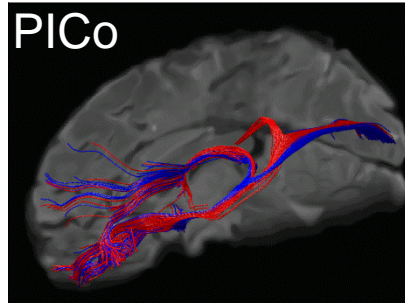
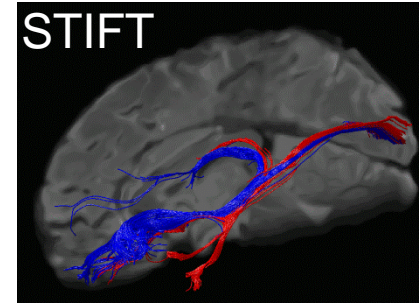




Donders Institute
for Brain, Cognition and Behaviour



PICo



STIFT

ISMRM 2010, # 4021

SWI-informed Diffusion Tensor Tractography

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Medical Centre

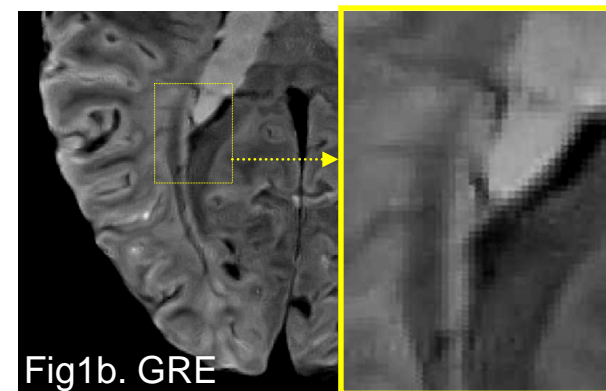
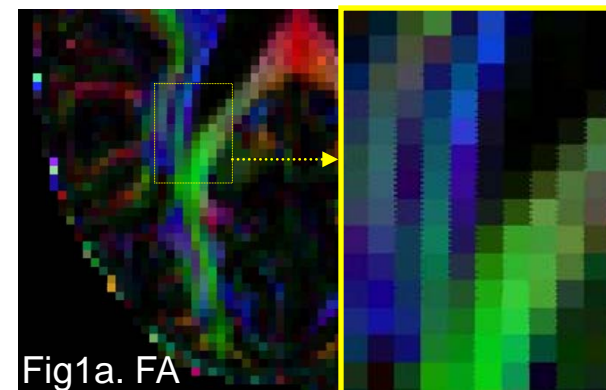
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Problem description

Fibre tractography in diffusion weighted images (DWI) suffers from partial volume effects

- Typical DWI voxels (8 ml) contain multiple tracts at e.g. tract borders (Fig.1a)
- High resolution (0.125 ml) long TE gradient echo images (GRE) also show contrast within the white matter at high field¹ (Fig.1b)

Tractography can benefit from combining the diffusion tensor with information from high resolution volumes



¹ Li et al., NI 2006



Structure Tensor Informed Fibre Tractography (STIFT)

The structure tensor (ST) is a suitable representation of scalar images to incorporate in tractography algorithms

- The ST captures local image features; the principal structure direction (PSD) is given by the first eigenvector of the ST (Fig.2)
- In anisotropic diffusion filtering the ST is used to enhance certain features of the image
- The ST calculated by edge-enhancing diffusion¹ is most useful for enhancing the sheet-like fiber bundles in the GRE image

The PSD at tract borders in GRE images is expected to be orthogonal to the principal diffusion direction (PDD) (Fig.2)

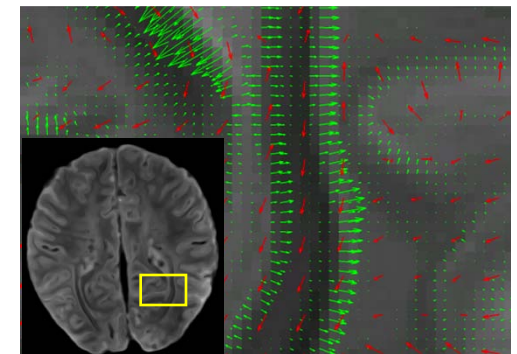


Fig2. PDD (red) & PSD (green)

¹ Weickert, PhD thesis 1996



Fibre tracking with Camino PICO¹ and STIFT

- Adapted PICO informed by the structure tensor
 - Tracking direction (TD) is found by rotating the PDD towards the plane orthogonal to the PSD proportional to its normalized first eigenvalue λ_1^{ST}

$$TD = \left\| \lambda_1^{ST} \left\| PSD \times (PDD \times PSD) \right\| + (1 - \lambda_1^{ST}) PDD \right\| \quad (\text{Eq.1})$$

