Identification of biomarkers related to metabolically unhealthy obesity in Korean obese adolescents: A cross-sectional study

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Research Article

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Abstract

Background

Obesity is classified as metabolically unhealthy obesity (MUO) and metabolically healthy obesity (MHO). The current study aimed to screen for relationships and different potential metabolic biomarkers involved between MHO and MUO in adolescents.

Methods

The study included 148 obese adolescents aged between 14 and 16. The study participants were divided into MUO and MHO groups based on the age-specific adolescent metabolic syndrome (MetS) criteria of the International Diabetes Federation. The current study was conducted to investigate the clinical and metabolic differences (AbsoluteIDQ™ p180 kit) between adolescents in the MHO group and those in the MUO group. Multivariate analyses were conducted to investigate the metabolites as independent predictors for the odds ratio and the presence of the MetS in adolescents.

Results

There were significant differences in the 3 acylcarnitines, 5 amino acids, glutamine/glutamate ratio, 3 biogenic amines, and 2 glycerophospholipids between the obese adolescents in the MUO group and those in the MHO group. Moreover, several metabolites were associated with the prevalence of MUO in adolescents. Additionally, several metabolites were inversely correlated with MHO in adolescents of the MUO group.

Conclusions

We observed that histidine, lysine, PCaaC34:1, and several clinical factors in adolescents of the MUO group were reverse correlated with the results in adolescents of the MHO group. In addition, the triglyceride-glucose index was related to MUO in adolescents, compared with the homeostasis model assessment of insulin resistance. Thus, the biomarkers found in this study have a potential to reflect the clinical outcomes of MUO in adolescents. These biomarkers will lead to a better understanding of MetS in obese adolescents.

Background

Despite growing health concerns, the prevalence of obesity in adults and children remains high worldwide (1). Obesity is defined as excessive body fat accumulation (2). Obesity is associated with cardiovascular disease (CVD) (3) and metabolic syndrome (MetS; such as diabetes mellitus (DM), hypertension (HTN), and dyslipidemia) In addition, MetS is associated with obesity (4, 5) and insulin resistance (IR) (6). However, obesity and MetS have different clinical features (7, 8). Therefore, it is important to understand and control obesity and MetS.

The prevalence of obese adolescents has an increasing trend (9). According to the 2013-2017 obesity fact sheet of the Korea Health Promotion Institute, the prevalence of Korean obese adolescents increased from 11.6% in 2007 to 17.3% in 2017 (10). Furthermore, obese adolescents are at a higher risk of MetS and CVD than non-obese adolescents (11, 12). Clinical factors (such as body fat composition and laboratory measurements) are very important factors related to obesity and atherosclerosis (13, 14). Several previous studies have suggested that metabolic biomarkers are related to obesity (15), MetS (16), and CVD in adolescents (17). However, periodic studies in adolescents are required because adolescence is a growing period that is sensitive to environmental and temporal changes (18). Thus, more detailed research about sensitive biomarkers and regulatory pathways concerning obesity and MetS in adolescents is required. There are a few studies about biomarkers or regulatory pathways that are atherogenic or related to obesity. However, not all obese populations develop MetS (8). In general, height and weight measurements are considered in obesity diagnosis (2). Therefore, the diagnosis of MetS in the obese population requires consideration of other factors such as body fat composition, laboratory measurements, and metabolites. In this regard, obesity is classified as metabolically unhealthy obesity (MUO) and metabolically healthy obesity (MHO) (19, 20).

Metabolites are crucial biomarkers that can be potentially used to observe dynamic physiological conditions corresponding to the clinical outcome of the patient (21, 22). Therefore, it is material to identify and/or quantify tools in adolescents to investigate the difference between MHO and MUO of biomarkers and regulatory pathways in obese adolescents.

The current study aimed to screen for relationships and different potential metabolic biomarkers involved between MHO and MUO in adolescents. We analyzed the difference between the two groups after adjusting for factors, such as age, sex, and body mass index (BMI) to exclude the effect of severe change by growth phase. Thus, the current study was conducted to investigate the clinical and metabolic differences between the adolescents in the MHO group and the adolescents in the MUO group.

Methods

Study participants

In the present study, 148 adolescents aged 14–16 years were included. The overall objective of this cross-sectional study was to identify early risk factors for obesity and associated metabolic disease in Korean adolescents. The current study excluded participants who did not have the data on confounding variables, such as nutrition intakes, biochemicals, and metabolites. The study participants were divided into MUO and MHO groups based on the age-specific adolescent MetS criteria of the International Diabetes Federation (IDF) (23, 24). The Institutional Review Board (IRB) of Korea National Institute of Health (KNIH; IRB No. 2020-07-05-P-A) approved the study protocol which complied with the Declaration of Helsinki.

