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ARTICLE Metabolomics facilitates the discovery of metabolic profiles and pathways for myopia: A systematic review

Xiao-Wen Hou¹, Ying Wang \mathbb{D}^1 , Chaofu Ke¹ and Chen-Wei Pan $\mathbb{D}^{1 \boxtimes}$

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BACKGROUND: Myopia is one of the major eve disorders and the global burden is increasing rapidly. Our purpose is to systematically summarize potential metabolic biomarkers and pathways in myopia to facilitate the understanding of disease mechanisms as well as the discovery of novel therapeutic measures.

METHODS: Myopia-related metabolomics studies were searched in electronic databases of PubMed and Web of Science until June 2021. Information regarding clinical and demographic characteristics of included studies and metabolomics findings were extracted. Myopia-related metabolic pathways were analysed for differential metabolic profiles, and the guality of included studies was assessed based on the QUADOMICS tool. Pathway analyses of differential metabolites were performed using bioinformatics tools and online software such as the Metaboanalyst 5.0.

RESULTS: The myopia-related metabolomics studies included in this study consisted of seven human and two animal studies. The results of the study quality assessment showed that studies were all phase I studies and all met the evaluation criteria of 70% or more. The myopia-control serum study identified 23 differential metabolites with the Sphingolipid metabolism pathway beings enriched. The high myopia-cataract aqueous humour study identified 40 differential metabolites with the Arginine biosynthesis pathway being enriched. The high myopia-control serum study identified 43 differential metabolites and 4 pathways were significantly associated with metabolites including Citrate cycle; Alanine, aspartate and glutamate metabolism; Glyoxylate and dicarboxylate metabolism; Biosynthesis of unsaturated fatty acids (all P value < 0.05).

CONCLUSIONS: This study summarizes potential metabolic biomarkers and pathways in myopia, providing new clues to elucidate disease mechanisms.

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INTRODUCTION

Myopia is the most common eye abnormality and the global burden of myopia is increasing rapidly in past decades [1]. Holden et al. predicts that the number of people with myopia will reach 4.76 billion worldwide in the year 2050, and individuals affected by high myopia (HM) will increase to nearly 1 billion [2]. As myopia progresses, the anterior and posterior diameters of the eye increase and the wall of the eye thins, resulting in a series of macular changes, including tessellated fundus and further progression of pathological myopia (diffuse chorioretinal atrophy, patchy chorioretinal atrophy, and even macular atrophy) [3]. In addition, individuals with myopia, especially HM, are particularly vulnerable to vision-threatening ocular complications such as open-angle glaucoma and age-related cataract [4]. Thus, clarifying the mechanism of myopia development is highly important for formulating effective myopia prevention strategies. Although the aetiology of myopia has been investigated for more than hundreds of years, the exact mechanisms remain unclear. Currently, it is generally accepted that the development of myopia is influenced by both genetic and environmental factors and joint effect of genetic and environmental exposures also plays a role. Therefore, myopia research needs to find new tools that can elucidate the combined genetic and environmental effects.

Metabolomics, as the end of the "omics" technologies, transmits the final response signals from the interaction of "genetics and environment" and provides direct "functional readouts" of the physiological state of organs, contributing to the understanding of disease complex phenotypes [5, 6]. In ophthalmology, metabolomics has successfully been used for the description of disease progression and mechanism exploration, including age-related macular degeneration (AMD), glaucoma, and diabetic retinopathy (DR), contributing to the understanding of ocular disease mechanisms [7-9]. In this study, we aimed to summarize myopia-related metabolite changes and explore the metabolic pathways involved in the pathogenesis and progression of myopia according to the existing metabolomics studies on myopia. These findings may contribute to the understanding of the mechanisms of myopia and provide new ideas for delaying or treating myopia and its associated complications.