- The adapted tracking direction is used in white matter only
- Seed point pairs were placed in the GRE image in adjacent voxels within and outside conspicuous fiber bundles

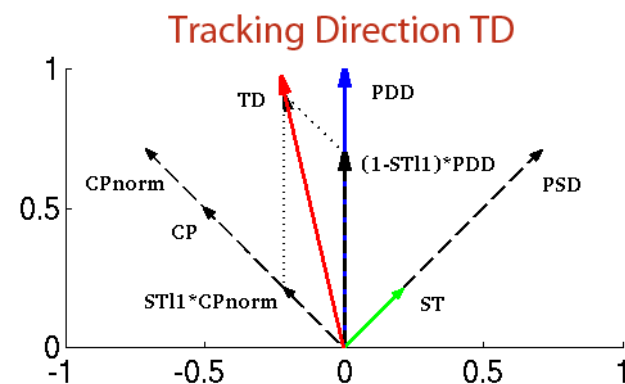


Fig3. Tracking Direction

¹ Parker et al., JMRI 2003

Optic radiation (OR)

Fig4a. PICO NB inferior view

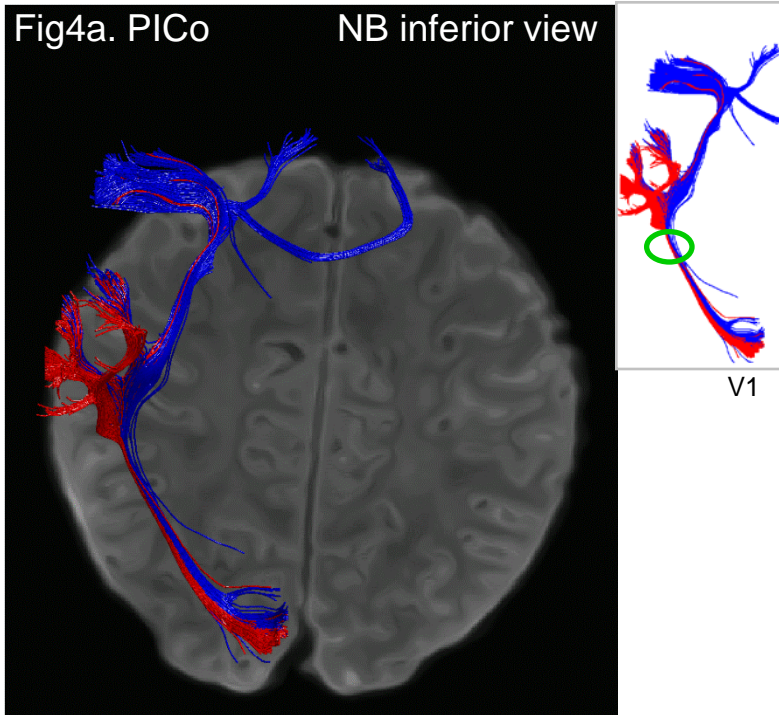
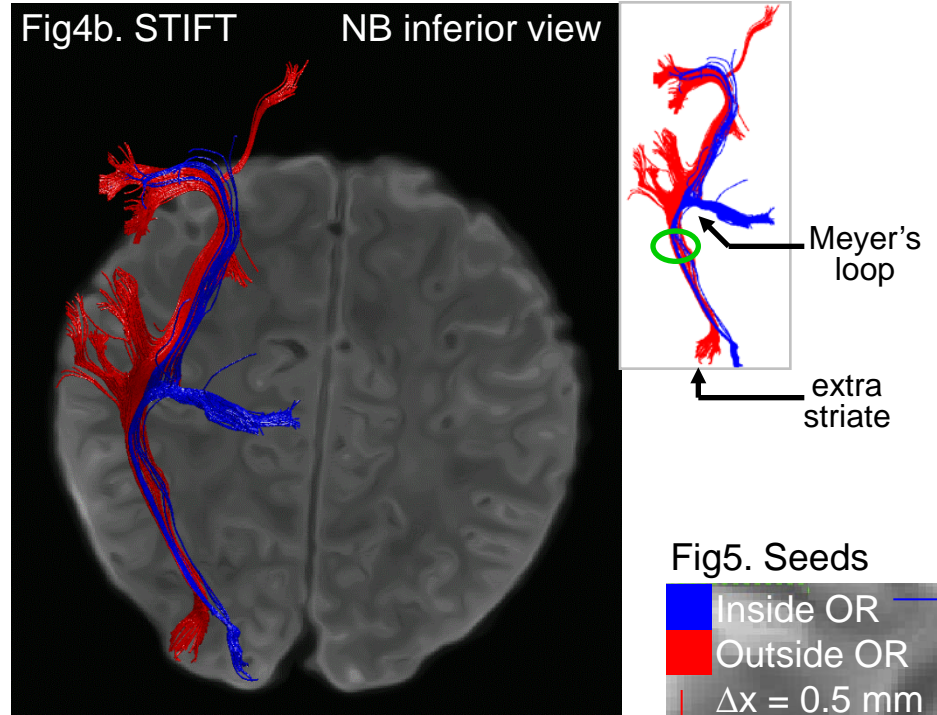



