Plasma Metabolic Profile Determination in Young ST-segment Elevation Myocardial Infarction Patients with Ischemia and Reperfusion: Ultra-performance Liquid Chromatography and Mass Spectrometry for Pathway Analysis

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Abstract

Background: This study was to establish a disease differentiation model for ST-segment elevation myocardial infarction (STEMI) youth patients experiencing ischemia and reperfusion via ultra-performance liquid chromatography and mass spectrometry (UPLC/MS) platform, which searches for closely related characteristic metabolites and metabolic pathways to evaluate their predictive value in the prognosis after discharge.

Methods: Forty-seven consecutive STEMI patients (23 patients under 45 years of age, referred to here as "youth," and 24 "elderly" patients) and 48 healthy control group members (24 youth, 24 elderly) were registered prospectively. The youth patients were required to provide a second blood draw during a follow-up visit one year after morbidity (n = 22, one lost). Characteristic metabolites and relative metabolic pathways were screened via UPLC/MS platform base on the Kyoto encyclopedia of genes and genomes (KEGG) and Human Metabolome Database. Receiver operating characteristic (ROC) curves were drawn to evaluate the predictive value of characteristic metabolites in the prognosis after discharge.

Results: We successfully established an orthogonal partial least squares discriminated analysis model ($R_2X = 71.2\%$, $R_2Y = 79.6\%$, and $Q_2 = 55.9\%$) and screened out 24 ions; the sphingolipid metabolism pathway showed the most drastic change. The ROC curve analysis showed that ceramide [Cer(d18:0/16:0), Cer(t18:0/12:0)] and sphinganine in the sphingolipid pathway have high sensitivity and specificity on the prognosis related to major adverse cardiovascular events after youth patients were discharged. The area under curve (AUC) was 0.671, 0.750, and 0.711, respectively. A follow-up validation one year after morbidity showed corresponding AUC of 0.778, 0.833, and 0.806. **Conclusions:** By analyzing the plasma metabolism of myocardial infarction patients, we successfully established a model that can distinguish two different factors simultaneously: pathological conditions and age. Sphingolipid metabolism is the top most altered pathway in young STEMI patients and as such may represent a valuable prognostic factor and potential therapeutic target.

Key words: Early-onset Myocardial Infarction; Metabolomics; Potential Biomarker; Sphingolipid Metabolism; Ultra-performance Liquid Chromatography-Mass Spectrometry

INTRODUCTION

Compared to older patients, acute myocardial infarction (AMI) is quite rare in individuals younger than 45 years. Epidemiological data also show that younger survivors have different cardiovascular risk factors and clinical manifestations than the elderly.^[1] Many previous studies have focused on the relationship between genetic susceptibility and plaque formation, rupture, and thrombosis;^[2] however, it

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is now commonly accepted that differing genetic factors are not the single factor in morbidity in young AMI patients.^[3] Metabolomics techniques, which are relatively new, can provide more and better information for researching AMI pathogenesis.^[4]

Metabolomics is a biological research method that, in essence, measures changes in the patient when certain stimuli or interference factors are introduced; this method focuses on the biological system as well as the end products affected by disparate genotypes and environments.^[5] Advanced technologies such as ultra-performance liquid chromatography and mass spectrometry (UPLC/MS) help researchers identify a disease or monitor its progression from blood and urine samples, and fosters more comprehensive, real-time understanding of the related phenotypes of disease statuses.^[6] In a study we conducted previously, we utilized UPLC/MS technology to find that compared to healthy volunteers, multiple metabolite levels were already changed significantly before ischemia/reperfusion (I/R) in the plasma of ST-segment elevation myocardial infarction (STEMI) patients; further, these characteristic metabolites showed a high diagnostic ability.^[7]

The aim of the present study was to identify the plasma characteristic metabolites and related metabolic pathways of young STEMI patients experiencing I/R via UPLC/ MS platform and to evaluate their prognostic value after discharge.

