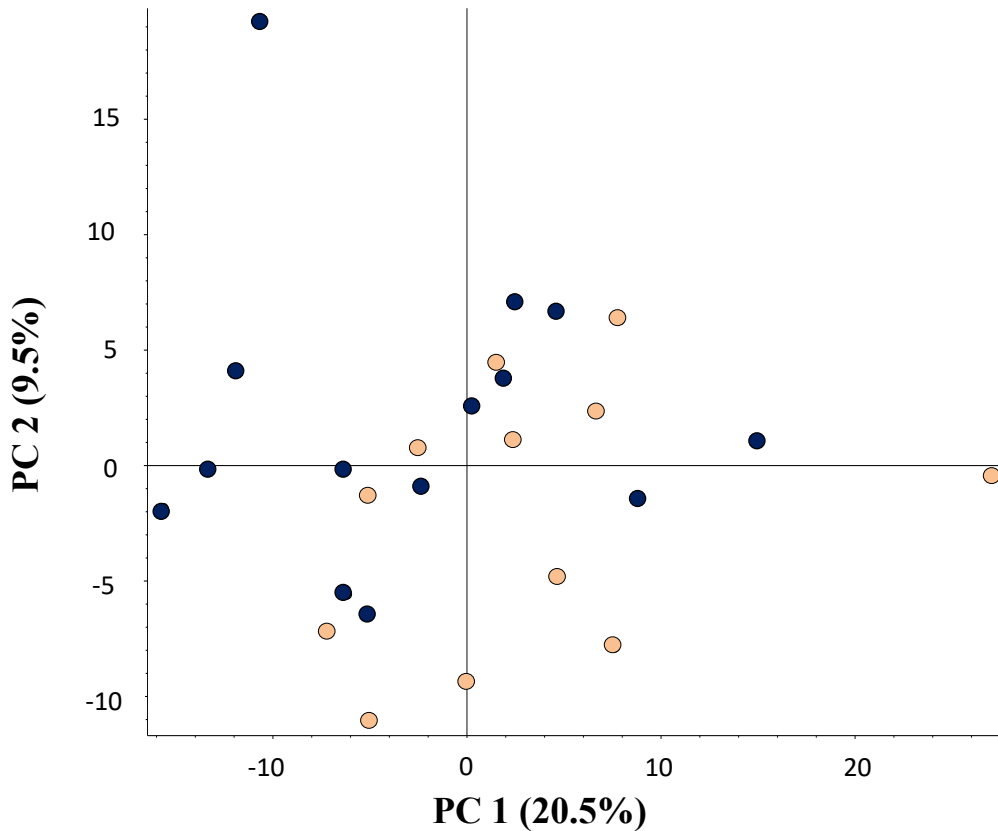


Figure 14 Principal component analysis (PCA) plot of all detected features from baseline samples prior to intervention. Each point represents all the metabolites detected in that sample. Blue and beige represent the LK (low carbohydrate) and KL (habitual diet) group respectively. The plot shows no clustering between the two groups.

The PCA plot confirmed a randomized population, thus further metabolomics analysis that compared the KD and control diet were conducted. The PQC samples were centered in the PCA plot, meaning that the variation observed for the individual participant samples represents the biological variation in the population and possible preanalytical differences (sample preparation, storage etc).

Because the PCA plot shown in **figure 14** displayed no obvious differences between the two groups, the population was collectively grouped as one with baseline V1-samples representing the habitual diet metabolomes for all participants in the following analyses.

4.2 Comparison of ketogenic and habitual diet

4.2.1 PCA plot shows a separation between ketogenic and habitual diet

Figure 15 shows the PCA plot from the analysis run in negative ionization mode of all samples taken at baseline (habitual diet) and immediately following intervention with KD among all the participants. The PCA plot shows distinct clustering discriminating the ketogenic and the habitual diet groups with a few exceptions (p18, p22 and p1). Principal component 1 (PC1) represented 19.4 % of the variability among the samples, while principal component 2 (PC2) explained 8.6 %. The distance in the PC1 direction represented a larger difference than in the PC2 direction. One point represents all the features detected in one sample, which implies that the closer the samples are located in the plot, the more similar their metabolomes are. The features that contribute to the largest variation were therefore relatively similar within each diet group, and differs between the groups.

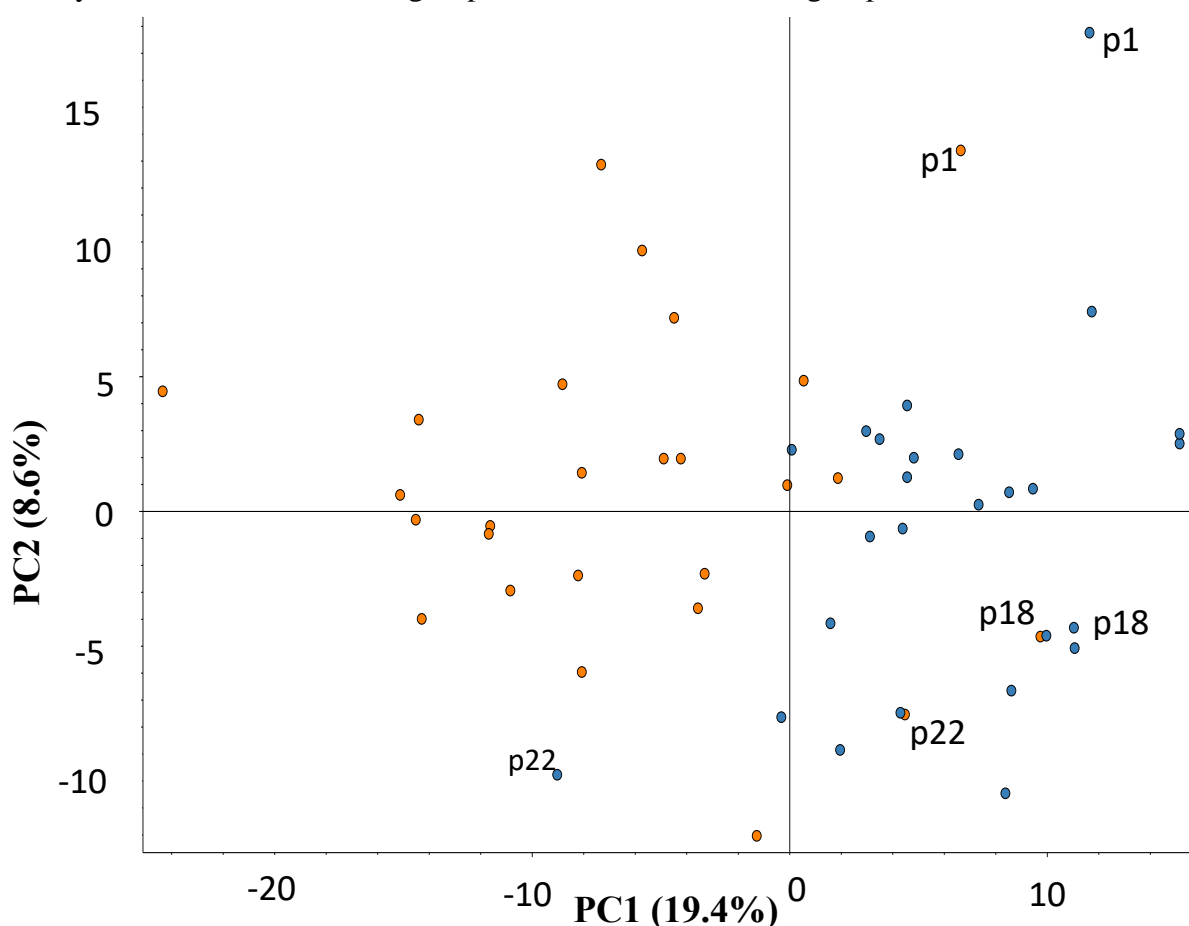


Figure 15 PCA plot from the analysis run in negative mode that display a separation between the habitual diet and the ketogenic diet, except from some individuals (p18,p22, p1). A point represents all the metabolites detected in that sample. An orange colored point represents the metabolome with a ketogenic diet and a blue colored point represent the metabolome at baseline (habitual diet).

See **appendix section 7.2.3** for the PCA plot in positive ionization mode with PQC samples.

To further investigate the separation seen in **figure 15**, volcano plot was used to look at the individual metabolites that were significantly increased or decreased following the KD intervention.

4.2.2 Volcano plot revealed numerous features that were significantly quantitatively changed following a ketogenic diet

From 26 young adults on KD, an average of 1085 of 2637 (1552 filtered out in the preprocessing) features in positive mode and 419 of 1116 (697 filtered out) features in negative ionization mode were detected.

Statistical analysis using volcano plot, see **Figures 16 and 17**, revealed that the quantity of numerous features was significantly changed in the participants at the end of the three week period on KD. The x-axis represents the generated ratio between the ketogenic and habitual diet in log₂ scale. The y-axis represents the significance of the generated ratio in negative log₁₀ scale. $P < 0.05$ was used as the cut off value for significance and log₂ fold change values was set to >1 and <1 , which represents a hundred percent increase or a fifty percent reduction in the ratio, respectively. In positive ionization mode 48 features were significantly increased and had a log₂ fold change greater than or equal to 1 in the KD (red in the volcano plot), with seven metabolites successfully identified: hydroxybutyrate, trihomomethionine, acetoacetate, stearoylcarnitine, acetyl-l-carnitine, hexanoylcarnitine and cis-5-tetradecanoylcarnitine. See **appendix section 7.2.4 table 15 and table 16** for m/z and rt for the remaining, unidentified metabolites. 277 additional features were significantly increased ($p < 0.05$), but the increase was less than 100 %.

Nine features were significantly decreased (green in the volcano plot), with four successful metabolites identifications: stachydrine, methyl indole 3-acetate, caffeine and N-acetylvaline. 130 additional features were significantly decreased but the reduction was less than 50 %. 11 had a reduction of more than 50 %, but the change was not significant.

Measured peak areas of identified metabolites that changed significantly in the KD samples and habitual samples are shown in section 4.2.3 in **table 7**.

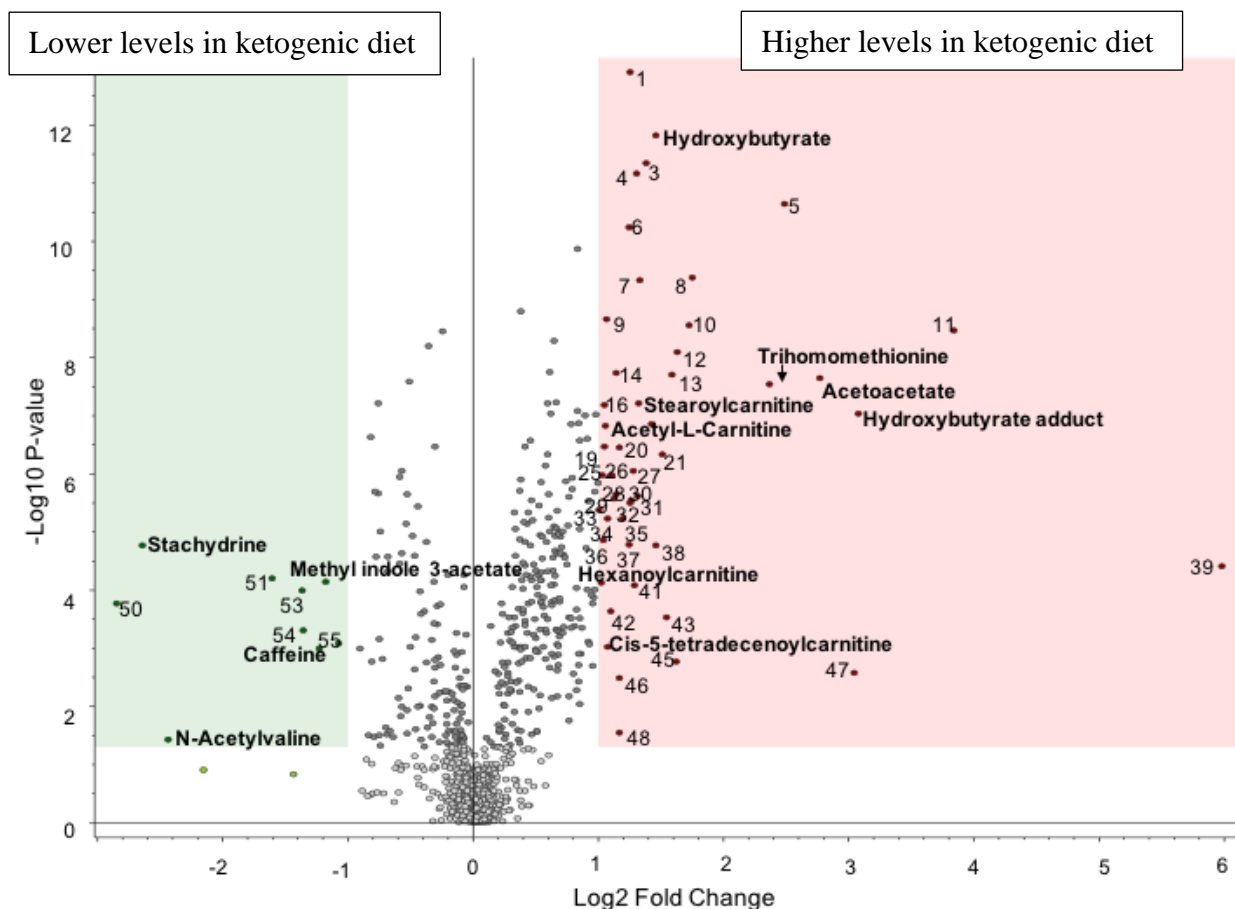


Figure 16 Differential analysis using volcano plot of data obtained from the positive ionization mode analysis of samples taken following ketogenic diet intervention. Metabolites identified are presented in bold text. The points in the green and red areas display metabolites with both fold changes values as well as statistical significance.

The volcano plot with negative ionization mode is shown in **figure 17**. 32 compounds were detected in significantly higher levels in samples taken after KD versus habitual diet and had a log₂ fold change greater than or equal to 1, with five successful metabolite identifications: hydroxybutyrate, methyl octadecenoate, arachidic acid, hydroxyvaleric acid and acetylglycine. 129 features were significantly increased, but the fold change was less than 1. One feature was more than doubled, but the increase was not significant. Four compounds had significantly lower levels in samples taken after KD versus habitual diet, but without any successful metabolite identifications. 57 features were significantly decreased, but the reduction was less than 50%. Two features were reduced more than 50%, but the change was not statistically significant.

Measured peak area of identified compounds that changed significantly in the KD samples and habitual diet samples are shown in section 4.2.3 in **table 7**.