METHODS

Search strategy

Studies using metabolomics techniques to analyse metabolic changes associated with myopia were investigated. We searched the electronic database of PubMed and Web of Science for relevant metabolomics

¹School of Public Health, Medical College of Soochow University, Suzhou, China. [⊠]email: pcwonly@gmail.com

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Fig. 1 Flow diagram of literature search and study selection.

studies on myopia published up to June 2021, with the following search terms: ("metabolomics" or "metabolomics" or "metabolome" or "metabolic profiling") AND ("myopia" or "refractive error"). All the relevant studies, regardless of the use of study species and biological samples, were included in this systematic review. Additional sources were obtained through searching the reference lists of the included studies that met the inclusion criteria. The overall goal was to guarantee the reliability of the search while including a wide range of original studies related to the topic of interest. Additionally, results were limited to papers written in English. Drugs evaluating reports, review articles and abstracts without full texts were excluded. All articles were screened via abstract and full text to determine their suitability for inclusion in the review. Two authors (XWH and YW) assessed the studies and provided independent reports based on their quality and applicability for inclusion. Any differences that emerged were addressed through the discussions with a senior author (CWP).

Quality assessment

Because the quality of the included studies affects the results and conclusions of the review, assessing the quality of the included studies before investigating the relationship between metabolites and myopia using the selected studies is critical to the success of this review. QUADOMICS is an improved quality assessment tool based on QUADAS tool (a quality assessment tool used to systematically evaluate diagnostic accuracy studies), specifically designed for 'omics'-based diagnostic reviews. QUADOMICS divides the specific challenges in systematic reviews. QUADOMICS divides the study into four clinical validation phases according to the study population and contains four study items, covering sample characteristics, pre-analytical differential conditions, clinical and physiological characteristics, and overfitting during the study and analysis. Two authors (XWH and YW) assessed the quality of included studies using QUADOMICS and addressed discrepancies through discussion.

Data extraction and analyses

For each study, information regarding the title, author, publication date, detection and analysis platforms, sample size, methods of statistical analysis, and repeated reports of biomarkers were extracted and summarized after reading the full articles and supplementary materials. Biochemical interpretation of all altered metabolites was performed using MetaboAnalyst 5.0 (http://www.metaboanalyst.ca/), a software that allows to analyse the impact of particular compounds on biochemical pathways. Pathway analysis was derived from integrating differential metabolites based on Kyoto Encyclopedia of Genes and Genomes and Human Metabolome Database (HMDB). *P*-values were calculated on the basis of

path enrichment analysis, and pathway impact value was derived on the basis of pathway topology analysis.

RESULTS Study characteristics

A total of 12482 records were identified by searching electronic databases and references included in the selected studies. After removing duplicate articles (n = 294), the abstracts of the remaining 12188 records were reviewed by two independent authors, excluding 12179 articles that were not relevant to the topic of this study or did not meet the inclusion criteria (review articles, studies that did not utilize metabolomics techniques to detect metabolites). The remaining twelve studies were subjected to full-text review. Finally, nine articles were included in the review. A simplified flow chart of study selection was shown in Fig. 1.

All studies described myopia-related metabolic profiles. Seven studies focused on humans [10-16] and two were animal studies (chicks and guinea pigs) [17, 18]. All studies were performed on appropriately sized groups and the sample size ranged from 38 to 211 subjects. The study with the largest sample size was a schoolbased case-control study of myopia and non-myopia in Chinese children and adolescents [13]. With regard to seven human studies, four analysed metabolic alterations in HM [10, 11, 14, 15], two focused on metabolic differences in any myopia [12, 13] and the other one investigated myopia-related retinal changes [16]. The most common definition of HM is spherical equivalent -6.0 dioptre (D) or worse, or axial length longer than 26 millimetres. Table 1 shows the basic information of all included studies in this review. A total of four types of biological tissues were involved. Of them, serum was analysed in five studies [12-16], aqueous humour (AH) samples were analysed in two studies [10, 11], and retinas were analysed in two studies (some studies analysed two types of biospecimens, vitreous and retinas) [17, 18]. With regard to the analytical platforms, mass-spectrometry based metabolomics studies were reported in all studies, including liquid chromatogram-mass spectrometry (LC-MS)(n = 5) [10, 12, 13, 15, 18], gas chromatography - mass spectrometry (GC-MS)(n = 4) [11, 14, 16, 17], whereas one study employed the capillary electrophoresis (CE) & LC-MS platform.

