



Anno 2012
Protocollo: RBFR12JF2Z



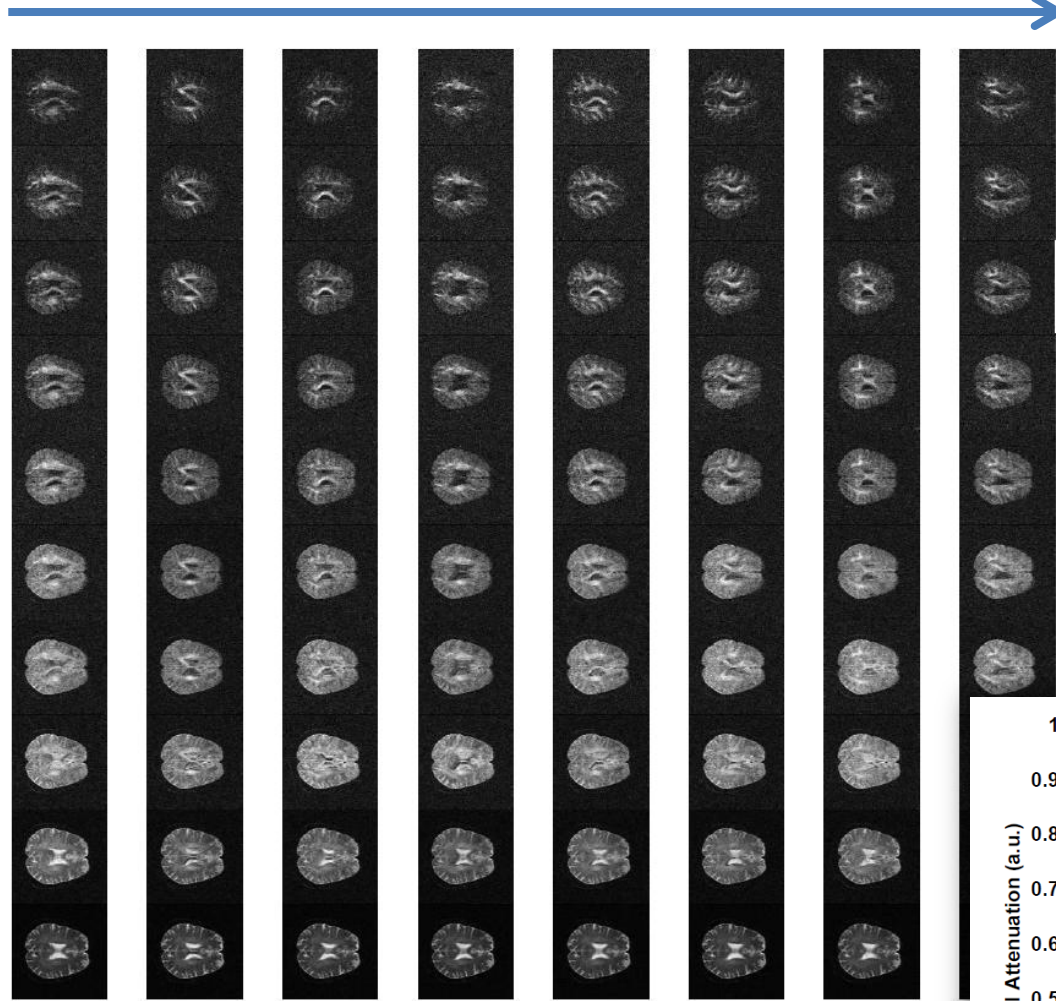
Unità A2: Imaging NMR, Roma

Accelerazione della ricostruzione di immagini
NMR ottenute con il metodo del contrasto in
diffusione