Anthropometric parameters

The anthropometric parameters of the study participants were measured in absence of clothing and shoes in the morning after overnight fasting for 12 h. The body weight and body fat percentage (BF%) were measured using a body composition analyzer (BC418; Tanita, Tokyo, Japan). The height was measured to the nearest 0.1 cm with a wall-mounted stadiometer (DS-102; Jenix, Seoul, Korea). The BMI was calculated as weight in kilograms divided by height in meters squared (kg/m^2). The BMI z-score was calculated by BMI-for-age percentiles together with the Lambda-Mu-Sigma method of Cole and Green, which provides a way of obtaining normalized growth percentile standards based on 2017 Korean National Growth Charts for Children and Adolescents. The waist circumference (WC) was measured, with participants standing straight, using a flexible tapeline at the midpoint of the lower rib and the iliac crest to the nearest 0.1 cm. The hip circumference (HC) was measured using a flexible tapeline at the horizontal circumference of the highest point of the buttocks to the nearest 0.1 cm. The waist-hip ratio was calculated based on recorded WC and HC measurements. The systolic blood pressure (SBP) and diastolic blood pressure (DBP) were measured twice on the arm of participants in a seated position following a rest of at least 5 min using an automatic sphygmomanometer (HEM-907, OMRON Healthcare Co., Kyoto, Japan); the two measurements were then averaged.

Biochemical analysis

The blood samples were collected after an overnight fast of at least 12 h. The antecubital vein blood was collected in a vacutainer tube. The blood samples were centrifuged to obtain the plasma and serum samples, which were then stored at -80 ℃. The levels of triglycerides (TGs), total cholesterol (T-cholesterol), high-density lipoprotein cholesterol (HDL-cholesterol), alanine aminotransferase (ALT), aspartate aminotransferase (AST), and glucose were measured using an autoanalyzer (model 7600; Hitachi, Tokyo, Japan). The low-density lipoprotein cholesterol (LDL-cholesterol) levels calculated using the equation of the Friedewald formula: LDL-cholesterol = total cholesterol - [HDL-cholesterol + (TG/5)].

MetS diagnosis

In the present study, MetS was diagnosed using adolescent criteria of IDF (23, 24); the definition ages were 10-16 years. MetS was defined as central obesity (defined as WC but can assumed if BMI > 30 kg/m^2) plus at least 2 out of 4 criteria (No. 2–5). The MetS diagnostic criteria provided by IDF is as follows:

1. WC ≥ 90^{th} percentile or adult cut-off if lower
2. SBP of 130 mmHg or DBP of 85 mmHg or treatment with anti-hypertensive medication
3. TG ≥ 150 mg/dL
4. HDL-cholesterol < 40 mg/dL
5. Fasting plasma glucose ≥ 100 mg/dL or known type 2 (T2) DM

Assessment of nutrition intake

In this study, participants were interviewed to obtain information about their nutrient intake. The dietary intake was assessed using a 3-day food record method. The data of food records were validated through their parents or caregiver. The study participants were asked to record their intake of meals and snacks, including beverages during a nonconsecutive period of 3 days (1 day of the weekend and 2 days of the weekdays). Furthermore, ingredients of meals as well as portion sizes were also recorded. A well-trained dietician reviewed and confirmed the written 3-day food record using food models during face-to-face interviews to increase precision in reporting. The research-based typical food intake data were used to calculate averages of energy, macronutrients, and micronutrients were analyzed through a food consumption survey database of the Korean National Health and Nutrition Examination Survey.

Metabolite measurements

A total of 186 metabolites in the plasma of 148 participants were measured using AbsoluteIDQ™ p180 kit (Biocrates Life Sciences AG, Innsbruck, Austria). The data quality of each metabolite was checked based on the following criteria: (1) half of the analyzed metabolite concentrations in the reference standards > limit of detection and (2) half of the analyzed metabolite concentrations in the experimental samples. We excluded 32 metabolites that failed the quality criteria. Finally, a total of 154 plasma samples (37 acylcarnitines (ACs), 20 amino acids (AAs), 8 biogenic amines (BAs), 81 glycerophospholipids (GPLs), and 8 sphingolipids (SPLs)) were analyzed with the AbsoluteIDQ™ p180 kit using the protocol described in the AbsoluteIDQ™ p180 user manual. The ACs, GPLs, and SPLs were quantified by flow injection analysis by tandem mass spectrometry using an ABI 4000 Q-Trap mass spectrometer (Applied Biosystems/MDS Sciex, Fosster city, CA). The AAs and BAs were quantified by stable isotope dilution in a liquid chromatography-tandem mass spectrometry. The biocrates MetIQ software was used to control the entire assay workflow, from sample registration to automated calculation of metabolite concentrations to the export of data into other data analysis programs. The metabolite concentration measurements in μmol/L (μM) units were automatically carried out with the MetVal™ software package (Biocrates Life Sciences AG, Innsbruck, Austria).

The calculation of homeostasis model assessment of insulin resistance (HOMA-IR) index and triglyceride-glucose (TyG) index

The HOMA-IR was calculated as the fasting glucose (mg/dL) × fasting insulin (μIU/mL)/405 (25). The TyG index was calculated as the Ln [fasting glucose (mg/dL) × fasting TG (mg/dL)/2] (26).

Data processing and statistical analysis

The statistical analyses were conducted using the Statistical Analysis System software version 9.4 (SAS Institute, Cary, NC, USA). The skewed variables were transformed logarithmically. For the comparison of categorical variables, a chi-squared test was conducted. The general linear model with a Bonferroni correction adjusting for confounding factors was used to compare the parameters collected from each group. Multivariate analyses were conducted to investigate the metabolites as independent predictors for the odds ratio (OR) and the presence of the MetS in adolescents. The results are expressed as the means

± standard errors (SE). A two-tailed *P*-value of < 0.05 was considered to be statistically significant. The Pearson's correlation coefficients were calculated to measure the extent of correlation between pairs of variables.