METHODS

Chemical reagents and instruments

All solvents were high-performance liquid chromatography grade and used without modification. Formic acid and acetonitrile (ACN) were obtained from Merck (KGaA Merck, Germany). Distilled water was produced using a Milli-Q Reagent Water System (Millipore, Billerica, MA, USA). UPLC was performed on a Thermo Fisher Accela System (Thermo Fisher Scientific, Franklin, MA, USA). MS was performed on a Thermo Fisher (Thermo Fisher Scientific, Franklin, MA, USA) LTQ Orbitrap XL hybrid mass spectrometer. Other equipment included a Multifuge X1R High-speed Centrifuge (Thermo Fisher Scientific, USA).

Subjects

For the purposes of this prospective study, we enrolled 95 patients who were divided into four groups: the youth patient group (18–44 years old, AY group, n = 23) and elderly patient group (65–80 years old, AO group, n = 24), patients were screened continuously from 405 STEMI patients treated from August 2013 to August 2014. Patients from two control groups (24 youth, NY group; 24 elderly, NO group) were selected at the same time and within approximately the same age range after a physical examination. The youth patients were required to provide a second blood draw during a follow-up visit 1 year after morbidity (SY group, n = 22, one lost). Because male patients comprised a dominant

proportion of AMI morbidity in young adults (21/23, 91.3%), we selected a matched proportion of males in the healthy youth control group, healthy elderly control group, and AMI group.

Entry criteria

The definition of STEMI was established according to the third universal definition of myocardial infarction symptoms.^[8] Patients with the following conditions were excluded: (1) secondary to nonatherosclerotic disease, (2) history of AMI, and (3) combination with cardiac shock, infection, liver or renal insufficiency, malignant tumor, or endocrine or metabolic diseases.

Ethics statement

All samples were collected in accordance with the ethical guidelines and written consent protocols mandated by our hospital.

Sample collection and pretreatment

Three milliliters of venous blood were drawn from the peripheral vein immediately after successful primary percutaneous coronary intervention, defined as the recovered blood flow of the culprit vessel to TIMI3, and preserved in an ethylene diamine tetra acetic acid anticoagulant tube. The blood was then centrifuged to separate the plasma and preserved at -80° C until analysis. Just before analysis, the samples were thawed at room temperature, mixed with 100 µl plasma and 400 µl methanol, intensely vibrated for 30 s, and incubated at 4°C for 5 min to precipitate the protein; finally, the mixture was centrifuged at 15,000 ×*g* for 30 min at 4°C, then the supernatant was filtrated through a 0.22 µm membrane and analyzed. Equivalent volumes of some of the fluid from each sample were mixed for quality control (QC).

Sample analysis

Chromatography conditions were as follows. Mobile phase: Phase A: 0.1% formic acid (volume ratio); Phase B: 95% ACN and 0.1% formic acid. Chromatographic separation was performed isocratically within 15 min, and the injection volume was 10 μ l. The flow rate was set at 200 μ l/min. The sample manager and column oven temperature were set at 4°C and 20°C, respectively. The chromatographic elution gradient was initialized at 5% Phase B and held for 3 min. In consecutive 10 min periods, Phase B was gradually escalated to 50%, and then a rapid increase in Phase B to 95% was completed within 3 min. After 4 min of maintaining the high volume of organic phase gradient, Phase B was immediately reduced to 5% and this elution gradient was used to balance the analytical column for the final 4 min.

MS was operating in the positive ion mode with an ion source voltage of 4.5 kV, a capillary voltage of 30 V, cone voltage of 150 V, desolvation temperature of 275°C, sheath gas flow of 30 arb, and assistant gas flow of 5 arb (99.999% nitrogen). Data were collected over 15 min in centroid mode over the mass range 50–1000 m/z. The MS resolution was at 100,000 full-width half maximum, and the calibration standards were provided by Thermo Fisher Scientific (Caffeine,

Ultramark 1621 and MRFA). MS/MS analysis was carried out with collision-induced dissociation collision energy 35 (normalization collision energy) and the collision gas was 99.999% helium.

There were 22 QC samples throughout the test (equal volume mixture of each analyzed sample). Before the samples were tested, ten QC samples were analyzed continuously and the remaining QC samples were inserted into the sequence after every ten samples were analyzed.

The sequence of samples was randomly generated by the Excel function before and after sample analysis (including QC), and cross-contamination was avoided by inserting a blank between adjacent samples. The whole experiment lasted 2100 min.