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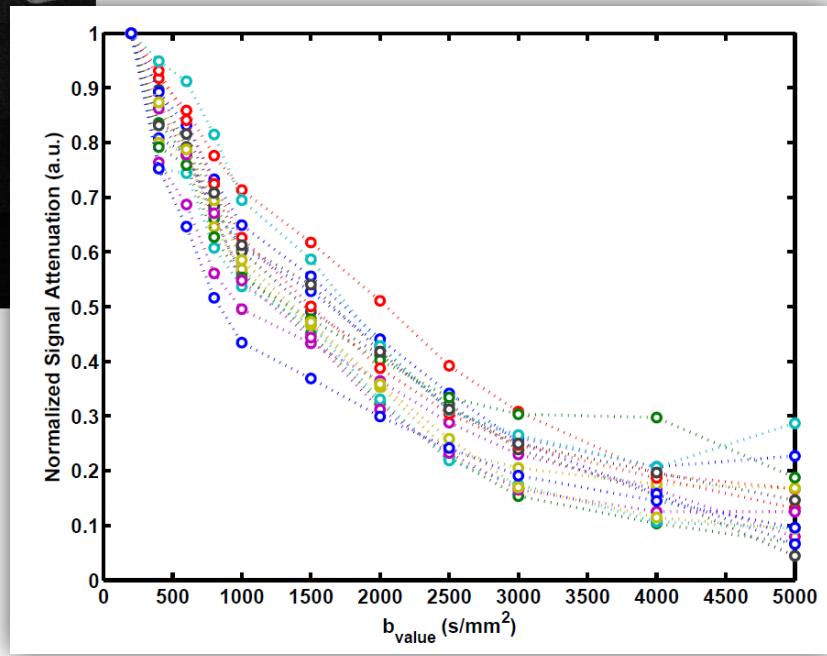
direction



Typical NMR
imaging
post-processing

128x128 pixel image
x
number of slice (~10-40)
x

number of diffusion gradient directions (≥ 15)



Cumulant expansion approach: kurtosis tensor measurement

$$S(b) \propto e^{-D_{app} b + \frac{1}{6} K_{app} [D_{app} b]^2 + \dots}$$



$$S(b) = S(0) \exp \left[-b \sum_{i,j} n_i n_j D_{ij} + \frac{1}{6} b^2 \left(\frac{1}{3} \sum_i D_{ii} \right)^2 \sum_{i,j,k,l} n_i n_j n_k n_l W_{ijkl} + O(b^3) \right]$$

$N \geq 15$ independent measures to get W_{ijkl}

$$K(\mathbf{n}) = \frac{\bar{D}^2}{[D(\mathbf{n})]^2} \sum_{i,j,k,l=1}^3 n_i n_j n_k n_l W_{ijkl}.$$



$$MK = \frac{1}{N} \sum_{i=1}^N K(\mathbf{n}_i)$$

$$\begin{aligned}\bar{K} = & F_1(\lambda_1, \lambda_2, \lambda_3)\tilde{W}_{1111} + F_1(\lambda_2, \lambda_1, \lambda_3)\tilde{W}_{2222} \\ & + F_1(\lambda_3, \lambda_2, \lambda_1)\tilde{W}_{3333} + F_2(\lambda_1, \lambda_2, \lambda_3)\tilde{W}_{2233} \\ & + F_2(\lambda_2, \lambda_1, \lambda_3)\tilde{W}_{1133} + F_2(\lambda_3, \lambda_2, \lambda_1)\tilde{W}_{1122},\end{aligned}$$

$$K_{\parallel} = \frac{(\lambda_1 + \lambda_2 + \lambda_3)^2}{9\lambda_1^2} \tilde{W}_{1111},$$

$$\begin{aligned}K_{\perp} = & G_1(\lambda_1, \lambda_2, \lambda_3)\tilde{W}_{2222} + G_1(\lambda_1, \lambda_3, \lambda_2)\tilde{W}_{3333} \\ & + G_2(\lambda_1, \lambda_2, \lambda_3)\tilde{W}_{2233},\end{aligned}$$

$$G_1(\lambda_1, \lambda_2, \lambda_3) = \frac{(\lambda_1 + \lambda_2 + \lambda_3)^2}{18\lambda_2(\lambda_2 - \lambda_3)^2} \left(2\lambda_2 + \frac{\lambda_3^2 - 3\lambda_2\lambda_3}{\sqrt{\lambda_2\lambda_3}} \right),$$

$$G_2(\lambda_1, \lambda_2, \lambda_3) = \frac{(\lambda_1 + \lambda_2 + \lambda_3)^2}{3(\lambda_2 - \lambda_3)^2} \left(\frac{\lambda_2 + \lambda_3}{\sqrt{\lambda_2\lambda_3}} - 2 \right).$$

$$F_1(\lambda_1, \lambda_2, \lambda_3) \equiv \frac{(\lambda_1 + \lambda_2 + \lambda_3)^2}{18(\lambda_1 - \lambda_2)(\lambda_1 - \lambda_3)} \left[\frac{\sqrt{\lambda_2\lambda_3}}{\lambda_1} R_F\left(\frac{\lambda_1}{\lambda_2}, \frac{\lambda_1}{\lambda_3}, 1\right) + \frac{3\lambda_1^2 - \lambda_1\lambda_2 - \lambda_2\lambda_3 - \lambda_1\lambda_3}{3\lambda_1\sqrt{\lambda_2\lambda_3}} R_D\left(\frac{\lambda_1}{\lambda_2}, \frac{\lambda_1}{\lambda_3}, 1\right) - 1 \right],$$

