An NMR Metabolomics Approach for the Diagnosis of Leptomeningeal Carcinomatosis

Hye Rim Cho1,2, He Wen3, Young Jin Ryu1, Yong Jin An5, Hyo Cheol Kim1,4, Woo Kyung Moon1,4, Moon Hee Han1,4, Sunghyouk Park3, and Seung Hong Choi1,4

Abstract

Leptomeningeal carcinomatosis (LC) is the third most common metastatic complication of the central nervous system. However, the current modalities to reliably diagnose this condition are not satisfactory. Here, we report a preclinical proof of concept for a metabolomics-based diagnostic strategy using a rat LC model incorporating glioma cells that stably express green fluorescent protein. Cytologic diagnoses gave 66.7% sensitivity for the 7-day LC group and 0% for the 3-day LC group. MR imaging could not diagnose LC at these stages. In contrast, nuclear magnetic resonance–based metabolomics on cerebrospinal fluid detected marked differences between the normal and LC groups. Predictions based on the multivariate model provided sensitivity, specificity, and overall accuracy of 88% to 89% in both groups for LC diagnosis. Further statistical analyses identified lactate, acetate, and creatine as specific for the 7-day LC group, with glucose a specific marker of the normal group. Overall, we showed that the metabolomics approach provided both earlier and more accurate diagnostic results than cytology and MR imaging in current use. Cancer Res 72(20): 5179–87. ©2012 AACR.

Introduction

Leptomeningeal carcinomatosis (LC) is a disorder caused by the seeding of the leptomeninges, the pia, arachnoid, and cerebrospinal fluid (CSF) within the subarachnoid space, by malignant cells. This disorder is the third most common metastatic complication of the central nervous system and increasingly common in cancer patients (1, 2). Although LC most often presents in patients with widely disseminated and progressive systemic cancer (>70%), it can present after a disease-free interval (20%) and even be the first manifestation of cancer (5%–10%), occasionally occurring even in the absence of other evidence of systemic disease (3). For patients diagnosed with LC, the median survival time of untreated patients is known to be 4 to 6 weeks, which can extend to 3 to 6 months with treatment (4, 5). Treatment can also improve or stabilize a patient’s neurological status and maintain their neurological quality of life (3). Thus, an early diagnosis of LC is important to alert the oncologist to begin therapy before neurologic deterioration.

The most useful laboratory test in the diagnosis of LC has been the examination of CSF obtained from lumbar puncture (6–8). However, abnormalities in the CSF, although suggestive of LC, are not diagnostic. The presence of malignant cells in the CSF detected in a cytologic examination is diagnostic of LC. However, the assignment to a particular tumor is often not possible, which is true for most cytologic analyses (6–8). Of the patients with final results that reveal positive CSF cytology, up to 45% will be cytologically negative upon initial examination (6, 7, 9). The yield increases to 80% with a second round of CSF examination, but there is little benefit for further rounds of examination (6, 7, 9). The low sensitivity of CSF cytology makes it difficult not only to diagnose leptomeningeal metastasis but also to assess the patient’s response to treatment (2). Biochemical markers, immunohistochemistry, and molecular biology techniques have been applied to CSF in an attempt to find reliable biological markers of the disease. Numerous biochemical markers have been evaluated, but their use has been generally limited by poor sensitivity and specificity (3). Therefore, it is important to develop a new method that improves the diagnostic sensitivity of LC, which can enhance the evaluation of prognosis and reduce unnecessary medical procedures.

