

# Data Evaluation Algorithms: Bayesian DOSY and ROSY Transforms (BDT and BRT)



MESTRELAB RESEARCH  
NMR Solutions

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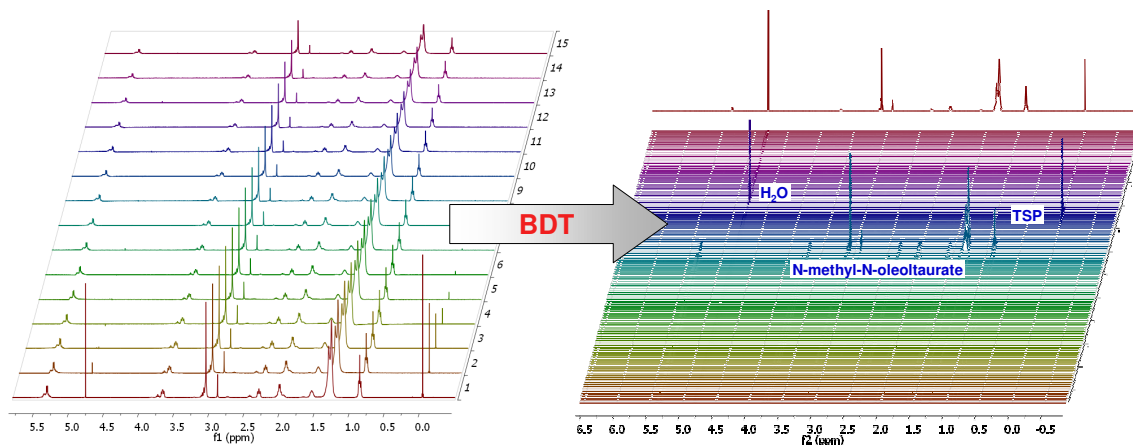
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## BDT & BRT

A novel Bayesian-based approach to DOSY & ROSY Transformation

A well known problem with several types of **multi-array NMR techniques** regards the best visual presentation of the associations between physico-chemical parameters and individual spectral peaks. Typical examples are **DOSY**, in which the desired second parameter is the diffusion coefficient, and **ROSY**, for which it is the relaxation time. In such cases, the transformation of the original data set to a suitable final 2D graph (the DOSY or ROSY transform, respectively) is conceptually and mathematically difficult to manage. We propose a **Bayesian approach** which is computationally very efficient and physically eminently meaningful, and gives very satisfactory and artifact-free results. Applied specifically to the DOSY and ROSY data sets, it leads to what we call the **BDT** (Bayesian DOSY transform) and **BRT** (Bayesian ROSY transform) algorithms



BDT of an aqueous solution of potassium N-methyl-N-oleolate (a surfactant) with TSP at 23 C  
The original Varian FID file has been obtained from the *VARIAN NMR USER GROUP LIBRARY* (submitted by Brian Antalek as a sample for this DECRA algorithm)

## The Bayesian approach in a nutshell

Let  $z$  be the gradient-dependent decay parameter associated with each DOSY spectrum and  $d$  the diffusion coefficient associated with a spectral line at some frequency  $f$ . Then the essence of a Bayesian approach to the problem of transforming the experimental  $S[f,z]$  map to a probabilistic  $[f,d]$  map is embodied in the following question:

If one selects an arbitrary point in the  $[f,d]$  map, can one assign to it an amplitude and a probability which gives an idea of how compatible (congruent) that particular point in the  $[f,d]$  map is with the experimental  $[f,z]$  map?