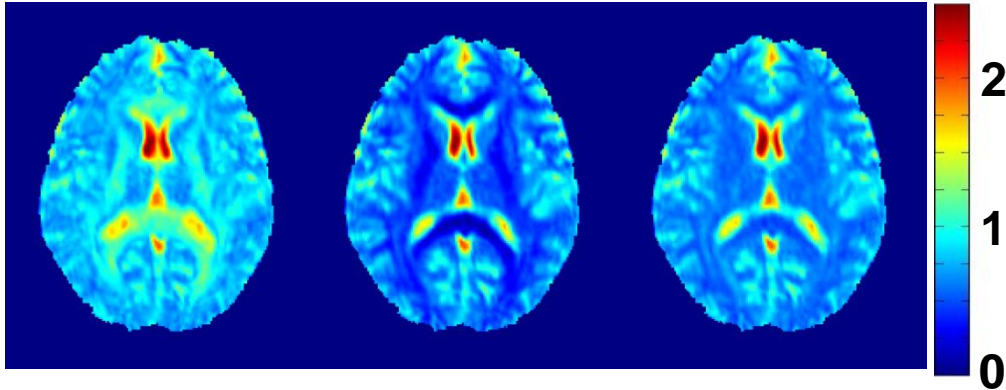
and

$$F_2(\lambda_1, \lambda_2, \lambda_3) \equiv \frac{(\lambda_1 + \lambda_2 + \lambda_3)^2}{3(\lambda_2 - \lambda_3)^2} \left[\frac{\lambda_2 + \lambda_3}{\sqrt{\lambda_2\lambda_3}} R_F\left(\frac{\lambda_1}{\lambda_2}, \frac{\lambda_1}{\lambda_3}, 1\right) + \frac{2\lambda_1 - \lambda_2 - \lambda_3}{3\sqrt{\lambda_2\lambda_3}} R_D\left(\frac{\lambda_1}{\lambda_2}, \frac{\lambda_1}{\lambda_3}, 1\right) - 2 \right].$$

LD

RD

MD ($\mu\text{m}^2/\text{ms}$)



Conventional DTI

$128 \times 128 \times 32 \times 15$

=

$7.864320 \cdot 10^6$

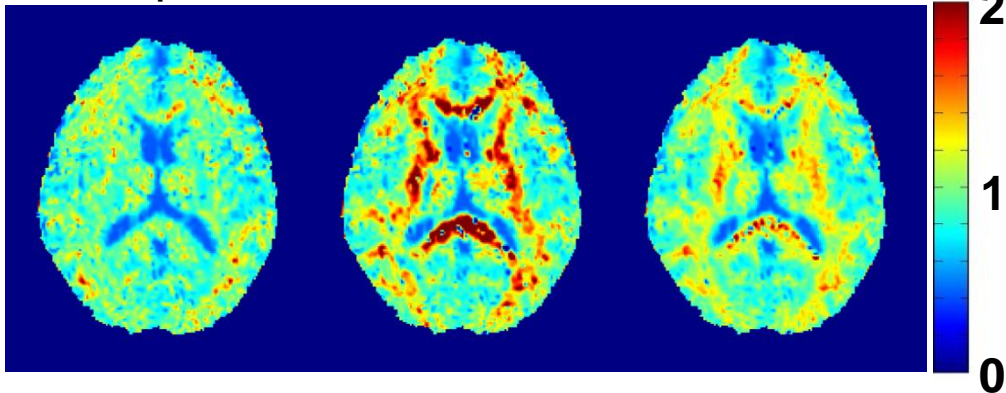
linear systems to solve

~ **15 s** on CPU*

K_{par}

K_{ort}

\bar{K}



Kurtosis tensor imaging

$128 \times 128 \times 32 \times 15$

=

$7.864320 \cdot 10^6$

non-linear fit to perform

~ **7200 s** on CPU*

* 8 threads on an Intel Xeon E5-2609 CPU at 2.4 GHz

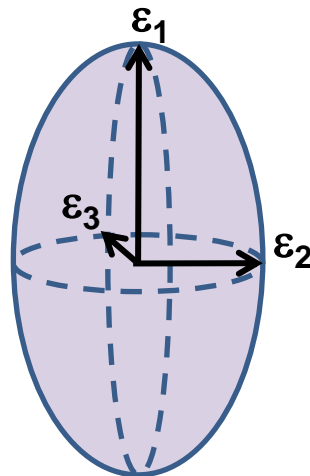
Stretched exponential model

$$S(\mathbf{b}) = S(0)e^{-A_1 (b_1^*)^{\gamma_1} - A_2 (b_2^*)^{\gamma_2} - A_3 (b_3^*)^{\gamma_3}}$$