While genomic and proteomic approaches focus on upstream genetic and protein variations, metabolomics is increasingly being used to understand the global metabolic changes that occur in response to alterations in nutrition, genetics, and environment (10, 11), which can complement the information obtained from genomics and proteomics. In addition, metabolomics has been used to identify metabolite-based biomarkers for a variety of diseases conditions (see ref. 12 for a...
recent review). Metabolic profiles can be investigated using high-throughput analytical tools, such as nuclear magnetic resonance (1H-NMR) spectroscopy. 1H-NMR spectroscopy of biofluids such as urine, serum, and bile can give comprehensive small molecular profiles of metabolites that are regulated by various physiologic signals involved in maintaining homeostasis. Homeostasis can be disrupted by changes in a patient’s physiologic condition, resulting in the perturbation of metabolite levels. Thus, monitoring these perturbations can provide not only novel information of the patient’s physiologic condition but also new biomarkers for disease diagnosis.

Because the metabolite composition of CSF is directly related to the biological processes of the brain, profiling CSF metabolites can provide important information for diagnosing various brain abnormalities, such as brain injury (13), Huntington’s disease (14), and Parkinson’s disease (15). Himmelreich and colleagues (16) showed a practical application of rapidly identifying the bacterial species in bacterial meningitis based on CSF metabolite profiles, which may help appropriate antibiotic selection in a clinical setting. With the recent report by Wishart and colleagues (17) detecting approximately 70 CSF metabolites and building a public database, we expect CSF metabolomics studies to be applied to a wider variety of systems.

The purpose of our study was to develop a new NMR spectroscopic method that could be used to improve the diagnosis of LC. We first designed a rat LC model with glioma cells stably expressing GFP and used NMR-based metabolomics to characterize CSF metabolites. As a result, we were able to obtain earlier and significantly improved diagnostic results compared with currently used cytologic or radiological approaches.

Materials and Methods

This experiment was approved by the animal care committee at Seoul National University Hospital.

Rat LC model and sham operation

To produce LC rats, GFP-expressing F-98 cells (Supplementary Materials and Methods and Supplementary Fig. S1) were prepared in 50 μl of serum-free Dulbecco’s Modified Eagle’s Media and intradurally transplanted into the parietal area of 6-week-old Fisher 344 rats (1 x 10^6 cells/rat; Supplementary Materials and Methods). For the sham operation, rats (n = 4) were intradurally injected with 50 μl of serum-free Dulbecco’s Modified Eagle’s Media. In vivo MR imaging and CSF collection (Supplementary Materials and Methods) were conducted 3 (n = 8: 3-day LC group) and 7 days (n = 9: 7-day LC group) after cell implantation. Rats in the sham group also underwent in vivo MR imaging 3 days after the operation.

NMR spectroscopy data acquisition

All NMR spectra were acquired at 19° C with a 500 MHz NMR spectrometer (Agilent, VNMRS 500 system, 11.7 T) equipped with an HR-MAS nano-probe (Agilent). A CSF sample (40 μl) was placed into the probe with inverse-detection and single Z-gradient capability. The receiver gain was adjusted carefully not to cause signal overload. The sample was spun at 5 KHz by a stream of nitrogen gas with 54.7° (the magic angle) relative to the applied magnetic field. The spectra were acquired with a total number of complex points of 16 K, sweep width of 7,961 Hz, and 1,024 transients. The 90° pulse was calibrated with each sample on the water resonance. The water signal was saturated using weak power continuous wave during the recycle delay.

Data processing

The time-domain spectra were apodized with an exponential function (1 Hz). Fourier transformed, phased, and manually baseline corrected. The signals were referenced to the TSP signal at 0.00 ppm and normalized against the total integration values. Water and ethanol peaks were removed to prevent artifacts in downstream analyses. To reduce the complexity of the NMR data for pattern recognition, the spectra were binned at every 0.0025 ppm interval. The normalization and binning were conducted using an in-house built Perl program.

Statistical analysis

Statistical analysis was conducted with R (from R Project for Statistical Computing) and SIMCA-P 11.0 (Umetrics). Che-nomx (Spectral database) and an in-house built database were used for metabolite identification. Partial least square regression and OPLS-DA (see Supplementary information S2) were conducted to identify latent patterns and compare the overall metabolite profile. Class discrimination models were built until the cross-validated predictability value did not significantly increase to avoid overfitting of the statistical model. Diagnostic performance was obtained by prediction of one left-out sample based on the distinction model constructed with the rest of the samples. The method was repeated until all the samples were left out once. An a priori cut-off value of 0.5 was used to evaluate the prediction results (18). Signals specific for each group were identified by conducting the Wilcoxon rank sum test on all the ppm variables using the in-house written R script.