Table 1. Clinical and den	nographic characteristics of myopia metabolomi	ics studies.					
References	Comparison	Diagnostic criteria	SE	Age	Male %	Source	
Liu(2020) [16]	57 PM			55.32 ± 14.49	28.08	China	
	81 cataract patients			65.83 ± 11.94	35.80		
Ke(2020) [14]	40 HM	HM(SE < - 0.6D), mild myopia(0 < SE <	-7.7D	≥ 60	22.00	Weitang, China	
	40 mild myopes	– 3.0D)	-1.7D	match	22.00		
Du(2020) [13]	108 myopes	myopic(SE < - 0.50D)	−2.20 ± 1.59(L), − 2.51 ± 1.28(R)	10.15 ± 0.99	50.93	7 schools in a district, China	
	103control(without myopia)		3.36 ± 1.15(L), 0.40 ± 0.75 (R)	10.09 ± 1.01	60.19		
Dai(2019) [15]	30 HM(Discoery)	myopia(SE ≤ − 6.00 D)	-7.8 ± 1.5 (L), -7.9 ± 1.2 (R)	21.0 ± 5.2	60.00	China	
	30 controls (without myopia)(Discoery)		-0.5~0.5D(L, R)	23.2±6.9	45.00		
	20HM(Test)		-7.2 ± 1.2(L), -7.3 ± 1.1(R)	20.8 ± 3.3	36.80		
	19 controls (without myopia)(Test)		-0.5~0.5D(L,R)	22.6 ± 5.5	45.00		
[11] (2017) ju	20 HM	HM(SE > −6.00D or AL ≥ 26 mm)		60.35 ± 8.92	35.00	China	
	20 cataract patients			62.75 ± 9.20	30.00		
Barbas-Bernardos(2016)	12 HM	HM(AL > 26 mm), low myopia(AL ≤ 26				Valencia(Spain)	
[10]	26 with low myopia	mm)					
Kearney(2017) [12]	25 myopes(Phase 1)	myopia(SE ≤ −0.5D)	-2.37 ± 1.27D	19.1 ± 0.81		England	
	29 controls(Phase 1)		$+0.62 \pm 0.89D$				
	22 myopes(Phase 2)		$-2.34 \pm 1.12D$				
	23 controls(Phase 2)		+0.78 ± 1.16D				
Yang(2017) [17]	24 FD guinea pigs(3 d) and 10 controls		$FD(3d) - 2.73 \pm 2.35D$			Wenzhou, China	
	24 FD guinea pigs(2 w) and 12 controls		FD(2 weeks)-5.01 ± 2.53D				
Najjar(2021) [<mark>18</mark>]	18 chicks with monocular FD(3900 K SW LED light)					Singapore	
	18 chicks with monocular FD(BEW light)						
PM Pathological myopia, H SF Snherical equivalent. D	M High myopia, FD Form deprivation, Diopter: Al Axis lenoth.						

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Fin radiological myopid, mit mgn myopid, ru rom SE Spherical equivalent, *D* Diopter, *AL* Axis length, *L* Left, *R* Right.

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Quality assessment

The quality of the studies was assessed with the tool described by Lumbreras et al. according to the QUADOMICS [19] and detailed information is presented in Supplementary Table 1. The second and fourteenth items of the QUADOMICS were not applicable in practice to the included studies and, therefore, were not presented. All of the studies included in the review were phase I and all met the evaluation criteria of 70% or more, with one study meeting the criteria of 13/14 (92.86%) [17]. All studies described the sample type (item 3) and selection criteria (item 1), but none met item 12 in the tool, indicating that they interpreted the index test results with knowledge of the reference standard. With regard to item 16, three studies lacked measures to avoid overfitting [12, 13, 15] and one study was not relevant to the item [10].