Results

Participants

A total of 148 obese adolescents between the ages of 14–16 years and both sexes (male; *n* = 82 and female; *n* = 66) participated in the current study. Among the total participants, 74 adolescents belonged to the MUO group, while the other 74 adolescents belonged to the MHO group.

Anthropometric parameters, characteristics, and laboratory measurements of MUO and MHO in adolescents

The anthropometric parameters, characteristics, and laboratory measurements of MUO and MHO in adolescents are shown in Table 1. Sex was not significantly different between the MUO and MHO groups. Weight, BMI, WC, HC, body fat in the MUO group was higher than in the MHO group. However, the difference was not significant. The mean of height, SBP, DBP, fat free mass (FFM), BMI percentage (BMI%), and weight per age percentile (weightth) in the MUO group was significantly higher than in the MHO group ($P = 0.0148$, $P < 0.0001$, $P < 0.0001$, $P = 0.0169$, $P = 0.0164$, and $P = 0.0073$, respectively). The mean value of AST, ALT, red blood cell (RBC), hemoglobin (Hb), hematocrit (Hct), platelet (Plt), and insulin was not significantly different between the two groups. The white blood cell count and glucose levels in the MUO group were significantly higher than in the MHO group; however, these levels were in a normal range for adolescents. We adjusted for factors such as age, sex, and BMI variable, except for self-correction.

Table 1
Anthropometric parameters, characteristics, laboratory measurements, IR assessment index, and plasma lipid profiles of MUO and MHO in adolescents

		Total participants (*n*=148)				
		MHO (n=74)		MUO (n=74)		P
Anthropometric and characteristics$^{\phi}$						
Male/Female	*n*, (%)	82(55.41)/66(44.59)				
		36(48.65)/38(51.35)		46(62.16)/28(37.84)		*0.0982*
Age		14.08	±0.097	13.93	±0.089	*0.2625*
Height	cm	163.7	±0.809	166.9	±0.944	0.0148
Weight	kg	91.12	±1.267	95.19	±1.816	*0.0983*
Body Mass Index	kg/m^2	33.92	±0.329	34.06	±0.453	*0.9531*
Waist Circumference	cm	102.3	±0.912	104.60	±1.113	*0.1402*
Hip Circumference	cm	112.6	±0.616	113.2	±0.925	*0.6343*
Systolic Blood Pressure	mmHg	118.2	±1.168	133.7	±1.453	<0.0001
Diastolic Blood Pressure	mmHg	75.34	±0.772	84.19	±1.050	<0.0001
Body fat	%	45.42	±0.626	44.22	±0.741	*0.1739*
Fat Mass	kg	41.56	±0.966	42.43	±1.239	*0.8330*
Free Fat Mass	kg	49.57	±0.760	52.76	±1.014	0.0169
BMI percentage	%	99.59	±0.039	99.23	±0.146	0.0164
Weight per age percentile	n^{th}	99.97	±0.004	99.77	±0.073	0.0073
Laboratory measurements$^{\phi}$						
Asparate aminotransferase	IU/L	27.68	±2.089	29.57	±1.866	*0.4439*
Alanine aminotransferase	IU/L	36.81	±4.257	43.28	±4.336	*0.1794*
White Blood Cell	$\times 10^3/mm^3$	7.013	±0.202	7.485	±0.164	0.0490
Red Blood Cell	$\times 10^3/mm^3$	4.949	±0.036	5.043	±0.034	*0.1841*

Mean ± SE.$^{\phi}$tested in the following logarithmic transformation, *P*-values were derived from a general linear model with a Bonferroni correction adjusting for age, sex and body mass index of the adolescents in the metabolically unhealthy obesity (MUO) group and of the adolescents in the metabolically healthy obesity (MHO) group.

		Total participants (*n*=148)				
		MHO (n=74)		MUO (n=74)		P
Hemoglobin	g/dL	14.19	±0.115	14.38	±0.116	*0.6055*
Hematocrit	g/dL	43.59	±0.309	44.20	±0.314	*0.3160*
Platele	$\times 10^3/mm^3$	311.9	±7.332	326.6	±5.733	*0.0714*
Insulin	*n*g/mL	27.06	±2.697	29.53	±1.944	*0.2221*
Glucose	mg/dL	92.50	±0.744	98.65	±2.439	0.0146
Insulin resistance (IR) assessment index$^{\phi}$						
Homeostasis model assessment of insulin resistance (HOMA-IR)		6.218	±0.630	7.389	±0.616	*0.2705*
Triglyceride-Glucose Index (TyG index)		8.417	±0.052	8.792	±0.059	<0.0001
Plasma lipid profiles$^{\phi}$						
Triglycerides	mg/dL	107.1	±5.358	150.4	±8.202	<0.0001
Total cholesterol	mg/dL	173.5	±3.119	180.5	±3.416	*0.0878*
LDL-cholesterol	mg/dL	106.2	±2.834	105.9	±3.071	*0.9564*
HDL-cholesterol	mg/dL	45.84	±0.657	44.51	±1.167	*0.1281*

Mean ± SE.$^{\$}$tested in the following logarithmic transformation, *P*-values were derived from a general linear model with a Bonferroni correction adjusting for age, sex and body mass index of the adolescents in the metabolically unhealthy obesity (MUO) group and of the adolescents in the metabolically healthy obesity (MHO) group.

Ir Assessment Index Of Muo And Mho In Adolescents

The TyG index of IR assessment index was significantly higher in the MUO group than in the MHO group ($P < 0.0001$, Table 1). However, the HOMA-IR was not significantly different between the two groups. We adjusted for factors such as age, sex, and BMI variable.

