Online Submissions: http://www.wjgnet.com/esps/bpgoffice@wjgnet.com doi:10.3748/wjg.v20.i1.163

World J Gastroenterol 2014 January 7; 20(1): 163-174 ISSN 1007-9327 (print) ISSN 2219-2840 (online) © 2014 Baishideng Publishing Group Co., Limited. All rights reserved.

ORIGINAL ARTICLE

Serum and urine metabolomic fingerprinting in diagnostics of inflammatory bowel diseases

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Supported by Wroclaw Medical University Research Program, Grant Number 15/Pbmn

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Telephone: +48-71-7332002 Fax: +48-71-7332009 Received: October 24, 2013 Revised: November 21, 2013

Accepted: December 5, 2013 Published online: January 7, 2014

Abstract

AIM: To evaluate the utility of serum and urine metabolomic analysis in diagnosing and monitoring of inflammatory bowel diseases (IBD).

METHODS: Serum and urine samples were collected from 24 patients with ulcerative colitis (UC), 19 patients with the Crohn's disease (CD) and 17 healthy controls. The activity of UC was assessed with the Simple Clinical Colitis Activity Index, while the activity of CD was determined using the Harvey-Bradshaw Index. The analysis of serum and urine samples was performed using proton nuclear magnetic resonance (NMR) spectroscopy. All spectra were exported to Matlab for preprocessing which resulted in two data matrixes for serum and urine. Prior to the chemometric analysis, both data sets were unit variance scaled. The differences in metabolite fingerprints were assessed using partial least-squaresdiscriminant analysis (PLS-DA). Receiver operating characteristic curves and area under curves were used to evaluate the quality and prediction performance of the obtained PLS-DA models. Metabolites responsible for separation in models were tested using STATISTICA 10 with the Mann-Whitney-Wilcoxon test and the Student's t test ($\alpha = 0.05$).

RESULTS: The comparison between the group of patients with active IBD and the group with IBD in remission provided good PLS-DA models (*P* value 0.002 for serum and 0.003 for urine). The metabolites that



allowed to distinguish these groups were: N-acetylated compounds and phenylalanine (up-regulated in serum), low-density lipoproteins and very low-density lipoproteins (decreased in serum) as well as glycine (increased in urine) and acetoacetate (decreased in urine). The significant differences in metabolomic profiles were also found between the group of patients with active IBD and healthy control subjects providing the PLS-DA models with a very good separation (P value < 0.001 for serum and 0.003 for urine). The metabolites that were found to be the strongest biomarkers included in this case: leucine, isoleucine, 3-hydroxybutyric acid, N-acetylated compounds, acetoacetate, glycine, phenylalanine and lactate (increased in serum), creatine, dimethyl sulfone, histidine, choline and its derivatives (decreased in serum), as well as citrate, hippurate, trigonelline, taurine, succinate and 2-hydroxyisobutyrate (decreased in urine). No clear separation in PLS-DA models was found between CD and UC patients based on the analysis of serum and urine samples, although one metabolite (formate) in univariate statistical analysis was significantly lower in serum of patients with active CD, and two metabolites (alanine and N-acetylated compounds) were significantly higher in serum of patients with CD when comparing jointly patients in the remission and active phase of the diseases. Contrary to the results obtained from the serum samples, the analysis of urine samples allowed to distinguish patients with IBD in remission from healthy control subjects. The metabolites of importance included in this case up-regulated acetoacetate and down-regulated citrate, hippurate, taurine, succinate, glycine, alanine and formate.

CONCLUSION: NMR-based metabolomic fingerprinting of serum and urine has the potential to be a useful tool in distinguishing patients with active IBD from those in remission.

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Key words: Metabolomics; Inflammatory bowel disease; Ulcerative colitis; Crohn's disease; Proton nuclear magnetic resonance spectroscopy; Urine; Serum; Partial least-squares-discriminant analysis

Core tip: This study demonstrates that proton nuclear magnetic resonance -based metabolic fingerprinting of human serum and urine combined with multivariate data analysis could be a useful tool for differentiation between the active and remission phase of the disease that is of great importance in inflammatory bowel diseases (IBD) monitoring. At the same time, the results indicate that this diagnostic method has rather a weak potential in the differential diagnosis of IBD.

Dawiskiba T, Deja S, Mulak A, Ząbek A, Jawień E, Pawełka D, Banasik M, Mastalerz-Migas A, Balcerzak W, Kaliszewski K, Skóra J, Barć P, Korta K, Pormańczuk K, Szyber P, Litarski A, Młynarz P. Serum and urine metabolomic fingerprinting in

diagnostics of inflammatory bowel diseases. *World J Gastroenterol* 2014; 20(1): 163-174 Available from: URL: http://www.wjgnet.com/1007-9327/full/v20/i1/163.htm DOI: http://dx.doi.org/10.3748/wjg.v20.i1.163

INTRODUCTION

Ulcerative colitis (UC) and the Crohn's disease (CD) are the main well-characterized forms of inflammatory bowel diseases (IBD). Although the etiology of IBD remains not fully explained, it is widely accepted that this multifactorial condition is associated with an inadequate immune response to environmental factors in genetically predisposed individuals^[1,2]. The diagnosis of IBD is based on the combination of clinical, endoscopic, histopathological, serological and radiological features but the precise distinction between UC and CD still remains a challenge in 5%-20% cases^[3-6]. The diagnosis may be difficult due to the clinical overlapping between these two conditions, an incomplete expression of IBD or deficiency in histological response of the intestinal mucosa^[4]. Moreover, the serological markers like anti-Saccharomyces cerevisiae antibodies and perinuclear antineutrophilic cytoplasmic antibodies recently implemented in IBD diagnostics have been found to have a high specificity in some studies, but a relatively low sensitivity [5,7]. The differential diagnosis between UC and CD is also essential regarding distinct approaches with respect to surgical treatment, moreover the diseases differ from each other in terms of prognosis, a tendency for recurrence and risks of cancer^[3,4]. This situation as well as the limited number of IBD monitoring tools (the commonly used erythrocyte sedimentation rate and C-reactive protein do not always correlate with disease activity and are not specific to IBD inflammation) force researchers to look for new approaches in IBD diagnostics^[8].

The use of metabolomics in medical diagnostics has recently become a very promising idea that is extensively studied in oncological conditions, diabetes, cardiovascular diseases, rheumatoid arthritis or multiple sclerosis [9-17]. Unlike the genomic and proteomic studies that examine genes and more or less complex proteins, metabolomic analysis allows to assess the simplest low molecular weight metabolites that are involved in disease processes (the basic forms and various derivatives of amino acids, ketones, fatty acids, amines, organic acids, nucleosides, aromatic compounds, sterols, saccharides etc.)[3,18-20]. There are several thousands of such metabolites in tissues and fluids of living organisms like blood, urine, cerebrospinal fluid or saliva, and most of those metabolites have a welldefined function. At the same time, the highly informative metabolic profiles as a potential source of biomarkers can be used not only for prognostic purposes, but also as an input for discriminant analysis, even when there are only premorphological biochemical changes that are not evident on histopathological examination [4,21].

So far there are only a few research studies that evalu-



Table 1 Demographic data and clinical profile of patients

	Ulcerative colitis	Crohn's disease	Healthy controls
Number of patients	24	19	17
Sex (male/female)	15/9	11/8	8/9
Average age	40.7 (18-70)	35.8 (18-61)	44.7 (24-62)
(mean/range)			
Mean activity score	6.1 (0-13)	6.7 (1-18)	-
SCCAI/HBI (range)			
Number of patients	12/12	16/3	-
in active/remission			
phase of disease			
Average disease	8.2 (0.1-32)	5.5 (0.3-21)	-
duration (yr)			

SCCAI: Simple clinical colitis activity index; HBI: Harvey-bradshaw index.

ate the usefulness of metabolomics in IBD diagnostics in humans. Fecal extracts^[22], intestinal tissue biopsies^[1,3,4,23], urine^[1,9,24,25] and blood^[21,24,26,27] were assessed as the source of potential biomarkers. It should be noted that the results of some investigations were contradictory, at the same time, to date there has been no study in which proton nuclear magnetic resonance (¹H NMR)-based combined serum and urine metabolomic analysis had been used to differentiate patients in the active and remission phase of IBD. The aim of our study was to answer the question if the ¹H NMR spectroscopy-based metabolic fingerprinting of human serum and urine combined with multivariate data analysis could be useful diagnostic tool in diagnosing and monitoring of IBD.