Metabolic markers and disease prediction

We identified seven studies that compared the metabolic profiles of myopia and controls, most of which used a combination of univariate analysis and multivariate statistical methods as evaluation criteria to identify differential metabolic characteristics (Table 2). Notably, Du et al. identified 256 differential features in metabolomics and lipidomics based on the U test only, and further qualifying criteria for differential features were added in this review, including quantitative identification of endogenous metabolites and fold change over 1.5 or less than 0.5, resulting in 21 metabolic features included in this review for further study [13]. Liu et al. conducted a study related to macular neovascularization in which 14 differential metabolic features were identified between pathological myopia and cataract controls without fundus lesions, and with the help of the backward stepwiseregression selection procedure, the best combination of biomarkers for diagnosis between groups was selected with a model area under receiver operating characteristic value of 0.906 [16]. A summary of studies assessing the ability of differential metabolic characteristics to distinguish myopia-related disorders is shown in Table 3. In the two studies analysing AH, 29 and 21 differential features without duplication between HM and cataract control groups, respectively, were identified, probably due to the inconsistencies in other characteristics of the study demographics between the two groups [10, 11]. Two other studies on myopia and controls were done with targeted and untargeted serum metabolomics, respectively [12, 13]. After standardization of names and assignment of HMDB identity documents, the full study identified 87 compounds with differential variations.

In addition, animal studies are an important part of myopiarelated metabolomics research, and animal models make it possible to track changes in retinal metabolic profiles and changes in ocular tissue metabolism associated with environmental alterations. Two animal model studies were included in the review, in which Yang et al. established a retinal metabolism model for form-deprived guinea pigs and found differences in retinal metabolism at different time points [17]; Najjar et al. studied the effects of moderate levels of ambient standard white (Standard white light: 233.1 lux, 3900 K) and blue-enriched white (Blue-enriched white: 223.8 lux, 9700 K) lights on the growth and metabolism of the chicken-model of form-deprivation myopia [18]. This study showed that the metabolic structure of the vitreous and retina in restored form-deprived eyes differed from that of control eyes and was dependent on the spectral content of ambient light, providing new insights into light-dependent regulation in ocular growth and metabolomics.

Pathway analysis

Based on the source of the biological samples and study subjects, we divided the included studies into three parts: including serum metabolomics study of myopia-controls [12, 13], AH metabolomics study of HM-cataract controls [10, 11], and serum metabolomics study of HM-controls [14-16] for separate pathway analysis. Only the sphingolipid metabolic pathway was statistically significantly enriched in the myopia-control serum metabolomics study, whereas only the arginine biosynthetic pathway was statistically significantly enriched in the AH metabolomics study in the HMcataract control. In a HM-control serum metabolomics study, 39 differential metabolites were imported in MetaboAnalyst 5.0 software for pathway analysis. As is shown in Fig. 2, myopia-altered metabolites belonged to 29 biochemical pathways, four of which had statistically significant effects (Supplementary Fig. 1). In this pathway analysis, the Citrate cycle (TCA cycle) pathway had the largest effect values and was the most statistically significant pathway, which included four altered metabolites (Oxaloacetate; cis-Aconitate; Citrate; Pyruvate). Other pathways were also enriched (P < 0.05), including Alanine, aspartate, and glutamate metabolism; Glyoxylate and dicarboxylate metabolism; Biosynthesis of unsaturated fatty acids. Additional details are available in the Supplementary Tables 2, 3.

DISCUSSION

Nine myopia metabolomics studies were included in this review. including both human and animal studies. The human studies included three types of study samples: comparison of highly myopic and non-myopic controls, comparison of highly myopic and mildly myopic, and comparison of myopic and non-myopic, covering both serum and AH biological samples. Limitations in the number of included review studies and the use of different methods (GC, LC, etc) and differential metabolite evaluation criteria resulted in duplicate biomarkers not being obtained. However, the inclusion of diversity studies was complementary to each other. Pathway analysis revealed four statistically significant metabolic pathways enriched in the HM serum metabolomics study. Only the sphingolipid metabolic pathway was statistically significantly affected in the myopia-control serum metabolomics study, whereas only the arginine biosynthetic pathway was statistically significantly affected in the AH metabolomics study in the HM-cataract controls.