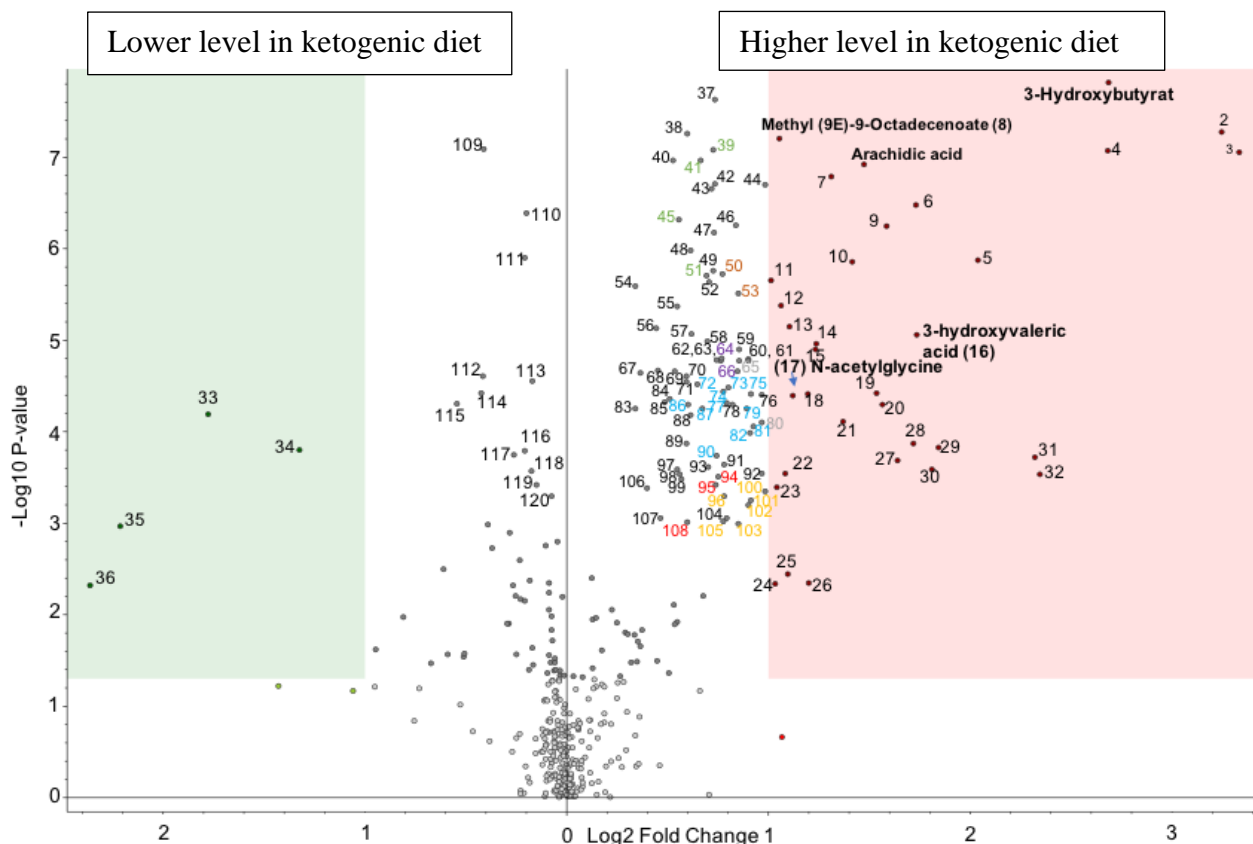


Figure 17
Differential analysis using volcano plot of data obtained from the negative ionization mode analysis of samples following ketogenic diet. Metabolites identified are marked in bold text. Numbers with the same colors had the same pattern which means they can be fragments or adducts from the same compound.

4.2.3 Peak area quantification for identification of significantly altered metabolites

Table 7 Metabolites identified that were markedly and significantly altered in quantity following 3 weeks on ketogenic diet

Compound	Group Areas		Ratio Ketogenic diet/baseline	P-value
	Baseline	Ketogenic diet		
Positive mode				
Hydroxybutyrate	5.01E+07	5.59E+08	11.16	9.30E-08
Acetoacetic acid	2.32E+07	1.53E+08	6.59	2.32E-08
Trihomomethionine	2.43E+05	1.11E+06	4.58	2.88E-08
3-Hydroxyisobutyric acid	2.04E+07	6.34E+07	3.11	1.53E-12
Stearoylcarnitine	9.90E+05	2.91E+06	2.94	6.14E-08
Acetyl-L-carnitine	1.18E+08	2.43E+08	2.06	1.52E-07
Hexanoylcarnitine	2.99E+06	5.86E+06	1.96	7.49E-05
Caffeine	1.27E+08	5.74E+07	0.45	1.02E-03
Methyl indole-3-acetate	4.08E+07	1.63E+07	0.40	7.44E-05
N-Acetylvaline	2.47E+06	4.15E+05	0.17	3.75E-02
Stachydrine	2.20E+08	2.58E+07	0.12	1.75E-05

Compound	Group Areas		Ratio	P-value
	Baseline	Ketogenic diet	Ketogenic diet/baseline	
Negative mode				
Hydroxybutyric acid	4.32E+05	2.70E+06	6.26	8.74E-05
Acetoacetate	1.11E+04	6.52E+04	5.85	6.84E-04
Arachidic acid	5.40E+04	1.42E+05	2.63	3.56E-05
Hydroxyvaleric acid	2.00E+05	5.08E+05	2.54	6.95E-04
Adrenic Acid	4.56E+04	1.04E+05	2.28	5.27E-06
N-Acetylglycine	4.88E+04	1.07E+05	2.20	4.31E-04
Octadecanedioic acid	1.37E+05	2.79E+05	2.04	6.58E-04

Figure 18 shows a box and whiskers plot of the measured peak area of hydroxybutyrate and n-acetylvaline, two compounds that were significantly changed after a KD.

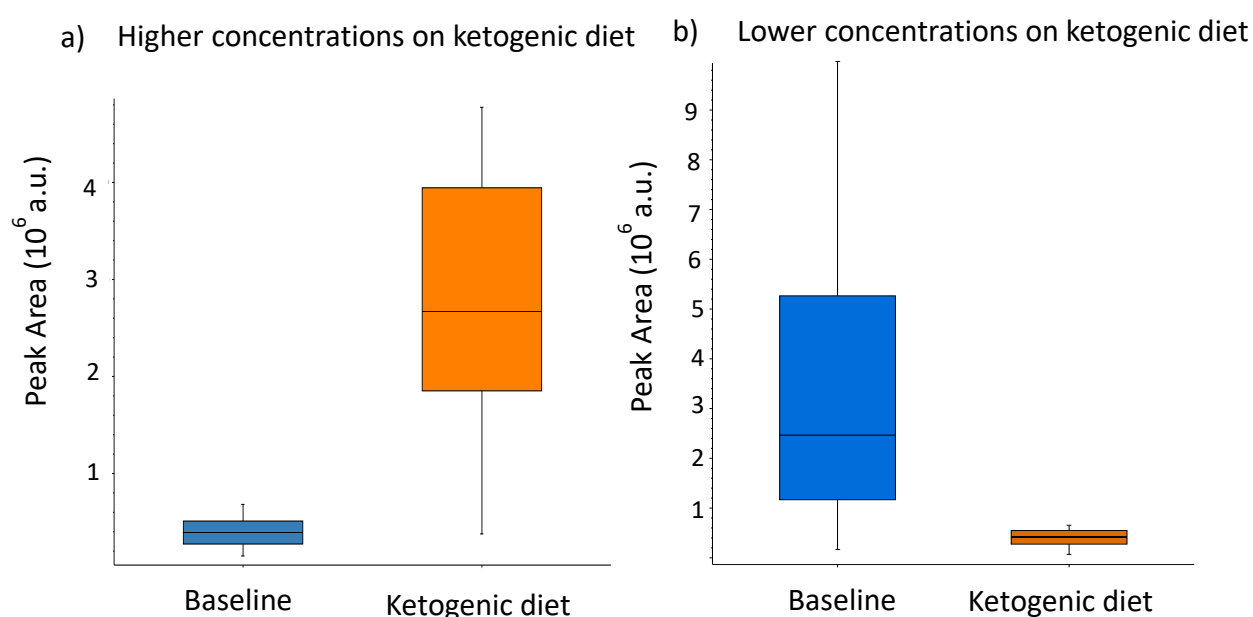


Figure 18 Box plot of changes in a) hydroxybutyrate and b) N-acetylvaline following three weeks on a ketogenic diet.

See appendix section 7.2.4 for box and whisker plots of acetoacetate, hydroxyvaleric acid, acetylglycine, arachidic acid, n-acetylvaline and caffeine.

To further characterize the human metabolism after a KD, a more targeted approach was conducted to look at specific metabolite abundances, such as fatty acids related to LDL-C, carnitine-bound metabolites necessary for lipid transport (10) and amino acids (73).

4.2.4 Fatty acids

Levels of fatty acids related to possible increase or decrease in LDL-C are shown in **table 8**.

Table 8 Saturated fatty acids that were markedly and significantly altered in quantity following 3 weeks on ketogenic diet

Compound	Group Areas		Ratio	P-value
	Baseline	Ketogenic		
Oleic acid	2.64E+07	5.45E+07	2.1	1.20E-04
Arachidonic acid	4.68E+05	7.92E+05	1.7	1.80E-05
Myristic acid	2.58E+06	4.43E+06	1.7	0.01
Palmitic acid	2.51E+07	4.09E+07	1.6	0.01
Lauric acid	3.29E+06	3.88E+06	1.2	0.05

4.2.5 Carnitine-bound metabolites were elevated in plasma on ketogenic diet

The identification of the carnitine-bound metabolites was not completed due to lack of time, and the carnitine conjugates listed in **table 9** are only the most likely compounds based on software searches in available databases. Higher levels of carnitine conjugates from baseline to 3 weeks on the KD are shown. Acylcarnitines linked to fatty acid metabolism showed higher levels, whereas acylcarnitines linked to amino acid metabolism were not detected, except for propionylcarnitine.

Table 9 Alterations in carnitine and carnitine conjugates from baseline to after ketogenic diet

Compound	Group Areas		Ratio	P-value
	Baseline	Ketogenic		
C0 L(-)-Carnitine	2.72E+06	2.77E+06	1.0	0.153
C2 Acetyl-L-carnitine	2.01E+06	3.88E+06	1.9	0.004
C3 Propionylcarnitine	1.83E+05	1.94E+05	1.1	0.849
C6 Hexanoylcarnitine 1	4.56E+04	9.13E+04	2.0	0.015
C8 Octanoylcarnitine 2	1.76E+05	3.55E+05	2.0	0.063
C14 Myristoylcarnitine*	2.14E+04	5.91E+04	2.8	<0.001
C16 L-Palmitoylcarnitine*	7.63E+04	1.24E+05	1.6	<0.001
C18 Stearoylcarnitine	1.58E+04	5.76E+04	3.6	<0.001
3-hydroxyoctanoylcarnitine*	3.85E+04	5.87E+04	1.5	0.019

Bold text; full match, *no mzcloud, no chemspider zinvalid mass

4.2.6 Markers of energy metabolism

Table 10 shows detected and identified metabolites involved in energy metabolism. Citric, malic and alpha-ketoglutaric acid, as intermediate in citric acid cycle, were significant upregulated. Creatine were significantly elevated.

Table 10 Energy intermediates Alterations in energy intermediates from baseline to after ketogenic diet

Compound	Group Areas		Ratio	P-value
	Baseline	Ketogenic		
Citric acid	5.65E+07	8.60E07	1.5	6.19E-08
Malic acid	2.36E+07	3.13E+07	1.3	3.33E-04
Alpha-ketoglutaric acid	3.03E+07	4.61E+07	1.4	1.71E-05
Creatine	2.82E+06	4.09E+06	1.45	0.001
Lactic acid	7.38E+07	8.34E+07	1.13	0.972
Creatinine	6.60E+08	6.82E+08	1.03	0.585

4.2.7 Alterations in amino acid profiles on ketogenic diet

Examination of the amino acid profiles from baseline to the end of three weeks on a KD revealed significantly decreased levels of threonine, asparagine, tryptophan, proline, β -alanine, leucine, arginine, methionine, glutamine and alanine. However, the average level of valine was elevated 43 % compared to baseline levels. The average peak areas and ratios of amino acids after 3 weeks on a KD versus habitual diet are presented in **table 11**.

Table 11 Average peak areas and ratios of amino acids from baseline and after a ketogenic diet

Name	Group Areas		Ratio	P-value
	Baseline	Ketogenic diet	Ketogenic diet/Baseline	
y-aminobutyric acid (GABA)	2.40E+06	4.09E+06	1.65	1.75E-06
Valine	8.86E+06	1.26E+07	1.43	4.41E-07
Isoleucine	4.20E+07	5.06E+07	1.21	1.27E-05
Citrulline	4.62E+06	4.79E+06	1.04	4.23E-01
Phenylalanine	2.10E+07	2.05E+07	0.98	2.06E-01
Lysine	4.03E+06	3.92E+06	0.97	5.64E-01
Cystine	1.89E+05	1.79E+05	0.95	1.93E-01
Tyrosine	1.06E+07	1.00E+07	0.95	1.05E-01
Glutamic acid	2.34E+06	2.21E+06	0.95	8.65E-01
Histidine	1.00E+06	9.30E+05	0.93	3.43E-01
Serine	3.74E+06	3.47E+06	0.93	1.61E-01
Arginine	1.96E+06	1.79E+06	0.92	7.95E-03
Glycine	8.27E+04	7.55E+04	0.91	1.49E-01
Asparagine	3.72E+06	3.36E+06	0.90	1.74E-05
Proline	6.35E+06	5.58E+06	0.88	1.54E-03
Alanine	3.30E+05	2.88E+05	0.87	7.09E-03
Leucine	2.52E+04	2.16E+04	0.86	3.15E-03
β-Alanine	1.02E+07	8.67E+06	0.85	2.66E-03
Methionine	6.52E+06	5.45E+06	0.84	8.11E-03
Glutamine	3.32E+06	2.78E+06	0.84	2.73E-04
Tryptophan	2.21E+07	1.77E+07	0.80	3.88E-04
Threonine	1.02E+07	7.76E+06	0.76	7.70E-06

4.2.8 Summary of metabolic data from the comparison of ketogenic and habitual diet

Table 12 summarizes the quantitative changes in metabolite groups and single metabolites following a three week period on a KD based on an untargeted and targeted approach.

Table 12 Biochemical effects of a ketogenic diet for three weeks

Elevated after ketogenic diet	Reduced after ketogenic diet
Ketone bodies, lipids, trihomomethionine, 3-hydroxyvaleric acid, acetylglycine, acylcarnitines, amino acids; valine, isoleucine, citric acid, malic acid, alpha-ketoglutaric acid, creatine	Acetylvaline, Caffeine, Methyl Indole 3-acetate, Stachydrine, amino acids; threonine, asparagine, tryptophan, proline, b-alanine, leucine, arginine, methionine, glutamine, alanine

Because the individual increase in LDL-C from baseline varied between 5 % to 128% following a KD for three weeks, the last analysis conducted by LC-ESI-MS global metabolomics in this thesis was based on the relative change of LDL levels.

4.1 Comparison of ketogenic and habitual diet metabolomes among those with lowest and highest relative LDL-C increase

Changes in plasma levels of LDL-C from baseline to the end of three weeks on a KD were determined for each participant based on data collected in 2012, see **appendix** section 7.3 **table 17** for further details. The participants were divided into either a low or a high LDL increase group. **Figure 19** displays the PCA plot in negative ionization mode with the participants that had the highest and the lowest relative LDL-C increase from the habitual diet baseline to the end of intervention with three weeks on a KD. The two groups clustered in the PC3 direction, which describes 10.1 % of the variability in the sample metabolomes. Two of the individuals in the high LDL increase group were clustered with samples from the low LDL increase group, which implied that these metabolomes were more similar to each other.

