Proton nuclear magnetic resonance-based metabonomic models for non-invasive diagnosis of liver fibrosis in chronic hepatitis C: Optimizing the classification of intermediate fibrosis

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Abstract

AIM

To develop metabonomic models (MMs), using $^1$H nuclear magnetic resonance (NMR) spectra of serum, to predict significant liver fibrosis (SF: METAVIR ≥ F2), advanced liver fibrosis (AF: METAVIR ≥ F3) and cirrhosis (C: METAVIR = F4 or clinical cirrhosis) in chronic hepatitis C (CHC) patients. Additionally, to compare the accuracy of the MMs with the aspartate aminotransferase to platelet ratio index (APRI) and fibrosis index based on four factors (FIB-4).

METHODS

Sixty-nine patients who had undergone biopsy in the previous 12 mo or had clinical cirrhosis were included. The presence of any other liver disease was a criterion for exclusion. The MMs, constructed using partial least squares discriminant analysis and linear discriminant analysis formalisms, were tested by cross-validation, considering SF, AF and C.

RESULTS

Results showed that forty-two patients (61%) presented SF, 28 (40%) AF and 18 (26%) C. The MMs showed sensitivity and specificity of 97.6% and 92.6% to predict SF; 96.4% and 95.1% to predict AF; and 100% and 98.0% to predict C. Besides that, the MMs correctly classified all 27 (39.7%) and 25 (38.8%) patients with intermediate values of APRI and FIB-4, respectively.

CONCLUSION

The metabonomic strategy performed excellently in predicting significant and advanced liver fibrosis in CHC patients, including those in the gray zone of APRI and FIB-4, which may contribute to reducing the need for these patients to undergo liver biopsy.

Key words: Metabolomics; Nuclear magnetic resonance spectroscopy; Chronic hepatitis C; Liver fibrosis; Surrogate markers

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Core tip: The assessment of liver fibrosis in chronic hepatitis C patients is important to make therapeutic decisions and predict clinical outcomes. Due to various drawbacks related to the use of liver biopsy, individual markers and scores have been validated with feeble accuracy to assess intermediate stages of fibrosis. Our study showed promising results for the metabolomics strategy as a non-invasive tool to distinguish patients with significant fibrosis, advanced fibrosis, and cirrhosis, with sensitivity and specificity values above 95% and high accuracy in the gray zone of aspartate aminotransferase to platelet ratio index and fibrosis index based on four factors, which could avoid a large number of biopsies in these patients.

INTRODUCTION

Approximately 130-170 million people worldwide are chronically infected with the hepatitis C virus (HCV) and an estimated 500000 individuals died in 2010 from virus-related illnesses. In Brazil, the estimated prevalence of hepatitis C is 1450000 cases and about 8000 deaths were due to HCV-related diseases in 2013.

The accurate diagnosis of significant fibrosis (SF), advanced fibrosis (AF) and cirrhosis (C) in the liver is important to determine the urgency of treatment, and to monitor complications of the disease. Hepatic histopathology assessment by liver biopsy is still considered the gold standard for evaluating liver fibrosis, but the procedure is invasive and may lead to major complications and death. Sampling error, intra- and inter-observer variability leads to discordant results in 33% of cases, thus making biopsy an imperfect reference standard. For these reasons, several non-invasive serological biomarkers of fibrosis have been evaluated. Among them, aminotransferase to platelet ratio index (APRI) and fibrosis index based on four factors (FIB-4) stand out because they are based on readily available laboratory tests in clinical practice. These surrogate markers have good accuracy as to excluding significant fibrosis and confirming cirrhosis. However they fail to diagnose intermediate fibrosis.

Metabonomics is defined as “a quantitative measure of the dynamic and multiparametric metabolic response of living organisms to pathological or genetic modifications. As a result of homeostasis, the presence of a pathological condition changes the profile of endogenous metabolites, which can be monitored by $^1$H nuclear magnetic resonance (NMR) spectroscopy. This method seeks to discriminate the samples in
groups according to their biochemical status, thereby associating this status to a given condition\(^1\)\(^{-12}\). The method has been studied in various liver diseases\(^1\)\(^{-13}\), and has been shown to be useful in distinguishing between patients with viral hepatitis and healthy volunteers\(^14\)\(^{-15}\) and performing well at identifying complications of liver cirrhosis\(^16\)\(^{-18}\). Recently, our group showed that a partial least squares discriminant analysis (PLS-DA) metabolomic model (MM), based on the H-1 NMR spectroscopy of serum samples, presented a clear separation between 18 patients coinfected with schistosomiasis mansoni and hepatitis B virus (HBV) or HCV and 22 HBV or HCV mono-infected patients, with an accuracy, a predictive ability (Q²) and a coefficient of determination (R²) of 100%, 98.1% and 97.5%, respectively\(^1\)\(^{-19}\). Therefore, the aim of this study was to develop and evaluate MMs, using \(^1\)H NMR spectrum of serum samples, as non-invasive markers of significant liver fibrosis, advanced liver fibrosis and cirrhosis in patients with chronic hepatitis C (CHC), and to compare their performance with the APRI and FIB-4.