Data process

The data were treated as previously described.^[9] Briefly, with the help of MZmine 2.0 and SIMCA-P + 12.0.1.0 (Umetrics AB, Umea, Sweden), the orthogonal partial least squares discriminated analysis (OPLS-DA) supervised model was established. The variances were evaluated by SPSS 16.0 software (IBM, Chicago, USA), using two independent-sample nonparametric tests.

RESULTS

Comparison of clinical and biochemical parameters between groups

There were no significant differences in indexes between AO and NO group. Compared to the NY group, the AY group showed higher rates of diabetes, active smoking, and body mass index and lower density lipoprotein cholesterol levels. Compared to the AO group, the AY group showed higher body mass index, active smoking, hemoglobin, urea nitrogen, myocardial enzyme peak value, cholesterol, and triglyceride levels and lower high-density lipoprotein cholesterol (HDL-C) and Gensini scores [Table 1].

Data pretreatment and quality control analysis

Figure 1 shows the total ion chromatograms obtained from the UPLC/MS platform where differences in content between dozens of metabolites were identified. The data were imported into MZmine 2.0 software for pretreatment and standardization where 380 and 412 integral peaks were, respectively, detected in the ion chromatography extracted from QC and test samples.

The stability of the UPLC/MS system was assessed by analyzing the QC samples throughout the entire period. The principal component analysis (PCA) of 22 data sets of QC samples was used to establish a PCA model with three principal components. Figure 2 shows the score plot of the QC sample sequence versus the first principal component [t(1)]. In the test sample sequence, a QC sample was inserted after every ten test samples to evaluate the stability of the system during the entire analytical process; results showed that the detection system was indeed stable throughout the experiment after the first ten QC samples were injected (we set our QC standard according to a previously published article).^[10] All the QC samples we inserted qualified, which allowed us to confirm that the analytical results were valid.

Different metabolic profiles among groups

A PCA model with six principal components ($R_2X = 30.3\%$, $Q_2 = 13.2\%$) was established and the score plot of its first two principal components drawn as shown in Figure 3a. The clustering trends among groups were not obvious in the direction of the principal components, so we established an OPLS-DA model [Figure 3b] with four predictive principal components and eight orthogonal principal components ($R_2X = 71.2\%$, $R_2Y = 79.6\%$, $Q_2 = 55.9\%$). There were significant distinguishing clustering trends between each group in the score plot of its principal component – in short, the OPLS-DA model made sufficient distinction between groups.

Table 1: Clinical and biochemical characteristics in patient and control groups						
Characteristics	AY $(n = 23)$	NY $(n = 24)$	A0 (<i>n</i> = 24)	NO (<i>n</i> = 24)		
BMI (kg/m ²)	$27.03 \pm 3.77^{*,\dagger}$	24.58 ± 3.04	23.14 ± 2.96	23.98 ± 4.02		
S2B (h)	4.8 ± 2.6	-	5.3 ± 2.1	_		
Hypertension, n (%)	11 (47.8)	5 (20.8)	14 (58.3)	10 (41.7)		
Diabetes mellitus, n (%)	9 (39.1)*	0 (0)	5 (20.8)	8 (33.3)		
Active smoke, n (%)	17 (73.9)*,†	8 (33.3)	9 (37.5)	12 (50)		
Gensini score	$42.6\pm23.6^{\dagger}$	-	62.6 ± 27.1	_		
Hemoglobin (g/L)	$155.52 \pm 13.16^{\dagger}$	152.88 ± 9.66	133.65 ± 21.15	138.83 ± 11.40		
Platelet count ($\times 10^9/L$)	238.35 ± 57.09	216.13 ± 53.25	209.70 ± 79.44	197.79 ± 54.40		
Peak CKMB value (U/L)	$178.74 \pm 51.70^{\dagger}$		135.38 ± 55.37			
Urea nitrogen (mmol/L)	$5.49 \pm 1.06^{+}$	4.85 ± 1.14	6.81 ± 2.44	5.73 ± 1.03		
Serum creatinine (µmol/L)	73.10 ± 15.06	72.91 ± 9.40	72.51 ± 23.66	69.35 ± 12.10		
Triglyceride (mmol/L)	$2.63\pm2.88^{\dagger}$	1.60 ± 0.82	1.21 ± 0.74	1.28 ± 0.60		
Cholesterol (mmol/L)	$5.10 \pm 1.49^{\dagger}$	4.37 ± 0.69	4.72 ± 1.38	4.74 ± 0.84		
HDL-C (mmol/L)	$0.97\pm0.31^{\dagger}$	1.12 ± 0.26	1.25 ± 0.36	1.37 ± 0.45		
LDL-C (mmol/L)	$3.06 \pm 1.05*$	2.40 ± 0.53	3.06 ± 1.06	2.69 ± 0.69		
LVEF (%)	48.59 ± 8.94	_	45.04 ± 7.59	_		