Results

CSF cytology, MR imaging, and histology

Among the 9 CSF samples obtained 7 days after F-98 cell implantation (7-day LC), 6 (66.7%) were found to have malignant cells in the CSF cytologic examination as evidenced by the expression of GFP (Fig. 1). In contrast, no samples collected 3 days after F-98 cell implantation (3-day LC) exhibited positive CSF cytology. MR imaging revealed meningeal enhancement in all 17 rats that underwent glioma cell implantation (Fig. 2A and B), but it also showed the enhancement in the 4 rats from the sham group (Fig. 2C). All implanted rats (n = 17) had F-98 cells infiltrating the leptomeninges as evidenced by GFP expression, which confirmed the proper establishment of LC in both 7-day (Fig. 3A) and 3-day (Fig. 3B) groups. Still, leptomeningeal infiltration of F-98 cells was more prominent in 7-day LC rats than in 3-day LC rats, suggesting that 3-day LC group is in an earlier phase of the LC development. Overall, among the 17 histology-confirmed LC rats, 6 in the 7-day LC group could be diagnosed with cytology, and MR imaging was not useful due to the false positive effects.
HR-MAS MR spectra and multivariate analysis

We acquired the NMR data with a high-resolution magic angle-spinning (HR-MAS) probe because the CSF volume obtained from a rat (typically ~ 40 µL) is far smaller than that required for a usual 5-mm probe (~ 500 µL). HR-MAS MR spectra of the 26 CSF samples (9, 8, and 9 samples for normal, 3-day LC, and 7-day LC rats, respectively) from the animal model gave reasonable signals, even though the sample volumes were less than one-tenth of that used for 5-mm NMR probe (Fig. 4 and Supplementary Fig. S3). Chemical shifts and scalar coupling values were used to identify a number of metabolites known to be present in CSF, such as lactate (1.35 and 4.12 ppm) and glucose (multiple signals between 3 and 4 ppm). As the intragroup variation hindered the assignment of the spectral characteristics of each group by simple visual inspection, we applied a more holistic approach of multivariate statistical analysis to the entire NMR data set. First, partial least square (PLS) analysis was conducted with all 3 groups to determine the differences with time (Fig. 5A). Our results show that spectral features move toward the right side along the first component axis as time increases. The metabolite profiles of 7-day LC rat samples were quite different from those of normal rats. In comparison, the profiles of 3-day LC rats were between the normal and 7-day LC rat samples in the direction of the first component. Still, the directional changes in the second component were opposite between the 3-day and 7-day LC rat samples, suggesting that the overall metabolite profiles may not change in a simple linear fashion with time.

To maximize the difference between a given pair and obtain quantitative information about the diagnostic performance at each time point, orthogonal projections to latent structure-discriminant analysis (OPLS-DA) modeling was conducted for samples on normal versus 3-day or 7-day LC rats. Both pairs yielded differentiation models with 1 predictive and 2 orthogonal components (Fig. 5B and C). The models featured 0.965 (0.963) of goodness of fit and 0.759 (0.717) of predictive values for normal versus 3-day LC rat (normal vs. 7-day LC rat) differentiation. Both 3-day and 7-day LC rat samples were well separated from normal rat samples, suggesting that relatively early metabolic changes at day 3 may be sufficient to distinguish these samples from those of normal rats.