**MATERIALS AND METHODS**

**Design of the study and patient selection**

This was a cross-sectional phase II validation diagnostic study\(^2\)\(^{0}\), with prospective inclusion, by spontaneous demand, of CHC adult outpatients (anti-HCV and HCV-RNA detectable in serum) attended to at the hepatology clinic of the Hospital das Clínicas/Universidade Federal de Pernambuco (HC/UFPE) between October/2012 and December/2015. Patients who had undergone a percutaneous liver biopsy in the previous 12 mo or had been clinically diagnosed with liver cirrhosis were included.

The clinical diagnosis of cirrhosis was based on characteristic symptoms and signals and/or according to evidence of chronic liver disease and/or portal hypertension on ultrasound (US), such as liver parenchymal heterogeneity, straight borders, reduced liver size, enhanced portal vein dimensions, presence of collateral vessels, splenomegaly, and/or signals of portal hypertension observed on upper gastrointestinal endoscopy, such as the presence of esophageal/gastric varices and/or hypertensive gastropathy.

Those undergoing antiviral treatment or diagnosed with periportal fibrosis induced by schistosomiasis, metabolic, autoimmune or cholestatic liver disease, HBV or HIV co-infection, neoplasia, or with ethanol ingestion > 20 g/d for women and > 30 g/d for men were excluded. All patients signed an informed consent form and the study was approved by the Ethics Committee of the Institution.

**Laboratory analysis, determination of APRI and FIB-4**

Fasting blood samples were collected from all patients by peripheral vein puncture. The laboratory tests were performed by an automated method. The HCV RNA was detected and the genotype was determined by real-time polymerase chain reaction, using COBAS\(^\circledR\) AmpliPrep/COBAS\(^\circledR\) TaqMan\(^\circledR\) (version 2, Roche, Pleasanton, CA, United States) with a detection limit of 15 IU/mL. The APRI and FIB-4 were calculated as described by Wai et al\(^2\)\(^{1}\) (2003) and Sterling et al\(^2\)\(^{0}\) (2006).

**\(^1\)H nuclear magnetic resonance spectroscopy**

Serum samples, stored at -20 °C, were thawed at room temperature and prepared by adding 200 μL of D2O to 400 μL of serum. \(^1\)H NMR spectra were obtained using a Varian Unity Plus 300 spectrometer, operating at 299.95 MHz, at 300 K. The samples were analyzed using a pulse sequence with suppression of the resonance of water and T2 filter (PRESAT-CPMG), as follows: Pre-saturation time of 2.0 s, acquisition time of 1.704 s, 128 repetitions and spectral width of 4.8 kHz. Spectra were processed using line broadening equal to 0.3 Hz. The signal attributed to the methyl group of the lactate, in δ 1.33 ppm, was used as the internal reference of chemical shift. The baseline of the spectra was corrected manually.

**Liver biopsies and allocation of patients**

The ultrasonography-guided percutaneous liver biopsies were performed using a 16 G x 90 mm Menghini needle in, at most, 2 punctures. Fragments with at least 15 mm and/or 6 complete portal tracts were included in the analysis. Fibrosis was classified as F0 to F4, in accordance with METAVIR\(^2\)\(^{1}\), by two experienced pathologists, blinded to the clinical and serological results. The patients were allocated into three groups: SF (METAVIR F ≥ 2), AF (METAVIR F ≥ 3) and C (METAVIR F = 4 or clinical cirrhosis).

**Multivariate statistical analysis of spectral data and MM development**

All spectra were processed using MestreNova software (version 9.0.1, MestreLab Research). The spectra were divided into 250 regions of 0.04 ppm, called bins, used to construct a dataset. The region containing the bins centered between δ 4.52 - 5.12 ppm was excluded so as to eliminate the residual signal of water. The spectra were normalized using the following expression:

\[
x = \frac{x_i - \bar{x}}{\sigma}
\]

Where \(x_i\) is the intensity in each bin, while \(\bar{x} = (x_1 + x_2 + ... + x_i)/n\) is the arithmetic mean of the intensities observed in the bins and \(\sigma\) is the standard deviation.

Principal components analysis (PCA) formalism was initially applied to the dataset, to explore inherent clusters and to identify the presence of outliers. PCA did not present natural grouping in the classes of interest. Then, PLS-DA and linear discriminant analysis (LDA) supervised formalisms were used. Three PLS-DA models were constructed to predict FS, FA and C, respectively, using the MetaboAnalyt 3.0 platform\(^2\)\(^{2}\). The models were validated by leave-one-out-cross-
validation (LOOCV) and by permutations tests. In addition, three LDA models were constructed, using the PCA matrix as input data, so as to predict SF, with five principal components (PCs); AF, with four PCs; and C, with five PCs. LDA models were validated by LOOCV, using STATISTICA software (version 10.0, Quest software).