MATERIALS AND METHODS

Patient recruitment

Samples were collected from patients who were hospitalized at the three cooperating clinics of Wroclaw Medical University: the Department of Gastroenterology and Hepatology, the First Department and Clinic of General, Gastroenterological and Endocrinological Surgery and the Department of Vascular, General and Transplantation Surgery. The protocol of this study was approved by the Commission of Bioethics at Wroclaw Medical University (Approval no. KB-597/2011) and written informed consents were obtained from all the patients before enrolment in the study.

Serum and urine samples were collected from adult individuals with established diagnosis of UC (n = 24) and CD (n = 19). Reference samples were taken from healthy control individuals (n = 17). No patients with indeterminate colitis were included. The activity of UC was assessed with the Simple Clinical Colitis Activity Index (SCCAI)^[28], while the activity of CD was determined using the Harvey-Bradshaw Index (HBI)^[29]. Based on the individual SCCAI/HBI scores the patients were classified into the subgroup with the active or remission phase of the disease (remission was defined as SCCAI < 5 for patients with UC and HBI < 5 for patients with CD).

Patients aged over 75 years or below 18 years, with severe mental illness, infectious diseases, structural ab-

normalities of the gastrointestinal tract, diabetes mellitus, renal failure, hepatic dysfunction, evidence of malignancy and other severe diseases were excluded from our study. To avoid the dietary influence on the metabolic fingerprints of serum and urine the samples were taken after overnight fasting. Signals related to medications (5-aminosalicylate, azathioprine and acetaminophen) were eliminated from the statistical and chemometric evaluation. The characteristics of patients and healthy control subjects are shown in Table 1.

Sample handling and NMR spectroscopy

Serum samples were thawed in room temperature and mixed using a vortex mixer. According to a well-established protocol^[30], aliquots of 200 µL of serum were transferred into a polypropylene Eppendorf tube and 400 μL of 0.9% saline solution in 15% D2O, containing 3 mmol/L 3-(trimethylsilyl)-2,2',3,3'-tetradeuteropropionate sodium salt TSP-d4 (TSP) were added. The samples were centrifuged for 10 min at $12000 \times g$ and 550 µL of each sample supernatant was subsequently moved into a 5 mm NMR tube. ¹H NMR experiments were performed with the following parameters: relaxation delay, 3.5 s; acquisition time, 2.73 s; number of transients (scans), 128; number of points, 65536; pulse program, cpmgpr1d (Bruker notation); number of loops, 80; spin echo delay, 400 ms; spectral width, 20 ppm; line-broadening factor, 0.3 Hz. The spectra were manually corrected for phase and baseline distortions and were referenced to α -glucose signal (δ = 5.225 ppm).

Urine exhibits high pH and ionic strength variability which can dramatically affect NMR spectra quality by introducing an unwanted variance in the chemical shifts of resonances. In order to overcome this issue we used a modified method of Jiang et al^[31] utilizing the additional step of precipitation of the most abundant urine salts: Mg²⁺ and Ca²⁺. All urine samples were thawed in room temperature and mixed using a vortex mixer. Aliquots of 570 µL of urine were transferred into polypropylene Eppendorf tubes and 30 µL of KF solution was added, and samples were mixed again creating insoluble in water MgF2 and CaF2 salts. The samples were centrifuged for 10 min at $12000 \times g$ and $540 \mu L$ of supernatant was transferred into a new Eppendorf tube. Next, the samples were mixed with 60 µL of 2.5 mol/L phosphate buffer solution with a pH of 7.00 in 99.8% D2O containing TSP as an internal standard. The final concentration of TSP in the sample was 1 mmol/L. The samples were mixed again and finally 550 µL was moved into a 5 mm NMR tube. ¹H NMR experiments were performed with the following parameters: relaxation delay, 3.5 s; acquisition time, 1.36 s; number of transients, 64; number of points, 32768; pulse program, noesypr1d (Bruker notation); spectral width, 20 ppm, line-broadening factor, 0.3 Hz. The spectra were manually corrected for phase and baseline distortions and were referenced to the TSP resonance at $\delta = 0.0$ ppm.

All NMR spectra were recorded at 300 K using a Bruker Avance II 600 spectrometer (Bruker GmBH,



Germany) operating at proton frequency of 600.58 MHz.

Data processing and multivariate statistical data analysis

All spectra were exported to Matlab (Matlab v. 8.1, Mathwork Inc.) for preprocessing which resulted in two data matrixes for serum and urine, respectively. Regions affected by solvent suppression were excluded (4.645-5.100 ppm for serum and 4.660-5.055 ppm for urine) and alignment procedures involving correlation optimized warping (COW)^[32] and interval correlation shifting (icoshift) algorithms were applied^[33].

Spectra of urine were aligned in two steps: first a global correction was performed using COW algorithm, while in the second step a custom interval approach was applied to adjust peak positions in baseline separated regions defined by the researcher. Urine spectra consisted of 24609 data points and were normalized using the Probabilistic Quotient method (PQ) to overcome the issue of dilution^[34].

Serum samples are much more stable compared with urine and usually demonstrate only small variation in resonance chemical shift. However, alignment may be required especially for the fingerprinting approach where even small errors could affect the multivariate analysis and bilinear data are required. For that reason correction of peak shifts was done using only the icoshift algorithm. Serum spectra consisted of 24446 data points and were normalized using the PQ to overcome the issue of dilution [34].

Prior to the chemometric analysis, both data sets were unit variance (UV) scaled. The differences in metabolite fingerprints were assessed using partial least-squares-discriminant analysis (PLS-DA). A default 7-fold cross validation (CV) was applied to each PLS-DA model (1/7 of the samples being excluded from calculations in each round).

Receiver Operating Characteristic (ROC) Curves and Area Under Curves (AUC) were used to evaluate the prediction performance of the obtained PLS-DA models. A performance curve algorithm from the Matlab statistics toolbox (Matlab v. 8.1, Mathwork Inc.) was utilized for calculations. Specificity and sensitivity were determined based on sample class prediction: predicted values of the fitted Y for observations (in SIMCA-P+ software: Y-predcv computed based on the 7-fold cross validation procedure).

Metabolites responsible for separation in models were tested using STATISTICA 10 with the Mann-Whitney-Wilcoxon test (MWW) and the Student's *t* test. A 0.05 level of probability was used as the criterion for statistical significance.

RESULTS

Metabolic fingerprinting of serum samples from IBD patients

Median spectra of serum from IBD patients and healthy control subjects are shown in Figure 1, while the parameters of PLS-DA models obtained from ¹H NMR analysis of serum samples are shown in Table 2.

In total, 43 low molecular weight compounds were assigned in serum spectra. The comparison between the group of patients with active IBD and the group with IBD in remission provided a good PLS-DA model. The metabolites that allowed to distinguish these groups were down-regulated low-density lipoproteins (LDL) and very low-density lipoproteins (VLDL) as well as up-regulated N-acetylated compounds, phenylalanine and the unknown metabolite with signal at δ 1.39 ppm. The significant differences were also found between the group of patients with active IBD and healthy control subjects providing the PLS-DA model with a very good separation that passed validation test (CV-ANOVA, P < 0.05). In this case the following metabolites were increased in IBD patients: leucine, isoleucine, 3-hydroxybutyric acid, N-acetylated compounds, acetoacetate, glycine, phenylalanine, lactate and one unknown (Unk) metabolite with signal at δ 1.39 ppm mentioned above. Simultaneously lower levels of creatine, dimethyl sulfone, choline and its derivatives glycerophosphocholine (GPC) and phosphocholine (PC) as well as histidine were observed (Table 3).

