Metabolic profiling of RG2, F98 and C6 glioma models using ¹H-MRS and ex-vivo ¹H HRMAS MRS

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Introduction.

¹H MRS is a powerful analytical method to investigate the complex metabolic consequences of cancer disease. *In vivo* ¹H MRS can provide information regarding glioma growth and response to treatment. A wider range of metabolites can be obtained ex vivo in biopsies using ¹H High Resolution Magic Angle Spinning (¹H RMAS) MRS. Comparison of metabolic profiles between in vivo and ¹H RMAS MRS is essential [1] and the ability of both approaches to discriminate tumoral from normal tissue with statistical tools such as PLS-DA [2] might be of great help for diagnosis. Here, we have used this approach to analyze in vivo and ex vivo spectra in the rat RG2, F98 and C6 models of glioma.

Materials and methods.

Fischer rats (RG2 model, n=8; F98 model, n=12) and Wistar rats (C6 model, n=12), 8 weeks old at tumor inoculation, were used. After decapitation under 5 % isoflurane, tumor bulk and contralateral striatum were isolated (<3 minutes) and frozen in liquid N₂.

In vivo Data acquisition: ¹H MRS experiments were performed on a 7 Tesla Bruker BioSpec. A sequence (TE/TR=20/2500 ms) with water suppression and outer volume suppression was used to record two data sets, tumor and contralateral tissue. For each data set the volume of interest was 3x3x3mm³ and the acquisition time was 16m40s. The localization of the voxels of interest was based on T2-weighted RARE.

Ex vivo Data Acquisition: Approximately 15 mg of biopsy were introduced in 50 µL rotors and spun at 4 kHz. ¹H HRMAS NMR spectra were acquired on a Bruker Avance III 400 spectrometer (9.4T) at 4°C with a CPMG pulse sequence (TE=30 ms), water presaturation, 3.5 s total repetition time, 256 scan, 15 min total acquisition time.

Quantification: Metabolite quantification was performed with the software package iMRUI (http://www.mrui.uab.es/mrui/) using the "subtract-QUEST" procedure [3] and a simulated metabolite database set. For ¹HR MAS quantification, 16 metabolites were included in the database : acetate (Ace), alanine (Ala), creatine and phosphocreatine (Cr), choline (Cho), g-amino-butyric acid (GABA), glutamate (Glu), glutamine (Gln), glutathione (Gsh), glycerophosphocholine (GPC), glycine (Gly), lactate (Lac), myo-inositol (M-ins), N-acetylaspartate (NAA), phosphoethanolamine (PE), phosphocholine (PC) and taurine (Tau). Hypotaurine (Hyp) and betaine (Bet) were detected only in tumor tissue and then were added in the database of tumor. For in vivo spectra, 11 metabolites were quantified: creatine and phosphocreatine (tCr), total choline (tCho), GABA, Glu, Gln, Lac, M-ins, NAA , Asppartate (Asp), Hyp and Tau. Metabolite concentrations are given relative to the sum of all metabolites, both in vivo and ex vivo data.

Statistical analysis: multivariate statistical analysis was performed using the SIMCAP V12 software (Umetrics AB, Umera, Sweden), using quantified data as X variables. PCA (Principal component analysis) allowed to verify the quality of data and to identify outliers. PLS-DA (Partial Least Square-Discriminant analysis) models were then built using the tumor models as Y classes. PLS-DA models are calculated in order to minimize intra-group variance and to maximize inter-group variance. R2Y and Q2 are produced to evaluate respectively the quality of fitting and the predictability of the statistical model. R2Y>0.5 and Q2>0.5 are considered as good models.

Results/Discussion. PCA analysis allows the separation of the 3 glioma models and their respective contralateral hemisphere (Fig.1). The PLS-DA statistical model built with ex vivo data shows a clear separation between the 3 glioma models (Fig.2A) whereas it is less obvious with in vivo data (Fig.2B). Metabolites that are statistically different between glioma models (data not shown) are also the most important for the separation of classes in the PLS-DA model (Fig.3). However, some metabolites such as NAA, lactate and betaine that are modulated in tumor tissue contribute less to the separation of tumor models in the PLS-DA analysis. On the other hand, hypotaurine is only observed in tumor samples [4] and participates to tumor model separation, with a higher level in RG2 and F98 than in C6.



Figure 1: PCA analysis of ex vivo ¹H RMAS NMR Data (A.) show a good separation between metabolic profile of contralateral and tumor biobises. 2 outliers are found ¹H RMAS data (A.). Moreover, the model has a relatively in the contralateral region. In vivo data (B.) are more scattered but contralateral and tumor profiles remain separated.

Figure 2: PLS-DA analysis using the 3 glioma models as Y classes shows a very good separation when using ex vivo good predictability (Q2>0.5). The in vivo PLS-DA model (B.) has R2Y and O2 values <0.5 and is then less reliable.

Figure 3: Variable Importance Plot (VIP). GABA, glutamine, myo-inositol, taurine, glutathion, total creatine and hypotaurine have VIP>1 and then are the most important for separation of the 3 tumour models, using ex vivo ¹H RMAS data (A.) and in vivo data (B.), total creatine, glutamate, myo-inositol, total choline and GABA have the highest VIP.

Conclusion: A robust PLS-DA model (high Q2 values) could be built with ¹H RMAS MRS quantitative data despite inter-individual variability. References: [1] Opstad et al., J. Mag. Reson. Imaging, 2010; [2] Holmes et al., NeuroRx, 2006; [3] Ratiney et al., NMR biomed., 2005; [4] Rémy C. et al., J Neurochem. 1994.