**Analysis of the performance of the MMs and of the APRI and FIB-4**

For each LDA MM, and for APRI and FIB-4, a 2 x 2 contingency matrix was used to calculate sensitivity (SN), specificity (SP), the positive likelihood ratio (LR+), the negative likelihood ratio (LR-) and accuracy (A). A Receiver Operating Characteristic (ROC) curve was also constructed for APRI and FIB-4.

**Statistical analysis**

The descriptive and comparative analysis of the data was carried out using STATA (version 12.0, StataCorp, College Station, Texas) and GraphPad Prism software (version 5.0 for Windows, GraphPad Software, La Jolla, California). The qualitative variables were presented as absolute and relative frequencies, and the quantitative variables as means and standard deviation or medians and 25th and 75th percentile. Categorical variables were compared using the $\chi^2$ test, applying Fisher’s exact test, when necessary. The Mann-Whitney and Student’s $t$-test were used to compare non-parametric and parametric continuous measurements, respectively. All tests were applied with 95% confidence ($P$ value $\leq 0.05$).

**RESULTS**

The group studied consisted of 80 CHC patients initially selected. Eleven of the subjects (14%) were excluded due to: diagnosis of periportal fibrosis induced by schistosomiasis in 5 patients, hepatocellular carcinoma in 1, abuse of ethanol in 1 and inadequate liver fragment in 4. Therefore, 69 patients were evaluated, of whom 59.4% were female, with a mean age of 57 ± 12 years. The HCV genotype was determined in 67 patients, while 1b was the most frequent genotype in 36 (53.7%) patients, followed by genotype 3 in 16 (23.9%), genotype 1/1a in 13 (19.4%), genotype 2 in 1 (1.5%) and genotype 4 in 1 (1.5%) patients. The main characteristics of the casuistry are described in Table 1.

Liver biopsy was performed on 54 (78%) patients. There were no serious complications or deaths related to the procedure. The median fragment was 15 mm in length (P 75: 13; P 25: 18 mm), and there were 15 portal tracts (P 75: 12; P 25: 20 mm). The METAVIR fibrosis stage was distributed as follows: F0 in 2; F1 in 25; F2 in 14; F3 in 10 and F4 in 3 patients. The diagnosis of cirrhosis was clinically established in 15 (21.7%), thus classified according to the Child-Pugh score: 10 patients Child-Pugh A and 5 Child-Pugh B.

Therefore, 42 (60.9%) patients were classified as SF, 28 (40.6%) as AF and 18 (26.1%) as C. Patients with SF and AF presented a higher mean age, a higher mean value of INR and a higher median value of bilirubin, gamma-glutamyl transferase, APRI and FIB-4, as well as a lower mean platelet count and albumin serum level, when compared with the groups of a lower stage of fibrosis (Table 2).

The PLS-DA MM for SF showed a clear discrimination between the samples with three latent components (Figure 1A). The model presented 100% accuracy, $R^2$ and $Q^2$ of 0.98 and 0.91, respectively, when five latent components were used (Figure 1D). The permutation tests, using up to 1000 permutations of the model, indicated that there was no permuted model better than the original one, with observed statistics at $P < 0.001$ (Figure 1G). The results of the LDA MM for SF are presented in Table 3 and were compared with APRI, with a lower and upper cut-off point of 0.5 and 1.5. The MM shown SN of 97.6% (95%CI: 87.4%-99.9%), similar to APRI cut-off of 0.5, which presented SN of 85.7% (95%CI: 71.5%-94.6%). The LR- of the MM was 0.03 (95%CI: 0.004-0.2), whereas the LR+ value was 0.5 was 0.3 (95%CI: 0.1-0.7). The MM presented SP of 92.6% (95%CI: 75.7%-99.1%) and LR+ 13.2 (95%CI: 3.5-50.1), similar to APRI cut-off of 1.5, which showed SP of 92.3% (95%CI: 74.9%-99.9%), with LR+ of 5.9 (95%CI: 1.5-23.2) for values > 1.5.

**Table 1** Main demographic and laboratory characteristics of 69 chronic hepatitis C patients from Pernambuco/Brazil

| Characteristics | Total (n = 69) |
|-----------------|--------------|
| Gender (n, %)   |              |
| Male            | 28           |
| Female          | 41           |
| Age (yr)        | 57.5 ± 11.9  |
| BMI (kg/m²)     | 27.7 ± 4.7   |
| AST (U/L)       | 51.5 (33.7-88)|
| ALT (U/L)       | 54 (32.6-106)|
| AST/ALT         | 1.03 ± 0.44  |
| GGTT (U/L)      | 81.8 (46.5-155)|
| Platelets (10³/mm³) | 191 ± 78 |
| Albumin (g/dL)  | 4.08 ± 0.56  |
| Total bilirubin (mg/dL) | 0.7 (0.50-1.09) |
| Alkaline phosphatase (U/L) | 94.6 ± 49.2 |
| INR             | 1.09 ± 0.14  |
| APRI            | 0.8 (0.44-2.18)|
| FIB-4           | 2.18 (1.29-4.72)|
| Liver fragment length (cm) | 1.5 (1.30-1.80) |
| Number of portal tracts | 15 (12.0-20.0)|
| Stage of fibrosis (n, %) |              |
| F0              | 2 ± 2.90%    |
| F1              | 25 ± 36.20%  |
| F2              | 14 ± 20.30%  |
| F3              | 10 ± 14.50%  |
| F4              | 18 ± 26.10%  |