No clear separation in PLS-DA models was found between CD and UC patients (all the patients as well as only those in the active phase of the disease). Only one metabolite (formate) in univariate statistical analysis was significantly lower in patients with active CD, and two metabolites (alanine and *N*-acetylated compounds) were significantly higher in patients with CD when comparing jointly patients in the remission and active phase of the diseases. It was also not possible to distinguish IBD patients in remission from healthy control subjects, although there were two statistically relevant differences in serum metabolic fingerprints between those cohorts of patients. The IBD patients in remission were characterized by a significantly higher concentration of dimethylamine and a lower level of dimethyl sulfone.

Metabolic fingerprinting of urine samples from IBD patients

Median spectra of urine from IBD patients and healthy control subjects are shown in Figure 2, while the parameters of PLS-DA models obtained from urine samples based on ¹H NMR analysis are shown in Table 4.

As in the case of the serum samples, the most significant differences in urine were found between the group of patients with active IBD and the group with IBD in remission, and between the group of patients with active IBD and healthy control subjects. In the first case a lower concentration of acetoacetate and a higher concentration of glycine were detected, while in the second case citrate, hippurate, trigonelline, taurine, succinate, 2-hydroxyisobutyrate (down-regulated) and an unknown metabolite with 4-hydroxyphenyl group and signal at δ 6.85 ppm (up-regulated) were found to be the strongest biomarker candidates (Table 5).

Contrary to the results obtained from the serum sam-



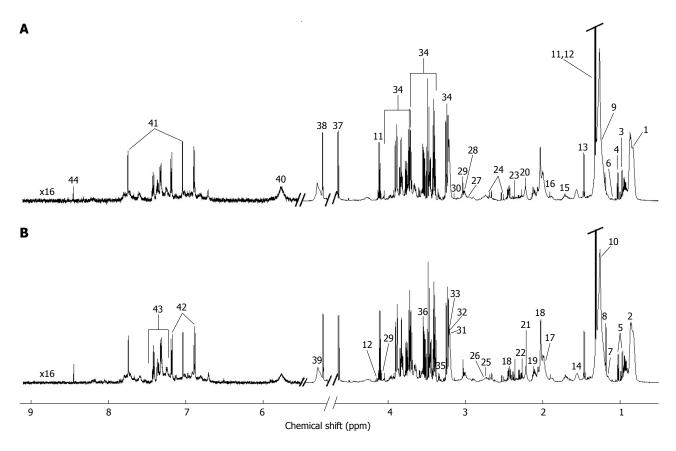


Figure 1 Median 600 MHz proton nuclear magnetic resonance noesy spectra of serum samples. A: Healthy control subjects; B: Patients with inflammatory bowel diseases. 1: Lipids LDL CH₃-(CH₂)_{n-}; 2: Lipids VLDL CH₃-(CH₂)_{n-}; 3: Leucine; 4: Isoleucine; 5: Valine; 6: 3-Methyl-2-oxovalerate; 7: Isobutyrate; 8: 3-Hydroxybutyrate; 9: Lipids LDL CH₃-(CH₂)_{n-}; 10: Lipids VLDL CH₃-(CH₂)_{n-}; 11: Lactate; 12: Threonine; 13: Alanine; 14: Lipids VLDL CH₂-C

Table 2 The partial least-squares-discriminant analysis models obtained from proton nuclear magnetic resonance-based analysis of serum samples

Comparison	Model	<i>P</i> value	Q ² (cum)	Number of latent variables
CD vs UC	PLS-DA	0.069	0.43	2
CD active vs UC active	PLS-DA	0.798	0.46	2
IBD active vs Control	PLS-DA	< 0.001	0.79	4
IBD remission vs Control	PLS-DA	0.372	0.26	2
IBD active vs IBD remission	PLS-DA	0.002	0.76	2

CD: Crohn's disease; UC: Ulcerative colitis; IBD: Inflammatory bowel diseases; PLS-DA: Partial least-squares-discriminant analysis.

ples, the analysis of urine samples allowed to distinguish patients with IBD in remission from healthy control subjects. The following metabolites were down-regulated in IBD patients in remission: citrate, hippurate, taurine, succinate, glycine, alanine and formate, and only acetoacetate was up-regulated (Table 5). It is worth noticing that the first four metabolites listed above (citrate, hippurate, taurine, succinate) significantly changed in the same direction in patients in both the active and remission phase of IBD.

As in the case of serum samples, the PLS-DA models between CD and UC patients did not pass the validation test (all the patients as well as only those in the active phase of the disease). There were no statistically relevant differences in urine metabolic fingerprints between those two cohorts of patients.

DISCUSSION

In the present study we used a ¹H NMR-based metabolomic approach to demonstrate metabolic differences between patients with active IBD and those in remission. Additionally, we investigated the differences between CD and UC patients as well as IBD subjects and healthy controls. The obtained results confirm the potential of this method in IBD monitoring, simultaneously it may provide the understanding of pathogenic mechanisms of the diseases. Notably, to date there are only few research studies that evaluated the metabolomic based diagnostic tests of serum^[21,24,26,27] and urine^[1,9,24,25] in patients with IBD. The earlier studies in animals (most often using a model of acute colitis in interleukin-10 gene deficient mouses or dextran sulfate sodium-induced colitis) provided only the comparison of the disease state *vs* healthy

Table 3 Significantly changed serum metabolites

Metabolite	Percentage difference					Relative Standard Deviation (%)							
	CD vs UC	CD active <i>vs</i> UC active	IBD active vs control	IBD remission vs control	IBD active vs IBD remission	CD	CD active	uc	UC active	IBD active	IBD remission	Control	
Lipids LDL	-7.2	-4.0	-11.6	0.4	-11.9ª	20.7	20.5	20.4	18.1	19.2	20.9	24.8	
CH3-(CH2)n-													
Lipids VLDL	-6.4	-6.0	-12.9	-2.2	-10.9°	18.3	18.9	20.9	26.0	22.2	13.7	24.0	
CH3-(CH2)n-													
Leucine	1.2	0.7	15.3ª	11.6	3.3	25.1	25.9	19.9	21.2	23.6	19.3	10.8	
Isoleucine	6.2	6.5	23.6 ^{a,c}	12.4	9.9	22.2	20.5	22.5	24.8	22.1	22.0	13.9	
3-Hydroxybutyrate	-6.5	3.8	52.6 ^{a,c}	45.9	4.6	50.2	48.1	76.8	53.9	49.6	94.4	34.1	
Unk1 δ 1.39 ppm	4.9	1.8	9.3 ^{a,c}	-3.0	12.7 ^{a,c}	11.3	10.4	10.4	7.5	9.2	10.0	10.8	
Alanine	12.2°	13.7	8.5	6.0	2.4	22.1	24.1	24.3	27.5	25.8	19.8	16.2	
N-acetylated	$8.9^{a,c}$	3.2	16.1 ^{a,c}	1.0	15.0 ^{a,c}	11.1	11.3	14.1	13.1	12.0	11.1	8.4	
compounds													
Acetoacetate	-2.8	4.4	77.3ª	50.7	17.7	79.8	76.7	97.9	79.8	76.6	118.6	49.1	
Citrate	2.6	2.5	-3.9	-10.9	7.9	20.5	20.2	23.9	26.1	22.4	21.5	27.5	
Dimethylamine	-4.5	- 6.0	14.5	16.1 ^{a,c}	-1.5	23.4	25.4	16.3	16.9	21.8	15.0	22.7	
Creatine	5.5	-4.1	-14.7°	-13.7	-1.1	24.9	22.6	31.4	31.0	26.2	32.9	28.4	
Dimethyl sulfone	-0.9	4.0	-31.0 ^{a,c}	-22.9 ^a	-10.5	31.2	33.1	30.1	25.6	29.8	30.4	36.1	
Choline + PC + GPC	-5.0	-0.6	-15.0°	-2.2	-13.0	21.5	21.9	24.0	21.2	21.2	23.3	27.9	
Glycine	6.8	0.8	26.6 ^{a,c}	13.7	11.3	20.3	21.2	24.6	26.8	23.3	19.9	21.2	
Phenylalanine	6.8	0.1	19.8 ^{a,c}	-0.6	20.5 ^{a,c}	25.9	26.4	22.3	20.7	23.7	18.8	18.4	
Histidine	2.6	7.8	-16.4 ^{a,c}	-6.7	-10.4	33.2	36.3	22.6	26.4	32.5	16.6	19.0	
Formate	-10.1	-22.7°	27.3	32.4	-3.8	48.5	51.1	32.5	30.6	42.7	34.2	45.6	
Lactate	4.1	10.3	17.8°	13.2	4.1	20.7	19.9	28.1	29.4	24.0	26.8	36.5	