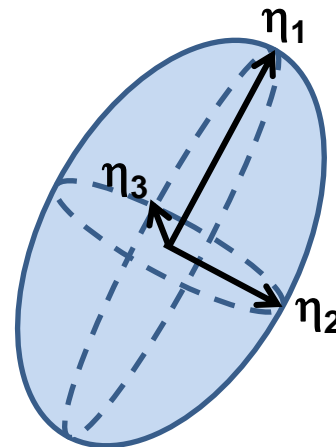
$$b_i^* = \vec{b} \cdot \vec{\eta}_i$$

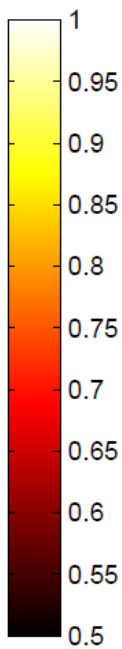
**Anomalous diffusion
reference frame**

**DTI
reference frame**



**Anomalous
Diffusion
reference frame**





$M\gamma$



γ_{par}



γ_{ort}



Conventional DTI

$128 \times 128 \times 32 \times 15$

=

$7.864320 \cdot 10^6$

linear systems to solve

~ **15 s** on CPU*

Stretched exponential imaging

$128 \times 128 \times 32 \times 15$

=

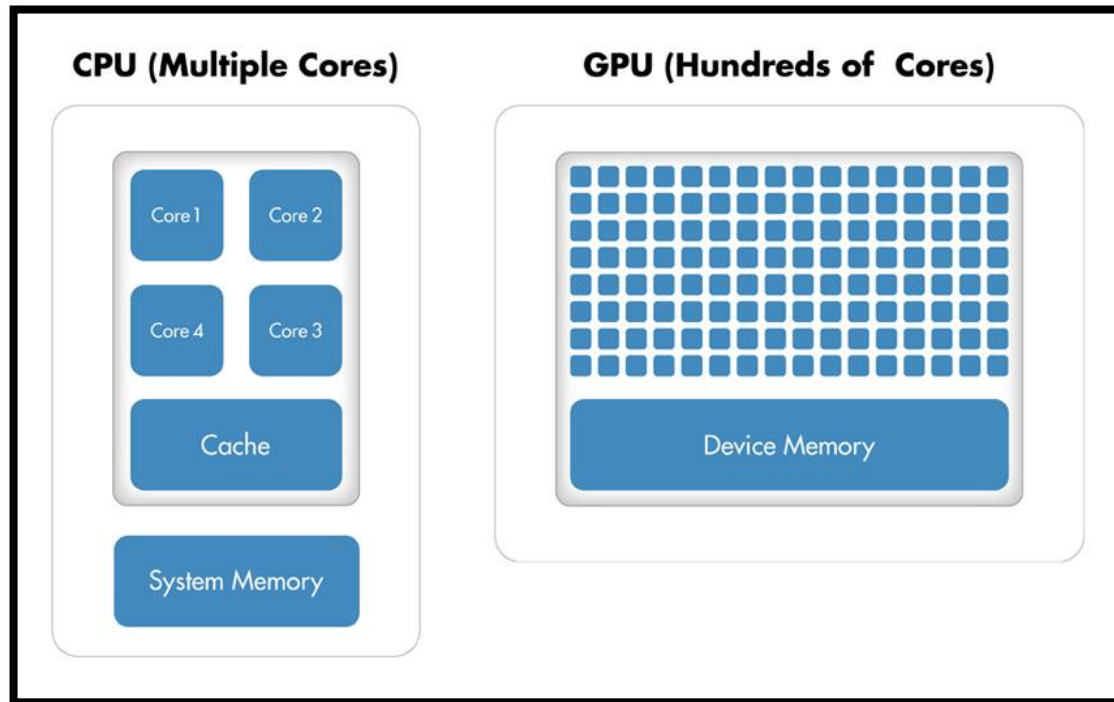
$7.864320 \cdot 10^6$

non-linear fit to perform

~ **6600 s** on CPU*

* 8 threads on an Intel Xeon E5-2609 CPU at 2.4 GHz

When GPU can accelerate an application



- **Computationally intensive** — The time spent on computation significantly exceeds the time spent on transferring data to and from GPU memory.
- **Massively parallel** — The computations can be broken down into hundreds or thousands of independent units of work.