¹Mean ± standard deviation; ²Median (P 25-P 75). ALT: Alanine aminotransferase; AST: Aspartate aminotransferase; BMI: Body mass index; GGTT: Gamma glutamyl transferase; INR: International normalized ratio; APRI: Aspartate aminotransferase to platelet ratio index; FIB-4: Fibrosis index based on four factors.
The PLS-DA MM for AF, using three latent components, discriminated all the samples, with 100% accuracy, R² of 0.98 and Q² of 0.93, with five latent components. Permutation tests indicated that the permuted models are not better than the original model (Figure 1B, E and H). The LDA MM for AF showed SN of 96.4% (95%CI: 81.7%-99.1%) and LR+ of 0.04 (95%CI: 0.005-0.3), similar to FIB-4 cut-off of 1.45, which showed SN of 89.3% (95%CI: 71.8%-97.7%) and LR- of 0.3 (95%CI: 0.1-0.8) for values ≤ 1.5. The two methods also presented high SP and LR+, there being observed SP of 95.1% (95%CI: 83.5%-99.4%) and LR+ of 19.8 (95%CI: 5.1-76.5) for MM, and SP of 92.5% (95%CI: 79.6%-98.4%) and LR+ of 10 (95%CI: 3.3-30.3) for FIB-4 cut-off of 3.25 (Table 3).

The PLS-DA MM for C, using three latent components, also adequately discriminated the samples, attaining an accuracy of 84.0% with five latent components. The permutation tests indicated that the original model was not exceeded by any of the permuted models (Figure 1C, F and I). The LDA MM for C showed SN of 100% (95%CI: 81.5%-100%) and LR- of 0.03 (95%CI: 0.002-0.4), similar to APR1 cut-off of 1.0, which presented SN of 77.8% (95%CI: 52.4-93.6), with

Table 2  Main demographic and laboratory characteristics of 69 chronic hepatitis C patients, in accordance with the fibrosis group, from Pernambuco/Brazil

| Gender (M/F, %) | Significant (≥ F2) n = 42 | Non-significant (< F2) n = 27 | P value |
|-----------------|--------------------------|-------------------------------|--------|
| F0-2            |                          |                               |        |
| Gender (M/F, %) | 16/26 (40.6/59.4)        | 12/15 (38.1/61.9)             | 0.60*  |
| Age (yr)        | 62.16 ± 9.46              | 50.22 ± 11.86                 | < 0.0001* |
| BMI (kg/m²)     | 28.6 ± 4.9                | 26.3 ± 3.8                    | 0.051* |
| AST (U/L)       | 61.5 (45.5-120)           | 34 (29.6-67)                  | < 0.0001* |
| ALT (U/L)       | 57.5 (41.7-108)           | 39.1 (26.8-81)                | 0.04  |
| platelets x 10⁴ | 216 ± 72                  | 239 ± 61                      | < 0.0001* |
| Albumin (g/dL)  | 3.91 ± 0.62               | 4.35 ± 0.30                   | 0.001* |
| Total bilirubin | 0.72 (0.60-1.30)          | 0.5 (0.40-0.80)               | 0.005* |
| Alkaline phosphatase (U/L) | 103.3 ± 55.6 | 79.2 ± 30.3 | 0.059* |
| INR             | 1.13 ± 0.17               | 1.03 ± 0.06                   | 0.002* |
| APR1            | 1.14 (0.75-2.84)          | 0.48 (0.27-0.79)              | < 0.0001* |
| FIB-4           | 3.45 (2.04-6.19)          | 1.37 (0.84-1.89)              | < 0.0001* |
| Cirrhosis (F4)  | 7/11 (58.9/61.1)          | 21/30 (41.2/58.8)             | 0.865* |
| Gender (M/F, %) | 63.7 ± 11.11              | 55.29 ± 11.51                 | 0.009* |
| BMI (kg/m²)     | 27.6 ± 3.9                | 27.9 ± 5.2                    | 0.082* |
| AST (U/L)       | 78 (52.4-140)             | 40 (31.4-63)                  | 0.001* |
| ALT (U/L)       | 62 (42.5-113.3)           | 43 (29.8-94.8)                | 0.078* |
| GGT (U/L)       | 1.19 ± 0.54               | 0.93 ± 0.33                   | 0.031* |
| Platelets x 10⁴ | 110.5 (76.1-159.3)        | 66.5 (36.2-137.8)             | 0.005* |
| Albumin (g/dL)  | 3.75 ± 0.66               | 4.33 ± 0.29                   | < 0.0001* |
| Total bilirubin | 0.94 (0.70-1.48)          | 0.5 (0.44-0.70)               | < 0.0001* |
| Alkaline phosphatase (U/L) | 99.5 ± 38.9 | 90.9 ± 56.1 | 0.490* |
| INR             | 1.18 ± 0.17               | 1.03 ± 0.07                   | < 0.0001* |
| APR1            | 2.13 (0.99-3.75)          | 0.59 (0.39-0.92)              | < 0.0001* |
| FIB-4           | 4.8 (2.56-9.30)           | 1.72 (1.01-2.21)              | < 0.0001* |

