Large-Scale Analysis of Circulating Amino Acids and Adipose Gene Expression in Relation to Abdominal Obesity

Running Head: Amino acids and abdominal obesity

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ABSTRACT

Background: Circulating level of the amino acid glutamate is associated with central fat

accumulation, yet the pathophysiology of this relationship remains unknown. We aimed to: i)

refine and validate the association between circulating glutamate and abdominal obesity in a

large population-based twin cohort; and ii) investigate whether transcriptomic profiles in adipose

tissue could provide insight into biological mechanisms underlying the association. **Methods:**

First, in a cohort of 4,665 individuals from the TwinsUK resource, we identified individuals with

abdominal obesity and compared prevalence of the latter across circulating glutamate quintiles.

Second, we used transcriptomic signatures generated from adipose tissue, both subcutaneous and

visceral, to investigate associations with circulating glutamate levels. **Results:** Individuals in the

top circulating glutamate quintile had a 7-fold higher prevalence of abdominal obesity compared

to those in the bottom quintile. The adipose tissue transcriptomic analyses identified *GLUL*,

encoding Glutamate-Ammonia Ligase, as being associated with circulating glutamate and

abdominal obesity, with pronounced signatures in the visceral depot. **Conclusion:** Circulating

glutamate is positively associated with the prevalence of abdominal obesity which relates to

dysregulated GLUL expression specifically in visceral adipose tissue.

INTRODUCTION

In the past decade, studies based on metabolomics have allowed investigators to establish that obesity is associated not only with alterations in circulating carbohydrates and lipids, but also in circulating amino acids (1). The most studied are branched-chain amino acids (BCAAs), namely leucine, isoleucine and valine. Indeed, they have been repeatedly demonstrated to be positively associated with obesity, insulin resistance and the long-term risk of type 2 diabetes (T2D) as well as cardiovascular diseases (1).

In studies investigating central fat accumulation specifically, rather then overall adiposity, another amino acid has stood out: glutamate. We have shown that, out of nearly 200 metabolites, glutamate was the most strongly associated with visceral adipose tissue (VAT) accumulation in a sample of 59 middle-aged women (2). Although glutamate is much less studied than the BCAAs, studies that did measure it generally found that it is more strongly correlated with central fat accumulation than the BCAAs themselves, as shown in the Framingham Heart Study Generation 3 cohort for waist circumference (WC) (3) and in a sample of 1,499 Japanese participants for VAT area measured by computed tomography (4). In the TwinsUK cohort, Menni et al. investigated whether the relationship between circulating metabolites and T2D or blood pressure was mediated by trunk fat mass (5). They found that the strongest independent association with trunk fat was seen for glutamate. Of note, the three BCAAs were not identified as being independently associated with trunk fat in this analysis. However, this was not the central analysis of the study and as such, those results were not further explored.

Considering the strength of the association between circulating glutamate and abdominal obesity and its consistency across cohorts as well as measurement tools of body fat distribution, we aimed to explore this relationship further. In this study, we first confirm the association between

circulating glutamate and abdominal obesity. Second, we identify the adipose tissue transcripts

associated with circulating glutamate to uncover pathways that relate to this association.

METHODS

Study population and samples

We used participant of the TwinsUK resource (6) where metabolomic and metabolic phenotypes

were available for analysis. The vast majority (94%) of participants were females, and average

body mass index (BMI), WC, total cholesterol, triglycerides and homeostatic model assessment

of insulin resistance (HOMA-IR) were within normal ranges (see **Table 1**). To obtain

measurements of body fat distribution, we used the dual-energy absorptiometry (DXA)-derived

measurement of trunk fat and calculated the percentage of fat in the trunk region. Because no

official threshold for abdominal obesity exists for this measurement, we determined one using a

WC threshold of 95 cm (7). From the linear regression between trunk fat (%) and WC, we

identified 36.7% trunk fat as a threshold for abdominal obesity. According to this definition,

46% of the cohort (n=2139) presented abdominal obesity. Participants from the TwinsUK cohort

all provided informed consent. Secondary analysis of the TwinsUK cohort was approved by the

ethics committee of Laval University.