Computationally intensive

Kurtosis tensor imaging

non-linear fit

~7 ms per pixel

vs

~ 10-80 ns to read data
from memory

Stretched exponential imaging

non-linear fit

~6 ms per pixel

vs

~10-80 ns to read data from
memory

Massively parallel

Kurtosis tensor imaging

~ 10^6 - 10^7 independent non-
linear fit

Stretched exponential imaging

10^6 - 10^7 independent non-
linear fit

Exemple: GPU acceleration of the solution of a wave equation in 2D*

$$\frac{\partial^2 u}{\partial t^2} = \frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2}$$

Solution in space

Chebyshev spectral method



Solution in time

Second-order central finite difference method (leap-frog method)

2048 x 2048 grid x 50 iterations ~ **80 s** on CPU\$

* from:

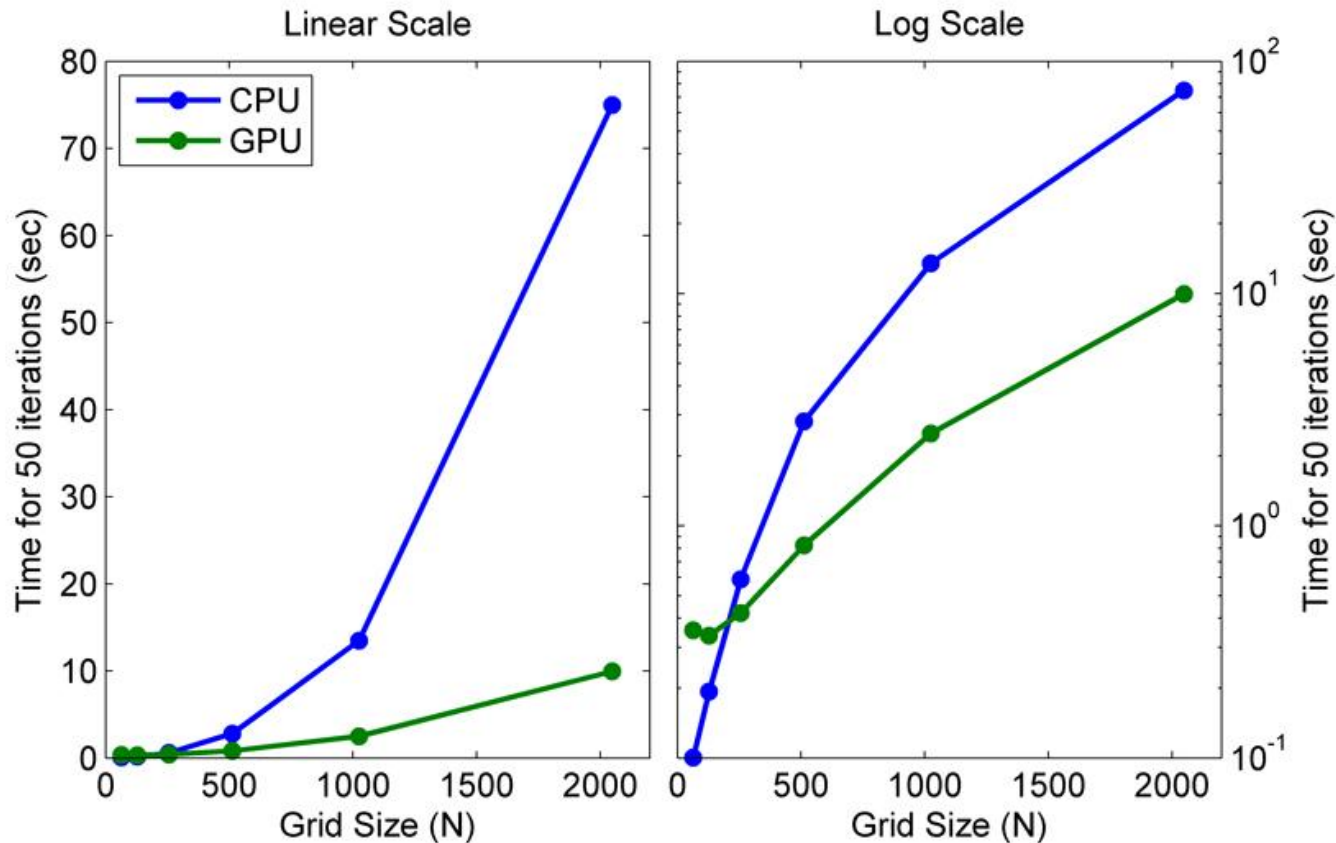
<http://www.mathworks.it/company/newsletters/articles/gpu-programming-in-matlab.html>

