

# MRI IVIM Parameter Estimation by Model weighting

fitting pixel selection

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# Outline

- 1 MRI Diffusion
  - monoexponential decay
- 2 MRI Diffusion: IVIM model
  - IVIM
- 3 The project: parametric maps
  - Methods
- 4 Results
  - Anatomical images
  - Results

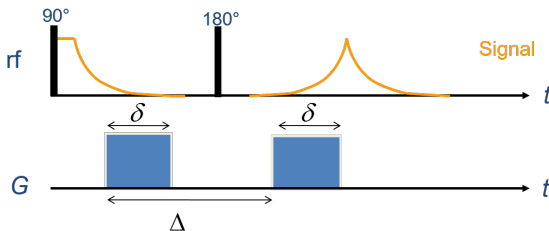


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# NMR measurement of diffusion



Stejskal-Tanner equation (1965)

$$\frac{S(G, \delta, \Delta)}{S(0, \delta, \Delta)} = \exp(-\gamma^2 G^2 \delta^2 D (\Delta - \delta/3)) = \exp(-bD) \quad (1)$$

$b$ -value depends on gradient intensity and shape



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# Bi-exponential and more

$$\text{IVIM model} \rightarrow \frac{S(b)}{S(0)} = (1 - f)e^{-bD} + f \cdot e^{-bD^*}$$

$$\frac{S(b)}{S(0)} \approx (1 - f)e^{-bD}$$

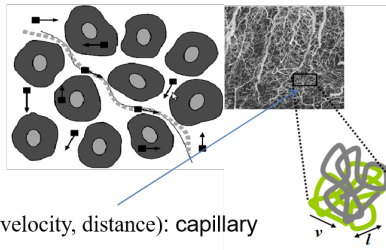
$D$ : Diffusion

$D^*$ : pseudo-diff, perfusion

$f$ : Perfusion frac. ( $\#D^*$  vs.  $\#D^*$ )

$K$ : Kurtosis, non Gaussian models.

$$\ln\left(\frac{S(b)}{S(0)}\right) = -f - bD + b^2 D^2 K / 6$$



$D^*$  (velocity, distance): capillary

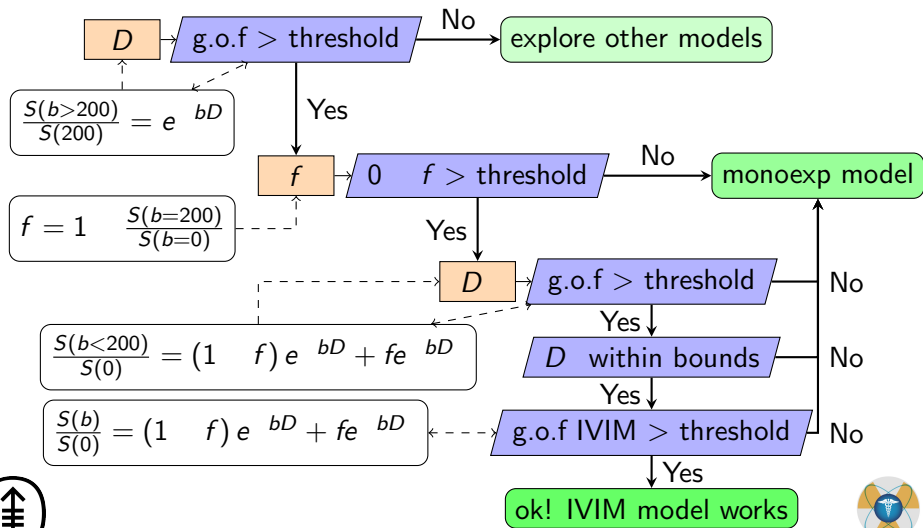


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# Decisions for parametric error category mapping exclusion



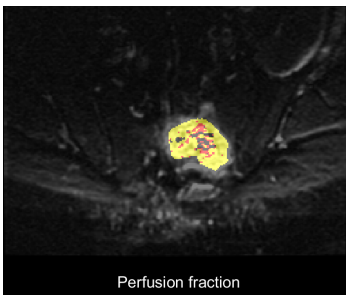
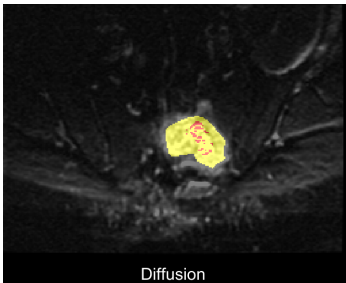