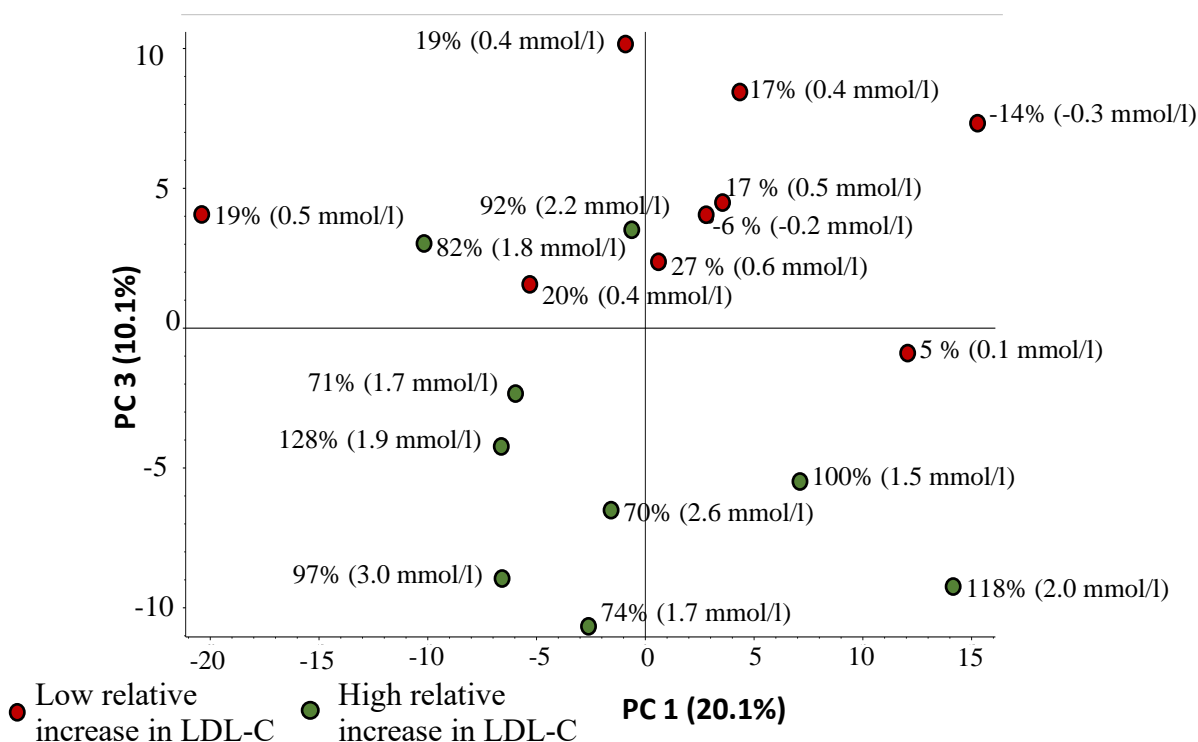


Figure 19 Principal component analysis plot of all detected compounds among those with high relative LDL-C increase (green) and low relative LDL-C increase (red) after three weeks on a ketogenic diet. The numbers represent percentage increase from baseline to end of ketogenic diet, and LDL-C increase levels in mmol/l. PC1 describes 20.1 % of the variability of the samples, and PC3 explains 10.1 %.

Since the PCA plot gave a pattern compatible with a high relative LDL-C increase group and a low relative LDL-C increase group, differential analysis was conducted to determine which components that changed the most between these two groups.

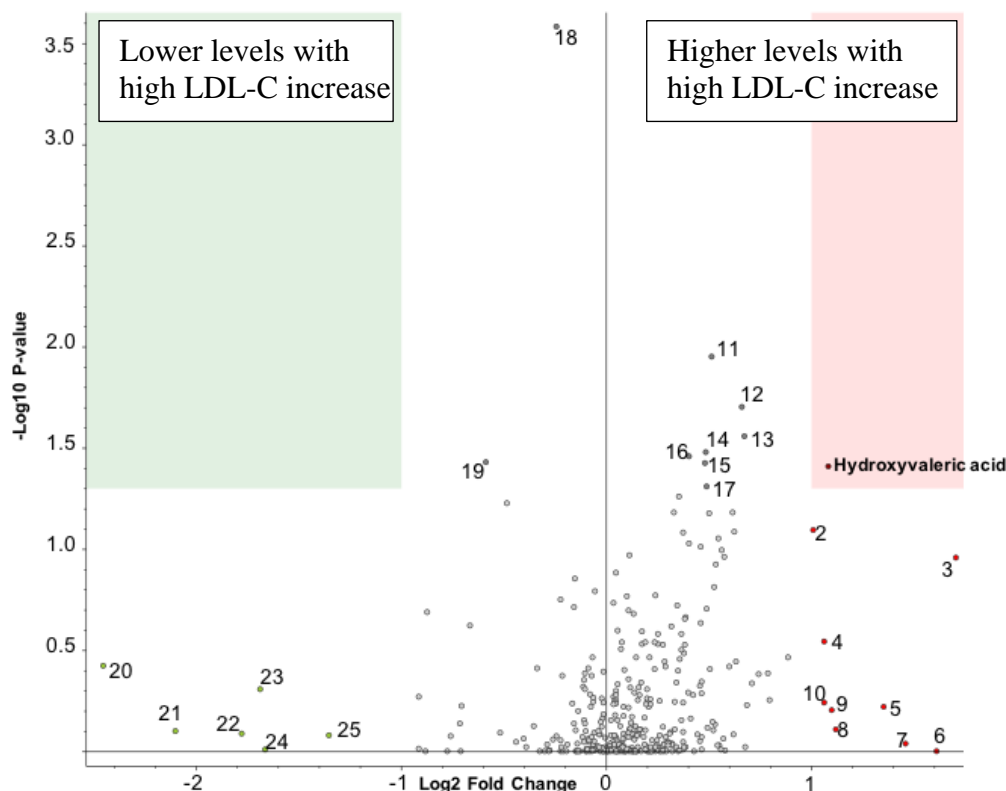


Figure 20 Differential analysis using volcano plot of data obtained from the negative ionization mode analysis of samples following ketogenic diet intervention among the participants with highest and lowest percent increase in LDL-C levels. The points in the green and red areas display metabolites with both fold change values as well as statistical significance. Metabolites identified are presented in bold text. Numbers with the same colors can be fragments or adducts of the same compound.

As shown in **figure 20**, the differential analysis revealed one features detected in significantly higher levels with a log2 fold change >1 among the high LDL-C increase group compared to the low LDL-C increase group, successfully identified as hydroxyvaleric acid. See **table 18** in **appendix** section 7.3.2 for molecular weight and retention time for the non-identified compounds. **Figure 21** display box and whiskers plot for hydroxyvaleric acid.

See **appendix** section 7.3.3 for volcano plot in positive ionization mode.

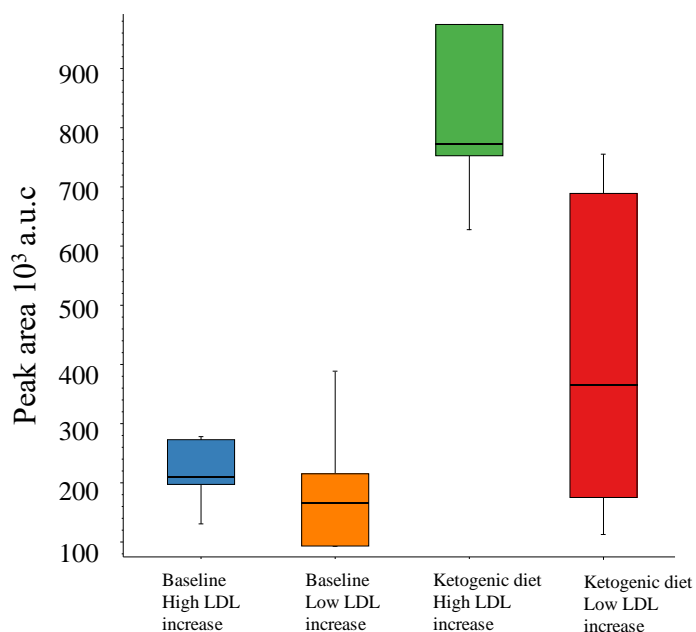


Figure 21 box and whiskers plot for hydroxyvaleric acid. There were no significant differences between the groups at baseline. The group with highest LDL increase had a greatest elevation of hydroxyvaleric acid following the ketogenic diet

Association between amino acids levels and LDL-C

Due to the interesting finding of altered amino acids levels following three weeks on a KD, shown in 4.2.7 **table 11**, correlation analyses were performed to check the association between LDL-C and amino acid levels in blood.

Scatter plots of change in LDL-C levels (mmol/l) on the x-axis vs. change in the amino acid peak area on the y-axis (ratio) following a three week KD are shown in **figure 22**.

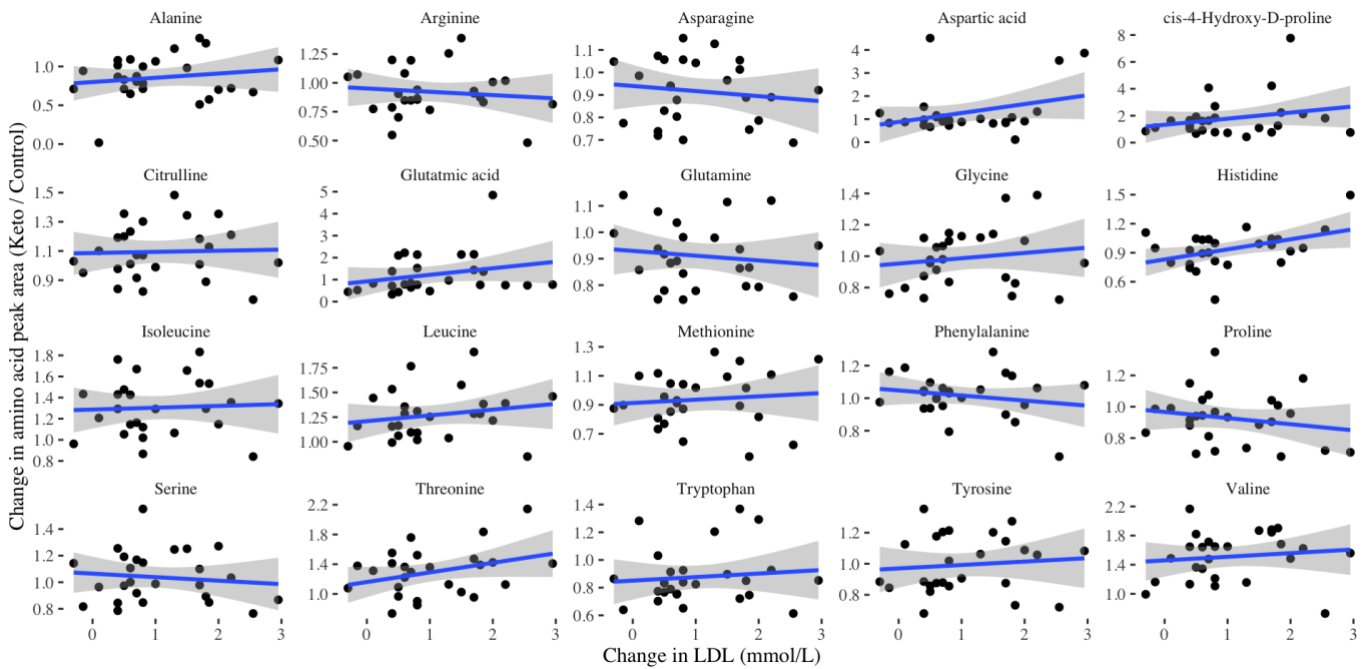


Figure 22 shows scatter plots of the change in LDL(mmol/L) on the x-axis and change in amino acid peak area (keto/habitual diet(control)) on the y-axis.

Univariate correlation matrix for change in LDL-C with change in amino acids showed positive correlations between change in LDL-C and change in histidine, threonine and aspartic acid. Histidine was the only amino acid that had a significantly (p-value: 0.03) positive correlation with change in LDL-C with a correlation coefficient of 0.44 (CI 0.06, 0.71). **Table 13** displays correlation coefficients between change in LDL-C and histidine, threonine and aspartic acid.

Table 13 Correlation coefficients between change in LDL-C and change in the three amino acids histidine, threonine and aspartic acid

	Correlation coefficient	CI	P-value ¹
	LDL-C		
Histidine	0.44	0.06, 0.71	0.03
Threonine	0.34	-0.06, 0.64	0.09
Aspartic acid	0.31	-0.09, 0.62	0.12

¹Pearson's correlation

5 Discussion

This thesis is an extension of the low carbohydrate high fat study from 2011/2012 (70) and investigates the effect on the metabolome of following KD for three weeks using a global metabolomics platform. To the author`s knowledge, this is the first study studying the KD effect with a global metabolomics approach in normal weighed humans. Previous studies have investigated the effect of KD as a weight loss method or as a treatment for different diseases; whereas we have looked at the global effect of three weeks on KD on healthy individuals in a study with no specific goals with respect to health or weight reduction.

5.1 Discussion of subjects and methods

5.1.1 Subject and study design

A great strength of this thesis is the randomized, cross over design in the study of Retterstøl et al. (70). Each participant acted as his or her own control; thus, the KD intervention was evaluated within the same subject and accordingly between subject variability was removed. Furthermore, both groups were combined into one group for evaluating the changes in the metabolomes from the baseline on habitual diet to the end of KD intervention. The two groups completed the study at different time points; however, we showed that this did not influence the results.

The population in the study was almost exclusively clinical nutrition students and employees from The University of Oslo, which makes it a homogenous group. The sample size of 26 participants was relatively small due to exclusion and withdrawal rates. Despite this fact, a highly statistical major change in the metabolomes after following a KD was seen. There was a relatively large variation in abundance of the different metabolites between the individuals, and it is difficult to determine if changes are due to random variation or real biological differences. Furthermore, since global metabolomics gives an enormous amount of complex data, a smaller sample size has the advantage that it takes less data processing time and is more manageable within the time limit of this thesis.

Compliance to the diet regimen

The risk of recall bias was minimized since the food record diet was collected prospectively. However, bias may have been induced by under- or overreporting of participants´ dietary

intakes since the data was self-reported. This is a major challenge in collecting dietary data in nutrition research (74). Weighed food record requires time and effort by the subjects, but since the subjects were primarily students and employees at the University of Oslo, it is likely that these weighed food records were performed meticulously. Due to the limited amount of time a comprehensive comparison of the dietary data and metabolomics data was not conducted. Long-term compliance is typically low and can be a big issue with KD as well as other interventions that require major lifestyle changes. Nine of the participants had adverse effects on the KD, whereas two were serious adverse events and seven experienced common side effects like headache and fatigue. We speculate whether the significantly lower levels of caffeine after KD compared to habitual diet may have caused these reported side effects.

Despite no significantly reported changes in energy intake between habitual and KD, all the participants lost weight during the ketogenic diet intervention. The participants were encouraged to eat according to their energy needs and the same level of physical activity (70). Some of the weight reduction can be related to water loss as a consequence of rapid glycogen depletion (21). In addition, the participants may have taken healthier choices in general or unintentionally reduced their energy intake when following an intervention study. Furthermore, ketosis is suggested to suppress the increase in appetite (75).

5.1.2 Laboratory and LC-MS method

Factors that may have affected the outcome includes variations both in the manual laboratory work and other technical challenges related to sample handling and storage, instrumentation, software etc. The blood samples had been stored in a -80°C freezer since 2012. This results in uncertainty regarding how the long-term storage period may have influenced the samples and potentially reduced or elevated the amount of different analytes in the samples (76).

Dried blood spot (DBS), is blood drops spotted onto a filter paper and dried. We could have used DBS as sample preparation material instead of plasma, since the LC-MS method is optimized for DBS and is used in the routine for the newborn screening because the paper makes the analytes less reactive and more stable. However, plasma was chosen because it was readily available and because there is higher concentrations in undiluted plasma samples than

in plasma applied to and diluted from DBS filter paper. Furthermore, there is a risk that some components are retained on the filter paper and thereby lost for metabolomics analyses.

A strength of the laboratory preparation design was the implementation of a pilot with a limited number of samples to evaluate how the samples were handled and how the whole sample handling process worked. The samples were partitioned in batches over several days, thus theoretically increasing the probability of day to day variation in the sample preparation. However, the only real possible error is pipetting errors, and the platform uses a semi-quantitative method that is not markedly affected by such small pipetting errors.