The Metabolites Of Muo And Mho In Adolescents

When individual metabolites were considered, 3 ACs, 5 AAs, glutamine/glutamate (Gln/Glu) ratio, 3 BAs, and 2 GPLs were observed, which were significantly associated with a higher incidence of MUO in adolescents (Table 2). The acetylcarnitine (C2), hydroxy propionyl carnitine (C3-OH), and methyl glutaryl carnitine (C5-M-DC) of ACs were significantly different between the adolescents of the MUO group and those of the MHO group ($P = 0.0184$, $P < 0.0001$, and $P = 0.0273$, respectively). AAs such as alanine (Ala),

glutamine (Gln), histidine (His), lysine (Lys), and serine (Ser)were significantly different between the adolescents of the MUO group and those of the MHO group (P = 0.0125, P = 0.0041, P = 0.0322, P = 0.0371, and P = 0.0061, respectively). The Gln/Glu ratio in the MUO group was significantly higher than in the MHO group (P = 0.0179). The Bas such as kynurenine, methionine sulfoxide (Met-SO), and spermidine were significantly different between the adolescents of the MUO group and the adolescents of the MHO group (P = 0.0003, P = 0.0194, and P = 0.0346, respectively). The phosphatidylcholine (PC) diacyl (aa) (C32:2 and C34:1) of GPLs were significantly different between the adolescents of the MUO group and the adolescents of the MHO group (P = 0.0397 and P = 0.0402, respectively). The SPLs were not significantly different between the adolescents of the two groups.

Table 2
The metabolites of MUO and MHO in adolescents

		Total participants (*n*=148)				
		MHO (n=74)		MUO (n=74)		P
Acylcarnitines$^{\phi}$						
Acetylcarnitine	C2	8.486	±0.371	9.796	±0.362	0.0184
Hydroxypropionylcarnitine	C3-OH	0.072	±0.004	0.077	±0.003	<0.0001
Methylglutarylcarnitine	C5-M-DC	0.033	±0.001	0.035	±0.001	0.0273
Amino Acids$^{\phi}$						
Alanine		421.3	±10.45	462.3	±80.65	0.0125
Glutamine		569.4	±13.19	545.1	±13.38	0.0041
Glutamate		108.3	±4.889	120.00	±4.877	*0.1244*
Glutamine/Glutamate ratio		5.926	±0.252	5.098	±0.230	0.0179
Histidine		94.64	±1.645	92.43	±1.539	0.0322
Lysine		229.7	±4.593	218.9	±4.463	0.0371
Serine		134.6	±2.921	126.9	±2.714	0.0061
Biogenic Amines$^{\phi}$						
Kynurenine		2.168	±0.073	2.612	±0.086	0.0003
Methioninesulfoxide		0.657	±0.028	0.761	±0.026	0.0194
Spermidine		0.202	±0.015	0.236	±0.012	0.0346
Glycerophospholipids$^{\phi}$						
PC aa C32:2		2.901	±0.109	3.180	±0.117	0.0397
PC aa C34:1		163.1	±4.379	179.3	±4.703	0.0402

Mean ± SE.$^{\phi}$tested in the following logarithmic transformation, *P*-values were derived from a general linear model with a Bonferroni correction adjusting for age, sex and body mass index (BMI) of the adolescents in the metabolically unhealthy obesity (MUO) group and of the adolescents in the metabolically healthy obesity (MHO) group. PC, phosphatidylcholine; aa, diacyl.

The correlation between metabolites associated with clinical parameters and IR assessment index of MUO and MHO in adolescents

The current study expressed a correlation heatmap containing the metabolites, clinical parameters and IR assessment index of MUO and MHO in adolescents (Figure 1 (A)–(D)). For example, C3-OH correlated positively with the weightth, BMI%, BMI, HC, BF%, fat mass, TGs, T-cholesterol, and TyG index and correlated negatively with height, FFM, RBC, Hb, and Hct (Figure 1 (A)) in the adolescents of the MHO group. In addition, Figure 1 (A) shows that C3-OH significantly positively correlated with BMI%, BMI, HC, BF%, and body fat mass (r = 0.27735, *P* = 0.0167, r = 0.27335, *P* = 0.0184, r = 0.28242, *P* = 0.0148, r = 0.30879, *P* = 0.0074, and r = 0.23006, *P* = 0.0486, respectively) and correlated negatively with RBC and Hb in the adolescents of the MUO group (r = -0.26086, *P* = 0.0248, and r = 0.28790, *P* = 0.0129, respectively). Moreover, Figure 1 (A) shows that C2 significantly positively correlated with DBP (r = 0.23090, *P* = 0.0478, respectively), Lys correlated negatively with HOMA-IR (Figure 1 (B); r = -0.26173, *P* = 0.0243), and PCaaC34:1 correlated negatively with Plt (Figure 1 (D); r = -0.23320, *P* = 0.0455) in the adolescents of the MUO group. His correlated positively with AST and ALT in the adolescents of the MHO group (Figure 1 (B); r = 0.31157, *P* = 0.0069 and r = -0.27001, *P* = 0.0200, respectively). These were inverse correlations in each group.