Data are shown as mean \pm SD.*AY versus NY, P < 0.05; [†]AY versus AO, P < 0.05. BMI: Body mass index; S2B: Duration of syndrome to balloon; HDL-C: High-density lipoprotein cholesterol; LDL-C: Low-density lipoprotein cholesterol; LVEF: Left ventricular eject fraction; SD: Standard deviation; AY: youth patient group; NY: youth control group: AO: elderly patient group; NO: elderly control group.



Figure 1: Total ion chromatogram of metabolic profiles in different groups (one sample chosen randomly). AY: Youth AMI group; AO: Elderly AMI group; NY: Youth control group; NO: Elderly control group; SY: One year after AY morbidity.



Figure 2: Score plot of principal component [t(1)] for quality control principal component analysis model. Each point represents a quality control sample. The detection system was stable throughout the experiment after the first 10 quality control samples were injected (no outliers exceeding \pm 2 standard deviation were detected in the quality control samples), the icon of t(1) represents the first principal component.

Metabolic profiling distinguishes youth patient group disease states

An OPLS-DA model ($R_2X = 61.6\%$, $R_2Y = 88.7\%$, $Q_2 = 70.7\%$) with two predictive principal components and five orthogonal principal components was established for the 69 specimens from youth patient groups (AY, NY, SY). Figure 3c shows the score plot of the first predictive

principal component and first orthogonal principal component where significant distinguishing clustering trends between the three groups were found in the first predictive principal component direction. The SY group was located between AY and NY groups, suggesting that the OPLS-DA model successfully distinguished the evolution of AMI in the youth patients.

Screening and identification of characteristic metabolites Again, metabolites with distinguishing characteristics were selected in the established OPLS-DA model.^[9] A total of 24 identified metabolites were identified as listed in Table 2.

Plasma metabolites among groups

Two independent-sample nonparametric tests were used to compare differences in metabolic profiles between two patient groups [Table 2]. The levels of phospholipid (glycerophosphatide, sphingolipid) metabolites, as well as some free fatty acids and amino acids, were significantly up- or down-regulated. Compared to the AO group, guanidoacetic acid and 9-decenoylcarnitine, threoninyl-lysine, and 2-hydroxyphytanic acid were the unique differential metabolites of youth STEMI patients after I/R. Compared to the NY group, the levels of threoninyl-lysine, phosphatidic acid, catelaidic acid, and ceramide (Cer) in the SY group also showed significant differences.



Figure 3: Metabolic profiles of plasma samples distinguish different groups. (a) Score plot of the first two components [t(1)/t(2)] of the plasma metabolic profiling principal component analysis model; (b) score plot of metabolic profiling orthogonal partial least squares discriminated analysis model; (c) score plot of metabolic profiling orthogonal partial least squares discriminated analysis model; not metabolic profiling orthogonal partial least squares discriminated analysis model; not metabolic profiling orthogonal partial least squares discriminated analysis model; not metabolic profiling orthogonal partial least squares discriminated analysis model in youth patients. Every point in the figure represents a sample. The icons of t(1) and t(2) represent the first and second principal component, respectively.