We also identified a subgroup of monozygotic (MZ) twins discordant for abdominal obesity as

follows. First, we calculated the intra-pair difference in trunk fat (%) between individuals of the

MZ twin pairs. Next, we ranked the pairs based on the intra-pair difference and selected those

among the highest (5th) percentile which were consequently considered as discordant for

abdominal adiposity, representing 54 pairs. Age was calculated at the time blood was drawn for

metabolomic measurements, and because in some cases the twins were not tested on the same

day, there is a small, non-significant difference in mean age of the 2 subgroups.

A subset (n=776) of the TwinsUK cohort was also part of the MuTHER project, where

subcutaneous adipose tissue samples were collected and the transcriptome was profiled as

described previously (8, 9). In all, transcriptomics, metabolomics and DXA measurements were

available for 689 of the 776 women. In this subgroup, 57% of women (n=395) had abdominal

obesity. Their complete characteristics are presented in **Table 1**.

Finally, to investigate the adipose tissue transcriptome and its variance by depot, we used a

sample of bariatric surgery patients from the Quebec Heart and Lung Institute (IUCPQ, Laval

University, Quebec City, Canada). Informed consent was obtained from each participant through

the management framework of the IUCPQ Obesity Biobank. The protocol was approved by the

Research Ethics Committee of IUCPQ (Protocol # 21320). Subcutaneous adipose tissue (SAT)

and omental VAT samples from 43 individuals were included in this study with BMI > 35.5

kg/m^2 and age from 29 to 65 years. The majority of the participants were females (n=22)

(**Supplemental Table 4**).

Amino acid measurement

In the TwinsUK cohort (including the MuTHER sample), amino acid measurements were

extracted from untargeted metabolomic performed by Metabolon Inc. as described

elsewhere (10). Briefly, plasma and serum samples were analyzed by ultra-high-performance

liquid chromatography-tandem mass spectrometry and metabolites were identified by automated

comparison. Metabolite concentrations were divided by the median of the day and inverse-

normalized.

For the sample of bariatric participants, all amino acids were quantified using a Water Acquity

UPLC system coupled to a Synapt G2-Si mass spectrometer (Waters) in tandem mode (LC-

MS/MS) using the EZ:faast amino acid sample testing kit (Phenomenex, 2003). Briefly, plasma

samples were mixed with an internal standard solution and amino acids were extracted using a

solid-phase support. Extracted amino acids were derivatized to increase stability before analysis

and purified using a two-phase liquid-liquid extraction. Samples were analyzed using the HPLC

column, method gradient, and multiple reaction monitoring (MRM) transitions provided by the

kit. Quantification was done using sample to internal standard ratios and a 5-point calibration

curve ranging from 20 nmol/mL to 200 nmol/mL. Our focus was on glutamate, but we also

performed all analyses on BCAA levels for comparison purposes.

RNA Sequencing (RNASeq) in bariatric samples

In the bariatric surgery sample, adipose tissue was digested within 30 minutes of collection using

collagenase adapting the Robdell method (11). The stroma-vascular fraction (SVF) cells,

adipocytes and whole tissue were cryopreserved and stored in liquid nitrogen. From the frozen

samples, 0.5 to 3 million cells were re-suspended in 500 uL TRIzol Reagent and total RNA was

extracted using the miRNeasy Mini Kit (Qiagen) according to the manufacturer's protocol. RNA

library preparations were carried out on 500 ng of RNA with RNA integrity number (RIN)>7

using the Illumina TruSeq Stranded Total RNA Sample preparation kit, according to the

manufacturer's protocol. Final libraries were analyzed on a Bioanalyzer and sequenced on the

Illumina HiSeq 2500. The sequenced reads were pair-ended and 100 – 125 bp in length.