\$ 12 thread on an Intel Xeon X5650 CPU at 2.66 GHz

By using GPU acceleration*

7.5x decrease in compute time

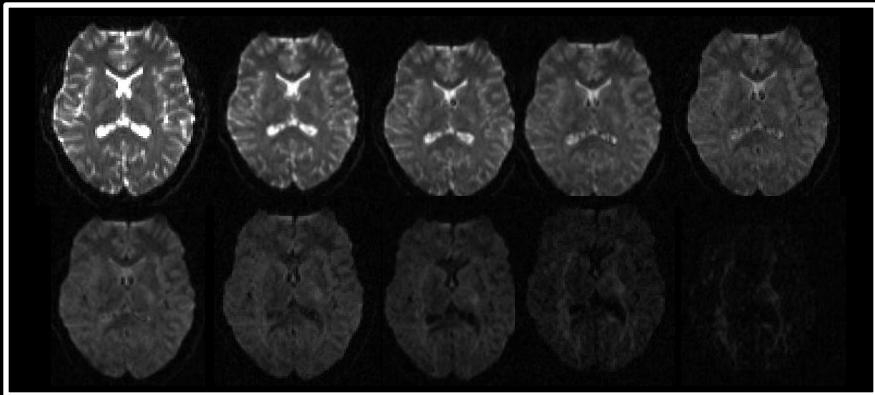
2048 x 2048 grid x 50 iterations ~ 10 s on GPU\$



* from:

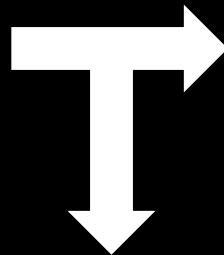
<http://www.mathworks.it/company/newsletters/articles/gpu-programming-in-matlab.html>

\$ NVIDIA Tesla C2050 GPU with 448 CUDA cores at 1.15 GHz

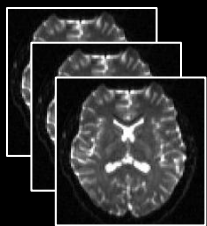
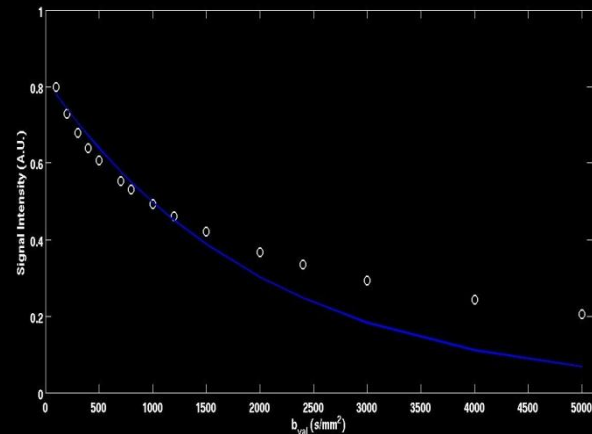


Pre-PROCESSING
(fsl, SPM)

PROCESSING
& ANALYSIS
(Matlab)



NON- GAUSSIAN DIFFUSION

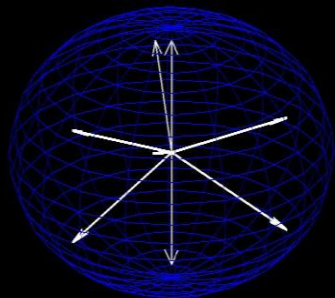


RAW DATA



MR scanner

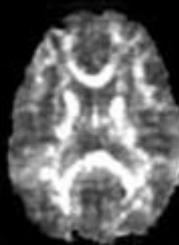
STRETCHED EXPONENTIAL
MODEL (6 directions)