Predictors of the MUO adolescents' prevalence ORs on significantly different metabolites

Predictors of the MUO adolescents' prevalence ORs on significantly different metabolites are shown in Table 3. The ORs of the adolescents in the MUO group, obtained using a logistic regression analysis showed that the prevalence ORs of the C2, Ala, Gln/Glu ratio, kynurenine, Met-SO, C34:1, HOMA-IR, and TyG index were significantly greater than ORs of the MUO occurrence, compared with each quartile of the metabolites (*P* = 0.0014, *P* = 0.0014, *P* = 0.0388, *P* = 0.00008, *P* = 0.0001, *P* = 0.0180, *P* = 0.0388, and *P* < 0.0001, respectively). On the other hand, every other metabolite, except for these metabolites, showed no significant difference between the adolescents in the MUO group and the adolescents in the MHO group.

Table 3
Predictors of the MUO adolescents' prevalence ORs on significantly different metabolites

Variables		Total subjects (*n*=148)	*P*
		ORs (95% CI)	
		For MUO adolescents	
Quartile of acylcarnitines			
Acetylcarnitine	C2	1.606 (1.191–2.165)	0.0014
Hydroxypropionylcarnitine	C3-OH	1.114 (0.835–1.488)	*0.4620*
Methylglutarylcarnitine	C5-M-DC	1.114 (0.835–1.488)	*0.4620*
Quartile of amino acids			
Alanine		1.621 (1.194–2.200)	0.0014
Glutamine		0.833 (0.624–1.112)	*0.2131*
Glutamate		1.364 (1.024–1.818)	0.0316
Glutamine/Glutamate ratio		0.735 (0.547–0.988)	0.0388
Histidine		0.948 (0.712–1.263)	*0.7146*
Lysine		0.839 (0.626–1.123)	*0.2363*
Serine		0.796 (0.599–1.058)	*0.1142*
Quartile of biogenic amines			
Kynurenine		1.661 (1.222–2.259)	0.0008
Methioninesulfoxide		1.983 (1.371–2.867)	0.0001
Spermidine		1.304 (0.971–1.751)	*0.0752*
Quartile of glycerophospholipids			
PC aa C32:2		1.258 (0.939–1.685)	*0.1214*
PC aa C34:1		1.424 (1.057–1.917)	0.0180
Quartile of insulin resistance assessment index			
Homeostasis model assessment of insulin resistance (HOMA-IR)		1.360 (1.012–1.827)	0.0388
TyG index		2.046 (1.476–2.835)	<0.0001

P-values derived from a logistic regression analysis on the metabolically unhealthy obesity (MUO).

The Nutrition Intakes Of Muo And Mho In Adolescents

The nutrition intakes of MUO and MHO in adolescents are shown in Table 4. All factors of nutrition intake were not significantly different between adolescents in the MUO group and those in the MHO group.

Table 4
The nutrition intakes of adolescents in MUO and MHO groups

		Total participants (*n*=148)				
		MHO (n=74)		MUO (n=74)		P
Nutrition intakes[ϕ]						
Energy	Kcal	1424.7	±45.72	1523.0	±40.64	*0.0600*
Carbohydrate	g	211.3	±6.588	229.1	±6.650	*0.0827*
Protein	g	56.44	±2.117	60.16	±1.751	*0.0698*
Fat	g	36.93	±1.801	38.24	±1.309	*0.2151*
Total fiber	g	9.163	±0.459	9.922	±0.459	*0.2348*
Soluble fiber	g	1.798	±0.129	2.020	±0.119	*0.1700*
Non-soluble fiber	g	6.703	±0.352	7.406	±0.353	*0.1336*
Cholesterol	mg	174.5	±10.41	186.0	±11.75	*0.3985*
Calcium	mg	366.7	±19.89	404.1	±17.10	*0.0677*
Potassium	mg	1615.1	±57.48	1776.7	±61.32	*0.0579*
Sodium	mg	2529.6	±100.3	2692.9	±111.4	*0.3018*
Total amino acids	mg	31383.4	±1340.8	32567.8	±1561.1	*0.5081*
Essencial amino acids	mg	14294.0	±627.4	14950.5	±744.5	*0.4711*
Non-essencial amino acids	mg	17089.4	±823.5	17617.3	±822.1	*0.5450*
Isoleucine	mg	1333.5	±59.16	1378.2	±67.69	*0.4235*
Leucine	mg	2533.9	±110.7	2661.2	±133.3	*0.5552*
Valine	mg	1554.5	±71.39	1621.5	±77.19	*0.4277*
Glutamic acid	mg	6212.1	±279.1	6319.9	±289.8	*0.5778*
Total fatty acids	g	28.75	±1.504	28.75	±1.227	*0.6784*
Total trans-fatty acids	g	0.368	±0.026	0.360	±0.018	*0.4523*
Total essential fatty acids	g	7.836	±0.350	8.425	±0.436	*0.4208*
Total saturated fatty acids	g	9.922	±0.669	9.702	±0.438	*0.5272*

Mean ± SE.[ϕ]tested in the following logarithmic transformation, *P*-values were derived from a general linear model with a Bonferroni correction adjusting for age, sex and body mass index of the adolescents in the metabolically unhealthy obesity (MUO) group and of the adolescents in the metabolically healthy obesity (MHO) group.