1Mean ± standard deviation; 2Median (IQR-IQR); 3χ² test; 4T test; 5Mann-Whitney test. ALT: Alanine aminotransferase; AST: Aspartate aminotransferase; BMI: Body mass index; GGT: Gamma glutamyl transferase; INR: International normalized ratio; APRI: Aspartate aminotransferase to platelet ratio index; FIB-4: Fibrosis index based on four factors.
Metabonomics for assessment of liver fibrosis

In general, the MM presented a high accuracy and performance, similar to the APRI to predict SF and C, and similar to FIB-4 to predict AF (Figure 2). The area under ROC curve (AUROC) of APRI to predict SF and C was 0.79 (95%CI: 0.68-0.90) and 0.76 (95%CI: 0.61-0.91), respectively, whereas the AUROC of FIB-4 to predict AF was 0.84 (95%CI: 0.74-0.95) (Supplementary Figure 1).

For the 68 patients with an APRI score, 27 (39.7%) had intermediate test values (> 0.5 and ≤ 1.5). Of these, 17 (63%) had SF by METAVIR. All these patients were correctly classified by the MM. Likewise, among 25 (36.8%) patients with FIB-4 values in the gray zone (> 1.45 and ≤ 3.25), the MM correctly identified 4 (16%) with AF. Figure 3 makes a comparison of the extent to which the MMs, APRI and FIB-4 would have correctly avoided the need for a biopsy.

DISCUSSION

In this study, it was observed that the higher the severity of liver fibrosis, the greater the mean age as well as the greater the impairment of liver function. In fact, this findings reflects the natural history of chronic hepatitis C, a fibrosing disease, which progresses to cirrhosis in about 30 years[23].

Our results suggest that the metabonomic strategy can discriminate F0-F1 from F2-F4 patients, F0-F2 from F3-F4 patients and F0-F3 from F4 patients. The MMs developed to predict SF, AF and C in CHC patients presented high performance, with values of SN, SP and an accuracy of above 90%. In practice, considering the confidence intervals, the results would be classified as similar to APRI and FIB-4. However, the MMs presented 100% accuracy for predicting SF and AF in the gray zone of APRI and FIB-4, when these indexes could not determine the absence or presence of significant and advanced fibrosis. If we consider the 39.7% of unclassified and 11.8% of incorrectly classified patients using APRI as a predictor of SF, liver biopsy would have been correctly avoided in 48.5%
of cases. On the other hand, if the MM were used to this end, the biopsy would have been correctly avoided in 95.7% of the patients. Regarding FIB-4 analysis, considering the 36.8% of unclassified and 8.8% of incorrectly classified patients, using this index as the only predictor of AF, biopsy would have been correctly avoided in 54.4% of patients, while the MM would have prevented biopsy in 95.7% of patients (Figure 3).

The most widely indirect methods for the assessment of liver fibrosis in CHC patients in routine clinical practice are the non-commercial serological indexes APRI and FIB-4, and the physical methods, such as liver stiffness measurement, by elastography based on US (transient liver elastography, acoustic radiation force impulse elastography and 2D-shear wave elastography) or based on magnetic resonance imaging. In general, these methods do not distinguish well intermediate stages of fibrosis, although they are increasingly useful in the exclusion of significant fibrosis (< F2) and presence of advanced fibrosis (≥ F3). The serum levels of the extracellular matrix protein osteopontin are promising for the diagnosis of intermediate fibrosis, with increasing concentration in different stages of fibrosis groups from F0 to F4, progressively and significantly different between the groups, and with an AUROC of 0.977 for the discrimination of F1-F2 from F3-F4 patients[24]. Bourbier et al[25] proposed the FibroMeter® + FibroScan® (FM+FS) algorithm, based on two fibrosis indexes (significant and advanced fibrosis indexes), from a combination of these two methods by logistic regression. Reliable diagnosis intervals of these two indexes were determined, resulting in a noninvasive classification of fibrosis in six classes. This classification showed an accuracy of 86.7% and using this algorithm, biopsy would be avoided in 100% of patients with significant and advanced fibrosis. However, these methods are based on high cost tests that are not always routinely available, especially in public health services in developing countries.