Percentage difference was calculated based on the mean values of relative signal integrals in each group. The calculations were made from left to right, e.g., a positive value in IBD active vs Control means that a particular metabolite exhibits higher concentration in patients with active IBD compared to the Control group, $^aP < 0.05$ using the Student's t test, $^cP < 0.05$ using the Mann-Whitney-Wilcoxon test. CD: Crohn's disease; UC: Ulcerative colitis; IBD: Inflammatory bowel diseases; LDL: Low-density lipoproteins; VLDL: Very low-density lipoproteins; PC: Phosphocholine; GPC: Glycerophosphocholine; Unk1: Unknown metabolite.

Table 4 The partial least-squares-discriminant analysis models obtained from proton nuclear magnetic resonance-based analysis of urine samples

Comparison	Model	<i>P</i> value	Q² (cum)	Number of latent variables
CD vs UC	PLS-DA	0.388	0.1	2
CD active vs UC active	PLS-DA	1.000	-0.2	2
IBD active vs control	PLS-DA	0.003	0.6	4
IBD remission vs control	PLS-DA	0.037	0.31	2
IBD active vs IBD remission	PLS-DA	0.003	0.38	3

CD: Crohn's disease; UC: Ulcerative colitis; IBD: Inflammatory bowel diseases; PLS-DA: Partial least-squares-discriminant analysis.

controls and did not allow to evaluate CD vs UC or to compare exacerbation vs remission [35-42].

To the best of our knowledge this is the first study in which ¹H NMR-based combined serum and urine metabolomic analysis is used to differentiate patients in the active and remission phase of IBD. The obtained results allowed to create very good cross-validated PLS-DA models with a high predictability for both biofluids (Figures 3 and 4). The metabolites that were found to be the strongest candidates for biomarkers were: phenylalanine and *N*-acetylated compounds in serum as well as glycine in urine (higher concentration), and LDL and VLDL in serum as well as acetoacetate in urine (lower concentration). The postulated mechanism of serum LDL and VLDL down-regulation demonstrated in our study is

the systemic activation of the inflammatory response with an increase in secretory phospholipase A2 activity that promotes hydrolysis of phospholipids in LDL and VLDL^[43]. Moreover, the effects of the proinflammatory cytokines like tumor necrosis factor-α, interleukin-1 and interferon-c, which affect the expression of lipoprotein lipase that hydrolyzes triglycerides in lipoproteins, are also considered [44,45]. In addition, it has been shown that lipoproteins in IBD patients are synthesized with abnormal apolipoprotein structure [46,47]. The up-regulation of serum N-acetylated compounds that was demonstrated in the active phase of the disease could be mainly associated with a1-acid glycoprotein that is an acute-phase protein^[26,48,49], while the increase in the concentration of serum phenylalanine is most likely caused by the impairment of phenylalanine-4-hydroxylase activity (a hepatic enzyme that converts phenylalanine into tyrosine) associated with immune activation and inflammation [50]. The increased glycine concentration in urine in patients in the active phase of IBD observed in our study seems to be related with its anti-inflammatory and cytoprotective effects including the inhibition of glycine-gated chloride channels, direct effects on target cells and/or inhibition of neutrophils and macrophages activation^[51,52]. At the same time, the decrease in urine acetoacetate concentration was rather unexpected. Acetoacetate as a ketone body reflects the demand of the body for energy, so it is naturally increased in the serum of patients with inflammatory disorders (just as in our study when comparing the group



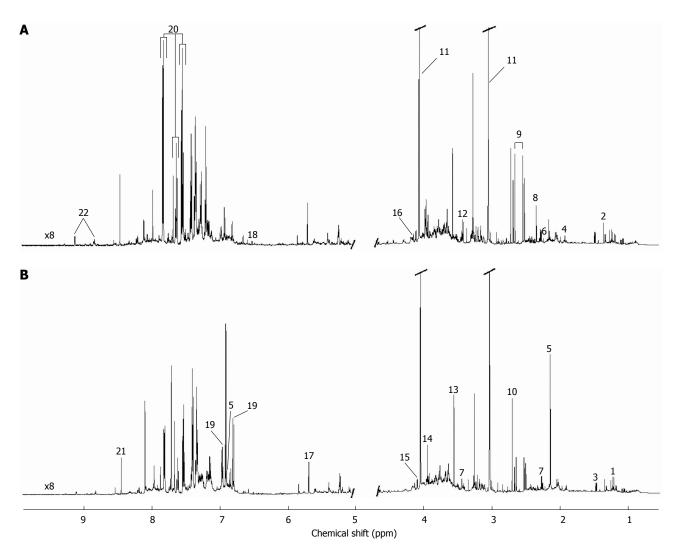


Figure 2 Median 600 MHz proton nuclear magnetic resonance noesy spectra of urine samples. A: Healthy control subjects; B: Patients with inflammatory bowel diseases. 1: 3-hydroxyisovalerate; 2: 2-hydroxyisobutyrate; 3: Alanine; 4: Acetate; 5: Acetaminophen; 6: Acetone; 7: Acetoacetate; 8: Succinate; 9: Citrate; 10: Dimethylamine; 11: Creatinine; 12: Taurine; 13: Glycine; 14: Creatine; 15: Lactate; 16: Threonine; 17: Cis-aconitate; 18: Fumarate; 19: 5-aminosialicylate; 20: Hippurate; 21: Formate; 22: Trigonelline.

of patients with active IBD and healthy controls). In only two found in literature research studies that compared the metabolomic profiles of patients in the active and remission phase of IBD no significant differences were found in one case (only urine evaluation of UC patients)^[1], while in the other one based on the chromatography data system and generation of a multivariate index using the "AminoIndexTM technology" (MIAI) a very good separation was achieved with a decrease in the concentration of tryptophan and histidine (only serum analysis)^[21].

Our results demonstrate that there are many metabolites that can be considered as potential biomarkers of IBD. At the same time, it is worth noticing that no comparison between IBD patients in the remission phase of the disease and healthy control subjects has been made so far. The metabolomic serum analysis in IBD patients in the active phase of the disease compared to healthy controls revealed an increase in the concentration of leucine, isoleucine, 3-hydroxybutyric acid, *N*-acetylated compounds, acetoacetate, glycine, phenylalanine, lactate and

a simultaneous decrease in the concentration of creatine, dimethyl sulfone, choline compounds, and histidine. At the same time, we found only two statistically relevant differences in serum metabolomic profiles between IBD patients with quiescent disease and healthy control subjects (a higher level of dimethylamine and a lower level of dimethyl sulfone) and in this case it was not possible to distinguish the cohorts by multivariate modeling. In contrast to the serum analysis, urine metabolomic tests proved that it is possible to distinguish IBD patients from healthy controls both in the active and remission phase of the disease. The active phase of IBD was associated with a higher concentration of an unknown metabolite with 4-hydroxyphenyl group and a lower concentration of citrate, hippurate, trigonelline, taurine, succinate, 2-hydroxyisobutyrate. Remission was characterized by a higher concentration of acetoacetate and a lower concentration of citrate, hippurate, taurine, succinate, glycine, alanine and formate. The obtained results reflect changes in cellular energy metabolism that occur in patients with