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# Anatomical images



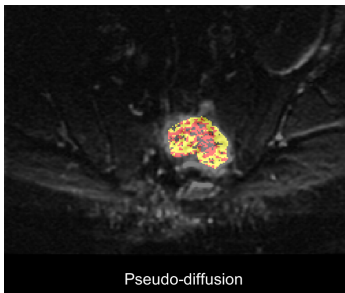
$D$  (top left),  $f$  (bottom left) and  $D$  (bottom right) spatial distribution

Red only physical thresholds are applied:

- $0 < D < 0.03\text{mm}^2/\text{s}$
- $0 < f < 1$
- $0 < D < 3\text{mm}^2/\text{s}$

Yellow pixels fill selection criteria based on g.o.f (normalized maximum absolute error [0 to 1]) and thresholds:

- $0 < D < 5$  std dev  $< 0.03\text{mm}^2/\text{s}$
- $0 < f < 5$  std dev  $< 1$
- $0 < D < 5$  std dev  $< 3\text{mm}^2/\text{s}$

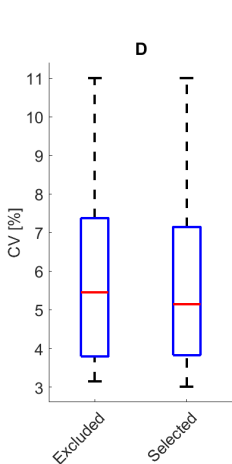


# Outline

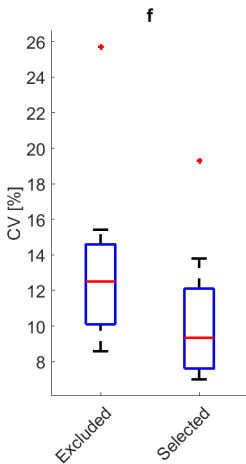
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# Coefficient of variance of all patients

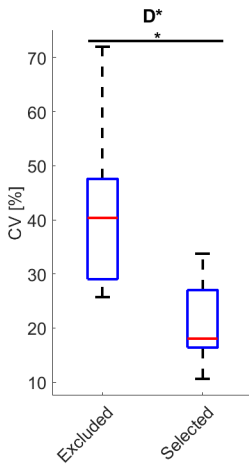


p-value 0.989



p-value 0.183

SO CLOSE...



p-value 0.001



P < 0.05

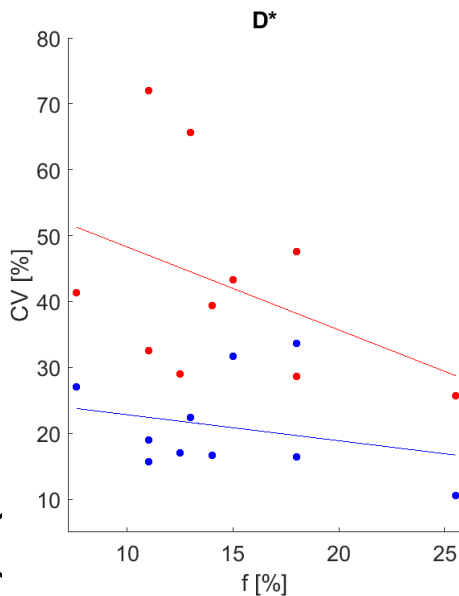
Any questions ?



# Questions?



# CV of pseudo diffusion dependence on perfusion fraction



CV of  $D$  vs.  $f$  average. CV of excluded data (red) and selected data (blue). Experimental data (full circle) is fitted to a linear regression.

$$S(b) = S(0) \left( (1 - f) e^{-bD} + f e^{-bD} \right)$$

Data	Excluded	Selected
Slope	-1.3 1	-0.4 0.5
Intercept	61 16	27 8
$r^2$	0.16	0.07
$p$ -value	0.25	0.47
Pearson	-0.40	-0.26



# Summary

Simulations test fitting robustness: SNR, precision and accuracy.