Diagnostic predictions

Good statistical characteristics of a differentiation model do not necessarily translate into good practical performance. Therefore, we conducted a diagnostic prediction test with the established differentiation models. We left out one CSF sample at a time, constructed a new model with the remaining samples, and predicted the disease status of the left-out sample. The procedure was repeated until all of the CSF samples were tested once, with a diagnosis based on a priori cut-off value of 0.5. The test was conducted on both 3-day (Fig. 6A) and 7-day LC samples (Fig. 6B) against normal rat samples. The prediction results showed that the models correctly predicted 16 samples out of a total of 18 for both groups. One identical normal sample (from case 3) was mispredicted as LC samples.
and one LC sample from each of the 3-day and 7-day LC groups was mispredicted as normal. Thus, our metabolomics model showed sensitivities of 89 (7-day LC group) or 88% (3-day LC group), a specificity of 89%, and an overall accuracy of 89% in the prediction of LC. These results suggest that this method could be useful for both relatively early (day 3) and late stage (day 7) LC diagnosis.

Analysis of marker signals

To identify the metabolites that are characteristic of LC, we conducted a Wilcoxon-rank sum test on all ppm variables (Fig. 7A). Although we could not identify all of the significant signals (blue dots) due to the relatively low signal-to-noise ratio and overlaps, we were able to identify glucose (3.40, 3.48, 3.73, and 3.83 ppm) as being specific for the normal group and lactate (1.35 and 4.12 ppm), acetate (1.93 ppm), and creatine (3.05 and 3.93 ppm) for the 7-day LC group by database analysis and comparisons with standard compounds (Fig. 7B; see Supplementary information S4). We also tried correlating the NMR spectral markers with the thickness of the LC infiltration in the 7-day LC animals. We built a partial least square regression model between the levels of the 4 markers as independent variables and the thickness as a dependent variable. The regression result gave good correlation with $R^2$ value of 0.95 between the actual values and regression–predicted values (Supplementary Fig. S5). The number and intensities of significant signals were smaller for the analysis of normal versus 3-day LC rat samples, and reliable identification of metabolites

Figure 2. Contrast-enhanced T1WI MR imaging of rat brain. A normal rat (A) shows no meningeal enhancement, whereas meningeal enhancement (arrows) was observed in both rats in LC model rats (B) and sham operations (C). Thus, meningeal enhancement was not a specific finding for the LC. CC, cerebral cortex; CL, cerebellum; OB, olfactory bulb; T, thalamus.

Figure 3. Histopathological examination showed the F-98 cells that expressed GFP infiltrating the leptomeninges. Leptomeningeal infiltration (indicated by arrows) of F-98 cells was more prominent in 7-day (A) than in 3-day (B) LC rats. Histopathology from the brains of 7-day (A) and 3-day (B) LC rats present the late and early LC status, respectively.
was difficult for 3-day LC rat samples (Fig. 7A bottom). Therefore, the distinction between normal and 3-day LC rat samples should be due to the sum of small and unassignable signals rather than several conspicuous metabolite signals. In addition, the pattern of the significant signals of 3-day and 7-day LC rat samples appeared to be different.

Discussion

Metabolomics is a global assessment of endogenous metabolites within a biological system. This assessment is gaining increased attention for its role in diagnosing cancer (19), refining preoperative differential diagnosis, detecting tumor progression, and monitoring responses to treatment (20). Magnetic resonance spectroscopy (MRS) is a method that detects metabolites within cells, tissues, or biofluids and has been used in medical fields along with MR imaging. Notably, proton MRS has been used for the detection of brain, breast, and prostate cancers, which led to the discovery of choline, lactate, and amino acid peaks as biomarkers for cancer. Analogously, in vitro NMR spectroscopy of fine-needle aspiration biopsy samples of primary tumors exhibited 93% accuracy in differentiating between benign and malignant samples by the measurement of choline peaks (21, 22). With these successful applications of NMR-based metabolomics to the medical field, we hypothesized that LC could also be diagnosed using the different metabolite profiles in CSF measured by in vitro NMR spectroscopy. In the present study, we used a rat LC model using F-98 glioma cells expressing GFP to enhance the detectability of cancer cells in cytology and histopathology.