The body abnormally alters its original metabolic state, and identifying metabolic differences between disease and healthy states may lead to a better understanding of metabolic changes associated with disease [20]. Pathway analysis is the combination of altered metabolites with biological a priori knowledge to simplify the interpretability of data analysis [21]. The 23 differential metabolites identified by the myopic-control serum metabolomics study were mainly enriched in the Sphingolipid metabolism pathway and contained three altered metabolites (Sphingosine 1phosphate; Sphingomyelin; N-Acylsphingosine) [12, 13]. Sphingolipids are one of the major components of the phospholipid bilayer of eukaryotic biological cell membranes and act as signalling molecules regulating inflammation, cell migration [22]. Sphingomyelin is a sphingolipid with phosphocholine as a polar headgroup, which can be hydrolyzed to ceramides and phosphocholine [23]. Ceramide regulates various cellular processes including apoptosis and senescence [24]. Ceramide hydrolysis is the only synthetic pathway for sphingosine, and then sphingosine undergoes phosphorylation to produce sphingosine 1-phosphate (S1P) [25]. Blood levels of S1P are relatively higher than intracellular levels, and the retina also produces S1P, which plays an important role in retinal angiogenesis [26]. As myopia progresses, HM is often accompanied by retinopathy. The present study found serum changes in sphingomyelin between myopiccontrol groups, which may be a pre-warning of retinal changes, and further validations are needed in future studies.

TCA cycle and Glyoxylate and dicarboxylate metabolism pathways were significantly enriched in the serum metabolism of HM, suggesting that glycometabolism disorder may play an important role in the pathogenesis of HM. This observation is supported by both genomics and proteomics. Expression of metabolic genes

