

# Generalized multivariate Student distributions : application to tumor characterization with multiparametric MRI

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① Study context

② Theoretical model

③ Finalized part

④ Current works

⑤ Future work

# Goal : improving brain tumor characterization

Multiple-scale  
distributions  
& MRI clustering

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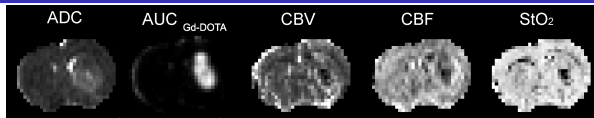
Study context

Theoretical  
model

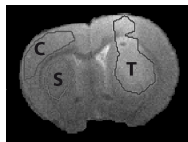
Finalized part

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- 5 physiological parameters
  - ▶ **ADC** : apparent diffusion coefficient
  - ▶ **CBV** : cerebral blood volume
  - ▶ **CBF** : cerebral blood flow
  - ▶ **AUC** : vessel permeability
  - ▶ **StO<sub>2</sub>** : tissue oxygen saturation
- 4 brain tumor models :
  - ▶ **C6, F98** (N. Coquery)
  - ▶ **C6, 9L** (B. Lemasson)
- 3 regions of interest :
  - ▶ **Cortex, Striatum**
  - ▶ **Tumor**



→ voxels clustering based on a mixture of heavy-tailed distributions (generalized multivariate Student distributions)

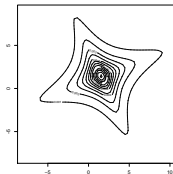
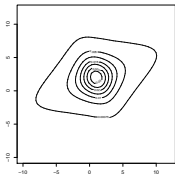
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Let  $\mathbf{y} \in \mathbb{R}^L$  a random variable of density :

$$p(\mathbf{y}) = \int_{\mathbb{R}^L} \mathcal{N}_L(\mathbf{y}; \boldsymbol{\mu}, \mathbf{D}\boldsymbol{\Delta}^{-1}\mathbf{A}^{-1}\mathbf{D}^t) f_{\boldsymbol{\Delta}}(\mathbf{w}) d\mathbf{w}$$

where  $\mathcal{N}_L(\cdot; \boldsymbol{\mu}, \mathbf{T}^{-1})$  denotes the L-dimensional Gaussian distribution with mean  $\boldsymbol{\mu}$  and precision matrix  $\mathbf{T} = \mathbf{D}\mathbf{A}\mathbf{D}^t$  ( $\mathbf{D}$  eigenvectors,  $\mathbf{A} = \text{diag}[\mathbf{a}_1, \dots, \mathbf{a}_L]$  eigenvalues).

$\boldsymbol{\Delta} = \text{diag}(\mathbf{w}_1, \dots, \mathbf{w}_L) \in \mathbb{R}^L$  is a diagonal weight matrix with density  $f_{\boldsymbol{\Delta}}$ .



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- ▶ detection and exclusion of atypical data :
  - choice of the number of classes with the Bayesian Information Criterion
  - unsupervised clustering of tumors
- ▶ tumor dictionary :
  - creation of tumor signatures
  - tumor prediction
  - dictionary relevance with a leave-one-out process
- ▶ theoretical development of a Variational Bayesian Approximation for the automatic choice of the number of classes within the mixture

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- ▶ improve speed and accuracy of numerical estimation of orthogonal matrices
- ▶ obtain an efficient parallelization of the R code
- ▶ rewrite the R code into (parallelized) C++ code
- ▶ choose priors for the decomposition of covariance matrix inside the Variational Bayesian Approximation

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- ▶ implementation of the Variational Bayesian Approximation
- ▶ clustering on whole brain data and raw scanner data
- ▶ taking to account spatial dependency with a hidden Markov field
- ▶ test of the parameters sensitivity

**Thanks !**