Figure 4. Representative HR-MAS spectra of CSF. NMR spectra of CSF samples from rat models from normal (top), 3-day LC (middle), and 7-day LC (bottom) groups. Specific metabolite peaks (marked in the top spectrum) were assigned with Chenomx (Spectral database). tCreatine, total creatine including creatine and phosphocreatine. All NMR spectra were acquired with a nano-probe installed in a 500 MHz NMR spectrometer (Agilent, VNMR 500 system).

Figure 5. Multivariate analysis of the spectral data. PLS analysis of the entire set of samples (A), where black, red, and blue represent normal, 3-day LC and 7-day LC groups, respectively. Mean values of the modeled scores are represented by square, triangle, and circle symbols. Vertical and horizontal error bars represent standard deviations. Binary discrimination with OPLS-DA between normal and 3-day LC samples (B), where black and red represent the normal and 3-day LC group, respectively. Binary discrimination with OPLS-DA between normal and 7-day LC rat samples (C), where black and blue represent normal and 7-day LC groups, respectively. Both OPLS-DA models were obtained with 1 predictive and 2 orthogonal components.
We applied the NMR spectroscopic method using an HR-MAS probe for the diagnosis of LC in a rat model. The approach showed a sensitivity and specificity of 89% and an overall accuracy of 89% for the 7-day LC model. In the 7-day LC group, we observed increases in lactate, creatine, and acetate levels and decreases in glucose levels compared with those of the normal group (P < 0.006 for all). Although we could not reliably identify marker signals in the model of an early stage (3-day LC rats), we were able to obtain similar diagnostic performance at day 3. This good performance constitutes the most significant results of our study because they were significantly better than those obtained using other currently available modalities on the same animals. For example, cytology, which is often used to confirm the presence of LC, failed to detect any glioma cells on day 3 and detected cancer cells in significantly less samples (66.7%) at a late stage (day 7). In addition, although in vivo MR imaging detected meningeal enhancement in LC model rats in both early and late stage groups (3-day and 7-day LC rats), enhancement in the sham operation group was also detected. Therefore, radiological enhancement is nonspecific and in vivo MR imaging has very low specificity, which may limit their use.

In many tumor tissues, more than 50% of cellular ATP can be generated from glycolysis, even in the presence of oxygen, which led to the term aerobic glycolysis (or Warburg Effect; ref. 23). By using aerobic glycolysis, even in the presence of oxygen and functional mitochondria, tumor cells divert a large portion of pyruvate to extramitochondrial breakdown to lactate, the typical product of anaerobic glycolysis (24). In vivo MRS of the brain showed increased aerobic glycolysis in gliomas, which was measured as increases in lactate levels and correlate to the grade of gliomas (25). Thus, we believe that the observed increase in lactate levels and decrease in glucose levels can be explained by the aerobic glycolysis of disseminated cancer cells in CSF.

In terms of the elevated levels of acetate or creatine in the CSF of 7-day LC rats, similar observations have been made in...
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Figure 7. Marker metabolites for the LC group. A, the Wilcoxon rank sum tests of 7-day LC group (top) and 3-day LC group (bottom) were conducted with an in-house written R script. The x and y axes represent the ppm variable axis and the log of 1-P values, respectively. Values with P values less than 0.01 are shown as blue dots. B, the actual levels of the LC marker metabolites (7-day LC group) are shown using the normalized average spectra of the corresponding peaks. Blue, normal; green, 3-day LC; red, 7-day LC. The P values for normal versus the 7-day LC group are also shown.