In fact, the NMR based metabonomics proved to be promising for evaluating the severity of liver disease, cirrhosis and its complications, which reflects the progression of fibrosis[26,27]. Amathieu et al[28] correlated the severity of hepatic impairment in 124 patients with chronic alcoholic liver disease, as measured by MELD, using the OPLS-DA model based on 1H NMR spectroscopy of serum samples. The same authors, using OPLS-DA model based on 1H NMR of serum samples, separated 93 patients with compensated alcoholic cirrhosis from 30 patients with acute on chronic liver failure, with good predictability[29]. Furthermore, Jiménez et al[30], using OPLS-DA MM, based on 1H NMR spectroscopy of serum samples from cirrhotic patients, were able to distinguish between 39 patients with minimal encephalopathy and 62 patients without encephalopathy, with R² = 0.68 and Q² = 0.63. A similar finding was described by Qi et al[31], who demonstrated that the OPLS-DA model, based on 1H

| Table 3 Linear discriminant analysis metabolomic models, APRI and FIB-4 performances to predict significant fibrosis, advanced fibrosis and cirrhosis in 69 chronic hepatitis C patients from Pernambuco/Brazil |
|---------------------------------------------------------------|
| Significant fibrosis | Biopsy | P value | Sensitivity (%) | Specificity (%) | LR+ | LR- | A (%) |
|----------------------|--------|---------|-----------------|-----------------|-----|-----|------|
| Model (n = 69) | F2-F4 | F0-F1 | ≥ F2 | 41 | 2 | < 0.001 | 97.6 | 87.4-99.9 | 92.6 | 75.7-99.1 | 13.2 | 3.5-50.1 | 0.03 | 0.004-0.2 | 95.7 |
| | F2-F4 | F0-F1 | < F2 | 1 | 25 | | | | | | | | | | 95.7 |
| APRI (n = 68) | > 0.5 | 0.5 | | 36 | 13 | 0.001 | 85.7 | 71.5-94.6 | 50 | 29.9-0.70 | 1.71 | 1.2-2.6 | 0.3 | 0.1-0.7 | 72 |
| ≤ 1.5 | 1.5 | 1 | 2 | 0.001 | 45.2 | 29.9-61.3 | 92.3 | 74.9-99.0 | 5.9 | 1.5-23.2 | 0.6 | 0.4-0.8 | 63.2 |
| Advanced fibrosis | Model (n = 69) | F3-F4 | F0-F2 | ≥ F3 | 27 | 2 | < 0.001 | 96.4 | 81.7-99.1 | 95.1 | 83.5-99.4 | 19.8 | 5.1-75.5 | 0.04 | 0.005-0.3 | 95.7 |
| | F3-F4 | F0-F2 | < F3 | 1 | 39 | | | | | | | | | | 95.7 |
| FIB-4 (n = 68) | > 1.45 | 1.45 | | 25 | 24 | 0.012 | 89.3 | 71.8-97.7 | 40 | 24.9-56.7 | 1.5 | 1.1-2.0 | 0.3 | 0.1-0.8 | 60.3 |
| ≤ 3.25 | 3.25 | 3 | 16 | 0.012 | 75 | 55.1-89.3 | 92.5 | 79.6-98.4 | 10 | 3.3-30.3 | 0.3 | 0.1-0.5 | 85.3 |
| Cirrhosis | Model (n = 69) | F4 | F0-F3 | ≥ F4 | 18 | 1 | < 0.001 | 100 | 81.5-100 | 98 | 89.6-99.9 | 33.8 | 6.9-163.7 | 0.03 | 0.002-0.4 | 98.6 |
| | F4 | F0-F3 | < F4 | 0 | 50 | | | | | | | | | | 98.6 |
| APRI (n = 68) | > 1.00 | 1.00 | | 14 | 16 | 0.002 | 77.8 | 52.4-93.6 | 68 | 53.3-80.5 | 2.4 | 1.5-3.9 | 0.3 | 0.1-0.8 | 70.6 |
| ≤ 1.00 | 1.00 | 4 | 34 | | | | | | | | | | | 70.6 |
| > 2.00 | 2.00 | 9 | 9 | 0.008 | 50 | 26.0-74.0 | 82 | 68.6-91.4 | 2.8 | 1.3-5.9 | 0.6 | 0.4-1.0 | 73.5 |