The LC-MS/MS analysis used is an optimized method to analyze as many compounds as possible, and is a well-established method at the Department of Medical Biochemistry, Oslo University Hospital, Rikshospitalet, Norway.

Since the consumption of fat changed enormously on a KD, the mobile phase for negative ionization was adjusted to increase lipidomic coverage (77, 78). Positive ionization mode detected notable more features than negative ionization. Some features are being ionized only in positive or negative modes, while some features are detected in both positive and negative ionization mode. Consequently, the two ionization modes are complementary.

For data processing, the different settings in the compound discover metabolomics processing workflow affect the results of the analysis and are thus chosen based on well-founded and well-documented criteria to detect and determine the possible identity of the metabolites. Despite this, the identification of metabolites is a comprehensive, complicated, and very time-consuming process that requires analytical chemistry expertise. Features could be generated from the same compounds (fragments and adducts) or related by other means, such as being a part of a shared metabolic pathway. Many of the detected features from the analyses have not been successfully identified. However, these compounds could be of high importance regarding the KD effects, and ideally this should be followed up in the future.

5.2 Discussion of the results

5.2.1 Comparison of ketogenic and habitual diet

An untargeted metabolomics approach reveals changes in a diverse set of compounds, some of which are anticipated and some are unexpected. The expected changes included elevated levels of ketones and lipids. Some unexpected findings includes decreased amino acid levels and correlation between histidine and LDL-C.

First, PCA plots visualized a difference in the metabolomes after following the KD for three weeks. However, the response to the KD varied considerably between the participants. Participant p1, p18 and p22 deviated from the separation seen in the other study participants in the PCA plot based on their diets. Both p1 and p18 moved in the same direction as the other participant`s metabolomes but their KD metabolome did not differ significantly from their habitual diet. Surprisingly, p22 on habitual diet was positioned with the KD metabolomes. There are many possible explanations to this, and this will be discussed later.

5.2.2 Elevated ketone bodies

As expected, levels of ketone bodies (hydroxybutyrate and acetoacetate) were significantly elevated after three weeks on a KD. These findings are in line with well-established knowledge on how the diet alters the energy metabolism from primarily glucose utilization to creating ketone bodies for energy (1, 10, 79). A study conducted by Ulven et al, the Noma study, showed significantly elevated levels of acetoacetate and acetate when saturated fat was replaced with polyunsaturated fat, and accordingly significant reduced LDL-C levels. However no change in hydroxybutyrate was observed. Three of the participants, p1, p18, and p22 (the participants mentioned in 5.2.1) did not have a high increase in ketone bodies (data not shown). The levels of ketone bodies may be influenced by many different factors. Higher consumption of carbohydrates than allowed can interrupt the physiological adaption to fat oxidation. However, food records of these participants showed a carbohydrate intake lower than 20 g per day. Errors in self reporting occurs and cannot be excluded. Another possible explanation is lowered ketone levels after exercise. Unfortunately, no registration of physical activity during the intervention was performed. Furthermore, body composition and variations in basal metabolic rate can also contribute to the metabolite levels observed (20).

Ketone bodies have been used as biomarkers for KD to identify noncompliance to the diet (1). Despite the lower levels of ketone bodies, these participants were not excluded due to the food records that were in agreement with the dietary regimen. In addition, p1 and p18 had a 5 % and -14 % change in LDL-C concentration, respectively, following KD. In contrast, p22 had a 118% increase in LDL-C concentration, thus being one of the participants with the highest LDL-C increase among all of the subjects. A relationship between the concentration of ketone bodies and LDL-C was not seen (data not shown).

More investigation revealed higher ketone bodies in subject p18 prior KD intervention. This finding may indicate that the participant started the KD intervention earlier than supposed. However, this was only seen in plasma and not serum samples and mislabeling or sample mix up prior to this project is therefore probable. This exemplifies that every extra step of manual handling increases the likelihood of mislabeling and sample mix up. Thus, p18 was mistakenly included in the analysis, and should have been excluded. This may have affected the results. Although, excluding one subject will not necessary affect the statistical results due to the sample size, it may weaken the precision of the estimates of the effect (80). At present, since there is not time enough to perform metabolic reanalyses, we can only speculate whether exclusion of this participant would have shown an even greater effect of the ketogenic diet, since both of the included samples from this participant most likely reflected the habitual diet.

The ketogenic diet itself may cause epigenetic modifications, and the effect of the ketone bodies produced have been linked to epigenetic effects (81). Epigenetic changes of KD, such as DNA methylation and histone modifications, contribute to the changes in gene transcription profiles that again lead to metabolic changes and altered metabolomes. The postulated anticonvulsant effect of ketone bodies themselves has been questioned (82), and this remains to be resolved. An animal study conducted by Kennedy et al evaluated the physiological and metabolic effect of a ketogenic diet in mice. They showed a unique gene expression pattern in KD, where expression of genes in fatty oxidation and ketogenic pathways were increased and expression of genes in lipid synthesis pathways were reduced (83). In this present study, Retterstøl et al found that gene expression of transcription factor SREBP-1 was significantly different between the KD and habitual diet, but no differences between the diets in the change of expression of the other lipid-related genes investigated were observed. However, several within-group changes in gene expression in both the

habitual and KD were shown. In addition, borderline reduction in expression of LDL receptor was observed in the KD (70). SREBP-1 is a key regulator of the lipogenic genes in the liver, and represents a possible mechanism by which dietary saturated fats can lead to hyperlipidemia and atherogenesis (84).

5.2.3 Raised fatty acids

Not surprisingly, plasma fatty acids were higher in KD, particularly hydroxyvaleric acid, saturated fatty acids; arachidic acid, medium-chain fatty acids; octadecanedioic acid, polyunsaturated fatty acids; arachidonic acid and adrenic acid and monounsaturated fatty acid; oleic acid. The elevation of free fatty acids as a consequence of lipolysis due to lower plasma glucose and insulin are characteristic of reduced carbohydrate availability. No restriction on the type of fats was given to the participants. Thus, fatty acid content and composition most likely differed between the participants explaining part of the differences observed.

Participants reported a higher intake of saturated fat while on KD. The saturated fatty acids that significantly increase total LDL-C, namely lauric, myristic and palmitic were all elevated at the end of the intervention groups. No correlations between relative LDL-C increase and peak area of the respectively saturated fatty acids were found (data not shown). Endogenously produced fat makes it challenging to use the saturated fatty acids as biomarkers of fat and fatty acid intake (85). Thus, the large variation in LDL-C response between the participants cannot be studied by using plasma levels of saturated fat. As earlier mentioned, cholesterol levels in some participants were almost unaffected by the KD with self-reported high intake of saturate fat, whereas other participants experienced a dramatic increase. This variation in response to a KD suggests that the diet needs to be implemented with caution if one wished to avoid large increases in LDL-C.

A higher intake of mono- and polyunsaturated fat than of saturated fat is preferable to decrease cardiovascular risk. Compared to the recommended guidelines of saturated fat intake < 10 E% (36), the participants had a mean intake of 29.5 E%. The saturated fatty acid arachidic acid (C20:0) had a 164% increase from habitual diet to ketogenic diet, thus being the fatty acid with the largest change. It is probable that the lipid profiles of the subjects in this study would have been improved by a higher proportional intake of foods with polyunsaturated fat. The monounsaturated fatty acid oleic acid, had a 110% increase from

habitual diet to KD. A study conducted by Ulven et al found a reduction in total cholesterol when exchanging a few regularly consumed food items with others with improved fat quality. This may have a potentially impact on future CVD risk (86).

5.2.4 Carnitine levels increased

Carnitine is involved in transport of long chain fatty acids into mitochondria for oxidation. Consistent with prior findings (23), we observed that those coupled to lipid metabolism were elevated, particularly myristoyl-, palmitoyl-, stearonyl- and hexanoylcarnitine. Unfortunately, those coupled to amino acid metabolism were not detected due to the properties of the analytical technique chosen. Requirements of carnitine may increase when ingesting a high fat diet (1). However, no change in free carnitine levels was seen. Carnitine is a reservoir for excess acyl residues, generated by high rates of beta-oxidation in mitochondria (10). Humans are able to synthesize carnitine from the essential amino acids lysine and methionine (10) and also gets it from dietary sources, mainly animal sources.

The Noma study showed significantly reduced levels of myristoyl-, and palmitoylcarnitine when replacing saturated fat with polyunsaturated fat (37). Accordingly, in this thesis, a higher intake of saturated fat significantly increased the levels of myristoyl-, and palmitoylcarnitine. Studies have suggested acylcarnitines to be biomarkers of CVD risk in obese and diabetes patients (87) and to be associated with cardiovascular events in the elderly (88).

5.2.5 Changes in markers of energy metabolism

Some metabolites involved in the energy metabolism were investigated to see if they were up- or down-regulated. Only citric acid, malic acid, alpha-ketoglutaric acid as intermediates in the citric acid cycle were successfully identified, all of which were significantly upregulated. These changes suggest an increased flux through the citric acid cycle. In addition, no significant differences in citrulline and ornithine were observed, indicating that there is no obvious elevation in amino acid degradation. However, degradation products were not investigated. Thus, these findings should be interpreted with care.

5.2.6 Alterations in amino acid levels

In this study, ten amino acids were significantly decreased in connection with the KD. Since amino acids levels were reduced, a possible explanation could be that these amino acids had been used in ketogenesis or gluconeogenesis. In contrast to these findings, the participants reported protein intake that increased significantly from 17.3 E% on habitual diet to 26.3 E% on KD.

Two clinical studies have reported significant increases in GABA levels in the brain following KD treatment of children with refractory epilepsy (89, 90). GABA is synthesized from glutamate, and is the main inhibitory neurotransmitter in the body. Consistent with these findings, plasma levels of GABA were elevated after a KD in our healthy population. However, these results might not be directly comparable due to different biological sample materials and different study populations.

Dietary amino acids have been related to serum lipid levels (73). On comparing the amino acid levels with LDL-C change levels, a significant correlation between histidine and LDL-C was observed. To the best of the author's knowledge, this is one of only a few previous studies that have investigated and compared change in plasma amino acids peak areas with change in LDL-C levels (mmol/L) (37). Similar to our findings, The Noma study with significant reduced levels of LDL-C when replacing saturated fat with polyunsaturated fat, found borderline elevated levels of histidine and glycine, and significantly higher levels of the amino acids serine, asparagine, proline and cystathionine. The clinical significance of this finding remains unanswered and should be further investigated. A study conducted by Teymoori et al investigated the association between dietary amino acids and prospective changes in the serum lipid profile in adults. In contrast to our study, they found no significant correlation with histidine. Despite weak correlations, they found that alanine and tryptophan were associated with higher and lower LDL-C changes respectively. Lysine, alanine, methionine, aspartic acid and alkaline amino acids showed positive association with changes in total cholesterol (91).

As in Cappuccio et al, the acetylated amino acid, acetylglycine, were significantly raised after KD. In this study, a 120 % increase was shown. Acetylglycine has been significantly associated with dietary fiber intake (92). In contrast to these findings, we found no correlation

between acetylglycine and reported dietary fiber intake (data not shown). The relevance of acetylglycine and the effect of ketogenic diet remain unclear.

5.2.7 Comparison of ketogenic and habitual diet metabolomes among those with lowest and highest relative LDL-C increase

Hydroxyvaleric acid was significantly elevated after KD. Hydroxyvaleric acid belongs to the class of fatty acyls known as hydroxy fatty acids. Interestingly, a greater elevation of hydroxyvaleric acid was detected in the group with highest relative LDL-C increase (>70%) compared to the low relative LDL-C increase group (<27%). The box and whiskers plot for hydroxyvaleric acid shows a large variation between the individual peak areas measured, especially in the low relative LDL-C increase group. In literature, too, there are studies suggesting a correlation between 2-hydroxyvaleric acid and LDL-C levels. Trupp et al, studied the effects on intermediary metabolism of reducing LDL-C with statin. 2-hydroxyvaleric acid was significantly decreased after treatment with statin. Furthermore, there was a correlation between low baseline 2-hydroxyvaleric acid levels and decreased LDL-C levels described (93). This finding is consistent with my results, although my study setup demonstrates the correlation between elevated levels of Hydroxyvaleric acid and LDL-C.

Why the change in LDL-C levels following KD varied so much between individuals remains unanswered. However, the metabolites that were found to change markedly and significantly on KD, but that were not identified, could be of importance related to the large inter-individual LDL-C response to KD.

6 Conclusion and future perspective

Metabolomic profiling using high-performance liquid chromatography coupled to mass spectrometry showed:

- Distinct clustering discriminating the ketogenic diet and the habitual diet using principal component analysis (PCA) plot. The features that contribute to the largest variation were therefore similar within each diet group and differed between the groups. However only a few of these features were successfully identified.
- Alterations in ketone bodies, lipids, acylcarnitine and amino acid profiles using the differential analysis; volcano plot.
 - Hydroxybutyrate, acetoacetic acid, trihomomethionine, steryoylcarnitine, acetyl-l-carnitine, hexanoylcarnitine, 3-hydroxyisobutyric acid, hydroxyvaleric acid, arachidic acid, octadecanedioic acid, adrenic acid, n-acetylglycine, arachidonic acid were successfully identified and significantly elevated after following KD diet with a log₂ fold change >1. GABA was significantly elevated, but with a log₂ fold change lesser than 1.
 - Four metabolites; stachydrine, methyl indole 3-acetate, caffeine and N-acetylvaline were successfully identified and significantly decreased after following KD diet with a log₂fold change >1 in positive ionization mode
 - The amino acid profiles among the participants after a KD were altered. Plasma levels of the amino acids threonine, asparagine, tryptophan, proline, b-alanine, leucine, arginine, methionine, glutamine and alanine were decreased, whereas levels of valine and isoleucine were significantly higher after KD.
- Intermediates in citric acid cycle were upregulated, in particular citric acid, malic acid and alpha-ketoglutarate. This may suggest an increased citric acid cycle flux.
- The metabolomic profiles among those with the lowest and highest relative LDL-C increase on KD differed from each other and were similar within each group based on the PCA plot.

- Significantly elevated levels of hydroxyvaleric acid among those with highest relative LDL-C increase.
- Histidine was positively associated with LDL-C

In conclusion, applying global metabolomics in this randomized controlled study identified substantial changes in the metabolome as a response to the dietary intervention. We found alterations in a diverse set of compounds, but unfortunately we were unable to explain why the individual LDL-C levels on KD varied so much.

Future perspective

The global metabolomics platform is implemented as a valuable tool in the diagnostic lab at the Department of Medical Biochemistry at OUS and is used in many different research projects. The analytical method used in this thesis is not optimized for lipids. Further research with a lipidomics platform could be of interest to increase the knowledge of how the ketogenic diet with high fat intake alters the metabolism. Furthermore, a longer intervention time would have been necessary to investigate the more long-term effects of a KD than that of a three week intervention.

In addition, a comprehensive comparison of the dietary data and metabolomics data can address an association between reported food intake and measured peak areas, and find potential biomarkers of a ketogenic diet and ideally also establish reference ranges for a normal metabolome and possibly even the ketogenic metabolome.

Metabolite identification is a difficult and time consuming task, and in this project only a few of the many compounds that changed significantly as a consequence of the dietary intervention were identified due to limited time. Further investigations into metabolite identification and statistical analyses of the results can be done to gain more knowledge of the effect of ketogenic diet on the metabolome.

Why the LDL-C levels varied a lot between the participants deserve further investigations. Ideally, metabolomics can be used to characterize similar metabolomic patterns to be able to identify the mechanism by which a ketogenic diet can cause a marked increase in LDL-C and find underlying susceptibility and resilience factors explaining the large individual variation in response, enabling a personalized way to predict if an individual will likely respond with a marked increase in LDL-C levels on a KD.