Table 2: Metabolite identification results and the difference between patient groups							
m/z	RT (min)	Metabolite	lonospheric models	AO versus NO‡	AY versus NY‡	AY versus AO [§]	SY versus NY‡
510.354	7.95221	LysoPC(17:0)	M+H	Down [†]	Down [†]	Down [†]	-
546.352	8.58423	LysoPC(18:0)	M+Na	Down [†]	Down^\dagger	_	Down [†]
508.375	8.03549	LysoPC(P-18:0)	M+H	Down [†]	Down^\dagger	_	Down [†]
506.359	8.58556	LysoPC(P-18:1(9Z))	M+H	Down^\dagger	Down^\dagger	_	Down [†]
522.354	7.66455	LysoPC(18:1(11Z))	M+H	Down [†]	Down^\dagger	_	_
550.386	8.77023	LysoPC(20:1(11Z))	M+H	Down [†]	Down^\dagger	_	Down [†]
540.305	6.80829	LysoPC(18:3(6Z,9Z,12Z))	M+Na	Down [†]	Down^\dagger	Up*	-
570.352	7.86944	LysoPC(20:2(11Z,14Z))	M+Na	Down [†]	Down^\dagger	-	Down [†]
487.286	8.57992	PA(20:4(5Z,8Z,11Z,14Z) e/2:0)	M+H	Down [†]	Down*	$\mathrm{U}\mathrm{p}^{\dagger}$	Down*
556.529	10.4299	Cer(t18:0/16:0)	M+H	Up^{\dagger}	$\mathrm{U}\mathrm{p}^\dagger$	Up*	Up*
544.338	7.11841	LysoPC(20:4(5Z,8Z,11Z,14Z))	M+H	Down*	-	-	-
302.305	6.72781	Sphinganine	M+H	Up*	Up*	-	-
356.351	7.40061	Catelaidic acid	M+NH4	Up^{\dagger}	$\mathrm{U}\mathrm{p}^\dagger$	_	Up*
312.326	7.28137	Phytal	M+NH4	Up^{\dagger}	Up*	Down*	-
495.248	4.54762	Leukotriene D5	M+H	Up^{\dagger}	-	-	-
484.464	9.44252	Cer(d18:0/12:0)	M+H	Up^{\dagger}	$\mathrm{U}\mathrm{p}^\dagger$	Up^{\dagger}	$\mathrm{U}\mathrm{p}^{\dagger}$
284.294	6.74211	Octadecanamide	M+H	$\mathrm{U}\mathrm{p}^\dagger$	$\mathrm{U}\mathrm{p}^\dagger$	-	-
346.331	6.74818	2-hydroxyphytanic acid	M+NH4	_	$\mathrm{U}\mathrm{p}^\dagger$	-	-
118.065	1.75456	Guanidoacetic acid	M+H	_	Down^\dagger	-	-
270.144	5.74974	Threoninyl-lysine	M+Na	-	Down^\dagger	-	Down [†]
314.228	5.75407	9-decenoylcarnitine	M+H	-	Down^\dagger	—	_

*P<0.05; †P<0.01; ‡Compared with control group, §Compared with AO group. Cer: Ceramide; -: No significant difference.

Metabolic pathways and functional analysis in youth patients

We used MetaboAnalyst 3.0 (The Wishart Research Group, Canada) to analyze the above most relevant metabolic pathways in greater detail. Results are shown in Figure 4 where the four principle pathways that most closely represented differences in metabolites between AY and NY groups were sphingolipid metabolism, glycerolphospholipid metabolism, glycine/serine/threonine metabolism, and arginine/proline metabolism; sphingolipid metabolism was the top most altered pathway. The results are also listed in Table 3, and the related pathway structures are shown in Figure 5.

Sphingolipids in youth patient prognoses

We next investigated the predictive value of sphingolipid metabolism on major adverse cardiovascular events (MACE), a summation of readmission caused by AMI, unstable angina, heart failure, malignant arrhythmia, or sudden cardiac death, in the AY group after discharge; a total of four patients had MACE at the 1-year follow-up. We also performed receiver operating characteristic (ROC) curve analysis on the predictive value of plasma Cer and sphinganine in the AY and SY groups as shown in Figures 6 and 7. All area under curves were larger than 0.7 excepted for Cer (d18:0/16:0) in the AY group, suggesting that the three biomarkers may be effective predictive factors for prognosis in youth patients after discharge [Table 4].

DISCUSSION

Even after receiving timely reperfusion and the best possible drug therapy, there is still a considerable portion





of young patients who fail to recover cardiac function after AMI diagnosis - or worse, whose cardiac health continue to decline after treatment.^[11] Early onset of myocardial infarction is most likely caused by a significant disruption in the body's metabolic system:^[12] thus, metabolomics can provide a direct molecular signature of cells that reflect changes in phenotypes and molecular physiology, and as such are very helpful for exploring pathogenesis and treatment efficacy.^[13] Laborde et al., for example, identified 15 metabolites with statistically significant differences from plasma samples of patients with non ST-segment elevation acute coronary syndrome and healthy controls using a gas chromatography MS platform; they found that the metabolites reflect myocardial cells subjected to oxidative stress and hypoxia.^[14] The metabolic pathway of STEMI patients (especially young patients) has seen relatively little research.