1Fisher’s exact test; 2χ² test; 3Estimated value. A: Accuracy; LR+: Positive likelihood ratio; LR-: Negative likelihood ratio; APRI: Aspartate aminotransferase to platelet ratio index; FIB-4: Fibrosis index based on four factors.
NMR of serum samples, distinguished between 30 compensated HBV cirrhotic patients from 30 patients with decompensated cirrhosis, with 85% accuracy. Regarding the evaluation of liver fibrosis, the metabonomic and metabolomic strategies have also shown promising results. In fact, Sands et al.\[32\] on analyzing $^1$H NMR plasma spectroscopy of healthy controls and CHC patients, constructed OPLS-DA MMs, using METAVIR and the Enhanced Liver Fibrosis (ELF) score as reference standards. The models distinguished between 6 healthy controls and 34 CHC patients with moderate fibrosis (F1-F2) and 22 with advanced fibrosis (F3-F4) in the validation cohort. Embade et al.\[33\] differentiated 27 cirrhotic patients (F4) from 30 patients without fibrosis (F0) using a PLS-DA model, through serum spectral analysis by $^1$H NMR, in CHC patients, with good predictive power ($R^2 = 0.8$ and $Q^2 = 0.6$). Sarfaraz et al.\[34\] using serum $^1$H NMR spectroscopy of 45 CHC patients, showed that F0-F2 and F3-F4 patients were well discriminated with the OPLS-DA MM ($R^2 = 0.67$ and $Q^2 = 0.28$). Moreover, Gabbani et al.\[35\], constructed PLS-models with canonical analysis (PLS-CA), based on serial analysis of serum/plasma and urine NMR spectra in 33 CHC patients, 10 compensated cirrhotic patients and 23 with mild or moderate fibrosis, as measured by biopsy or transient hepatic elastography. The authors demonstrated that the models recognized compensated cirrhosis with better accuracy in plasma/serum samples (68% in plasma, 56% in serum and 50% in urine samples). Herein, the main finding of our study was the correct classification of patients in the gray zone of APRI and FIB-4, who had mainly intermediate fibrosis (METAVIR classification equal to F2).

In conclusion, the metabonomic strategy was able to distinguish between significant liver fibrosis, advanced liver fibrosis and cirrhosis in CHC patients, showing promising results as a non-invasive tool to evaluate liver fibrosis. Furthermore, the method presented high accuracy in the gray zone of APRI and FIB-4, who had mainly intermediate fibrosis (METAVIR classification equal to F2).

Due to the small sample size, it was not possible to use one group for external validation. Therefore, further studies with a larger number of patients tested and external validation of the models are necessary, in order to confirm the performance of the MMs, for later incorporation into clinical practice.

**ARTICLE HIGHLIGHTS**

**Research background**

Liver fibrosis is the most important prognostic factor in chronic hepatitis C (CHC). The gold standard method for liver fibrosis evaluation is the liver biopsy, which
is invasive and subject to complications and misclassification.

**Research motivation**

The assessment of significant liver fibrosis is important to make therapeutic decisions and predict clinical outcomes. The serological and physical surrogate methods in hepatic fibrosis evaluation, such as APRI, FIB-4 and hepatic elastography, are good in excluding significant fibrosis and confirming advanced fibrosis, but they fail in diagnosing intermediate fibrosis. The metabolomics strategy is a method that seeks to discriminate biological samples, such as serum, through $^1$H nuclear magnetic resonance (NMR) spectroscopy, according to their biochemical status, thereby associating this status to a given condition. The method has been studied in various liver diseases, showing to be useful in identifying liver cirrhosis and its complications and with promising results in hepatic fibrosis evaluation in chronic liver diseases.

**Research objectives**

The aim of this study was to develop and evaluate metabolomic models (MMs), using $^1$H NMR spectrum of serum samples, as non-invasive tool for assessment of significant liver fibrosis, advanced liver fibrosis and cirrhosis in patients with CHC. Additionally, to compare the performance of the MMs with the serological surrogate indexes APRI and FIB-4 and to investigate the performance of MMs in the gray zone of these serological indexes.

**Research methods**

This was a cross-sectional phase II validation diagnostic study, with prospective inclusion of CHC adult outpatients who had undergone percutaneous liver biopsy in the previous 12 mo or had been clinically diagnosed with liver cirrhosis. The clinical diagnosis of cirrhosis was based on characteristic symptoms and signals and/or according to evidence of chronic liver disease and/or portal hypertension on ultrasound and/or signals of portal hypertension observed on upper gastrointestinal endoscopy. Those undergoing antiviral treatment or diagnosed with other liver disease were excluded. All patients signed an informed consent form and the study was approved by the Ethics Committee of the Institution. Fasting blood samples were collected from all patients by peripheral vein puncture. The laboratory tests were performed by an automated method. The HCV RNA was detected by real-time polymerase chain reaction. APRI and FIB-4 were calculated as described by Wai et al (2003) and Scarlino et al (2006). Serum samples, stored at -20 °C, were thawed at room temperature and prepared by adding 200 μL of D2O to 400 μL of serum. $^1$H NMR spectra were obtained using a Varian Unity Plus 300 spectrometer. The samples were analyzed using a pulse sequence with suppression of the resonance of water and T2 filter (PRESS-CPMG). Spectra were processed using line broadening equal to 0.3 Hz. The signal attributed to the methyl resonance of water and T2 filter (PRESAT-CPMG). Spectra were processed using MestreNova software. The spectra were divided into 250 regions of 0.04 ppm, called bins, used to construct a dataset, eliminating the residual signal of water. The spectra were normalized and MLS-DA and LDA supervised formals were used. Three MLS-DA models were constructed to predict SF, FA and C, respectively, using the MetaboAnalyst 3.0 platform. The models were validated by leave-one-out cross-validation (LOOCV) and by permutations tests. In addition, three LDA models were constructed, using the PCA matrix as input data, to predict SF, AF and Ccs, and validated by LOOCV using STATISTICA software. For each LDA MM, and for APRI and FIB-4, a 2 x 2 contingency matrix was used to calculate sensitivity (SN), specificity (SP), the positive likelihood ratio (LR+), the negative likelihood ratio (LR-) and accuracy (A), using liver biopsy as reference standard.