Statistical analyses

We evaluated the association between circulating amino acid levels and trunk fat using mixed

linear regressions. The models were adjusted for age, sex, metabolomic batch and twin-pair

clustering. We calculated the odds ratio (OR) of presenting abdominal obesity and metabolic

syndrome (MetS) features according to circulating amino acid quintiles using mixed logistic

regressions adjusted for age, sex, metabolomic batch and twin-pair clustering. We evaluated the

potential influence of genetics on the association between amino acids and trunk fat by

conducting a co-twin case-control analysis in MZ twin pairs discordant for abdominal obesity.

We compared circulating amino acid levels between leaner and heavier (in terms of trunk fat)

twins of each pair using paired and two-tailed Student' t-test.

Transcriptomic analysis contrasting with metabolite measurements in MuTHER

We used the transcriptome profile of healthy female twins from the MuTHER dataset (8) to

study changes in gene expression pattern in accordance to circulating glutamate levels.

Subcutaneous adipose tissue was profiled using Illumina Human HT12 array as previously

described (8, 9). We identified the individuals with circulating glutamate in quintiles 1 and 5

who had gene expression data (n=209). In order to find the differentially expressed genes, we

used linear mixed model and compared gene expression with the quintile adjusted for age, sex,

metabolomic batch and twin-pair clustering. Further, we expanded our analysis with the data

obtained from bariatric participants. Circulating glutamate levels were compared with RNASeq

data normalized using DeSeq2. We used linear model to identify varying gene expression

patterns against metabolite measurements adjusted for the batch effect.

Differential gene expression analysis of RNASeq from bariatric patients

The sequencing reads were trimmed for quality (phred33 $\geq$30) and length (n$\geq$32), and Illumina

adapters were clipped off using Trimmomatic v.0.35 (12). Filtered reads were then aligned to the

GRCh37 human reference genome using STAR v.2.5.3a (13). Raw read counts of genes were

obtained using htseq-count v.0.6.0 (14). Differential gene expression analysis was done using

DeSeq2   v.1.18.1   (15).   Ingenuity   Pathway   Analysis   (IPA: QIAGEN   Inc.,

https://www.qiagenbioinformatics.com/products/ingenuitypathway-analysis)   was   used   for

identifying enriched pathways. To identify VAT-specific SVF genes, we did differential gene

expression analysis between visceral-specific SVF and matched adipocyte RNA-Seq data,

identifying 9,023 SVF-specific genes (log2FC of >=2, Adj. p <= 0.05). The genes were further

filtered by restricting to those upregulated in VAT SVF in comparisons with SAT SVF using a

nominal p value $\leq$ 0.05 and it resulted in 1,820 genes (**Supplemental Table 6**) including 1,117

(61%) protein-coding genes.

RESULTS

Association between circulating amino acid concentrations and abdominal obesity

We first investigated the biomarker potential of glutamate for abdominal obesity in the TwinsUK

cohort (6). For this analysis, we considered individuals for whom metabolomic and body

composition data were available and taken within 6 months of each other, which corresponded to

4,665 individuals. We applied a linear mixed model linking circulating glutamate with

measurements of obesity and found a significant and positive association with trunk fat (%)

(n=4665, β=0.23, p-value=2.1e-70) and BMI (n=4596, β=0.27, p-value=3e-78). We also tested

the association between circulating glutamate and additional metabolic and T2D-related traits

and found positive associations with total triglycerides (n=4058, β=0.29, p-value=1.2e-62) and

the HOMA-IR index (n=3881, β=0.31, p-value=5.7e-38) and a negative association with high-

density-lipoprotein cholesterol (HDL-C) (n=2855, β=-0.23, p-value=2.1e-32). Associations with

these variables were all stronger for glutamate than for any of the three BCAAs (**Supplemental**

Table 1).