References

1. Shaw V. Clinical paediatric dietetics. 4th ed. ed. Oxford: Wiley Blackwell; 2015.
2. Wheless JW. History and Origin of the Ketogenic Diet. In: Stafstrom CE, Rho JM, editors. Epilepsy and the Ketogenic Diet. Totowa, NJ: Humana Press; 2004. p. 31-50.
3. Barañano KW, Hartman AL. The ketogenic diet: uses in epilepsy and other neurologic illnesses. *Curr Treat Options Neurol*. 2008;10(6):410-9.
4. Scholl-Burgi S, Holler A, Pichler K, Michel M, Haberlandt E, Karall D. Ketogenic diets in patients with inherited metabolic disorders. *Journal of inherited metabolic disease*. 2015;38(4):765-73.
5. Om medfødte stoffskiftesykdommer [Internet]. [06.05.20]. Available from: <https://oslo-universitetssykehus.no/fag-og-forskning/laboratorietjenester/-medisinsk-biokjemi/seksjon-for-medfodte-metabolske-sykdommer/om-medfodte-stoffskiftesykdommer>.
6. Kossoff EH, Zupec-Kania BA, Rho JM. Ketogenic diets: an update for child neurologists. *Journal of child neurology*. 2009;24(8):979-88.
7. Boison D. New insights into the mechanisms of the ketogenic diet. *Current opinion in neurology*. 2017;30(2):187-92.
8. Sondike SB, Copperman N, Jacobson MS. Effects of a low-carbohydrate diet on weight loss and cardiovascular risk factor in overweight adolescents. *The Journal of pediatrics*. 2003;142(3):253-8.
9. Yancy WS, Jr., Olsen MK, Guyton JR, Bakst RP, Westman EC. A low-carbohydrate, ketogenic diet versus a low-fat diet to treat obesity and hyperlipidemia: a randomized, controlled trial. *Annals of internal medicine*. 2004;140(10):769-77.
10. Ross CA. Modern nutrition in health and disease. 11th ed. ed. Philadelphia, Pa: Lippincott Williams & Wilkins; 2014.
11. Bye Dan J. The Gale Encyclopedia of Diets: A Guide to Health and Nutrition. *Reference Reviews*. 2009;23(1):42-4.
12. Kossoff E, Zupec-Kania B, Rho J. Ketogenic Diets: An Update for Child Neurologists. *Journal of child neurology*. 2009;24:979-88.
13. Kverneland M, Molteberg E, Haavardsholm KC, Pedersen S, Ramm-Pettersen A, Nakken KO. Diettbehandling av epilepsi [Internet]. [cited 20 03.03]. Available from: <https://tidsskriftet.no/2017/09/klinisk-oversikt/diettbehandling-av-epilepsi#ref4>.
14. Slavin JL, Lloyd B. Health benefits of fruits and vegetables. *Adv Nutr*. 2012;3(4):506-16.
15. Lee PR, Kossoff EH. Dietary treatments for epilepsy: Management guidelines for the general practitioner. *Epilepsy & Behavior*. 2011;21(2):115-21.
16. Nordic nutrition recommendations 2012. Copenhagen: Nordic council of ministers; 2013.
17. Prabhakar A, Quach A, Zhang H, Terrera M, Jackemeyer D, Xian X, et al. Acetone as biomarker for ketosis buildup capability--a study in healthy individuals under combined high fat and starvation diets. *Nutr J*. 2015;14:41-.
18. Vazquez JA, Kazi U. Lipolysis and gluconeogenesis from glycerol during weight reduction with very-low-calorie diets. *Metabolism*. 1994;43(10):1293-9.
19. Veldhorst MAB, Westerterp-Plantenga MS, Westerterp KR. Gluconeogenesis and energy expenditure after a high-protein, carbohydrate-free diet. *The American journal of clinical nutrition*. 2009;90(3):519-26.
20. Laffel L. Ketone bodies: a review of physiology, pathophysiology and application of monitoring to diabetes. *Diabetes/Metabolism Research and Reviews*. 1999;15(6):412-26.

21. O'Neill B, Raggi P. The ketogenic diet: Pros and cons. *Atherosclerosis*. 2020;292:119-26.
22. Kwiterovich PO, Jr., Vining EP, Pyzik P, Skolasky R, Jr., Freeman JM. Effect of a high-fat ketogenic diet on plasma levels of lipids, lipoproteins, and apolipoproteins in children. *Jama*. 2003;290(7):912-20.
23. Cappuccio G, Pinelli M, Alagia M, Donti T, Day-Salvatore DL, Veggiotti P, et al. Biochemical phenotyping unravels novel metabolic abnormalities and potential biomarkers associated with treatment of GLUT1 deficiency with ketogenic diet. *PloS one*. 2017;12(9):e0184022.
24. Urbain P, Strom L, Morawski L, Wehrle A, Deibert P, Bertz H. Impact of a 6-week non-energy-restricted ketogenic diet on physical fitness, body composition and biochemical parameters in healthy adults. *Nutr Metab (Lond)*. 2017;14:17-.
25. Nizamuddin J, Turner Z, Rubenstein JE, Pyzik PL, Kossoff EH. Management and risk factors for dyslipidemia with the ketogenic diet. *Journal of child neurology*. 2008;23(7):758-61.
26. McDonald TJW, Ratchford EV, Henry-Barron BJ, Kossoff EH, Cervenka MC. Impact of the modified Atkins diet on cardiovascular health in adults with epilepsy. *Epilepsy & behavior : E&B*. 2018;79:82-6.
27. Hartman AL, Vining EPG. Clinical Aspects of the Ketogenic Diet. *Epilepsia*. 2007;48(1):31-42.
28. Schaefer EJ, Tsunoda F, Diffenderfer M, Polisecki E, Thai N, Asztalos B. The Measurement of Lipids, Lipoproteins, Apolipoproteins, Fatty Acids, and Sterols, and Next Generation Sequencing for the Diagnosis and Treatment of Lipid Disorders. In: Feingold KR, Anawalt B, Boyce A, Chrousos G, Dungan K, Grossman A, et al., editors. *Endotext*. South Dartmouth (MA): MDText.com, Inc.; 2000.
29. Ference BA, Ginsberg HN, Graham I, Ray KK, Packard CJ, Bruckert E, et al. Low-density lipoproteins cause atherosclerotic cardiovascular disease. 1. Evidence from genetic, epidemiologic, and clinical studies. A consensus statement from the European Atherosclerosis Society Consensus Panel. *European Heart Journal*. 2017;38(32):2459-72.
30. T8.13 Hyperlipidemi [Internet]. [cited 2020 06.05]. Available from: <https://www.legemiddelhandboka.no/T8.13/Hyperlipidemi>.
31. Singh RB, Mengi SA, Xu Y-J, Arneja AS, Dhalla NS. Pathogenesis of atherosclerosis: A multifactorial process. *Exp Clin Cardiol*. 2002;7(1):40-53.
32. Knudsen AK TM, Haaland ØA, Kinge JM, Skirbekk V, Vollset SE. Disease Burden in Norway 2015. Results from the Global Burden of Diseases, Injuries, and Risk Factors Study 2015 (GBD 2015). Public Health Institute Norway May 2017. Report No.: 978-82-8082-840-8.
33. Kannel WB. Metabolic risk factors for coronary heart disease in women: Perspective from the Framingham Study. *American Heart Journal*. 1987;114(2):413-9.
34. ernæring Nrf. Kostråd om fett - en oppdatering og vurdering av kunnskapsgrunnlaget 05/17
35. Linton MRF, Yancey PG, Davies SS, Jerome WG, Linton EF, Song WL, et al. The Role of Lipids and Lipoproteins in Atherosclerosis. In: Feingold KR, Anawalt B, Boyce A, Chrousos G, Dungan K, Grossman A, et al., editors. *Endotext*. South Dartmouth (MA): MDText.com, Inc.; 2000.
36. Kostråd for å fremme folkehelsen og forebygge kroniske sykdommer, Metodologi og vitenskapelig kunnskapsgrunnlag, Nasjonalt råd for ernæring 2011 2011. Report No.: IS-1881.
37. Ulven SM, Christensen JJ, Nygard O, Svoldal A, Leder L, Ottestad I, et al. Using metabolic profiling and gene expression analyses to explore molecular effects of replacing

saturated fat with polyunsaturated fat—a randomized controlled dietary intervention study. *The American journal of clinical nutrition*. 2019;109(5):1239-50.

38. Kostråd om fett, en oppdatering og vurdering av kunnskapsgrunnlaget 05/17 Report No.: IS-2625.
39. Sacks FM, Bray GA, Carey VJ, Smith SR, Ryan DH, Anton SD, et al. Comparison of weight-loss diets with different compositions of fat, protein, and carbohydrates. *The New England journal of medicine*. 2009;360(9):859-73.
40. Shai I, Schwarzfuchs D, Henkin Y, Shahar DR, Witkow S, Greenberg I, et al. Weight Loss with a Low-Carbohydrate, Mediterranean, or Low-Fat Diet. *New England Journal of Medicine*. 2008;359(3):229-41.
41. Wishart DS. Current progress in computational metabolomics. *Briefings in bioinformatics*. 2007;8(5):279-93.
42. Bujak R, Struck-Lewicka W, Markuszewski MJ, Kaliszan R. Metabolomics for laboratory diagnostics. *Journal of Pharmaceutical and Biomedical Analysis*. 2015;113:108-20.
43. Phenotype/ phenotypes [Internet]. [cited 2020 08.05]. Available from: <https://www.nature.com/scitable/definition/phenotype-phenotypes-35/>.
44. Metabolomics: Understanding Metabolism in the 21st Century. In: Birmingham Uo, editor. 2019.
45. Patti GJ, Yanes O, Siuzdak G. Metabolomics: the apogee of the omics trilogy. *Nature Reviews Molecular Cell Biology*. 2012;13(4):263-9.
46. Jurowski K, Kochan K, Walczak-Skierska J, Nska Mg, Piekoszewski W, Buszewski Ba. Analytical Techniques in Lipidomics: State of the Art. 2017.
47. Zhou B, Xiao JF, Tuli L, Ransom HW. LC-MS-based metabolomics. *Mol Biosyst*. 2012;8(2):470-81.
48. Jacob M, Malkawi A, Albast N, Al Bougha S, Lopata A, Dasouki M, et al. A targeted metabolomics approach for clinical diagnosis of inborn errors of metabolism. *Analytica chimica acta*. 2018;1025:141-53.
49. Broadhurst D, Goodacre R, Reinke SN, Kuligowski J, Wilson ID, Lewis MR, et al. Guidelines and considerations for the use of system suitability and quality control samples in mass spectrometry assays applied in untargeted clinical metabolomic studies. *Metabolomics*. 2018;14(6):72.
50. Wilm M. Principles of Electrospray Ionization. *Molecular & Cellular Proteomics*. 2011;10(7):M111.009407.
51. Greaves J, Roboz J. *Mass Spectrometry for the Novice* CRC press; 2014.
52. Podwojski K, Fritsch A, Chamrad DC, Paul W, Sitek B, Stühler K, et al. Retention time alignment algorithms for LC/MS data must consider non-linear shifts. *Bioinformatics*. 2009;25(6):758-64.
53. Ho CS, Lam CW, Chan MH, Cheung RC, Law LK, Lit LC, et al. Electrospray ionisation mass spectrometry: principles and clinical applications. *The Clinical biochemist Reviews*. 2003;24(1):3-12.
54. Silva C, Cavaco C, Perestrelo R, Pereira J, Câmara SJ. Microextraction by Packed Sorbent (MEPS) and Solid-Phase Microextraction (SPME) as Sample Preparation Procedures for the Metabolomic Profiling of Urine. *Metabolites*. 2014;4(1).
55. Gika H, Theodoridis G. Sample preparation prior to the LC–MS-based metabolomics/metabonomics of blood-derived samples. *Bioanalysis*. 2011;3(14):1647-61.
56. Cicalini I, Rossi C, Pieragostino D, Agnifili L, Mastropasqua L, di Ioia M, et al. Integrated Lipidomics and Metabolomics Analysis of Tears in Multiple Sclerosis: An Insight into Diagnostic Potential of Lacrimal Fluid. *Int J Mol Sci*. 2019;20(6):1265.
57. Chepyala D, Kuo HC, Su KY, Liao HW, Wang SY, Chepyala SR, et al. Improved Dried Blood Spot-Based Metabolomics Analysis by a Postcolumn Infused-Internal Standard

- Assisted Liquid Chromatography-Electrospray Ionization Mass Spectrometry Method. *Analytical chemistry*. 2019;91(16):10702-12.
58. Weiss N, Barbier Saint Hilaire P, Colsch B, Isnard F, Attala S, Schaefer A, et al. Cerebrospinal fluid metabolomics highlights dysregulation of energy metabolism in overt hepatic encephalopathy. *Journal of hepatology*. 2016;65(6):1120-30.
 59. Sargent M, Sage A, Wolff C, Mussell C, Neville D, Lord G, et al. Guide to achieving reliable quantitative LC-MS measurements, RSC Analytical Methods Committee 2013.
 60. Xiao JF, Zhou B, Ransom HW. Metabolite identification and quantitation in LC-MS/MS-based metabolomics. *Trends Analyt Chem*. 2012;32:1-14.
 61. Franceschi P, Mylonas R, Shahaf N, Scholz M, Arapitsas P, Masuero D, et al. MetaDB a Data Processing Workflow in Untargeted MS-Based Metabolomics Experiments. *Front Bioeng Biotechnol*. 2014;2:72-.
 62. Bro R, Smilde AK. Principal component analysis. *Analytical Methods*. 2014;6(9):2812-31.
 63. McGarigal K, Stafford S, Cushman S. Ordination: Principal Components Analysis. In: McGarigal K, Stafford S, Cushman S, editors. *Multivariate Statistics for Wildlife and Ecology Research*. New York, NY: Springer New York; 2000. p. 19-80.
 64. Weckwerth W, Morgenthal K. Metabolomics: from pattern recognition to biological interpretation. *Drug Discovery Today*. 2005;10(22):1551-8.
 65. Compound Discoverer User Guide Software Version 3.1 [Internet]. July 2019 [cited 20 06.05]. Available from: <https://assets.thermofisher.com/TFS-Assets/CMD/manuals/man-xcali-98120-compound-discoverer-user-manxcali98120-en.pdf>.
 66. Gibney MJ, Walsh M, Brennan L, Roche HM, German B, van Ommen B. Metabolomics in human nutrition: opportunities and challenges. *The American journal of clinical nutrition*. 2005;82(3):497-503.
 67. Puiggros F, Sola R, Blade C, Salvado MJ, Arola L. Nutritional biomarkers and foodomic methodologies for qualitative and quantitative analysis of bioactive ingredients in dietary intervention studies. *Journal of chromatography A*. 2011;1218(42):7399-414.
 68. Esko T, Hirschhorn JN, Feldman HA, Hsu YH, Deik AA, Clish CB, et al. Metabolomic profiles as reliable biomarkers of dietary composition. *The American journal of clinical nutrition*. 2017;105(3):547-54.
 69. Andersen MB, Rinnan A, Manach C, Poulsen SK, Pujos-Guillot E, Larsen TM, et al. Untargeted metabolomics as a screening tool for estimating compliance to a dietary pattern. *Journal of proteome research*. 2014;13(3):1405-18.
 70. Retterstol K, Svendsen M, Narverud I, Holven KB. Effect of low carbohydrate high fat diet on LDL cholesterol and gene expression in normal-weight, young adults: A randomized controlled study. *Atherosclerosis*. 2018;279:52-61.
 71. M. D. Atkins RC. *Dr. Atkins' New Diet Revolution*: M. Evans; 2002.
 72. Milli-Q Reference Water Purification System [Internet]. 2020 [cited 2020 13.02]. Available from: https://www.merckmillipore.com/NO/en/product/Milli-Q-Reference-Water-Purification-System,MM_NF-C79625.
 73. Oda H. Functions of Sulfur-Containing Amino Acids in Lipid Metabolism. *The Journal of Nutrition*. 2006;136(6):1666S-9S.
 74. Institute of M, Food, Nutrition B, Committee on Dietary Risk Assessment in the WICP. *Dietary Risk Assessment in the WIC Program*: National Academies Press; 2002.
 75. Gibson A, Seimon R, Lee C, Ayre J, Franklin J, Markovic T, et al. Do ketogenic diets really suppress appetite? A systematic review and meta-analysis. *Obesity reviews : an official journal of the International Association for the Study of Obesity*. 2014;16.