Table 5 Significantly changed urine metabolites

Metabolite	Percentage difference					Relative Standard Deviation (%)							
	CD vs UC	CD active <i>vs</i> UC active	IBD active vs control	IBD remission vs control	IBD active vs IBD remission	CD	CD active	uc	UC active	IBD active	IBD remission	Control	
Citrate	6.9	-4.0	-32.7°	-47.6 ^{a,c}	28.3	73.2	72.8	91.4	111.7	90.4	45.1	57.8	
Hippurate	7.2	29.7	-53.3 ^{a,c}	-35.1°	-28.1	66.6	61.3	123.5	70.7	64.9	122.9	78.4	
Glycine	23.3	11.4	12.6	-29.8ª	60.3 ^{a,c}	43.6	40.1	47.9	45.9	42.0	36.3	50.3	
Alanine	-0.2	-12.2	-9.1	-30.3 ^{a,c}	30.3	36.8	36.0	47.7	49.4	42.8	34.9	51.7	
Formate	2.9	1.5	-17.0	-30.6 ^{a,c}	19.5	47.1	42.1	37.4	37.2	39.4	45.3	38.8	
Trigonelline	-10.2	139.6	-73.5 ^{a,c}	-51.7	-45.1	148.3	144.8	137.4	64.3	150.8	121.9	133.7	
Acetoacetate	-6.2	82.2	7.8	85.1°	-41.7°	125.9	133.8	122.5	58.0	129.3	108.4	92.3	
Taurine	7.0	21.7	-28.1 ^{a,c}	-31.8 ^{a,c}	5.5	66.5	66.1	35.1	35.0	57.8	38.3	41.4	
Succinate	2.3	-11.4	-63.7 ^{a,c}	-51.8 ^{a,c}	-24.7	128.9	120.5	93.6	95.4	107.4	111.1	74.1	
2-hydroxyisobutyrate	17.3	30.2	-68.1°	-66.8	-3.9	41.5	44.9	38.4	31.4	42.9	37.2	256.5	
Unk2 with	0.8	-5.0	78.5 ^{a,c}	33.1	34.1	63.8	62.2	73.0	81.9	70.4	53.3	45.3	
4-hydroxyphenyl group δ 6.85 ppm													

Percentage difference was calculated based on the mean values of relative signal integrals in each group. The calculations were made from left to right, e.g., a positive value in IBD active vs Control means that a particular metabolite exhibits higher concentration in patients with active IBD compared to the Control group, $^aP < 0.05$ using the Student's t test, $^cP < 0.05$ using the Mann-Whitney-Wilcoxon test. CD: Crohn's disease; UC: Ulcerative colitis; IBD: Inflammatory bowel diseases; Unk2: Unknown metabolite.

IBD. The increase in serum concentration of ketone bodies (like acetoacetate and 3-hydroxybutyric acid) used as an alternative source of energy when glucose level is insufficient^[27,53], as well as the decrease in the concentration of Krebs cycle intermediates (such as citrate and succinate) and creatine (that serves as an energy reservoir [54]) indicate a high demand of the organism for energy and rapid utilization of metabolites [9,24]. Similarly, the increased concentration of lactate that is probably secreted from inflamed colonic mucosa cells^[36] indicates the up-regulation of glycolysis, which occurs in case of higher cellular energy need and/or lack of energy sources under inflammatory conditions^[42,55]. The decreased level of membrane metabolites like choline and its derivatives (PC and GPC) that was demonstrated in the study is most probably caused by the increased use of choline in IBD patients [26] associated with a change in mucosal membrane integrity^[3,56], extensive apoptosis and turnover of intestinal cell membranes under inflammatory conditions^[26]. Additionally, choline is indirectly involved in the synthesis of the platelet-activating factor, which is a major inflammatory mediator^[54,57]. Finally, a lower urine concentration of gut microbial-host cometabolites like hippurate (its level positively correlates with the presence of Clostridia spp. [25,58]) and formate (formed in the colonic lumen by intestinal bacteria such as Enterobacteriaceae, in the process of unabsorbed carbohydrates fermentation [3,24,59]) may be associated with changes in gut microbiome that occur in IBD patients^[60].

Our observations concerning the discrimination of IBD patients from healthy controls correlate well with most of previously published results^[9,21,24-27], although, as it was mentioned before, the comparison between IBD patients in the remission phase of the disease and healthy control subjects was never made before. In the serum of IBD patients as most prominently increased there were reported: *N*-acetylated glycoprotein, lactate^[26], methanol,

mannose, formate, 3-methyl-2-oxovalerate, 2-hydroxybutyrate, lysine, creatine^[24], 3-hydroxybutyrate, phenylalanine, α and β -glucose^[27] while those decreased included: lipids [26,27], urea, \(\tau\)-methylhistidine, valine, tyrosine, citrate, choline, creatinine, dimethyl sulfone^[24], histidine, tryptophan^[21] and alanine^[26]. Concomitantly, as most prominently increased metabolites in urine were characterized: mannitol, allantoin, xylose, carnitine, lactate, threonine[24] and formate^[25], whereas decreased hippurate^[9,24,25], succinate, citrate, 1-methylhistidine, methanol, formate, trigonelline, urea, taurine [9,24], asparagine, lysine, histidine, acetate, methylamine, creatine, trans-aconitate^[9], betaine^[24] and 4-cresol sulfate^[25]. It should be noted that an increase in the concentration of leucine and acetoacetate in serum reported in our study was earlier described only in animal models [35,36] while a decrease in the concentration of 2-hydroxyisobutyrate in urine has never been demonstrated before. The elevated concentration of isoleucine in serum (conflicting results in the literature [24,26]) as well as a decrease in the concentration of creatine (once reported as elevated^[24]) should be also pointed out. There is only one study (applies only to the urine of UC patients) in which the obtained results did not allow to distinguish IBD patients from healthy individuals[1].

The utility of metabolomics in the differential diagnosis of IBD remains controversial. Our results indicate that no chemometric model for distinction between CD and UC could be constructed both for serum and urine samples, although there are statistically significant differences between some metabolites in these diseases. Comparing the combined group of CD patients with active disease and remission to UC patients in the active and remission phase of the disease we found a significantly increased serum concentration of alanine and N-acetylated compounds. Comparing the patients with active CD and active UC we observed a significantly lower serum level of formate in the first group (not reported before). There were no

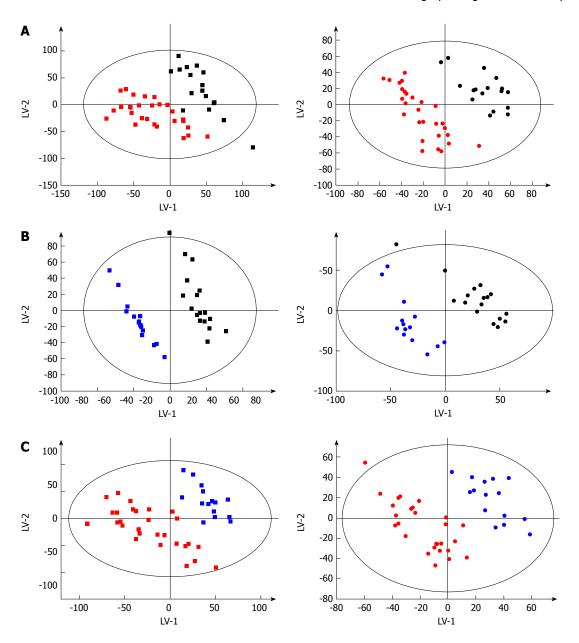
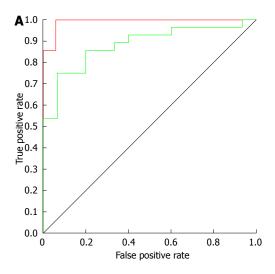