Next, we computed the odds of presenting with abdominal obesity in the 2nd, 3rd, 4th and 5th

circulating glutamate quintiles compared to the 1st with mixed logistic regression models. Results

are presented in **Figure 1**. The odds of presenting abdominal obesity were significantly higher

for the 3rd, 4th and 5th quintiles. Participants in the 5th glutamate quintile also had a greater

prevalence of most MetS features: high WC (OR: 9.78, 95%CI: 4.65, 20.58), high triglycerides

(OR: 7.28, 95%CI: 4.84, 10.94), low HDL-C (OR: 4.91, 95%CI: 2.94, 8.21) and high fasting

glucose (OR: 4.27, 95%CI: 2.73, 6.68). Individuals in the 5th quintile of any one of the BCAAs

also had higher prevalence of these metabolic dysfunction markers, but the risk was lower than

for glutamate (see **Supplemental Figure 1**). The risk of having high blood pressure was not

significantly higher in the 5th vs the 1st quintile of glutamate, leucine and isoleucine. It was

slightly elevated for the 5th vs the 1st circulating valine quintile (**Supplemental Figure 1**).

Non-genetic effect of circulating amino acids and the link with abdominal obesity

We then sought to investigate the contribution of genetic vs. non-genetic factors to the

relationship between glutamate and abdominal obesity. As such, we selected 108 MZ twins of

the 4665 TwinsUK participants where the co-twins were discordant for trunk fat. The

characteristics of these 54 MZ twin pairs are presented in **Table 1**. Mean intra-pair trunk fat

difference was 14.75 % and on average, the heavier co-twin had greater adiposity parameters,

triglycerides and HOMA-IR values as well as lower HDL-C levels.

As shown in **Figure 2**, heavier twins of discordant pairs had significantly higher levels of

circulating glutamate, indicating that the relationship between this amino acid and trunk fat is at

least partially independent from genetic background. Heavier twins also had higher circulating

levels of BCAAs than their leaner co-twins.

Adipose tissue transcriptome associated with circulating glutamate in MuTHER

Next, we attempted to disentangle tissue-specific molecular mechanisms potentially underlying

the observed relationship between circulating glutamate and abdominal obesity. To this end, we

used a subset of the TwinsUK cohort where SAT samples have been collected for transcriptome

profiling as part of the MuTHER study. We applied a linear mixed model linking circulating

glutamate with SAT gene expression and identified 224 genes for which mRNA abundance was

significantly (adjusted p<0.05) associated with circulating glutamate (**Supplemental Table 2).**

Interestingly, among the genes inversely associated with glutamate was *GLUL* (adjusted

p=0.0003, beta=0.089), encoding for Glutamate-Ammonia Ligase, a protein belonging to the

glutamine synthetase family and mapping to the Glutamine Biosynthesis I pathway (p=0.0204,

supplemental Table 3). We then tested the direct association between *GLUL* expression and

trunk fat mass (%) and found a strong negative relationship (beta=-0.03, adjusted p=4.4e-82)

which was maintained also after adjusting for BMI (beta=-0.03, adjusted p=1.4e-38). We also

used overlapping data from the similar sized METSIM Study including ~800 male samples

where VAT samples have been collected for transcriptome profiling (16). In this sample, *GLUL*

expression was negatively associated with both BMI (beta=-0.46, adjusted p=3.7e-38) and waist-

to-hip ratio (beta=-0.23, adjusted p=1.01e-8).

Adipose tissue transcriptome associated with circulating glutamate in bariatric patients

To further investigate the adipose tissue transcriptome associated with circulating glutamate, we

used VAT and SAT biopsy samples from bariatric surgery patients and performed analysis on

the SVF and adipocyte fractions. The complete list of genes associated with glutamate, leucine,

isoleucine and valine in each cell fraction of each depot is presented in **Supplemental Table 5.**

We did not find a significant association between circulating glutamate and *GLUL* expression, be

it in the SVF or adipocyte and in VAT or SAT. We used IPA to annotate the genes significantly

associated with glutamate to canonical pathways. All significant pathways (p<0.05) are shown in

Figure 3.