		Total participants (*n*=148)				
		MHO (n=74)		MUO (n=74)		P
Total mono-unsaturated fatty aicds	g	10.201	±0.628	9.902	±0.455	*0.8554*
Total poly-unsaturated fatty acids	g	8.260	±0.363	8.788	±0.451	*0.5087*
n-3 fatty acids	g	0.861	±0.044	0.952	±0.057	*0.2989*
n-6 fatty acids	g	7.161	±0.314	7.666	±0.388	*0.4401*

Mean ± SE.$^{\$}$tested in the following logarithmic transformation, *P*-values were derived from a general linear model with a Bonferroni correction adjusting for age, sex and body mass index of the adolescents in the metabolically unhealthy obesity (MUO) group and of the adolescents in the metabolically healthy obesity (MHO) group.

Discussion

This study investigated the difference or/and relationship of clinical features and metabolites between MHO and MUO in adolescents. To investigate the clinical features of MetS in obese adolescents, this study divided the participants into MHO and MUO groups, according to the IDF criteria. The results revealed that there were significant differences between the adolescents in the MHO group and those in the MUO group in terms of laboratory measurements, IR assessment index (*i.e.*, HOMA-IR and TyG index), and lipid profiles. Furthermore, there were significant differences in the 3 ACs, 5 AAs, Gln/Glu ratio, 3 BAs, and 2 GPLs between the adolescents in the MHO group and those in the MUO group. The results showed that several metabolites were associated with the prevalence of MUO in adolescents. Additionally, several metabolites were inversely correlated with MHO in adolescents of the MUO group. Thus, MUO adolescents have metabolic characteristics despite the same obesity.

Despite the same obese adolescents (even nutrition intake was not significantly different), the obesity or MetS diagnosis factors (*e.g.*, height, weight, FFM, BMI%, weightth, SBP, and DBP) were significantly different between the adolescents of the MHO group and the adolescents of the MUO group. These factors are known to be good biomarkers of obesity and MetS (27, 28). In the current study, AST and ALT were positively correlated with His in the adolescents of the MHO group, whereas in the adolescents of the MUO group, an inverse correlation was found. Obese individuals or those with T2DM or IR had a high level of AST and ALT (29). Some research has shown that although levels of AST and ALT are increased (30), His concentrations are lower (31) in patients with CVD. However, research reported that there was an increase in the liver AST following an intake of His supplementation by obese women with MetS (32). Further research is warranted to understand the correlation between AST, ALT, and His in obese or MetS adolescents.

Unlike the HOMA-IR, the TyG index was significantly higher in the adolescents of the MUO group than in the adolescents of the MHO group. Moreover, the MUO prevalence ORs of the TyG index were 2.046-fold higher for each quartile. Recently, the TyG index is being considered as an IR assessment index along

with HOMA-IR (33). In addition, studies have reported that the TyG index is more appropriate for the diagnosis of T2DM than weight gain (34). Thus, the TyG index could be a good biomarker of MetS in obese adolescents.

The C2, C3-OH, and C5-M-DC of ACs were significantly different between the adolescents of the MHO group and those of the MUO group. Unlike in adolescents of the MHO group, the C2 was positively correlated with DBP in adolescents of the MUO group. Additionally, the MUO prevalence OR of C2 was 1.606-fold higher for each quartile. ACs have a trend to increase the level of individuals who have a risk of obesity and MetS (35, 36). ACs are divided by the length of carbon chains in the molecular structure, such as free carnitine; C0, short-chain; C2–C5, medium-chain; C6–C12, and long-chain C14–C18 (37). C2 of short-chain ACs has a positive correlation with BMI in patients with T2DM (38, 39). Furthermore, C2 has been associated with SBP (40). Some research showed that the level of C3-OH was significantly high in patients with peripheral artery disease, diabetic nephropathy, and DM than that in a control group (41, 42). A study showed that C5-M-DC was associated with IR, in mice with DM (43). In addition, C5-M-DC has a strong relationship between glomerular filtrating rate and creatinine in patients with CVD (44). Furthermore, the current study investigated the predictors of the MUO adolescent's prevalence as ACs by the length. The results depicted that the prevalence ORs of the short-chain ACs were significantly greater than ORs of the MUO occurrence, compared with each quartile of the metabolites (Table S1, Additional file 1). Some research suggested that an increase of short-chain ACs was associated with MetS such as T2DM and obesity (45, 46). Therefore, C2, C3-OH, and C5-M-DC of ACs may be metabolic biomarkers of related MetS in obese adolescents.

The Gln, His, Lys, Ser, and Gln/Glu ratios were significantly lower in the adolescents of the MUO group than in the adolescents of the MHO group. On the contrary, Ala was significantly higher in the adolescents of the MUO group compared with the adolescents of the MHO group. According to a study, levels of Ala were high in obese individuals than normal-weight individuals (47). Furthermore, Ala is a gluconeogenic substrate secreted by skeletal muscle at higher levels in patients with DM (48). Some studies have suggested that Ser has therapeutic potential for DM (49), such as improved regulation of blood glucose (50) and insulin secretion (51). Lys levels were decreased in MetS individuals (in particular, cardio-metabolic features and inflammatory biomarkers), as per a study (52). Another study suggested that Lys has potential protective effects against MetS (53). These results are in line with the results of our study. Therefore, Ala, Gln, His, Lys, and Ser might be good biomarkers of MetS in obese adolescents. According to Cheng *et al.* (54), individuals with MetS have increased levels of the Gln/Glu ratio. Moreover, some researchers have suggested that Gln/Glu ratios (55, 56) are associated with MetS such as DM, CVD, and IR. In addition, in the present study, the level of Glu was higher in the MUO group than in the MHO group; however, it was not significant (Table 2). Thus, Glu, Gln, and Gln/Glu ratios are useful biomarkers of MetS in adolescents. However, Gln was significantly lower in hyperuricemia patients with MetS (57). Therefore, detailed studies and research regarding MetS and other diseases in individuals, are required in the future.