**Research results**

Sixty-nine patients were evaluated, of whom 59.4% were female, with a mean age of 57 ± 12 years. Liver biopsy was performed on 54 (78%) patients. The median fragment was 15 mm in length (P25: 13; P75: 18 mm), and there were 15 portal tracts (P25: 12; P75: 20). The METAVIR fibrosis stage was distributed as follows: F0 in 2; F1 in 25; F2 in 14; F3 in 10 and F4 in 3 patients. The diagnosis of cirrhosis was clinically established in 15 (21.7%), thus classified according to the Child-Pugh score: 10 patients Child-Pugh A and 5 Child-Pugh B. Therefore, 42 (60.9%) patients were classified as SF, 28 (40.6%) as AF and 18 (26.1%) as C. The MLS-DA MM for SF showed presented 100% accuracy, R2 of 0.98 and Q2 of 0.91, respectively. The results of the LDA MM for SF were compared to APRI and showed SN of 97.6% (95%CI: 87.4%-99.9%) and LR- of 0.03 (95%CI: 0.004-0.2), similar to APRI cut-off of 0.5, which presented SN of 85.7% (95%CI: 71.5%-94.6%) and LR- of 0.3 (95%CI: 0.1-0.7). The MM presented SF of 92.6% (95%CI: 75.7%-99.1%) and LR+ of 13.2 (95%CI: 3.5-50.1), similar to APRI cut-off of 1.5, which showed SP of 92.3% (95%CI: 74.9%-99.9%), with LR+ of 5.9 (95%CI: 1.5-23.2). The MLS-DA MM for SF discriminated all the samples, with 100% accuracy, R2 of 0.98 and Q2 of 0.93. LDA MM for AF showed SN of 96.4% (95%CI: 81.7%-99.1%) and LR- of 0.04 (95%CI: 0.005-0.3), similar to FIB-4 cut-off of 1.45, which showed SN of 89.3% (95%CI: 71.3%-97.7%) and LR- of 0.3 (95%CI: 0.1-0.8). The two methods also presented high SP and LR+, there being observed SP of 95.1% (95%CI: 83.5%-99.4%) and LR+ of 19.8 (95%CI: 5.1-76.5) for MM, and SP of 92.3% (95%CI: 79.6%-98.4%) and LR+ of 10 (95%CI: 3.3-30.3) for FIB-4 cut-off of 3.25. The MLS-DA MM for C also adequately discriminated the samples, attaining an accuracy of 84.0%. The LDA MM for C showed SN of 100% (95%CI: 81.5%-100%) and LR- of 0.03 (95%CI: 0.002-0.4), similar to APRI cut-off of 1.0, which presented SN of 77.8% (95%CI: 52.4 - 93.6), with LR- of 0.3 (95%CI: 0.1-0.8). The MM presented SP of 98% (95%CI: 89.6%-99.9%), similar to APRI cut-off of 2.0, which showed SP of 82% (95%CI: 68.6 - 91.4), with higher LR+ of 33.8 (95%CI: 6.9-163.7), compared to APRI cut-off of 2.0, which presented LR+ of 2.8 (95%CI: 1.3-5.9). In general, the MMs presented similar performance to APRI and FIB-4. However, their accuracy for predicting SF and AF in the gray zone of APRI and FIB-4 was 100%. Considering the 39.7% of unclassified and 11.8% of incorrectly classified patients using APRI as a predictor of SF, liver biopsy would have been correctly avoided in 46.5% of cases. On the other hand, if the MM were used to this end, the biopsy would have been correctly avoided in 95.7% of the patients. Regarding FIB-4 analysis, considering the 36.8% of unclassified and 8.8% of incorrectly classified patients, using this index as the only predictor of AF, biopsy would have been correctly avoided in 54.4% of patients, while the MM would have prevented biopsy in 95.7% of patients.

**Research conclusions**

The metabolomic strategy was able to distinguish between significant liver fibrosis, advanced liver fibrosis and cirrhosis in CHC patients, showing promising results as a non-invasive tool to evaluate liver fibrosis. The main finding of our study was the correct classification of patients in the gray zone of APRI and FIB-4, who had mainly intermediate fibrosis (METAVIR classification equal to F2), which could avoid a large number of biopsies in CHC patients.

**Research perspectives**

It is necessary to perform further studies testing a larger number of patients and with external validation of the models, in order to confirm the performance of the MMs, for later incorporation into clinical practice.

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