Figure 3 The partial least-squares-discriminant analysis score plots based on the metabolic fingerprints. A: IBD active vs control; B: IBD remission vs control; C: IBD active vs IBD remission. Boxes: Serum samples; Circles: Urine samples; Red: IBD active; Blue: IBD remission; Black: Control; IBD: Inflammatory bowel diseases.

statistically relevant differences in urine metabolic fingerprints between the cohorts of CD and UC patients. Our findings are substantially consistent with the results of Schicho et al²⁴ and differ from the results of Williams et al^{25,26]}. The first study revealed poorly pronounced differences between serum of the CD and UC cohorts, that allowed to create only a weak predictive chemometric model (AUROC 0.655 with Q2Y 0.024 for serum, and AUROC 0.7325 with Q²Y 0.115 for plasma), and no possibilities to create such model based on urine metabolomic analysis. Regarding to blood analysis, increased concentrations of arginine, 3-methyl-2-oxovalerate, τ-methylhistidine, lysine, leucine, glycerol in CD and elevated glutamine, citrate, tyrosine, glucose, mannose, carnitine, 2-hydroxybutyrate, betaine, creatynine, isoleucine in UC patients were reported^[24]. The results of Williams et al²⁶ revealed that on the basis of OSC-PLS-DA serum analysis it is possible to distinguish CD and UC cohorts with a significant predictive accuracy. The metabolites of importance included up-regulated N-acetylglycoprotein in CD patients, and down-regulated LDL and HDL cholesterol, unsaturated lipids, choline, isoleucine and alanine. Similarly, it has been shown that the differential diagnosis of IBD is possible on the basis of the urinary metabolic profile and the statistically significant differences concerned: hippurate, 4-cresol sulfate and citrate (decreased in CD) as well as formate (increased in CD)^[25]. Two other research groups estimated the usefulness of metabolomics in the differential diagnosis of IBD, but in these cases only one biofluid was evaluated. Stephens et al⁹ observed that CD could be differentiated from UC using the urine metabolomic profile but only when IBD



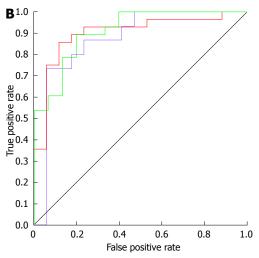


Figure 4 Receiver operating characteristic curves based on partial least-squares-discriminant analysis models for nuclear magnetic resonance spectra. A: Serum samples; B: Urine samples. Serum: Red - IBD active vs control (AUC = 0.992); Green - IBD active vs IBD remission (AUC = 0.883). Urine: Red - IBD active vs control (AUC = 0.900); Green - IBD active vs IBD remission (AUC = 0.900); Violet - IBD remission vs Control (AUC = 0.871). IBD: Inflammatory bowel diseases; AUC: Area under curves.

patients with intestinal resection were included in the cohorts (no differences were seen if the urine sample data from individuals with intestinal resection were removed from the analysis). Hisamatsu *et al*^[21] using MIAI technology, demonstrated a potential for discrimination between CD and UC patients with AUROC = 0.879 in the active phase of the disease and 0.744 in remission. The authors underlined the fact that although the alterations in amino acid metabolism in IBD patients could be found, it is difficult to establish a single biomarker for a clinical use.

The results of our study demonstrate that ¹H NMR-based metabolic fingerprinting of human serum and urine combined with multivariate data analysis could be a useful tool in distinguishing patients with active IBD from those in remission that is of great importance in IBD monitoring. At the same time, the results indicate that this diagnostic method has rather a weak potential in the differential diagnosis of IBD.

COMMENTS

Background

Difficulties in the differential diagnosis between ulcerative colitis and the Crohn's disease, as well as the limited number of monitoring tools in inflammatory bowel diseases (IBD) force searching for new approaches in IBD diagnostics.

Research frontiers

The use of metabolomics in medical diagnostics has recently become a very promising idea that is extensively studied in oncological conditions, diabetes, cardiovascular diseases, rheumatoid arthritis or multiple sclerosis. Unlike the genomic and proteomic studies that examine genes and more or less complex proteins, metabolomic analysis allows to assess the simplest low molecular weight metabolites that are involved in disease processes. The highly informative metabolic profiles as a potential source of biomarkers can be used for both diagnostic and prognostic purposes. So far the usefulness of metabolomics in IBD diagnostics has been evaluated only in a few studies.

Innovations and breakthroughs

In the present study the utility of serum and urine metabolomic analysis in diagnosing and monitoring of IBD has been valuated using proton nuclear magnetic resonance (¹H-NMR) spectroscopy. The significant differences in metabolomic

profiles were found between the group of patients with active IBD and the group with IBD in remission providing a good partial least-squares-discriminant analysis (PLS-DA) model.

Applications

NMR-based metabolomic fingerprinting of serum and urine has the potential to be a useful tool for differentiation between the active and remission phase of the disease that is of great importance in IBD monitoring.

Terminology

Metabolomics is the systematic study allowing the simultaneous and relative quantification of low molecular weight metabolites which form unique chemical fingerprints associated with the specific condition of the living system. PLS-DA was used to assess the differences in metabolite fingerprints. ¹H-NMR denotes proton nuclear magnetic resonance spectroscopy which allows to quantify the concentration and determine the structure of chemical compounds.

Peer review

The study is well conducted from the methodological point of view and presents interesting data concerning serum and urine metabolomic fingerprinting in the diagnostics of IBD. A new finding that metabolomics may be also helpful in differentiating between the active and remission phase of the disease is of particular importance in IBD monitoring.

REFERENCES

- Bjerrum JT, Nielsen OH, Hao F, Tang H, Nicholson JK, Wang Y, Olsen J. Metabonomics in ulcerative colitis: diagnostics, biomarker identification, and insight into the pathophysiology. J Proteome Res 2010; 9: 954-962 [PMID: 19860486 DOI: 10.1021/pr9008223]
- Strober W, Fuss I, Mannon P. The fundamental basis of inflammatory bowel disease. J Clin Invest 2007; 117: 514-521 [PMID: 17332878]
- Balasubramanian K, Kumar S, Singh RR, Sharma U, Ahuja V, Makharia GK, Jagannathan NR. Metabolism of the colonic mucosa in patients with inflammatory bowel diseases: an in vitro proton magnetic resonance spectroscopy study. Magn Reson Imaging 2009; 27: 79-86 [PMID: 18599242 DOI: 10.1016/j.mri.2008.05.014]
- **Bezabeh T**, Somorjai RL, Smith IC, Nikulin AE, Dolenko B, Bernstein CN. The use of 1H magnetic resonance spectroscopy in inflammatory bowel diseases: distinguishing ulcerative colitis from Crohn's disease. *Am J Gastroenterol* 2001; **96**: 442-448 [PMID: 11232688]
- 5 Mandal A, Mayberry J. Magnetic resonance spectroscopy: a