Glutamate signaling across adipose depots and cell populations

To further validate the origin of *GLUL* expression across depots and adipose tissue cell fractions,

we accessed untargeted genome-wide differential expression analysis derived from bariatric

patients (**Supplemental Table 4**). We first observed a trend towards higher *GLUL* expression in

SVF vs adipocyte fraction. Restricting to SVF only, we found that *GLUL* was significantly more

expressed in the visceral than the subcutaneous depot (log2Foldchange=0.47, p=0.035). We then

utilized our recent single-cell adipose tissue expression atlas (17) and were able to validate the

more pronounced expression of *GLUL* in VAT vs SAT specifically in progenitor cells (avg.

logFC=0.27, adjusted p=1.74E-07) (**Supplemental Figure 2**).

Finally, in order to explore potentially additional expression signatures related to *GLUL*

signaling, we identified 1,821 genes (**Supplemental Table 6**) that were upregulated in SVF

VAT. Interestingly, pathway analysis on these SVF-VAT genes identified "Glutamate Receptor

Signaling" as one of top-three enriched pathways (p-value=2.34e-05, **Supplemental Table 7**)

including 12 (out of 38) mapped genes *DLG4, GLS, GRIK1, GRIK5, GRIN1, GRIN2A, GRIN2C,*

GRIP1, GRM4, GRM8, HOMER1, SLC38A1 (**Figure 4**). In our single cell data, 18 out of 38

genes mapping to the Glutamate Receptor Signalling pathway were expressed in at least one

adipose progenitor/stem cell from either depot. Twelve of these showed differential expression

(nominal $p<0.05$) across depots with 9 being upregulated in visceral-derived progenitor cells,

including the aforementioned *GLUL* (**Supplemental Table 8**). The top visceral-specific genes

reaching statistical significance after correction for multiple testing are shown in **Supplemental**

Figure 2. This analysis indicates that expression of many genes related to glutamate are more

expressed in VAT compared to SAT and in SVF compared to adipocytes.

DISCUSSION

In this study, we aimed to confirm that circulating glutamate is associated with abdominal

obesity. Moreover, we wanted to investigate the adipose tissue transcriptomic profile associated

with circulating glutamate to explore potential pathways relevant to this association. In 4,665

individuals from the TwinsUK cohort, we report that people with the highest circulating

glutamate level have a significantly higher prevalence of abdominal obesity (OR: 7.00, CI: 5.21-

9.39) compared to those with the lowest glutamate level. Moreover, analyses on identical twins

discordant for abdominal adiposity showed that the positive relationship between plasma

glutamate and abdominal fat is at least partially independent of genetic background. In the

MuTHER sample, we found that circulating glutamate level was inversely associated with SAT

GLUL gene expression. Moreover, we showed that *GLUL* expression is greater in visceral than

subcutaneous adipose tissue.

Our observation that glutamate level is significantly associated with abdominal obesity

prevalence is concordant with our previous reports that circulating glutamate can correctly

identify 78% of women presenting visceral obesity assessed by CT scan (18) as well as 90% of women and men presenting abdominal obesity assessed with WC (19). It is also in line with various reports showing that circulating glutamate is correlated with measurements of central fat accumulation in different populations (3-5). The present study adds to the increasing number of studies now supporting this notion and provides clues regarding it potential bases.

The impact of genetics on circulating metabolites was studied by Shin et al. in 2014 (20). They reported that glutamate had a relatively low heritability (25%), compared for example to the BCAAs (43%, 49% and 41% for leucine, isoleucine and valine respectively). This is concordant with our results on MZ twins discordant for abdominal obesity.

To our knowledge, this is the first study describing the adipose tissue transcriptome associated with circulating glutamate. The *GLUL* gene codes for the enzyme Glutamate-Ammonia Ligase, which catalyses the addition of an amino group to glutamate to create glutamine. In 2020, Petrus et al. reported that expression of *GLUL* in SAT was lower in obese vs lean women (21). This was accompanied by a lower glutamine concentration in adipose tissue and in plasma. Moreover, *GLUL* expression was normalized (increased) after weight loss by bariatric surgery. In this article, the authors showed that low adipose tissue glutamine concentration was associated with increased expression of genes coding for pro-inflammatory cytokines (*CCL2*, *CD68* et *EMR1*). Their experiments suggest that this is mediated by an increased glycolysis which in turn increases the O-GlcNAcylation of chromatin and modifies expression of pro-inflammatory genes. In this framework, *GLUL* expression, abundance and activity could possibly play a causal role in adipose tissue inflammation, which requires further experimentation. If this is the case, and if plasma concentration of glutamate tracks with *GLUL*, it could potentially be a marker of adipose tissue inflammation.