Another malignancy, which were conducted using in vitro NMR spectroscopy. Recently, Imperiale and colleagues (26) reported the metabolic content of intact biopsy samples of adrenal neuroblastomas using HR-MAS NMR spectroscopy, which exhibited higher levels of creatine, glutamine/glutamate, acetate, and glycine in neuroblastomas. They also showed that acetate and creatine are characteristic of stage IV neuroblastomas, which led to the conclusion that high levels of acetate and creatine in tumors could be used in diagnosis and staging. High acetate concentrations in tumors are known to be mainly due to the enhanced lipid synthesis in the generation of cell membranes, which reflects the high growth activity of neoplasms (27). Thus, [1-13C]-acetate positron emission tomography has become a useful tool for detecting various types of malignant tumors, such as prostate cancer (28, 29), renal carcinoma (30), hepatocellular carcinoma (31), brain astrocytoma (32), and gliomas (33).

The elevation of creatine levels in the CSF of 7-day LC rats can be attributed to the changes in energy metabolism (34). A significant downregulation of creatine kinase has been reported in some cancers, such as oral squamous cell carcinoma, resulting in elevated levels of creatine (35). However, the creatine levels of the gliomas used in the present study are known to be lower than those of the cerebral cortex where creatine is highly produced. In addition, the total creatine concentration in gliomas tends to decrease according to the degree of malignancy (36), and the F-98 cell is derived from a high-grade glioma of a Fisher 344 rat. Thus, we believe that the elevation of creatine levels in the CSF of 7-day LC rats is due to the destruction of the cerebral cortex by seeded cancer cells in the CSF space (subarachnoid space).

Contrast-enhanced (e.g., gadolinium chelate agents) MR imaging is the technique of choice used to evaluate patients that are suspected to have LC (37). Some changes, such as cranial nerve enhancement and subarachnoid-enhancing nodules, can be considered diagnostic of LC in patients with cancer. Nonetheless, contrast-enhanced MR imaging has more than a 30% incidence of false-negative results. Thus, a normal study does not exclude the diagnosis of LC (3). In addition, other diseases, including infectious, inflammatory diseases, trauma, or subdural hematomas, can lead to similar MR imaging findings (37–41). Similar to the present study, surgically induced contrast enhancement in meninges is also well known (42). Thus, the detection of malignant cells in the CSF has been pursued to confirm the diagnosis of LC. However, our study further emphasizes that contrast-enhancement MR imaging may not be sufficient for the diagnosis of LC mainly due to its low specificity.

Although we obtained good diagnostic performance in both early (3 day) and late (day 7) stage LC models, we could not reliably identify the marker metabolites in the early stage LC group. Most of the contributing signals for 3-day LC rats were of low intensity or overlapped. Therefore, the distinction at the early stage may have been a result of the sum of small contributions. Still, the overall profiles of the significant signals were different between 3-day and 7-day LC groups and changed nonlinearly with time (see Fig. 5A and 7). For example, glucose and lactate, important and conspicuous marker signals on day 7, did not exhibit significantly different levels on day 3. As we failed to detect glioma cells on day 3 by cytology, a small number of cancer cells may have generated cancer-related metabolites just enough to make the overall profile different, but not enough...
to allow each metabolite identification. Future experiments with human samples with both a larger number of samples and increased sample volumes should enable metabolite detection with higher sensitivity.

As stated above, the diagnosis of LC can be conducted by MR imaging, CSF cytology, histology, and NMR-based metabolomics. Among these modalities, histology can provide a definite answer in animal models but cannot be conducted in clinical settings. From a practical perspective, NMR-based metabolomics showed the best performance among the aforementioned clinically applicable techniques. One theoretical advantage of NMR metabolomics over cytology is that NMR measures small molecules (typically those with MW < 400) which are highly diffusible, exhibiting uniform distribution regardless of the sampling positions used. In comparison, the cells observed in cytology are much less diffusible in the subarachnoid space where the CSF is collected; therefore, these cells may reflect rather local status. The significantly higher sensitivity of NMR method at day 7 (89%) over cytology (66.7%) may be due to these factors.