- new test for differentiating ulcerative colitis from Crohn's disease? *Am J Gastroenterol* 2001; **96**: 271-273 [PMID: 11232662]
- 6 Bach SP, Mortensen NJ. Ileal pouch surgery for ulcerative colitis. World J Gastroenterol 2007; 13: 3288-3300 [PMID: 17659667]
- 7 Papp M, Norman GL, Altorjay I, Lakatos PL. Utility of serological markers in inflammatory bowel diseases: gadget or magic? World J Gastroenterol 2007; 13: 2028-2036 [PMID: 17465443]
- 8 Lin HM, Helsby NA, Rowan DD, Ferguson LR. Using metabolomic analysis to understand inflammatory bowel diseases. *Inflamm Bowel Dis* 2011; 17: 1021-1029 [PMID: 20629098 DOI: 10.1002/ibd.21426]
- 9 Stephens NS, Siffledeen J, Su X, Murdoch TB, Fedorak RN, Slupsky CM. Urinary NMR metabolomic profiles discriminate inflammatory bowel disease from healthy. *J Crohns Colitis* 2013; 7: e42-e48 [PMID: 22626506 DOI: 10.1016/ j.crohns.2012.04.019]
- 10 Wang H, Tso VK, Slupsky CM, Fedorak RN. Metabolomics and detection of colorectal cancer in humans: a systematic review. Future Oncol 2010; 6: 1395-1406 [PMID: 20919825 DOI: 10.2217/fon.10.107]
- 11 DeFeo EM, Wu CL, McDougal WS, Cheng LL. A decade in prostate cancer: from NMR to metabolomics. *Nat Rev Urol* 2011; 8: 301-311 [PMID: 21587223 DOI: 10.1038/nru-rol.2011.53]
- 12 Bu Q, Huang Y, Yan G, Cen X, Zhao YL. Metabolomics: a revolution for novel cancer marker identification. *Comb Chem High Throughput Screen* 2012; 15: 266-275 [PMID: 22221059]
- Hori S, Nishiumi S, Kobayashi K, Shinohara M, Hatakeyama Y, Kotani Y, Hatano N, Maniwa Y, Nishio W, Bamba T, Fukusaki E, Azuma T, Takenawa T, Nishimura Y, Yoshida M. A metabolomic approach to lung cancer. *Lung Cancer* 2011; 74: 284-292 [PMID: 21411176 DOI: 10.1016/j.lungcan.2011.02.008]
- 14 Lu J, Xie G, Jia W, Jia W. Metabolomics in human type 2 diabetes research. Front Med 2013; 7: 4-13 [PMID: 23377891 DOI: 10.1007/s11684-013-0248-4]
- 15 Rasmiena AA, Ng TW, Meikle PJ. Metabolomics and ischaemic heart disease. *Clin Sci* (Lond) 2013; **124**: 289-306 [PMID: 23157406 DOI: 10.1042/CS20120268]
- Noga MJ, Dane A, Shi S, Attali A, van Aken H, Suidgeest E, Tuinstra T, Muilwijk B, Coulier L, Luider T, Reijmers TH, Vreeken RJ, Hankemeier T. Metabolomics of cerebrospinal fluid reveals changes in the central nervous system metabolism in a rat model of multiple sclerosis. *Metabolomics* 2012; 8: 253-263 [PMID: 22448154]
- Hügle T, Kovacs H, Heijnen IA, Daikeler T, Baisch U, Hicks JM, Valderrabano V. Synovial fluid metabolomics in different forms of arthritis assessed by nuclear magnetic resonance spectroscopy. Clin Exp Rheumatol 2012; 30: 240-245 [PMID: 22410098]
- 18 Bothwell JH, Griffin JL. An introduction to biological nuclear magnetic resonance spectroscopy. *Biol Rev Camb Philos Soc* 2011; 86: 493-510 [PMID: 20969720 DOI: 10.1111/j.1469-185X.2010.00157.x]
- Fiehn O. Metabolomics--the link between genotypes and phenotypes. *Plant Mol Biol* 2002; 48: 155-171 [PMID: 11860207]
- 20 Nicholson JK, Lindon JC, Holmes E. 'Metabonomics': understanding the metabolic responses of living systems to pathophysiological stimuli via multivariate statistical analysis of biological NMR spectroscopic data. *Xenobiotica* 1999; 29: 1181-1189 [PMID: 10598751]
- 21 **Hisamatsu** T, Okamoto S, Hashimoto M, Muramatsu T, Andou A, Uo M, Kitazume MT, Matsuoka K, Yajima T, Inoue N, Kanai T, Ogata H, Iwao Y, Yamakado M, Sakai R, Ono N, Ando T, Suzuki M, Hibi T. Novel, objective, multivariate biomarkers composed of plasma amino acid profiles for the diagnosis and assessment of inflammatory bowel disease.

- PLoS One 2012; 7: e31131 [PMID: 22303484 DOI: 10.1371/journal.pone.0031131]
- 22 Marchesi JR, Holmes E, Khan F, Kochhar S, Scanlan P, Shanahan F, Wilson ID, Wang Y. Rapid and noninvasive metabonomic characterization of inflammatory bowel disease. J Proteome Res 2007; 6: 546-551 [PMID: 17269711]
- Bi X, Walsh A, Mahadevan-Jansen A, Herline A. Development of spectral markers for the discrimination of ulcerative colitis and Crohn's disease using Raman spectroscopy. *Dis Colon Rectum* 2011; 54: 48-53 [PMID: 21160313 DOI: 10.1007/DCR.0b013e3181fcf68d]
- Schicho R, Shaykhutdinov R, Ngo J, Nazyrova A, Schneider C, Panaccione R, Kaplan GG, Vogel HJ, Storr M. Quantitative Metabolomic Profiling of Serum, Plasma, and Urine by (1)H NMR Spectroscopy Discriminates between Patients with Inflammatory Bowel Disease and Healthy Individuals. J Proteome Res 2012 May 17; Epub ahead of print [PMID: 22574726]
- Williams HR, Cox IJ, Walker DG, North BV, Patel VM, Marshall SE, Jewell DP, Ghosh S, Thomas HJ, Teare JP, Jakobovits S, Zeki S, Welsh KI, Taylor-Robinson SD, Orchard TR. Characterization of inflammatory bowel disease with urinary metabolic profiling. *Am J Gastroenterol* 2009; 104: 1435-1444 [PMID: 19491857 DOI: 10.1038/ajg.2009.175]
- 26 Williams HR, Willsmore JD, Cox IJ, Walker DG, Cobbold JF, Taylor-Robinson SD, Orchard TR. Serum metabolic profiling in inflammatory bowel disease. *Dig Dis Sci* 2012; 57: 2157-2165 [PMID: 22488632 DOI: 10.1007/s10620-012-2127-2]
- 27 Zhang Y, Lin L, Xu Y, Lin Y, Jin Y, Zheng C. 1H NMR-based spectroscopy detects metabolic alterations in serum of patients with early-stage ulcerative colitis. *Biochem Biophys Res Commun* 2013; 433: 547-551 [PMID: 23510994 DOI: 10.1016/j.bbrc.2013.03.012]
- Walmsley RS, Ayres RC, Pounder RE, Allan RN. A simple clinical colitis activity index. *Gut* 1998; 43: 29-32 [PMID: 9771402]
- 29 Harvey RF, Bradshaw JM. A simple index of Crohn's-disease activity. *Lancet* 1980; 1: 514 [PMID: 6102236 DOI: 10.1016/ S0140-6736(80)92767-1]
- 30 Beckonert O, Keun HC, Ebbels TM, Bundy J, Holmes E, Lindon JC, Nicholson JK. Metabolic profiling, metabolomic and metabonomic procedures for NMR spectroscopy of urine, plasma, serum and tissue extracts. *Nat Protoc* 2007; 2: 2692-2703 [PMID: 18007604]
- 31 **Jiang L**, Huang J, Wang Y, Tang H. Eliminating the dication-induced intersample chemical-shift variations for NMR-based biofluid metabonomic analysis. *Analyst* 2012; **137**: 4209-4219 [PMID: 22858594 DOI: 10.1039/c2an35392j]
- 32 Tomasi G, van den Berg F, Andersson C. Correlation optimized warping and dynamic time warping as preprocessing methods for chromatographic data. *J Chemometr* 2004; 18: 231-241 [DOI: 10.1002/cem.859]
- 33 Savorani F, Tomasi G, Engelsen SB. icoshift: A versatile tool for the rapid alignment of 1D NMR spectra. *J Magn Reson* 2010; 202: 190-202 [PMID: 20004603 DOI: 10.1016/j.jmr.2009.11.012]
- 34 Dieterle F, Ross A, Schlotterbeck G, Senn H. Probabilistic quotient normalization as robust method to account for dilution of complex biological mixtures. Application in 1H NMR metabonomics. Anal Chem 2006; 78: 4281-4290 [PMID: 16808434]
- Shiomi Y, Nishiumi S, Ooi M, Hatano N, Shinohara M, Yoshie T, Kondo Y, Furumatsu K, Shiomi H, Kutsumi H, Azuma T, Yoshida M. GCMS-based metabolomic study in mice with colitis induced by dextran sulfate sodium. *Inflamm Bowel Dis* 2011; 17: 2261-2274 [PMID: 21287666 DOI: 10.1002/ibd.21616]
- 36 Schicho R, Nazyrova A, Shaykhutdinov R, Duggan G, Vogel HJ, Storr M. Quantitative metabolomic profiling of serum and urine in DSS-induced ulcerative colitis of mice by