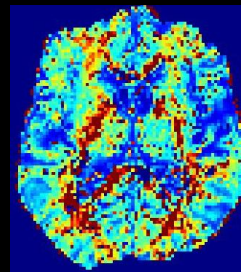
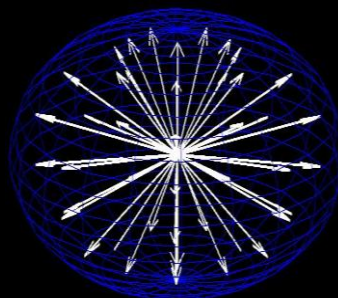
$M\gamma$



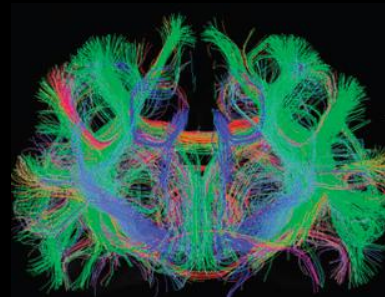
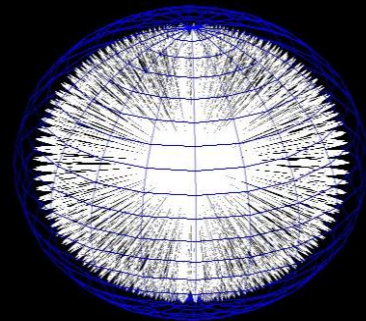
γA



KURTOSIS TENSOR
MODEL (15 directions)



DIFFUSION SPECTRUM
MODEL (515 directions)



- Development of highly parallelized non-conventional NMR image processing algorithms;

DONE

- Non-conventional NMR image processing algorithms code optimization;

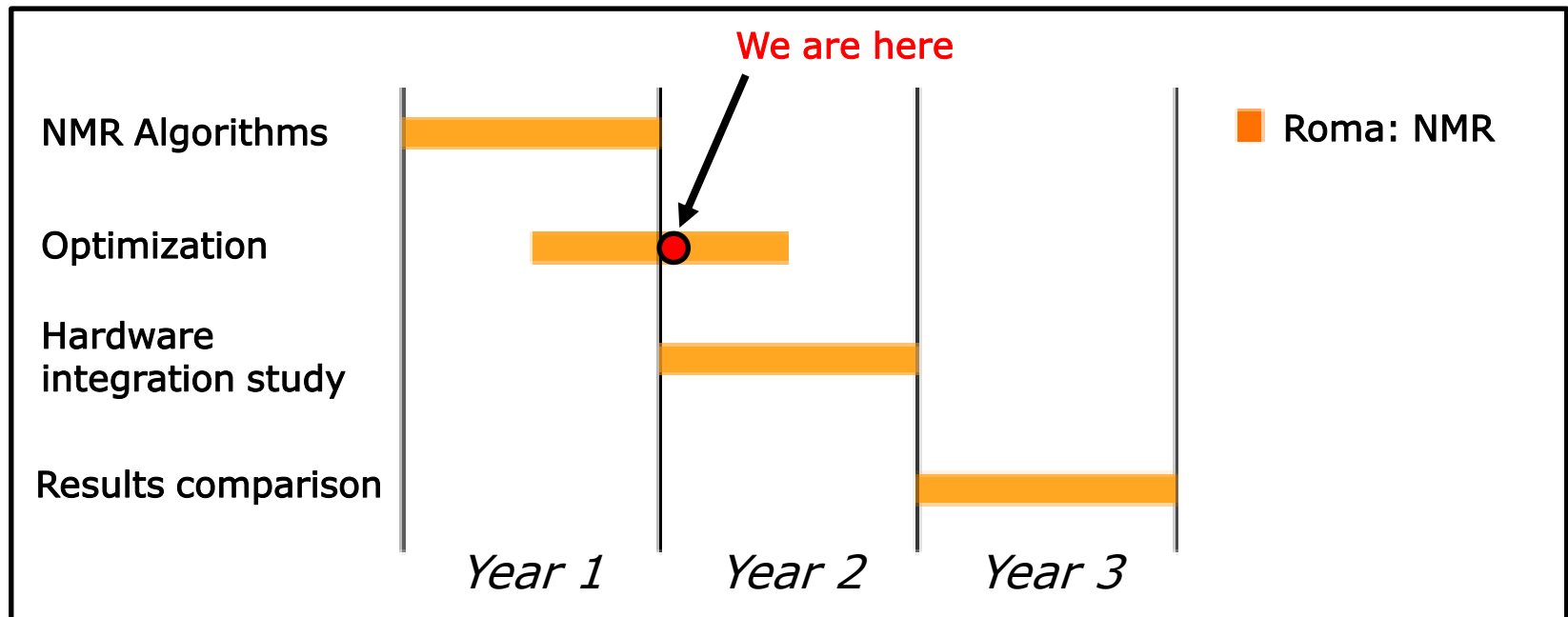
PARTIALLY DONE

- Algorithms porting on GPU and hardware integration study;

LESS THAN 1 YEAR REQUIRED

- Results comparison.