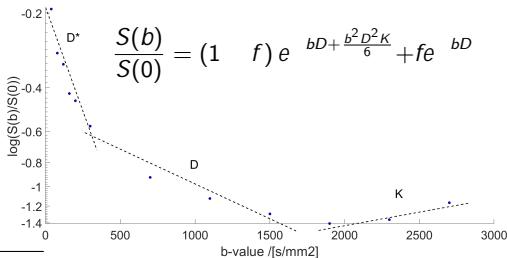
- Based on simulation, precision and accuracy depends on SNR ; real images, has a variety of SNR within the ROI.

Patient images test data robustness by measuring the CV.

- Selection of pixels leads to an enhance of the bi-exponential model data reliability.
- Psuedo-diffusion fitted values benefits from selected pixels more than diffusion and perfusion fraction.
- CV of  $D^*$  dependence on  $f$ , decreases if selected pixels based on residuals are considered.

## Outlook

- Extend the diffusion model to kurtosis.



# Outlook

The Akaike information criterion (AIC) is a measure of the relative quality of statistical models for a given set of data: it provides a means for model selection.

$$AIC = N \ln(SS) + 2p$$

$N$ : number of data

$SS$ : sum of square deviance

$p$ : number of estimated parameters

$$\frac{S(b)}{S(0)} = e^{-bD} \quad (2)$$

$$\frac{S(b)}{S(0)} = (1 - f) e^{-bD + \frac{b^2 D^2 K}{6}} + f e^{-bD} \quad (3)$$

$$\frac{S(b)}{S(0)} = e^{-bD + \frac{b^2 D^2 K}{6}} \quad (4)$$

$$\frac{S(b)}{S(0)} = (1 - f) e^{-bD + \frac{b^2 D^2 K}{6}} + f e^{-bD} \quad (5)$$





# Diffusion Models

Model	type	Gaussian	Non-Gaussian: Kurtosis
Mono-exp		$\frac{S(b)}{S(0)} = \exp(-bD)$	$\frac{S(b)}{S(0)} = \exp\left(-bD + b^2 D^2 \frac{K}{6}\right)$
IVIM	Bi-exp	$\frac{S(b)}{S(0)} = (1-f)\exp(-bD) + f\exp(-bD^*)$	$\frac{S(b)}{S(0)} = (1-f)\exp\left(-bD + b^2 D^2 \frac{K}{6}\right) + f\exp(-bD^*)$
	ADC-approx	$\frac{S(b)}{S(0)} = \exp(-f - bD)$	$\frac{S(b)}{S(0)} = \exp\left(-f - bD + b^2 D^2 \frac{K}{6}\right)$



# IVIM: Piece-wise fitting

$$\frac{S(b)}{S(0)} = (1 - f) e^{-bD} + f e^{-bD^*} \quad (6)$$

## Segmented fitting

1. Fit  $D$  from high  $b$ -values; usually  $b > 200 \text{ s/mm}^2$ .

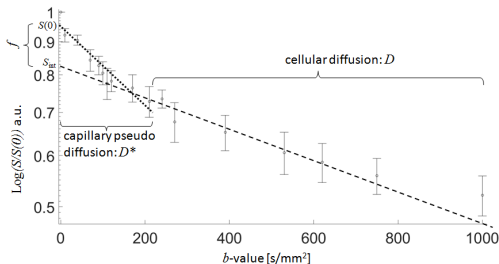
$$\frac{S(b > 200)}{S_{int}} = e^{-bD} \quad (7)$$

2. Set  $f$  from the difference of the intersection of the two slopes

$$f = 1 - \frac{S_{int}}{S(b=0)} \quad (8)$$

3. Fit  $D^*$  from small  $b$ -values; usually  $b < 200 \text{ s/mm}^2$ .

$$\frac{S(b < 200)}{S(b=0)} = (1 - f) e^{-bD} + f e^{-bD^*} \quad (9)$$



**Figure:** Perfusion fraction ( $f$ ) and signal intercept ( $S_{int}$ ) definition