Table 2.	Summary	' of	myopia	metabolomics	study	results.
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References	Biofluid/Technique/ Standard	Altered metabolites ↑	ţ	
Liu(2020) [16]	Serum	(8)Oleic acid, Pyruvic acid, Pipecolic acid,	(6)Maleic acid, Beta-Alanine, Ribonolactone, D- Maltose, Hypoxanthine, L-2-Amino-3-(1- pyrazolyl)propanoic acid	
	GC-TOF-MS	Petroselinic acid, Linoleic acid, Glycerol, Palmitic		
	OPLS-DA(VIP > 1.0), <i>t</i> - test(P < 0.05), and FC > 1.2 or <0.8			
Ke(2020) [14]	Serum	(12)Citric acid, Aminomalonic acid, Palmitoleic	(8)Alanine, Mannose, Itaconic acid, Aconitic acid, O-Acetylserine, Phthalic acid, Abietic acid, Salicin	
	GC-TOF-MS	4-Hydroxyphenylacetic acid, Hesperitin,		
	PLS-DA(VIP > 1.0) and <i>t</i> -test(<i>P</i> < 0.05)	Anandamide, Oxalacetic acid, Pimelic acid, 2- Ketoadipate, N-Ethylmaleamic acid		
Du(2020) [13]	Serum	(12)DG(16:1(9Z)/16:1(9Z)/0:0), Sphingosine 1-	(9)Uridine diphosphate-N-acetylglucosamine, PC (18:2(9Z,12Z)/P-18:1(11Z)), Taurocyamine, Phosphoserine, DG(18:2(9Z,12Z)/20:4 (5Z,8Z,11Z,14Z)/0:0), SM(d18:0/22:0), Tyramine, all-trans-Retinoic acid, Docosapentaenoic acid (22n-3)	
	UHPLC-MS	phosphate, PC(24:0/P-18:1(11Z)), Allopregnanolone, DG(18:1(9Z)/18:2(9Z 12Z)/0:0)		
	U test(<i>P</i> < 0.001),FC > 1.5 or <0.5,quantified, Endogenous	Leukotriene D4, Ceramide (d18:1/18:0), Thiamine monophosphate, TG(16:0/16:0/18:0), DG(16:0/20:4 (5Z,8Z,11Z,14Z)/0:0), Deoxyuridine, Hydrocinnamic acid		
Dai(2019) [15]	Serum	(7)seryltryptophan, DG (8:0/17:0/0:0),	(2)5-Methyltetrahydrofolic acid, 12-Oxo-20-	
	LC-QTOF/MS	phosphatidylethanolamine (PE) (20:3/22:6),	trihydroxy-leukotriene B4	
	PLS-DA(VIP > 1.5) and <i>t</i> -test(<i>P</i> < 0.05)	25-hydroxyvitamin D2-25-glucuronide, γ- glutamyltyrosine, and 5-hydroxytryptamine		
Ji(2017) [11]	AH	(27)glutamine 1, N-alpha-Acetyl-L-ornithine 3,	(2)Analyte 420, Analyte 76	
	GC/ TOF MS	Nicotinoylglycine 2, o-Hydroxyhippuric acid 2,		
	PLS-DA(VIP \geq 1) and <i>t</i> -test($p \leq 0.05$)	acid, Linoleic acid, oxalic acid, hibose, cis-gondolic acid, Linoleic acid methyl ester, thymidine 3, phosphate, indole-3-acetamide 4, 2-aminophenol 2, 2-ketoadipate 2, 3-Phenyllactic acid, cis-Phytol, conduritol b epoxide 2, salicin, 3 analytes and 6 unknowns		
Barbas- Bernardos (2016) [10]	АН	(9)HM vs LM: L-Arginine, Citrulline, Butyryl-L- carnitine, Pantothenic Acid (VITAMIN B5), Sphinganine, Histidinyl-Phenylalanine, PC(O-	(13)Aminocyclohexanecarboxylic acid, Aminooctanoic acid, Aminoundecanoic acid, Dodecanedioic acid, Trihydroxyphenyl-gamma-	
	CE-MS & LC-MS	32:2)//PC(P-32:1), C24 Sulfatide, LacCer(d40:0)	valerolactone, Didehydro-Retinoic acid, L-Cysteinylglycine disulfide, Dihydropteroic acid,	
	OPLS-DA(VIP > 1) & RSD		Dimethylnonanoyl carnitine, PC (42:6), PC(P- 42:2)//PC(O-42:3), Trihexosylceramide (d36:2), NeuAcaGalCer(d42:2)	
Kearney(2017) [12]	Serum	(1)Melatonin(Phase 1&2)	(1)Dopamine(Phase 2)	
	LC-SPE-MS/MS			
	Logistic regression model			
Yang(2017) [17]	Retina	2-Ketoglutaric acid, glucose, mannose	urea, arabinose, tyrosine, glutamic acid, Treonine, valine, isoleucine, alanine, malic acid, Arachidic acid (20:0), Octadecenoic acid (18:1), Octadecanoic acid (18:0), Arachidonic acid (20:4), Hexadecanoic acid (16:0), Tetradecanoic acid (14:0), Octadecadienoic acid (18:2), Cholesterol, Ethanolamine, GABA	
	GC-TOF/MS			
	PLS-DA、OPLS(VIP > 1.0)			
Najjar(2021) [18]	Vitreous & Retinas	dihydroxyphenylalanine, ornithine, methionine sulfoxide, total dimethylarginine, arginine, lysine,	asparagine, glycerophospholipids, carnitine, Spermine, acylcarnitines	
	LC-MS	tryptophan,sphingolipids, glycerophospholipids, symmetric dimethylargining, glutamate		
	PLS-DA、OPLS-DA	serotonin, taurine, trans-4-Hydroxyproline, tyrosine, proline, histidine, phenylalanine, alanine, threonine, valine and isoleucine, monosaccharides, glycerophospholipids, sphingolipids		

GC Gas chromatography, TOF Time of flight, UHPLC Ultra Performance Liquid Chromatography, CE Capillary electrophoresis, SPE Solid phase extraction, MS Mass spectrometry;

OPLS-DA Orthogonal partial least squares discriminant analysis, PLS-DA Partial least squares discriminant analysis, FC Fold change, VIP Variable importance projection, RSD Relative standard deviation,

AH Aqueous humor.