Some pathways identified in adipose tissue as being associated with circulating glutamate were

related to immunity and inflammation: complement system, dendritic cell maturation, death

receptor signaling and IL-15 production in SAT as well as interferon signaling in SAT SVF and

the inflammasome pathway in VAT SVF. Because excess visceral adiposity correlates with

adipose tissue inflammation in both SAT and VAT (22), an association between glutamate and

inflammation could potentially explain the association that we and others have reported between

circulating glutamate and abdominal/visceral fat accumulation.

We identified the Glutamate Receptor Signaling pathway as being upregulated in the SVF of

VAT. Studies have shown that VAT is more infiltrated by immune cells, such as macrophages,

compared to SAT (22). Moreover, SVF is the adipose tissue cell fraction containing these

immune cells (22). Therefore, finding genes related to glutamate metabolism to be most

expressed in this depot and cell fraction is concordant with the association we saw in other

analyses between circulating glutamate and adipose tissue inflammation genes/pathways. The

relevance of this pathway in adipose tissue remains to be established.

In all, our results suggest that the association between circulating glutamate and central obesity is

related to the expression of *GLUL* and inflammatory genes in adipose tissue. We acknowledge

that *GLUL* expression was not associated with circulating glutamate in our sample of bariatric

patients. This might be due to the smaller sample size or because the association is blunted in

patients with severe obesity. Further studies are needed to confirm the association between

circulating glutamate and *GLUL* expression in adipose tissue and whether obesity level

influences this association. Moreover, the direction and causality of the associations needs to be

elucidated. The strengths of this study include the large number of participants in the TwinsUK-

related analysis as well as the use of different adipose tissue depots and cell types in the RNAseq

analyses. One limitation is that the samples for both include mostly Caucasian women, making

extrapolation to other population subgroups difficult.

In conclusion, we have confirmed in a large cohort that elevated circulating glutamate is

associated with a higher prevalence of central obesity. Concomitant alterations in the expression

of *GLUL* and inflammatory genes are observed in adipose tissue, which may provide insight into

this association.

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FIGURES CAPTIONS

Figure 1. Trunk fat and odds ratios of having abdominal obesity according to circulating amino acid quintiles. Left panels: abdominal fat for each circulating amino acid quintile. The lower, middle and top borders of the boxplot represent the second quantile, median and third quantile, respectively. Right panel: Odds ratios of having abdominal obesity of the second, third, fourth and fifth circulating amino acid quintiles, compared to the first. Error bars represent the 95% confidence interval. Abdominal obesity was defined as having 37.7% of fat or more in the trunk region. Odds ratios were obtained for mixed logistic regression models, adjusted for age, sex, metabolomic batch and twin pair clustering.

Figure 2. Difference in circulating amino acids between heavier and leaner twins from monozygotic pairs discordant for trunk fat percentage (n=108, 54 pairs). Twin pairs were considered discordant if the intra-pair trunk fat difference was greater or equal to the 95th percentile of intra-pair trunk fat percentage difference for monozygotic twins, that is 11.61% of trunk fat difference. In each discordant pair, we identified the leaner and heavier (in terms of trunk fat) individual. Left panels show the circulating amino acid level of all individuals, according to whether they are the leaner of heavier twin of the pair. The right panels show the circulating amino acid difference between the heavier and leaner twin. P-values are from two-tailed and paired Student's t-tests.

Figure 3. Top canonical pathways in adipocytes and SVF from SAT and VAT associated with circulating glutamate. Genes significantly associated with glutamate were identified using a linear model using a p-value threshold of 5x10e-5 and annotated to canonical pathways using Ingenuity Pathway Analysis. SAT: subcutaneous adipose tissue, VAT: visceral adipose tissue, SVF: stroma-vascular fraction.