76. Enroth S, Hallmans G, Grankvist K, Gyllensten U. Effects of Long-Term Storage Time and Original Sampling Month on Biobank Plasma Protein Concentrations. *EBioMedicine*. 2016;12:309-14.
77. Cajka T, Fiehn O. Increasing lipidomic coverage by selecting optimal mobile-phase modifiers in LC–MS of blood plasma. *Metabolomics*. 2016;12(2):34.
78. Creydt M, Fischer M. Plant Metabolomics: Maximizing Metabolome Coverage by Optimizing Mobile Phase Additives for Nontargeted Mass Spectrometry in Positive and Negative Electrospray Ionization Mode. *Analytical chemistry*. 2017;89(19):10474-86.
79. Masood W, Annamaraju P, Uppaluri KR. *Ketogenic Diet*. StatPearls. Treasure Island (FL): StatPearls Publishing; 2020.
80. Fergusson D, Aaron SD, Guyatt G, Hébert P. Post-randomisation exclusions: the intention to treat principle and excluding patients from analysis. *BMJ (Clinical research ed)*. 2002;325(7365):652-4.
81. Ruan AH-B, Crawford AP. Ketone bodies as epigenetic modifiers. *Current Opinion in Clinical Nutrition and Metabolic Care*. 2018;21(4):260-6.
82. Nylen K, Likhodii S, Burnham WM. The ketogenic diet: proposed mechanisms of action. *Neurotherapeutics*. 2009;6(2):402-5.
83. Kennedy AR, Pissios P, Otu H, Xue B, Asakura K, Furukawa N, et al. A high-fat, ketogenic diet induces a unique metabolic state in mice. *American Journal of Physiology-Endocrinology and Metabolism*. 2007;292(6):E1724-E39.
84. Lin J, Yang R, Tarr PT, Wu PH, Handschin C, Li S, et al. Hyperlipidemic effects of dietary saturated fats mediated through PGC-1 β coactivation of SREBP. *Cell*. 2005;120(2):261-73.
85. Arab L. Biomarkers of fat and fatty acid intake. *J Nutr*. 2003;133 Suppl 3(3):925s-32s.
86. Ulven S, Leder L, Elind E, Ottestad I, Christensen J, Telle-Hansen V, et al. Exchanging a few commercial, regularly consumed food items with improved fat quality reduces total cholesterol and LDL-cholesterol: a double-blind, randomised controlled trial. *British Journal of Nutrition*. 2016;116:1-11.
87. Mihalik SJ, Goodpaster BH, Kelley DE, Chace DH, Vockley J, Toledo FG, et al. Increased levels of plasma acylcarnitines in obesity and type 2 diabetes and identification of a marker of glucolipotoxicity. *Obesity (Silver Spring, Md)*. 2010;18(9):1695-700.
88. Rizza S, Copetti M, Rossi C, Cianfarani MA, Zucchelli M, Luzi A, et al. Metabolomics signature improves the prediction of cardiovascular events in elderly subjects. *Atherosclerosis*. 2014;232(2):260-4.
89. Wang ZJ, Bergqvist C, Hunter JV, Jin D, Wang D-J, Wehrli S, et al. In vivo measurement of brain metabolites using two-dimensional double-quantum MR spectroscopy—exploration of GABA levels in a ketogenic diet. *Magnetic Resonance in Medicine*. 2003;49(4):615-9.
90. Dahlin M, Elfving A, Ungerstedt U, Amark P. The ketogenic diet influences the levels of excitatory and inhibitory amino acids in the CSF in children with refractory epilepsy. *Epilepsy research*. 2005;64(3):115-25.
91. Teymoori F, Asghari G, Salehi P, Sadeghian S, Mirmiran P, Azizi F. Are dietary amino acids prospectively predicts changes in serum lipid profile? *Diabetes & metabolic syndrome*. 2019;13(3):1837-43.
92. Lustgarten MS, Price LL, Chalé A, Fielding RA. Metabolites related to gut bacterial metabolism, peroxisome proliferator-activated receptor- α activation, and insulin sensitivity are associated with physical function in functionally-limited older adults. *Aging Cell*. 2014;13(5):918-25.

93. Trupp M, Zhu H, Wikoff WR, Baillie RA, Zeng ZB, Karp PD, et al. Metabolomics reveals amino acids contribute to variation in response to simvastatin treatment. PloS one. 2012;7(7):e38386.

Appendix

- Appendix 7.1 Sample preparation
- Appendix 7.2 Comparison of ketogenic and habitual diet
- Appendix 7.3 Comparison of ketogenic and habitual diet metabolomes among those
with lowest and highest relative LDL-C increase
- Appendix 7.4 Approval by the Regional Ethical Committee for Medical Research

7 Appendix

7.1 Sample preparation

7.1.1 Sample preparation set up

The sample preparations were distributed over four and three days for serum and plasma

Serum

90 samples - 4 days + PQC + processed blank

Day 1: 8 subjects x 3 visits =>24 samples

Day 2: 7 subjects x 3 visits=>21 samples

Day 3: 8 subjects x 3 visits =>24 samples

Day 4: 7 subjects x 3 visits =>21 samples

LC-MS analysis time: FullMS Pos + FullMS Neg = 6 days

Plasma

90 samples - 3 days + PQC + processed blank

Day 1:15 subjects x 3 visits => 45 samples

Day 2: 8 subjects x 3 visits => 24 samples

Day 3: 7 subjects x 3 visits => 21 samples

LC-MS analysis time: FullMS Pos + FullMS Neg = 6 days

7.1.2 PQC preparation

PQCmain (from day 1)

- Check precision in the PCA plot
- Correct for analytical drift for each metabolite using linear regression

PQCday (2,3,4)

- Pooled all samples from day 2-4
- MS/MS spectra to provide information about the chemical structure to identify metabolites

PQC procedure

PQCmain (day1)

- Collected all biological samples from day 1 in a 1,5 ml Eppendorf tube (24 biological samples)
- Mixed on a table vortex for 5 seconds
- Aliquoted 30 µl to five Eppendorf tubes
- Prepared the samples separately (see protein precipitation)
- Collected all the samples to a 1,5 ml Eppendorf tube
- Aliquoted to eight HPLC vials

PQCday2

- Collected all biological samples from day 2 in a 1,5 ml Eppendorf tube (21 biological samples)
- Mixed on a table vortex for 5 seconds
- See protein precipitation
- Transferred to a HPLC vial

PQCday3

- Collected all biological samples from day 3 in a 1,5 ml Eppendorf tube (24 biological samples)
- Mixed on a table vortex for 5 seconds
- See protein precipitation
- Transferred to a HPLC vial

PQCday4

- Collected all biological samples from day 4 in a 1,5 ml Eppendorf tube (21 biological samples)
- Mixed on a table vortex for 5 seconds
- See protein precipitation
- Transferred to a HPLC vial

PQCtotal

- Collected 190µl from PQCday1 (main), day2, day3 and day4
- Mixed on a table vortex for 5 seconds
- Aliquoted 30 µl to five Eppendorf tubes
- Prepared the samples separately (protein precipitation)
- Collected 30 µl of all the samples to a 1,5 ml Eppendorf tube
- Transferred to a HPLC vial

7.1.3 Analysis sequence

Table 14 The analysis sequence applied in the thesis. A blank sample was analyzed at the start of the analytical batch, a blank extraction sample was analyzed twice, followed by two system conditioning QC samples. Five biological samples were analyzed between each PQC. A set of PQC samples was applied separately at the end of the run for MS/MS data acquisition.

Table 14 The analysis sequence

Injection no.	Sample type	Comments
1	Blank sample	Water
2	Process blank sample	Extraction of background
3	System conditioning QC sample 1	Not included in data processing
4	System conditioning QC sample 2	Not included in data processing
5	Pooled QC main sample 1	
6	Pooled QC main sample 2	
7	Biological sample	
8	Biological sample	
9	Biological sample	
10	Biological sample	
11	Biological sample	
12	Pooled QC main sample 3	
13	Biological sample	
.	Biological sample	
.	Biological sample	
.	Biological sample	
35	Biological sample	
36	Pooled QC main sample 8	
37	Pooled QC day2 MSMS	
38	Biological sample	
.	Biological sample	
.	Biological sample	
.	Biological sample	
109	Biological sample	
110	Pooled QC main sample 19	
111	Biological sample	
112	Biological sample	
113	Biological sample	
114	Biological sample	
115	Biological sample	
116	Pooled QC main sample 20	
117	Pooled QC main sample 21	
118	Pooled QC main MSMS	For identification
119	Pooled QC day4 MSMS	For identification
120	Pooled QC total MSMS	For identification

7.2 Comparison of ketogenic and habitual diet

7.2.1 Exclusion of participants because of non-compliance

All three blood samples were required. Since p25 was missing a blood sample, this participant was excluded from the analysis.

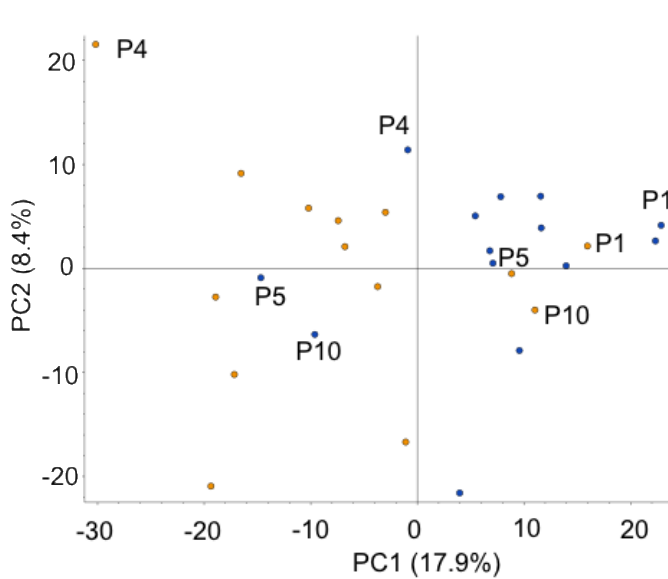


Figure 23 PCA plot of the LK group in positive ionization mode

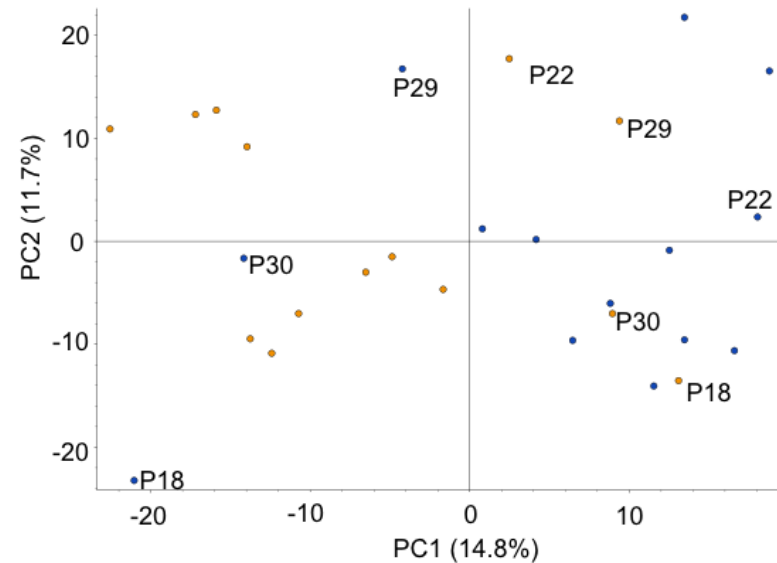


Figure 24 PCA plot of the KL group in positive ionization mode

As shown in **figure 23 and 24**; p5, p10, p29 and p30 deviated from the separation seen in the other study participants in the PCA plot based on their diets both in plasma and serum samples. High levels of b-hydroxybutyrate prior intervention (shown in figure 25 and 26) and low levels of b-hydroxybutyrate during the intervention identified non-compliance, or at least insufficient adherence, to the ketogenic diet (shown in figure 27 and 28). Further investigation revealed that p30 started the ketogenic diet earlier, and the participant was recoded and included in the analysis, while p5, p10 and p29 were excluded for further analysis.

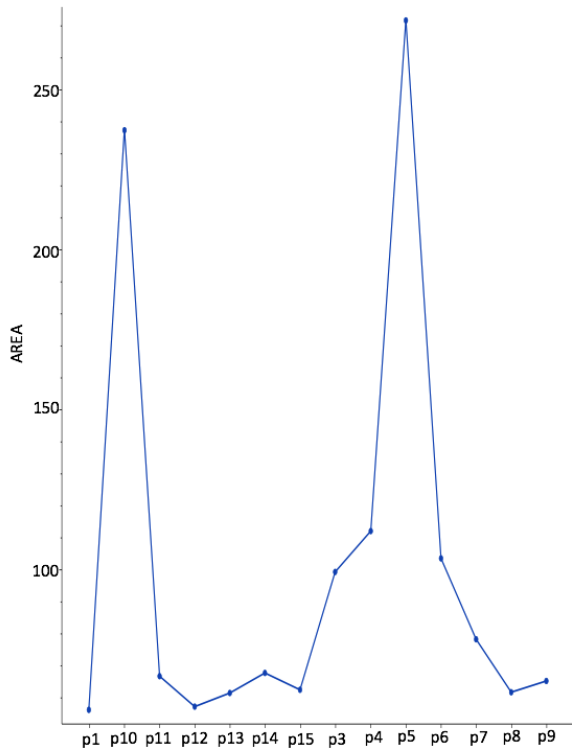


Figure 25 relative abundance of hydroxybutyrate prior to ketogenic diet in the LK group, where p10 and p5 shows elevated levels of the ketone body

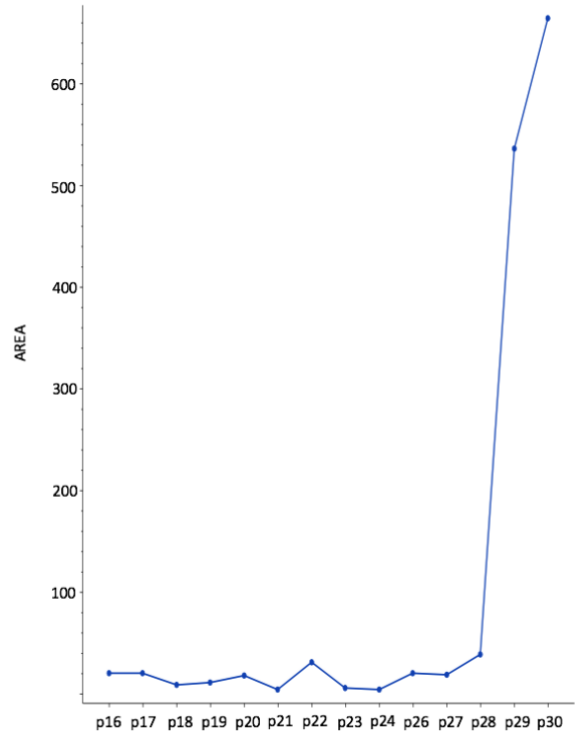


Figure 26 relative abundance of hydroxybutyrate prior to ketogenic diet in the KL group, where p29 and p30 have elevated levels of the ketone body