Table 3. Classification potential of the information of metabolomics.					
References	Method	Discriminant models	Discriminant group;Precision		
Liu(2020) [16]	Logistic Regression	Demographic characteristics and panel metabolites (hypoxanthine, L-2-amino-3-(1-pyrazolyl)propanoic acid, linoleic acid, maleic acid, ribonolactone)	AUC = 0.906; Sen = 0.877; Spe = 0.684		
Ke(2020) [14]	AUC	Altered metabolites	AUC: 0.59~0.71		
Dai(2019) [15]	Logistic regression	γ -glutamyltyrosine and 12-oxo-20-trihydroxy-leukotriene B4	(Discovery)AUC = 0.983; Sen = 97%; Spe = 90%; (Test)AUC = 0.958; Sen = 80%; Spe = 95%		

AUC Area under receiver operating characteristic, Sen Sensitivity, Spe Specificity.



Fig. 2 Results of the pathway analysis of high myopia-control serum metabolomics study. The colour of the circle indicates the significance level in the enrichment analysis, the darker the colour (more red) indicates more significant; the size of the circle reflects the pathway impact value in the topology analysis, the larger the circle, the larger the impact value. The x-axis is the pathway impact value calculated based on topology analysis.

associated with mitochondrial metabolism pathways have been demonstrated to be correlated with ocular axial length and refraction [27], providing evidence of the abnormal energy metabolism during myopia development. Yu et al. performed quantitative analysis of myopic retinal proteins and found that many glucose-related metabolic enzymes (including pyruvate kinase and serine/threonine phosphatase) were upregulated, which supports the hypothesis of a hypoxic environment in animal models of myopia [28]. Abnormal retinal glucose metabolism was also found in the Yang et al. animal model, with glucose accumulating in the form-deprived retina, indicating a decrease in aerobic glycolysis and an abnormal TCA cycle [17]. Briefly, hypoxia could result in hyperactivation of adenosine signalling [29], one of the key signal pathways related to energy metabolism. Therefore, it could be inferred that the development of myopia was related to energy metabolism disorder and oxidative stress induced by anoxic environment. In addition, in this review, we observed that the metabolic regulatory mechanisms controlling progression of myopia and DR and AMD appear to be related [9, 30], since AMD and DR have the same abnormal energy metabolism feature as HM. Because choroidal neovascularization is a common pathology in many vision-threatening retinal diseases [31, 32], biochemical changes in the choroid might underlie progression of eve diseases. Neovascular tissues are prone to be incompetent and leaky, thus presumably altering the availability of metabolites. One of the pathological mechanisms expected to be involved is the choroid energy metabolism. Emerging evidence has displayed that glucose metabolism could control endothelial cell neovascularization [33]. VEGF has emerged as a key molecular regulator of retinal and choroidal neovascularization. Over the past decade, the advent of injectable antiangiogenic agents into the vitreous cavity of the eye has achieved considerable success [34]. Many therapeutic compounds with parameters and antibodies that bind VEGF have placed emphasis on inhibiting the VEGF signalling pathway. In oxygen-induced retinopathy (OIR) mouse models, oral administration or intraperitoneal injections of arginine-glutamine increased retinal Docose Hexaenoie Acid (DHA) and inhibited retinal neovascularization, which was also associated with decreased retinal VEGF mRNA levels [35]. Vascular dysfunction may also lead to oxidative stress and excessive production of reactive oxygen species (ROS), resulting in oxidative damage [36]. The corresponding hypoxiainducible factor-1 α (HIF-1 α) is a specific mediator in adaptation to hypoxic environments and pathological responses [37]. Therefore, the link between the role of disorders of glucose metabolism, and VEGF-related hypoxic metabolism in retinopathy may be a common potential therapeutic target for these visionthreatening diseases and needs to be further investigated.