In this study, the main components of the OPLS-DA model we established involved AMI and age. In other words, these two factors created the most obvious distinction among the groups we tested. We comparatively analyzed the metabolic profiles of patients of different ages post-I/R as well as corresponding healthy controls, and identified metabolites that are closely related to the main component of the cell membrane, and therefore to cell proliferation, differentiation, apoptosis, and biological energy supply.^[15] By analyzing potential metabolites, as well as related interactions and pathways in youth STEMI patients based on the KEGG and Human Metabolome Database, we ultimately found that sphingolipid metabolism was the top most altered metabolic factor related to AMI prognosis.

By means of UPLC/MS technology, we found in a previous study that there are significant changes between the phospholipid levels of AMI patients before I/R treatment compared to healthy controls – specifically, glycerophospholipid is significantly down-regulated and most sphingomyelin are up-regulated.^[7] We found similar changes in phospholipid metabolic pathways after emergency reperfusion therapy in the infarction zone of youth STEMI patients, in accordance with our previously published results; moreover, as discussed above, sphingolipid metabolism was the top most altered metabolic factor. Sphingosine, Cer, and sphingosine-1-phosphate (S1P), the primary members of the metabolic pathway involved, are crucial cell membrane components that reciprocally transform to maintain homeostasis by enzymatic reaction.^[16]

Importantly, increases in sphingosine and Cer levels can cause reduction of S1P. Sphingosine also is a signal

Table 3: Pathway analysis results						
Items	Total	Expected	Hits	Raw P	-log(<i>P</i>)	Impact
Sphingolipid metabolism	25	0.04	1	4.09E-02	3.20E+00	0.104
Glycerophospholipid metabolism	39	0.06	1	6.33E-02	2.76E+00	0.00
Glycine, serine, and threonine metabolism	48	0.08	1	7.75E-02	2.56E+00	0.01
Arginine and proline metabolism	77	0.13	1	1.22E-01	2.10E+00	0.02



Figure 5: Systemic analysis of metabolic changes in acute myocardial infarction pathogenesis. Each of the disturbed metabolites detected was searched in the KEGG database, and each KEGG pathway was scored according to the influence of the metabolic pathway (this map was generated from the KEGG reference map). Green boxes represent currently recognized enzymes in human beings.

Table 4: ROC curves of sphingolipid biomarker levels for predicting MACE in young patients after discharge						
Marker	Cut-off value	Sensitivity (%)	Specificity (%)	AUC	Р	95% <i>CI</i> *
AY group						
Cer(d18:0/16:0)	1,385,000	100.00	42.11	0.671	0.292	0.411-0.931
Cer(t18:0/12:0)	465,214	75.00	78.95	0.750	0.123	0.421-1.079
Sphinganine	8,950,000	75.00	73.68	0.711	0.194	0.474-0.947
SY group						
Cer(d18:0/16:0)	1,170,000	100.00	66.67	0.778	0.089	0.589-0.967
Cer(t18:0/12:0)	462,514	100.00	72.22	0.833	0.041	0.654-1.103
Sphinganine	7,430,000	100.00	66.67	0.806	0.061	0.622-0.989

*Null hypothesis: True area: 0.5. CI: Confidence interval; ROC: Receiver operating characteristic; MACE: Major adverse cardiovascular events; AUC: Area under curve; Cer: Ceramide.