For a diagnostic measure, another important aspect is specificity. Although we cannot assess the specificity of LC diagnosis against all possible brain diseases, we may address the specificity against LC models using cancer cells from different tissues. It is expected that the distinction between related LC models based on different primary cancer cells should be more difficult than that between LC and other brain pathologies. In addition, it is clinically important to determine the tissue origin of a particular LC case. Therefore, we established another LC model using mammary adenocarcinoma cells (13762 MAT B III), measured its CSF metabolite profiles, and compared the results from the original model using brain cancer cells. For the specificity, we were able to observe a distinct group of significant marker signals (choline, glutamate, creatine, and lactate) of the 7-day LC developed with the breast cancer cells from those with the brain cancer cells (acetate, glucose, creatine, and lactate; Supplementary Fig S6). These results suggest that different metabolite profiles can be detected depending on the particular types of metastasizing cancer cells. We admit that this comparison of LC models is not a full resolution of the specificity concerns. Still, the results are encouraging in that we obtained differential metabolic profiles even for different subtypes of single brain pathology (LC). In actual human cases, additional information is available. Differentiation with meningitis, for example, can be also helped by clinical information such as medical history (e.g., presence of the underlying cancer or not), clinical picture (e.g., fever or not), and CSF findings (e.g., protein level). Therefore, we believe that the metabolomics approach is quite promising in human cases, and that our results warrant its human application. The diagnostic specificity can be ultimately addressed using patients with diverse brain pathologic conditions.

For humans, NMR techniques require lumbar puncture for detection with higher sensitivity. In addition, we only needed 0.04 mL of CSF, which can be readily provided from routine CSF samples (~ 10 mL) for cytology examinations. Thus, NMR samples can be obtained almost "for free" in terms of the reduced amount of labor of clinicians and suffering by patients. In addition, the NMR metabolomics approach does not require sample pretreatment and the data can be obtained in less than 5 minutes. As with other analytical approaches, there are also some difficulties in NMR metabolomics, including low sensitivity, resonance overlap, and possible processing biases. These may be addressed with better hardware and software in the near future with the availability of cryogenic microprobes, higher field magnets, and signal deconvolution algorithms. Particularly for the sensitivity of quantifying low-level metabolites, the NMR metabolomics is expected to perform much better for actual human samples that have the ultimate relevance to the diagnosis. That is due to the larger volume of attainable CSF volume (see above) that enables the use of standard 5-mm probes for enhanced sensitivity and/or concentration of the samples. Overall, we expect to see more examples of NMR metabolomics applied to diagnostic research with CSF as recently shown with pneumococcal or cryptococcal meningitis (16).

In conclusion, we have shown that the diagnosis of LC is possible using a metabolomics-based approach in a rat model. This approach was superior to the currently used CSF cytology and MR imaging procedures. It also enabled detection of LC at an early stage when CSF cytology failed. To the best of our knowledge, this study is the first report of a metabolomic diagnostic approach for LC, and it may become a useful clinical protocol that can augment or replace current diagnostic methods, if proven in human samples.

Disclosure of Potential Conflicts of Interest
No potential conflicts of interest were disclosed.

Authors' Contributions
Conception and design: H.R. Cho, H.C. Kim, W.K. Moon, M.H. Han, S. Park, S.H. Choi
Development of methodology: H.R. Cho, S.H. Choi
Acquisition of data (provided animals, acquired and managed patients, provided facilities, etc.): H.R. Cho, H. Wen, Y.J. Ryu, Y.J. An, S.H. Choi
Analysis and interpretation of data (e.g., statistical analysis, biostatistics, computational analysis): H.R. Cho, H. Wen, W.K. Moon, S. Park, S.H. Choi
Writing, review, and/or revision of the manuscript: H.R. Cho, H. Wen, Y.J. Ryu, H.C. Kim, W.K. Moon, S. Park, S.H. Choi
Administrative, technical, or material support (i.e., reporting or organizing data, constructing databases): H.R. Cho, M.H. Han, S.H. Choi
Study supervision: H.C. Kim, M.H. Han, S. Park, S.H. Choi

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