- (1)H NMR spectroscopy. *J Proteome Res* 2010; **9**: 6265-6273 [PMID: 20886908 DOI: 10.1021/pr100547y]
- 37 Dong F, Zhang L, Hao F, Tang H, Wang Y. Systemic responses of mice to dextran sulfate sodium-induced acute ulcerative colitis using 1H NMR spectroscopy. *J Proteome Res* 2013; 12: 2958-2966 [PMID: 23651354 DOI: 10.1021/pr4002383]
- 38 Lin HM, Edmunds SI, Helsby NA, Ferguson LR, Rowan DD. Nontargeted urinary metabolite profiling of a mouse model of Crohn's disease. *J Proteome Res* 2009; 8: 2045-2057 [PMID: 19275240 DOI: 10.1021/pr800999t]
- 39 Murdoch TB, Fu H, MacFarlane S, Sydora BC, Fedorak RN, Slupsky CM. Urinary metabolic profiles of inflammatory bowel disease in interleukin-10 gene-deficient mice. *Anal Chem* 2008; 80: 5524-5531 [PMID: 18558774 DOI: 10.1021/ac8005236]
- 40 Chen C, Shah YM, Morimura K, Krausz KW, Miyazaki M, Richardson TA, Morgan ET, Ntambi JM, Idle JR, Gonzalez FJ. Metabolomics reveals that hepatic stearoyl-CoA desaturase 1 downregulation exacerbates inflammation and acute colitis. *Cell Metab* 2008; 7: 135-147 [PMID: 18249173 DOI: 10.1016/j.cmet.2007.12.003]
- 41 **Lu K**, Knutson CG, Wishnok JS, Fox JG, Tannenbaum SR. Serum metabolomics in a Helicobacter hepaticus mouse model of inflammatory bowel disease reveal important changes in the microbiome, serum peptides, and intermediary metabolism. *J Proteome Res* 2012; **11**: 4916-4926 [PMID: 22957933 DOI: 10.1021/pr300429x]
- 42 Martin FP, Rezzi S, Philippe D, Tornier L, Messlik A, Hölzlwimmer G, Baur P, Quintanilla-Fend L, Loh G, Blaut M, Blum S, Kochhar S, Haller D. Metabolic assessment of gradual development of moderate experimental colitis in IL-10 deficient mice. *J Proteome Res* 2009; 8: 2376-2387 [PMID: 19323467 DOI: 10.1021/pr801006e]
- 43 Chalimoniuk M. [Secretory phospholipase A2 and its role in oxidative stress and inflammation]. *Postepy Biochem* 2012; 58: 204-208 [PMID: 23214144]
- 44 Spoettl T, Hausmann M, Klebl F, Dirmeier A, Klump B, Hoffmann J, Herfarth H, Timmer A, Rogler G. Serum soluble TNF receptor I and II levels correlate with disease activity in IBD patients. *Inflamm Bowel Dis* 2007; 13: 727-732 [PMID: 17260368]
- 45 Friedman G, Barak V, Chajek-Shaul T, Etienne J, Treves AJ, Stein O, Stein Y. Recombinant human interleukin-1 suppresses lipoprotein lipase activity, but not expression of lipoprotein lipase mRNA in mesenchymal rat heart cell cultures. *Biochim Biophys Acta* 1991; 1089: 83-87 [PMID: 2025651]
- 46 Ripollés Piquer B, Nazih H, Bourreille A, Segain JP, Huvelin JM, Galmiche JP, Bard JM. Altered lipid, apolipoprotein, and lipoprotein profiles in inflammatory bowel disease: consequences on the cholesterol efflux capacity of serum us-

- ing Fu5AH cell system. *Metabolism* 2006; **55**: 980-988 [PMID: 16784973]
- 47 Levy E, Rizwan Y, Thibault L, Lepage G, Brunet S, Bouthillier L, Seidman E. Altered lipid profile, lipoprotein composition, and oxidant and antioxidant status in pediatric Crohn disease. Am J Clin Nutr 2000; 71: 807-815 [PMID: 10702177]
- 48 Nicholson JK, Foxall PJ, Spraul M, Farrant RD, Lindon JC. 750 MHz 1H and 1H-13C NMR spectroscopy of human blood plasma. *Anal Chem* 1995; 67: 793-811 [PMID: 7762816]
- 49 Gabay C, Kushner I. Acute-phase proteins and other systemic responses to inflammation. N Engl J Med 1999; 340: 448-454 [PMID: 9971870]
- Scholl-Bürgi S, Schroecksnadel S, Jenny M, Karall D, Fuchs D. Chronic Immune Stimulation May Cause Moderate Impairment of Phenylalanine 4-hydroxylase. *Pteridines* 2013;
 22: 120-125 [DOI: 10.1515/pteridines.2011.22.1.120]
- 51 Tsune I, Ikejima K, Hirose M, Yoshikawa M, Enomoto N, Takei Y, Sato N. Dietary glycine prevents chemical-induced experimental colitis in the rat. *Gastroenterology* 2003; 125: 775-785 [PMID: 12949723]
- McCole DF. The epithelial glycine transporter GLYT1: protecting the gut from inflammation. *J Physiol* 2010; 588: 1033-1034 [PMID: 20360026 DOI: 10.1113/jphysiol.2010.188516]
- 53 Stone WJ, Balagura S, Pitts RF. Diffusion equilibrium for ammonia in the kidney of the acidotic dog. *J Clin Invest* 1967;
 46: 1603-1608 [PMID: 6061738 DOI: 10.1172/JCI105650]
- Varma S, Eskin MN, Bird R, Dolenko B, Raju J, Ijare OB, Bezabeh T. Potential of magnetic resonance spectroscopy in assessing the effect of fatty acids on inflammatory bowel disease in an animal model. *Lipids* 2010; 45: 843-854 [PMID: 20721632 DOI: 10.1007/s11745-010-3455-7]
- 55 Seeger K. Metabolic changes in autoimmune diseases. Curr Drug Discov Technol 2009; 6: 256-261 [PMID: 20025593]
- Michel V, Yuan Z, Ramsubir S, Bakovic M. Choline transport for phospholipid synthesis. *Exp Biol Med* (Maywood) 2006; 231: 490-504 [PMID: 16636297]
- 57 Zeisel SH, Blusztajn JK. Choline and human nutrition. Annu Rev Nutr 1994; 14: 269-296 [PMID: 7946521]
- 58 Li M, Wang B, Zhang M, Rantalainen M, Wang S, Zhou H, Zhang Y, Shen J, Pang X, Zhang M, Wei H, Chen Y, Lu H, Zuo J, Su M, Qiu Y, Jia W, Xiao C, Smith LM, Yang S, Holmes E, Tang H, Zhao G, Nicholson JK, Li L, Zhao L. Symbiotic gut microbes modulate human metabolic phenotypes. *Proc Natl Acad Sci USA* 2008; 105: 2117-2122 [PMID: 18252821 DOI: 10.1073/pnas.0712038105]
- 59 Leonhartsberger S, Korsa I, Böck A. The molecular biology of formate metabolism in enterobacteria. J Mol Microbiol Biotechnol 2002; 4: 269-276 [PMID: 11931558]
- 60 Sartor RB. Microbial influences in inflammatory bowel diseases. Gastroenterology 2008; 134: 577-594 [PMID: 18242222 DOI: 10.1053/j.gastro.2007.11.059]

P- Reviewers: Cheon JH, Giraudeau P, Kannen V S- Editor: Ma YJ L- Editor: A E- Editor: Ma S







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ISSN 1007 - 9327