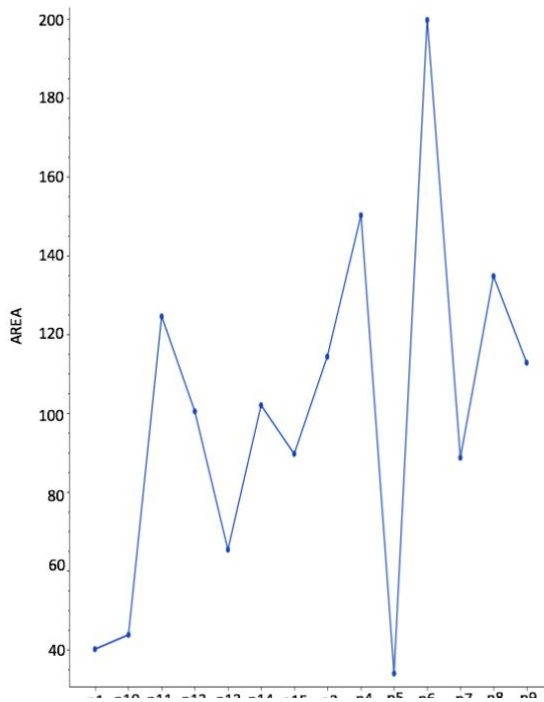


Figure 27 relative abundance of hydroxybutyrate on ketogenic diet in the LK group, where p1, p5 and p10 shows decreased levels of the ketone body

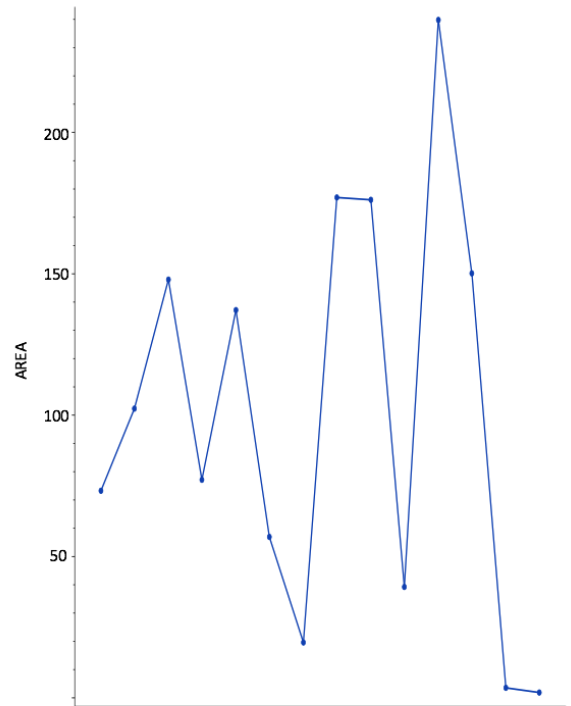


Figure 28 relative abundance of hydroxybutyrate on ketogenic diet in the KL group, where p29 and p30 shows decreased levels of the ketone body

7.2.2 Did study participation affect the habitual diet?

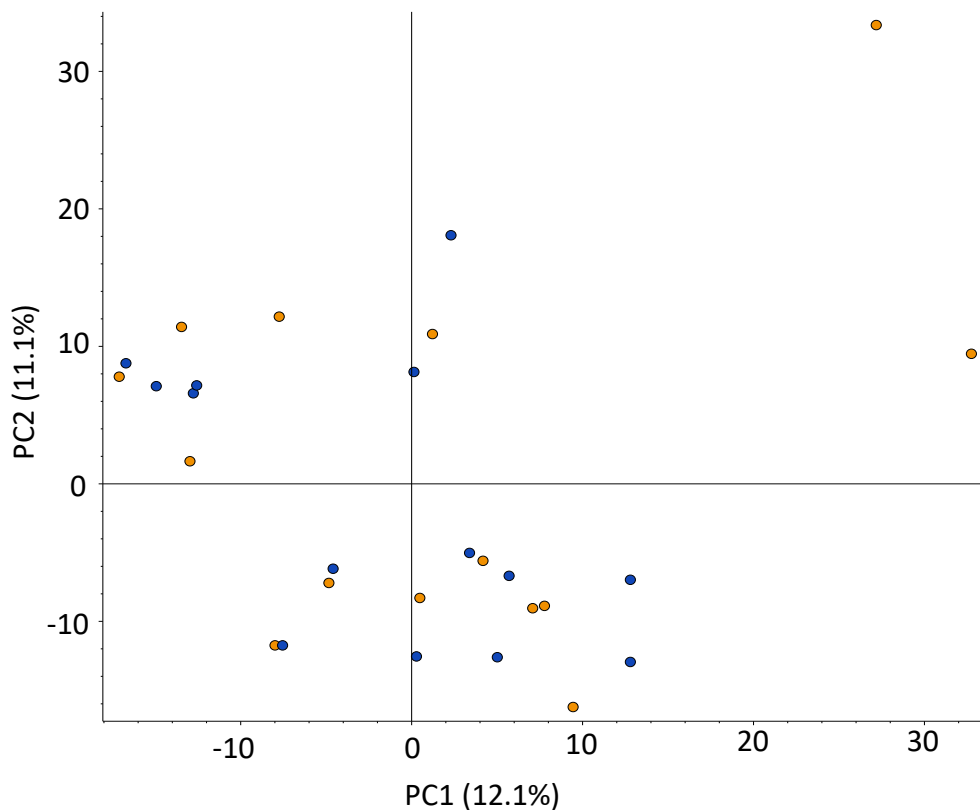


Figure 29 PCA plot of habitual diet at visit1 and visit2 for the KL group. Blue points represent visit1 and yellow points represents visit2.

The principal component analysis (PCA) plot showed no substantial difference between the metabolomes of the participants at these two visits on the habitual diet as shown in **figure 29**.

7.2.3 PCA plot with quality controls

Figure 30 shows the PCA in positive ionization mode. PCA plot from the analysis run in positive ionization mode of all samples taken at baseline (habitual diet) and immediately following intervention with KD among all the participants. The PCA plot shows the same pattern as negative ionization mode, with a distinct clustering discriminating the ketogenic and the habitual diet groups with a few exceptions (p18, p22 and p1). The pooled quality control (PQC) samples were focused and centered, indicating that the results were not affected by analytical drift.

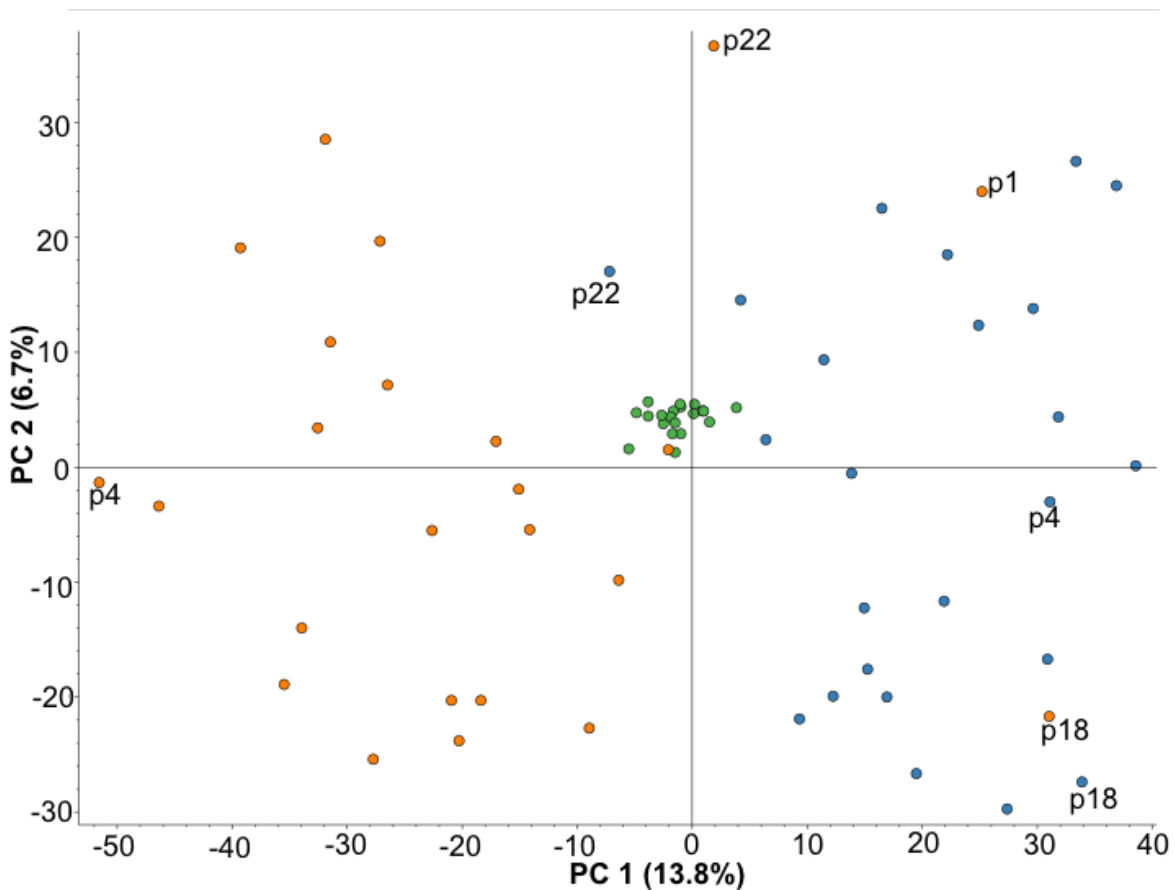


Figure 30 PCA plot from the analysis in positive mode that display a separation between the habitual diet and the ketogenic diet, except from some individuals (p18,p22, p1) . A point represents all the metabolites detected in that sample. An orange colored point represents the metabolome with a ketogenic diet and a blue colored point represent the metabolome at baseline (habitual diet). The green colored point represents the pooled quality control samples.

7.2.4 Molecular weight and retention time to volcano plot

Molecular weight and retention time to detected compounds in positive and negative ionization mode respectively are listed in **table 14** and **table 15**.

Table 15 Molecular weight and retention time to detected compounds in positive ionization mode

Molecular				Molecular			
no	weight	RT	Suggestions	no	weight	RT	Suggestions
2	186.05040	2.161	hydroxybutyrate adduct	46	270.25597	18.794	margaric acid
3	202.02446	2.148	hydroxybutyrate adduct	47	576.33050	19.694	
4	200.03534	2.155	hydroxybutyrate adduct	48	117.07901	2.701	5-aminovaleric acid γ-aminobutyric acid
5	298.12653	2.449		49	103.06347	2.439	(GABA)
6	568.54353	19.687		50	792.37102	18.466	similar to 53
7	448.13769	12.084		51	394.27536	19.697	similar to 41
8	296.27147	19.605		52	448.27791	19.689	
9	749.53691	29.397		53	304.24036	18.464	arachidonic acid
10	590.52547	19.685	similar to Stearic acid	54	131.09461	3.393	beta-leucine
11	616.46583	17.904	similar to nr 12	55	406.27522	18.724	
12	534.46245	17.921		56	836.36305	17.915	
13	156.07864	12.054	2,5- dimethyl-4 ethoxy-3(2H)- furanone	57	354.21707	17.925	
14	512.48037	17.929	similar to nr 12/11	58	574.31497	18.724	
15	188.14132	12.927	5-hydroxydecanoic acid	59	848.36372	18.723	
16	118.06305	2.475		60	342.27720	18.726	
17	117.04265	2.208		61	152.06856	2.569	xylitol
18	558.46271	17.974	similar to nr 12/11	62	308.27166	19.699	
19	586.49409	18.724	similar to oleic acid	63	436.30105	18.723	
20	564.51207	18.722		64	364.25917	18.723	similar to 66
21	536.48091	17.974	5-9Z-9-Hexadecenoyloxy Octadecanoic acid	65	184.14634	13.958	similar to 80
22	462.19001	13.523		66	282.25588	18.724	Oleic acid
23	350.21306	17.113		67	862.37931	18.727	
24	224.17783	15.327		68	592.30767	16.375	
25	581.36952	21.947	similar to sorbitan monooleate	69	770.38643	18.725	
26	534.05004	12.249		70	392.25968	18.725	
27	506.01921	11.933		71	588.33069	19.725	
28	582.46259	18.018		72	768.37082	18.024	
29	560.48094	18.026	similar to 9-nitrooleate	73	362.24356	18.019	
30	616.11061	11.969		74	586.31503	18.018	
31	184.14622	11.893		75	340.26143	18.017	
32	520.03423	12.115		76	378.24393	18.722	
33	481.13790	11.922	similar to IpA	77	404.25948	18.013	
34	189.07901	11.918	IpA	78	268.24025	17.891	
35	154.02670	11.164		79	280.24024	18.015	linoelaidic acid
36	397.04708	11.360		80	476.20525	13.956	Similar to 65
37	380.25956	19.689		81	376.22825	18.016	
38	772.40210	19.696		82	390.24403	18.020	
39	366.27469	19.696	similar to 41	83	227.08265	3.399	
40	174.01644	1.642		84	268.13108	13.163	
41	284.27156	19.696	stearic acid	85	850.37925	19.692	
42	283.12676	2.428	similar to gaba	86	444.24654	18.021	
43	408.29087	19.686		87	572.29923	18.020	
44	332.27179	20.044	adrenic acid	88	312.30290	21.564	arachidic acid dihomo-γ- linolenic acid
45	590.34624	19.696	Similar to 41	89	306.25603	19.009	

Molecular				Molecular			
no	weight	RT	Suggestions	no	weight	RT	Suggestions
90	846.34819	18.020		107	131.06951	2.710	creatine
91	226.19343	15.735	myristoleic acid	108	766.35511	17.492	
92	170.13089	13.521		109	190.04781	2.97	
93	742.35504	17.109		110	164.06858	2.605	
94	278.22458	17.491	alpha Eleostearic acid	111	224.08964	2.578	
95	360.22790	17.494		112	128.05863	2.657	dimethylhydantoin
96	254.22464	17.109	hexadecenoic acid	113	228.07249	2.371	
97	446.26220	18.724		114	119.05830	2.341	threonine
98	330.25610	19.481		115	255.12167	2.366	
99	134.02157	1.696	malic acid	116	234.08524	1.731	
100	314.24593	17.114		117	132.05350	2.305	asparagine
101	820.33228	17.116		118	146.06917	2.366	glutamine
102	336.22787	17.109		119	326.13263	2.379	
103	618.32349	17.114		120	1596.10110	2.858	
104	310.28731	20.499					
105	364.22840	17.109					
106	420.24660	17.924					

Table 16 Molecular weight and retention time to detected compounds in negative ionization mode