The kynurenine, Met-SO, and spermidine were significantly higher in the adolescents of the MUO group than in those of the MHO group. Some researchers reported that high levels of kynurenine are associated

with obesity (58), idiopathic pulmonary arterial hypertension (59), and IR (60). Met-SO was significantly lower in patients with HTN who were on a low-sodium diet (61) and significantly higher in patients with DM (62). Thus, the results of this study suggest that kynurenine and Met-SO are associated with MetS in obese adolescents. According to Choksomngam *et al*. (63), in animal models, spermidine supplementation has shown to protect against diet-induced obesity. Furthermore, in humans (64) and mice (65), spermidine intake correlates with reduced blood pressure and decreased risk of CVD. In the current study, the levels of spermidine were higher in the adolescents of the MUO group than in adolescents of the MHO group. The results of this study are opposite to the results of some studies. Therefore, further studies are needed to understand the effects of spermidine in MUO and MHO in adolescents.

The results of the present study show that 2 GPLs were significantly higher in the adolescents of the MUO group than in the adolescents of the MHO group. According to Jové *et al*. (66), levels of GPLs were higher in the overweight and obese group with MetS than in the overweight and obese group without MetS. Furthermore, GPLs are associated with related MetS factors such as LDL-cholesterol, glucose, and IR in humans (67) and rats (68). However, for each length or double bond, the GPLs may be different in the individuals with MetS. In this study, PCaaC34:1 as a significant biomarker had significantly higher prevalence OR of MUO in adolescents as against PCaaC32:2. There is research that PCaaC34:1 reduction in type 1 DM (69), increase in Alzheimer's disease (70) and sepsis in the event of community-acquired pneumonia (69). However, there is insufficient research regarding PCaaC34:1 in individuals with MetS or obesity. Despite the need for further studies of GPLs in adolescents with MetS, the results of this study suggest that GPLs have a potential as MetS biomarkers in obese adolescents.

In particular, this study focused on two results. First, MetS associated increase in short-chain ACs in obese adolescents. Second, the correlation results of Lys (with HOMA-IR) and PCaaC34:1 (with Plt) in adolescents in the MUO group and His (with AST and ALT) correlation results in adolescents in the MHO group. This is because these results suggest the relevant MetS biomarkers in adolescents. Furthermore, studies of biomarkers as a clinical feature of MetS in obese adolescents are still limited. In addition, the TyG index also indicated a deeper relationship with the adolescents in the MUO group than the adolescents in the MHO group. This study had certain limitations. Due to the cross-sectional design of the study, it was not possible to establish a causal relationship between metabolites and MetS in adolescents. Another limitation of this study is that it was conducted with a small number of adolescents aged 14–16 years. Thus, the application of general metabolite properties for MetS obtained in the present study may be limited in adolescents of other ages, such as preschool children or adolescents aged 17 years and older. Therefore, further studies using large sample sizes are required to investigate the relationship between metabolites and MetS in adolescents.

Nevertheless, based on the results of this study, several metabolites of MetS features in obese adolescents should be considered as biomarkers. Furthermore, these metabolites and the TyG index would prove to be good biomarkers that reflect the association of MetS clinical symptoms and the presence of IR-related diseases.

Conclusion

The discovered biomarkers were metabolites of MetS characteristics in obese adolescents. In particular, the current study observed that His, Lys, PCaaC34:1, and several clinical factors in adolescents of the MUO group were reverse correlated with the results in adolescents of the MHO group. In addition, the TyG index was related to MUO in adolescents compared with HOMA-IR. Thus, the biomarkers found in this study have a potential to reflect the clinical outcomes of MUO in adolescents. These biomarkers will lead to a better understanding of MetS in obese adolescents.

Abbreviations

AA	Amino acid
AC	Acylcarnitine
Ala	Alanine
ALT	Alanine aminotransferase
AST	Aspartate aminotransferase
BA	Biogenic amine
BF%	Body fat percentage
BMI	Body mass index
C2	Acetylcarnitine
C3-OH	Hydroxy propionyl carnitine
C5-M-DC	Methyl glutaryl carnitine
CVD	Cardiovascular disease
DBP	Diastolic blood pressure
DM	Diabetes mellitus
FFM	Free fat mass
Gln	Glutamine
Glu	Glutamate
GPL	Glycerophospholipid
HC	Hip circumference
Hct	Hematocrit
HDL-cholesterol	High-density lipoprotein cholesterol
Hb	Hemoglobin
His	Histidine

HOMA-IR	Homeostasis model assessment of insulin resistance
HTN	Hypertension
IDF	International Diabetes Foundation
IR	Insulin resistance
IRB	Institutional Review Board
KNIH	Korea National Institute of Health
LDL-cholesterol	Low-density lipoprotein cholesterol
Lys	Lysine
MetS	Metabolic syndrome
Met-SO	Methionine sulfoxide
MHO	Metabolically healthy obesity
MUO	Metabolically unhealthy obesity
OR	Odds ratio
PCaa	Phosphatidylcholine diacyl
Plt	Platelet
RBC	Red blood cell
SAS	Statistical Analysis System
SBP	Systolic blood pressure
SE	Standard errors
Ser	Serine
SPL	Sphingolipid
T2DM	Type 2 diabetes mellitus
T-cholesterol	Total cholesterol

TG	Triglyceride
TyG	Triglycerid-glucose
WC	Waist circumference
weightth	Weight per age percentile

Declarations

Ethics approval and consent to participate

The Institutional Review Board (IRB) of Korea National Institute of Health (KNIH; IRB No. 2020-07-05-P-A) approved the study protocol.