Limitations and research directions

Myopia is a risk factor for numerous ocular diseases, especially in patients with HM, and increases the risk of a range of secondary sequelae and irreversible vision loss, including myopic macular degeneration, glaucoma, retinal detachment, and cataract [38]. A summary of current myopia metabolomics studies shows that there are few reproducible validated results, mainly for the following reasons that limit the extrapolation of findings: First, due to the diverse metabolic status of the population [39], with age, gender, race, and lifestyle affecting metabolic levels, most studies do not describe quality control prior to sample collection, which may affect the extent to which differential metabolites are identified. Therefore, in experimental design sessions, study subjects should fast overnight or for at least one or two hours prior to sample collection; a complete study presentation should include a brief description of food intake over the past 12-24 h; [40] and reduce the possibility of differences due to sex, age, race, BMI and lifestyle factors. Most of the studies in this review matched subjects for age and gender, but did not describe in detail the pre-sampling medication and dietary status of the subjects. Myopia is a disease with a complex aetiology, and the study population includes patients with HM, low myopia, and concomitant retinal changes, etc., and because the eye is a relatively independent organ, the difficulty in obtaining ocular tissue, controls and myopes are often derived from non-healthy controls operated on for cataract or other eye diseases, which limits the range of myopia-related biomarkers. Second, the

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different evaluation criteria for metabolite detection platforms and differential metabolites in the study also contributed to the lack of duplicate identification of metabolites. For example, Liu et al. conducted experiments using a GC-TOF-MS metabolic detection platform, where biological samples need to be derivatized to be volatile for analysis [16], while the study by Barbas-Bernardos et al. adopted a dual platform combination strategy to increase metabolite coverage [10]. The characteristics contained in the preprocessed data are static during data analysis, so it is important to choose a suitable statistical method [41]. The quality of studies evaluated in this review found studies in which only five studies considered avoiding overfitting [11, 14, 16-18], except for the study by Barbas-Bernardos et al. which did not involve overfitting issues [10]. Our inability to access the original assay data for most of the studies also limited the reuse of the study results. Finally, the pathway-based analysis approach combines the results of data analysis with a priori knowledge of biology by combining a series of altered metabolic features revealed by statistical analysis in metabolomics studies [21]. However, due to the limited number of studies, only the Sphingolipid metabolism pathway was enriched in serum metabolomics studies in children and adolescents, and only the Arginine biosynthesis pathway was enriched in AH metabolomics studies in HM. We believe that NMR metabolomics studies on myopia will be a promising tool for analysis. In the future, metabolomics studies of complex diseases need to develop in a comprehensive (multi-platform approach, multi-omics, multitissue, and multi-time-point) direction to gradually map the complete myopia mechanism network and provide a reference basis for myopia prevention and control.

In conclusions, myopia has metabolic alterations in serum, AH, and retina, and the predictive value of a metabolic biomarker panel suggests the value of metabolomics in the management of myopia disease. Abnormal carbohydrate metabolism is present in HM, and the same pathways of change exist as in AMD and DR. In the future, standardized myopia metabolomics studies are needed to further unravel the mystery of myopia.

Supplementary information is available at Eye's website

SUMMARY TABLE

What was known before

 Myopia is the most common eye abnormality and the global burden of myopia is increasing rapidly in past decades. In ophthalmology, metabolomics has successfully been used for the description of disease progression and mechanism exploration, including age-related macular degeneration, glaucoma, and diabetic retinopathy, contributing to the unveiling of ophthalmic diseases.

What this study adds

 This study found significant metabolic alterations in serum, aqueous humorous fluid and retina in myopia, and found that the same metabolic changes exist in myopia as in other eye diseases.

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AUTHOR CONTRIBUTIONS

Conception or design of the work: XWH. Data collection: XWH and YW. Data analysis and interpretation: XWH and CK. Article drafting: XWH and YW. Critical revision of the article: CK and CWP. Final approval of the version to be published: XWH, YW, CK, and CWP.

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COMPETING INTERESTS

The authors declare no competing interests.

ADDITIONAL INFORMATION

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Correspondence and requests for materials should be addressed to Chen-Wei Pan.

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