Molecular				Molecular			
no	weight	RT	Suggestions	no	weight	RT	Suggestions
1	192.02135	4.883		30	750.53630	26.437	
2	104.04729	4.864		31	789.49773	26.761	
3	126.02920	4.900		32	723.51922	26.032	
4	206.03694	4.920		33	884.71328	20.312	
5	142.00226	4.440		34	751.54966	26.760	(1Z-octadecenyl)-2-aracgidonoyl-sn-glycero-3-phosphoethanolamine
6	206.01018	4.925		35	749.53469	26.421	
7	154.06046	11.968		36	806.66646	19.566	
8	86.03679	4.851	biacetyl/ succinaldehyd/ fragment acetoacetate	37	397.31900	13.870	
9	187.99959	4.917		38	118.06293	8.998	
10	204.07617	14.246		39	208.09446	4.443	
11	230.03687	4.819		40	259.17813	11.932	hexanoylcarnitine
12	331.25403	13.631		41	789.56389	28.941	
13	303.22300	13.175		42	285.19379	12.303	
14	273.17591	12.723		43	604.44382	19.686	
15	427.36595	15.005	stearoylcarnitine	44	369.28751	13.347	cis-5-tetradecenoylcarnitine
16	256.13936	2.708		45	546.39584	18.784	
17	203.11562	3.674	Acetyl-L-Carnitine	46	145.07377	2.468	
18	-	-		47	320.07765	8.979	
19	371.30317	13.553		48	246.15914	16.811	
20	178.06043	13.845	Similar to N-Acetyl-L-cysteine	49	143.09453	4.163	stachydrine
21	240.03076	13.854		50	466.26897	17.858	
22	191.09766	11.388	trihomomethionine	51	272.03113	13.439	
23	102.03163	4.756	acetoacetic acid	52	189.07884	13.440	methyl indole-3-acetate
24	104.04729	104.04729	hydroxybutyrate	53	129.05783	13.439	
25	201.07891	11.762	2-amino-2,3-biphenyldiol - PC(o-	54	280.16476	14.985	
26	767.58190	28.873	18:2(9Z,12Z)/18:2(9Z,12Z))	55	210.12291	12.931	
27	745.50122	26.032		56	194.08028	12.762	caffeine
28	771.51647	26.433		57	159.08939	2.736	N-Acetylvaline
29	556.44392	19.571					

7.2.5 Peak areas of significantly altered identified metabolites

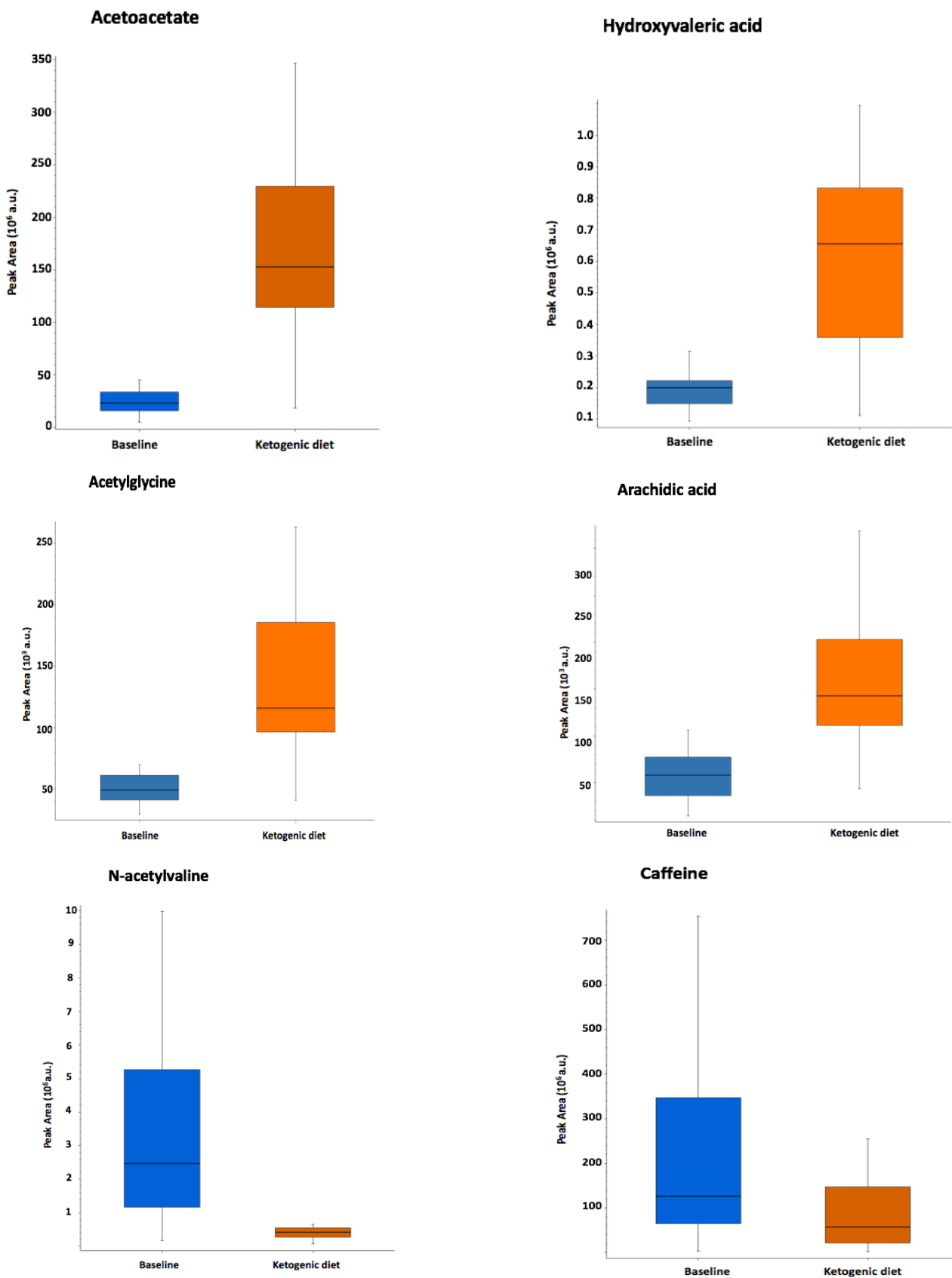


Figure 31 Peak areas of significantly altered identified metabolites

7.3 Comparison of ketogenic and habitual diet metabolomes among those with lowest and highest relative LDL-C increase

7.3.1 LDL-C change values for each participant

Table 17 gives the LDL-C change values estimated for each participant based on data collected in 2012. Changes are given both based on absolute values and LDL-C% change. “H” and “L” represent the group with high LDL-C change and low LDL-C change respectively.

Table 17 LDL-C change levels in mmol/l and percentage change

ID	LDL-C change mmol/L ¹	LDL-C %Change ²	Group
P17	1.9	128 %	H
P22	2.0	118 %	H
P21	1.5	100 %	H
P28	3.0	97 %	H
P11	2.2	92 %	H
P12	1.8	82 %	H
P27	1.7	74 %	H
P8	1.7	71 %	H
P26	2.6	70 %	H
P7	1.3	43 %	
P23	0.8	40 %	
P6	0.7	39 %	
P14	0.6	38 %	
P2	0.8	32 %	
P30	0.8	32 %	
P3	0.7	32 %	
P24	1	26 %	
P15	0.6	27 %	L
P20	0.4	20 %	L
P4	0.5	19 %	L
P13	0.4	19 %	L
P9	0.4	17 %	L
P16	0.5	17 %	L
P1	0.1	5 %	L
P19	-0.2	-6 %	L
P18	-0.3	-14 %	L

¹ Group1: $\text{visit2} - \text{visit1}$, Group2: $\text{visit 3} - \frac{\text{visit1} + \text{visit2}}{2}$

² Group1: $\frac{\text{LDLchange}}{\text{visit1}}$, Group2: $\text{LDLchange} \cdot \frac{\text{visit1} + \text{visit2}}{2}$

7.3.2 Table of features that were significantly changed in amount in volcano plot in negative ionization mode

Table 18 Features detected from volcano plot in negative ionization mode LDL-C comparison

no	Molecular weight	RT	Suggestions
2	298.12653	2.452	
3	477.28551	18.920	2-linoleoyl-sn-glycero-3-phosphoethanolamine
4	312.23023	14.385	(10E,12Z)-9-Hydroperoxy-10,12-octadecadienoic acid
5	568.54390	19.693	Stearic acid
6	590.52547	19.668	
7	214.03020	12.586	
8	179.05827	7.002	Hippuric acid
9	513.27633	13.919	
10	353.02354	6.991	
11	510.04712	11.225	
12	698.07895	11.518	
13	130.06307	4.589	(hydroxyethyl) methacrylate – ethyl-acetoacetate
14	579.35364	20.281	similar to linoleic acid
15	311.12181	2.435	Similar to 16
16	131.05824	2.443	cis-4-hydroxy-D-proline
17	298.97981	2.319	
18	227.08266	3.396	similar to N-acetyl- D-alloisoleucine
19	762.58978	28.760	
20	555.35394	21.003	
21	198.98060	1.990	
22	161.03248	1.694	
23	204.00946	11.524	
24	202.03007	12.346	
25	484.03180	11.078	
26	218.02397	12.143	
27	246.05604	11.922	
28	481.13797	11.928	

7.3.3 Volcano plot in positive ionization mode



Figure 32 Differential analysis using volcano plot of data obtained from the positive ionization mode analysis of samples following ketogenic diet intervention among the participants with highest and lowest percent change in LDL-C levels. The points in the green and red areas display metabolites with both fold change values as well as statistical significance. Numbers with the same colors can be fragments or adducts of the same compound.

Table 19 shows molecular weight and retention time to detected compounds in positive ionization mode

7.3.1 Table of features that were significantly changed in amount in volcano plot in positive ionization mode

Table 19 Molecular weight and retention time to detected compounds in positive ionization mode among the participants with highest and lowest percent change in LDL-C levels.

	Molecular weight	RT	Suggestions		Molecular weight	RT	Suggestions
1	1076.61122	19.231		40	131.05814	2.400	cis-4-hydroxy-D-proline/ 5-aminolevulinic acid tetraacetylenediamine (hydroxyprolylproline)
2	1058.62837	19.196		41	228.11080	2.741	
3	557.27894	19.235		42	845.58119	21.137	
4	542.31761	18.660		43	503.29845	18.915	
5	188.04481	12.800		44	1553.95868	19.178	
6	320.07765	8.979		45	990.66326	19.149	
7	192.05764	8.972	(cys-ala)	46	471.29595	16.254	
8	519.33230	19.226	(1-linoleoyl-sn-glycerol-3-phosphocholine	47	449.31382	16.228	
9	220.02578	8.992		48	1080.64255	20.015	
10	234.03178	8.991	Similar as 9	49	195.05306	11.777	
11	220.05248	8.990	Similar as 9	50	137.04756	3.711	4-aminobenzoic acid
12	140.04481	8.985	Similar as 7	51	1084.67383	21.080	
13	118.06293	8.999	Similar as 9	52	518.21256	13.719	
14	1012.64578	19.151		53	496.23036	13.704	
15	750.53630	26.445		54	270.03384	12.227	
16	817.59447	29.979		55	301.22496	12.433	
17	427.36595	15.003	stearoylcarnitine	56	367.06624	12.361	
18	704.58190	26.033			57	104.02614	12.361
19	232.14200	2.705		58	179.05807	12.361	Hippuric acid
20	676.55104	24.865		59	232.11855	12.910	
21	204.03977	11.845		60	263.01035	12.363	
22	773.5335	26.761		61	249.03106	12.363	
23	339.33445	14.195	palmitylcarnitine	62	332.19591	16.881	
24	1585.07178	25.719			63	167.03777	10.947
25	227.92603	1.647		64	150.13674	2.437	
26	792.53577	25.743		65	199.00676	11.275	
27	311.12114	2.404		66	-	-	
28	164.08378	24.989		67	430.26923	16.283	
29	767.58190	28.906		68	155.06925	3.151	L-Histidine
30	202.01514	8.986		69	356.19590	14.537	
31	477.28526	17.097		70	466.26922	18.564	
32	725.05014	25.712		71	316.20356	14.537	
33	379.24843	16.025	Sphingosine 1-phosphate	72	444.28715	18.564	
34	1449.09763	25.723			73	173.14152	12.644
35	703.15069	2.367		74	155.09461	12.432	
36	245.16238	-	2-methylbutyrylcarnitine	75	232.10686	15.124	Succinyl -L -Ornithine
37	85.08908	3.514			76	700.19416	
38	780.53293	28.327		77	498.31642	16.280	
39	557.28788	19.238		78	129.05782	13.439	



Region: REK sør-øst	Saksbehandler: Finn Skre Fjordholm	Telefon: 22845821	Vår dato: 18.06.2019	Vår referanse: 2011/1365 REK sør-øst D
			Deres dato: 23.05.2019	Deres referanse:

Vår referanse må oppgis ved alle henvendelser

Kjetil Retterstøl
Oslo universitetssykehus

2011/1365 Virkningen av høyfettdiett

Forskningsansvarlig: Oslo universitetssykehus HF
Prosjektleder: Kjetil Retterstøl

Vi viser til søknad om prosjektendring datert 23.05.2019 for ovennevnte forskningsprosjekt. Søknaden er behandlet av sekretariatet for REK sør-øst D på fullmakt, med hjemmel i helseforskningsloven § 11.

Endringen består i at tre nye medarbeidere skal delta i prosjektet:

- Helge Rootwelt, overlege
- Katja Benedicte Prestø Elgstøen, forsker
- Katrine Pettersen, masterstudent

Vurdering

Sekretariatet i REK har vurdert den omsøkte endringen, og har ingen forskningsetiske innvendinger til endringen slik den er beskrevet i skjema for prosjektendring.

Det opplyses at det er Katrine Pettersen som skal skrive masteroppgaven som er nevnt i endringsmeldingen av 23.04.2019. Protokollen for oppgaven er vedlagt søknaden og tas til orientering.

Videre er det lagt ved reviderte utgaver av informasjonsskriv med to samtykkeskjema og følgebrev. REK forutsatte i vedtak den 13.05.2019 at dato for sletting i disse ble korrigeret til 31.12.2028.

Dato er endret i korrekt i informasjonsskrivet og REK tar dette til orientering.

I bekreftelse av nektet samtykke og følgebrevet heter det fremdeles at dato for sletting er den 31.12.2023. Det er prosjektperioden som er utvidet til 2023. Sletting av studiematerialet skal ikke skje før i 2028.

REK viser til vedtak den 13.05.2019, og ber igjen om at dato for sletting endres i alt materiell som sendes ut til deltagerne. Korrekte versjoner må sendes REK til orientering.

REK ber om tilbakemelding dersom noe skulle være uklart.

Vedtak

Sekretariatet har gjort en forskningsetisk vurdering av endringen i prosjektet og godkjenner prosjektet slik det nå foreligger, jfr. helseforskningsloven § 11, annet ledd.

Tillatelsen er gitt under forutsetning av at prosjektet gjennomføres slik det er beskrevet i søknaden, endringssøknad, oppdatert protokoll og de bestemmelser som følger av helseforskningsloven med forskrifter.

Vi gjør samtidig oppmerksom på at etter ny personopplysningslov må det også foreligge et behandlingsgrunnlag etter personvernforordningen. Det må forankres i egen institusjon.

Klageadgang

Vedtaket kan påklages, jf. forvaltningslovens § 28 flg. Eventuell klage sendes til REK sør-øst. Klagefristen er tre uker fra du mottar dette brevet. Dersom vedtaket opprettholdes av REK sør-øst, sendes klagen videre til Den nasjonale forskningsetiske komité for medisin og helsefag for endelig vurdering.

Vennligst oppgi vårt referansenummer i korrespondansen.

Til informasjon bytter REK søknadsportal i sommer. Den nye portalen vil være klar i august. Se våre hjemmesider «Aktuelle meldinger» for oppdatert informasjon.

Henvendelser kan sendes inn på e-post post@helseforskning.etikkom.no.

Med vennlig hilsen

Knut Ruyter
avdelingsdirektør
REK sør-øst sekretariatet

Finn Skre Fjordholm
rådgiver

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