Having fixed  $f$  and  $d$ , one has only one parameter to fit, namely the decay-curve amplitude  $a_k(d)$ ,  $k$  being the data index corresponding to the frequency  $f$ . This is a linear, one parameter fit which can be done explicitly. Assuming the decay curve

$$y_{k\alpha} = a_k(d) \exp(-dz_{k\alpha}), \text{ where } \alpha \text{ is the index within the } z\text{-set,}$$

the value of  $a_k$  which minimizes the total square deviation  $\sum_{\alpha} (y_{k\alpha} - S_{k\alpha})^2$  is

$$a_k(d) = [\sum_{\alpha} S_{k\alpha} \exp(-dz_{k\alpha})] / [\sum_{\alpha} \exp(-2dz_{k\alpha})].$$

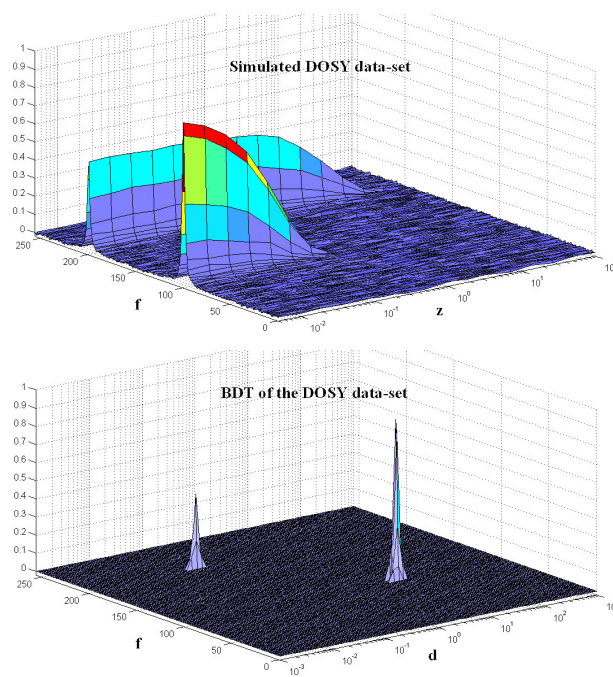
Even more important than the optimal amplitude is the Bayesian weight  $w_k$  associated with it. Denoting as  $\sigma$  the standard deviation of the experimental noise, the multiplicative contribution of the experimental data point  $S_{k\alpha}$  to  $w_k$  is equal to  $\exp(-(y_{k\alpha} - S_{k\alpha})^2 / (2\sigma^2))$ . Considering all  $z$ -points, we have

$$w_k = \prod_{\alpha} \exp(-(y_{k\alpha} - S_{k\alpha})^2 / \sigma^2) = \exp(-[\sum_{\alpha} (y_{k\alpha} - S_{k\alpha})^2] / \sigma^2),$$

Whose maximum  $w_k(d)$  coincides with the  $a_k(d)$  obtained by least-squares. Notice that, in principle, any  $d$  is legitimate, but the value of  $w_k(d)$  is appreciable only when  $d$  is close to the 'correct' value. When the chosen  $d$  differs from the correct one, there are large systematic deviations (residues) and  $w(d)$  becomes very small.  $w_k(d)$  is the proper 'vertical' value to be assigned to the  $(f,d)$  point in the DOSY 2D-plot, apart from the final normalization.

The final normalization, which is a characteristic step of any Bayesian evaluation, is very simple. Once the  $w_k(d)$  values were calculated for all possible  $d$ -values, they are scaled so as to make their sum (vertical projection) equal to the intensity of the original spectrum at  $z = 0$  (usually defined by  $S_{k0}$ ).

## Illustration of the algorithm using Matlab and simulated data



## Conclusions

Compared to other approaches, the **Bayesian method** appears extremely promising. It automatically avoids having exact, unnatural zeros anywhere in the resulting map since *every point of the 2D map has a well defined value of statistical congruence with the data*. Moreover, the **BDT maps** show 'normal' line widths in the  $f$ -direction, correctly positioned and resolved peaks in the  $d$ -direction and **quantitatively correct horizontal and vertical projections** – a combination difficult to achieved by any other means.

The approach can be easily extended to non-exponential cases arising from **overlapping lines** and, like all Bayesian methods, incorporate additional information available from other sources (*a-priori* knowledge). Likewise, it is possible to place a statistical premium on alignment of spectral peaks along horizontal lines in the  $[f,d]$  plot.