transduction factor of tumor necrosis factor-alpha (TNF- α), the level of synthesis and secretion of which is positively correlated with the amount of myocardial cell necrosis during AMI. As the mediator of TNF- α , sphingosine synthesized in cardiomvocytes can cause reduced myocardial contractility and increased myocardial cell apoptosis due to calcium overload.^[17,18] Cer is the hub of the sphingolipid metabolism pathway, which involves cell response to ischemic stress and I/R injury, as well as activation of endogenous and exogenous apoptotic pathways through receptor-independent mechanisms.^[19,20] It can be generated de novo or by hydrolysis of membrane sphingomyelin by sphingomyelinase.^[20] In a study on rat hearts, researchers discovered that Cer levels increased after I/R whether isolation perfusion occurred in vivo or in vitro[21-23] although the degree to which they increase can be partially mitigated by ischemic preconditioning.^[23,24] By contrast, its metabolite S1P has a protective effect on against ischemia and apoptosis

in the pathogenesis of myocardial infarction.^[21,25,26] The level of S1P in the heart after I/R is more than 50% lower than in control myocardium (which can also be partially prevented by preconditioning).^[22] Duan et al. found that transfer of adenovirus-mediated sphingosine kinase1 gene can prevent myocardial injury caused by I/R and attenuate ischemic heart failure,^[27] and Knapp et al. found that the ratio of S1P to Cer is significantly decreased at several time points after STEMI in rats, which is the cause of increased myocardial cell apoptosis in the noninfarct zone after myocardial infarction.^[28] In this study, as discussed above, we found that changes in Cer in young patients were much more dramatic than those in the elderly patient group - this may have been related to the lack of coronary collateral circulation for ischemia preconditioning and to more severe cardiomyocyte apoptosis after I/R injury in the younger patients. The peak of myocardial enzymes in the AY group was also higher than that in the AO group, which further supports this argument.



Figure 6: Receiver operating characteristic curve analysis of sphingolipid biomarkers ceramide and sphinganine to predict major adverse cardiovascular events of youth patients with acute myocardial infarction after discharge; Cer1: Cer (d18:0/16:0), Cer2: Cer (t18:0/12:0).

We also found that the proportion of abnormal lipid spectra in the AY group was higher than the corresponding control or NY group, reflecting lower HDL-C and higher low-density lipoprotein cholesterol (LDL-C) in these patients. The increase in LDL oxidation may have had a positive correlation with plasma Cer levels, representing a risk factor for atherosclerosis.^[29] The reduction of HDL had a significant relationship with lifestyle in the youth patients as well, including smoking habits, obesity, and larger carbohydrate intake.[30] Previous researchers found through in vivo experiments that HDL and its component S1P act quickly to protect the heart from I/R injury, with potential inhibitory mechanisms on inflammatory cell recruitment and myocardial cell apoptosis in the infarction area as related to S1P-mediated and nitric oxide (NO)-dependent pathways.^[31-34] Sphingosine kinase is a major rate-limiting enzyme in the cellular synthesis of S1P and regulates both Cer and S1P by reducing Cer to generate S1P; to this effect, when up-regulated, it can promote cell proliferation and alleviate apoptosis.[25,26] Acute administration of recombinant S1P-containing HDL has been shown to restore normal endothelial function in an NO-dependent manner;^[33,35,36] thus, a rapid therapeutic elevation of HDL plasma levels may benefit myocardial infarction patients.[37]

In this study, we also found that even after receiving timely reperfusion and optimized medication, the SY group still showed higher Cer levels than young healthy controls, indicating that even the most cutting-edge AMI treatment strategies still do not effectively restore balance to the sphingolipid metabolism pathway. This may be related to the recurrence of acute coronary events after discharge. ROC curve analysis also suggested that Cer and sphinganine are both sensitive and specific markers for MACE prediction in young patients. Taken together, these results suggest that sphingolipid metabolism pathway homeostasis is not only



Figure 7: Receiver operating characteristic curve analysis of sphingolipid biomarkers ceramide and sphinganine to validate predictive value in youth patients one year after morbidity.

a prognosis predictor but also a potential therapeutic target for improving the outcome of young AMI patients.

This study has some limitations. Our sample size was relatively small and the follow-up was relatively short. A larger sample size and lengthier follow-up after treatment remain necessary to confirm our findings. Further testing that includes metabolic urine profiling will also make the results more reliable.

In conclusion, we determined via UPLC/MS platform that sphingolipid metabolism is the top most altered pathway in youth STEMI patients experiencing I/R. Some key components in this pathway thus may be predictors for adverse cardiovascular events after discharge, and recovery of its homeostasis may be a promising therapeutic target.

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Conflicts of interest

There are no conflicts of interest.

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