Consent for publication

The datasets generated and/or analysed during the current study are not publicly available due [reason for reseaarch ethics] but are available from the corresponding author on reasonable request.

Authors' competing interests

The authors declare that they have no competing interests.

Authors' Contributions

SRJ conceptualized the study, curated and analyzed the data, wrote the original draft, reviewed and edited the manuscript, and administered the project. HBJ and DY reviewed and edited the manuscript and contributed to discussion. HJK contributed in the curation, investigation, acquisition, and analyzation of the data. HJL contributed to the conceptualization and interpretation of the data, project administration and supervision. All authors critically revised the manuscript, read and approved the final manuscript, and agreed to be held fully accountable for the integrity and accuracy of the work.

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Availability of data and materials

All data generated or analysed during this study are included in this article.

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Figures

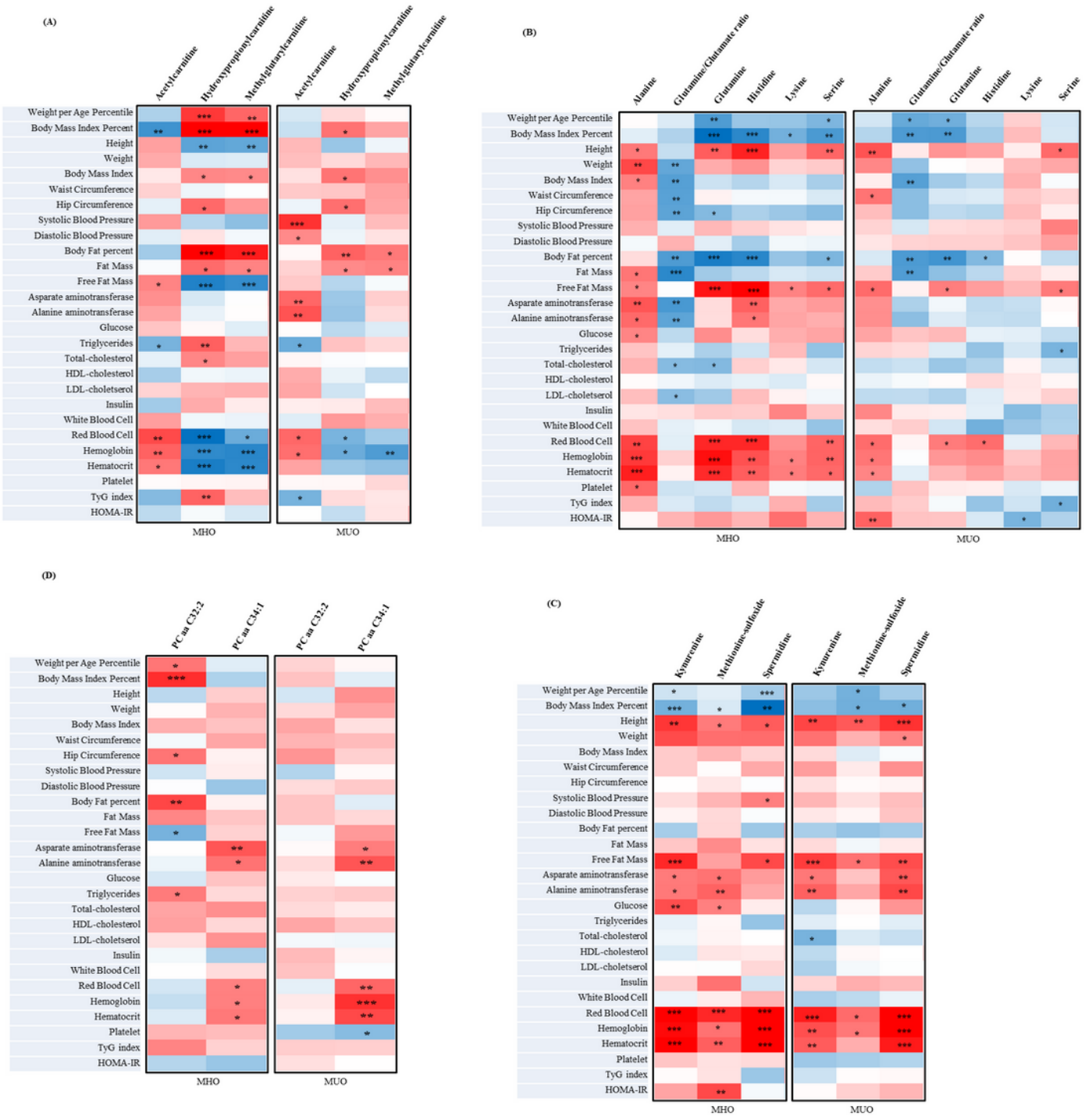

Figure 1

Correlation heatmap between metabolites associated with clinical parameters and IR assessment index in MHO and MUO

Supplementary Files

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- Supplementarydata.docx