Figure 4: Expression levels of glutamate receptor genes show increased expression in SVF derived from VAT. Adipocytes from both SAT and VAT had consistently lower expression in comparison with SVF.

TABLE

Table 1. Characteristics of the different samples derived from the TwinsUK cohort.

Variables	Whole sample	MZ twins discordant for trunk fat		MuTHER
		Leaner	Heavier	
Sex (Female : Male)	4400 : 265	53 :1	53 :1	689 : 0
Zygosity (MZ : DZ)	2247 : 2418	54 :0	54 :0	270 : 419
Age (years)	53.02 ± 13.87	52.03 ± 12.89	52.79 ± 13.17	58.77 ± 9.20
BMI (kg/m^2)	25.98 ± 4.74	22.70 ± 3.69	29.01 ± 4.82	26.47 ± 4.74
Trunk fat (%)	35.72 ± 8.76	27.23 ± 5.97	41.98 ± 5.76	38.57 ± 7.57
WC (cm)	80.56 ± 11.71	77.80 ± 8.95	86.17 ± 9.43	77.75 ± 10.99
Cholesterol (mmol/L)	5.54 ± 1.07	5.23 ± 0.95	5.65 ± 1.16	5.56 ± 1.05
HDL-C (mmol/L)	1.83 ± 0.48	1.99 ± 0.49	1.72 ± 0.36	1.82 ± 0.48
LDL-C (mmol/L)	3.23 ± 0.98	2.88 ± 0.78	3.35 ± 1.09	3.24 ± 0.98
Triglycerides (mmol/L)	1.09 ± 0.62	0.80 ± 0.25	1.23 ± 0.69	1.10 ± 0.58
SBP (mmHg)	122.31 ± 17.29	117.40 ± 16.77	125.20 ± 17.11	125.02 ± 19.33
DBP (mmHg)	76.25 ± 11.07	73.75 ± 11.87	79.60 ± 10.62	77.02 ± 9.75
HOMA-IR	1.21 (0.67-1.75)	0.63 (0.43-0.84)	1.44 (0.80-2.08)	1.24 (0.68-1.79)

Values are mean ± standard deviation, except for HOMA-IR index which is presented as median (interquantile range). MZ: monozygotic, DZ: dizygotic, BMI: body mass index, WC: waist circumference, HDL-C: high-density-lipoprotein cholesterol, LDL-C: low-density-lipoprotein cholesterol, SBP: systolic blood pressure, DBP: diastolic blood pressure, HOMA-IR: homeostasic model assessment of insulin resistance

a) Glutamate

b) Leucine

c) Isoleucine

d) Valine

a) SAT – Adipocyte

Nucleotide Excision Repair Pathway
Androgen Signaling
Flavin Biosynthesis IV (Mammalian)
Choline Degradation I
Assembly of RNA Polymerase II Complex
Anandamide Degradation
Arginine Degradation I (Arginase Pathway)
ERK5 Signaling
Arginine Degradation VI (Arginase 2 Pathway)
Urea Cycle
Citrulline Biosynthesis
NER Pathway
Axonal Guidance Signaling
Bile Acid Biosynthesis, Neutral Pathway
LXR/RXR Activation
Superpathway of Citrulline Metabolism
Human Embryonic Stem Cell Pluripotency
Adipogenesis pathway
Methylglyoxal Degradation III

b) SAT – SVF

Interferon Signaling
Flavin Biosynthesis IV (Mammalian)
Fatty Acid Î²–oxidation I
Oxidative Phosphorylation

c) VAT – Adipocyte

EIF2 Signaling
Oxidative Phosphorylation
Mitochondrial Dysfunction

−log(p)

d) VAT – SVF

Endocannabinoid Cancer Inhibition Pathway
NADH Repair
Osteoarthritis Pathway
Salvage Pathways of Pyrimidine Deoxyribonucleotides
Salvage Pathways of Pyrimidine Ribonucleotides
Sphingosine−1−phosphate Signaling
Granzyme B Signaling
Inflammasome pathway

−log(p)