LESS THAN 1 YEAR REQUIRED



- Development of highly parallelized non-conventional NMR image processing algorithms;

DONE

- Non-conventional NMR image processing algorithms code optimization;

PARTIALLY DONE

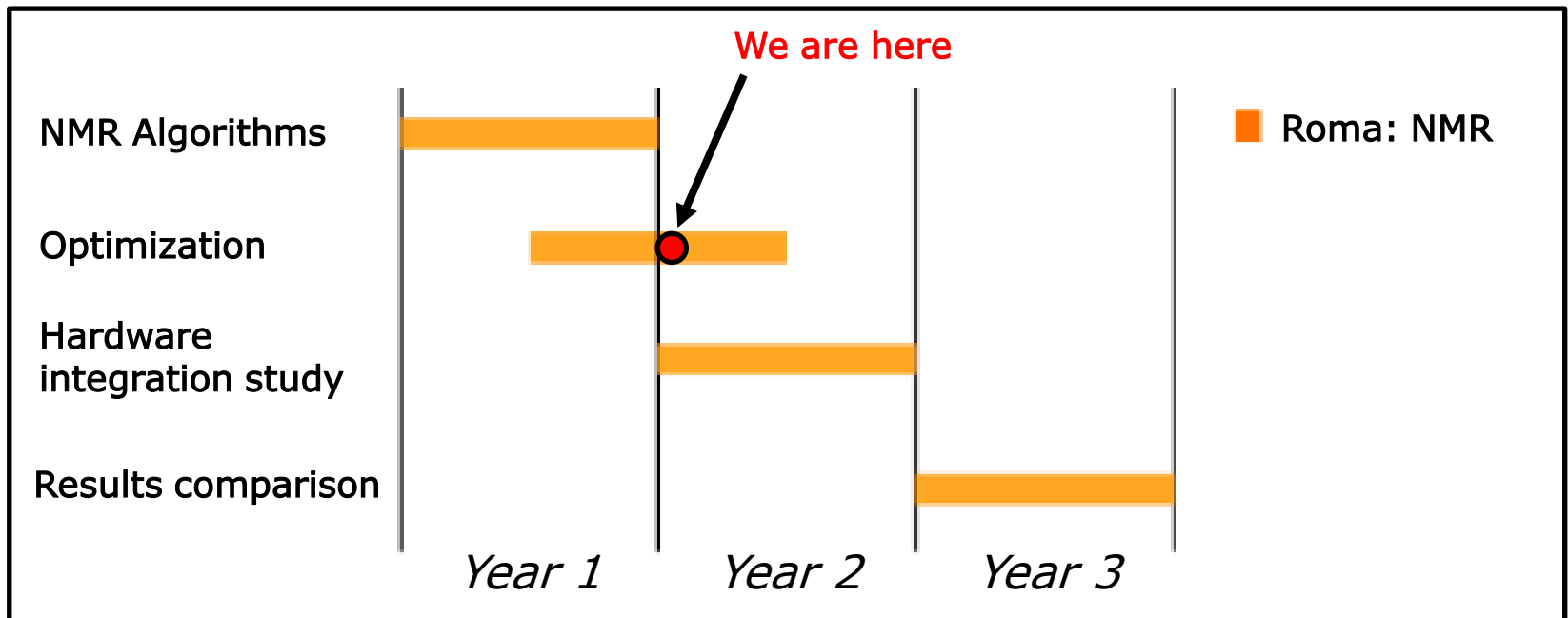
- Algorithms porting on GPU

and hardware integration study;

LESS THAN 1 YEAR REQUIRED

- Results comparison.

LESS THAN 1 YEAR REQUIRED



- Non-conventional NMR image processing algorithms code optimization;

Algorithms porting on GPU

MATLAB



CUDA

- Easier and faster (all our codes are written for Matlab)
- Bypass the study of the specific hardware architecture
- Only a single GPU can be used

- Slower (all our codes are written for Matlab)
- Require the study of the specific hardware architecture
- Multiple GPUs can be used