Fig4b. STIFT NB inferior view



PICO:


 Frontal¹ and temporal²
to V1

¹ Inferior occipito-frontal fasciculus (IOFF))

² Inferior longitudinal fasciculus (ILF)

³ Optic radiation (OR)

STIFT:

 Meyer's loop to V1³


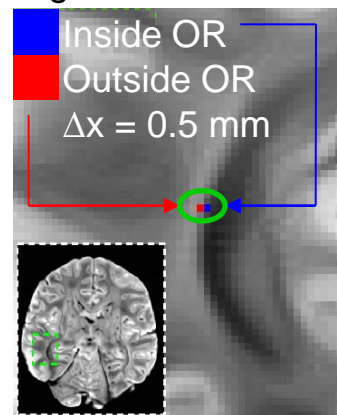
 Frontal¹ and temporal² to
extrastriate areas

Fig5. Seeds



Tapetum (TM)

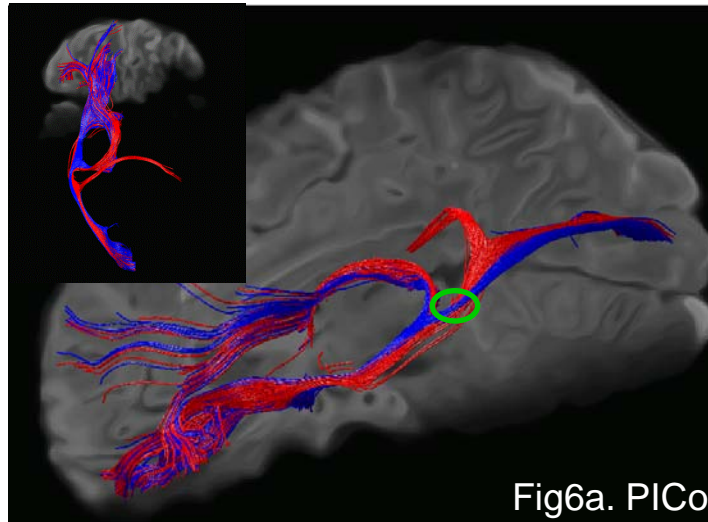


Fig6a. PICo

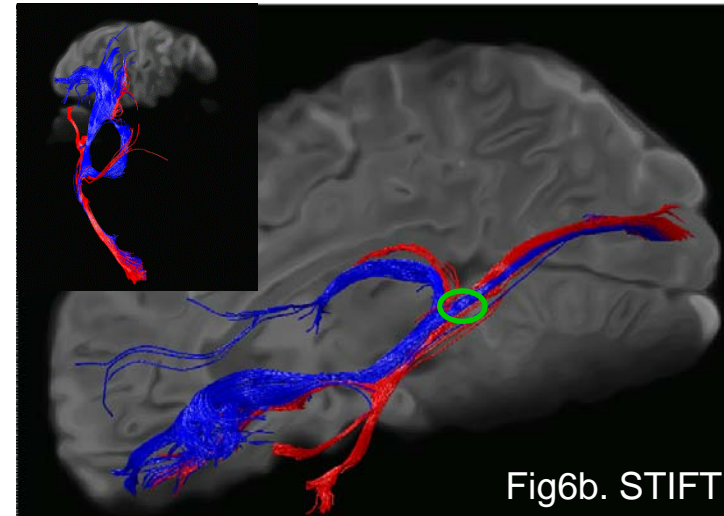


Fig6b. STIFT

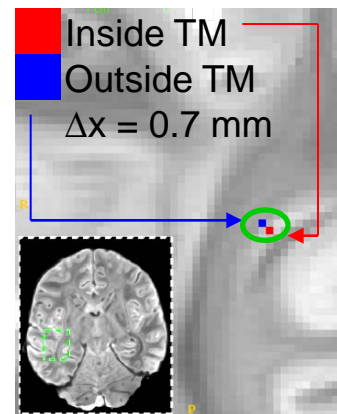
PICo:

- Frontal to V1; mixed for both seeds
- Tapetum corpus callosum specific to seed in TM

STIFT:

- Separation of frontal fibers
- Temporal tapetum specific to seed in TM

Fig7. Seeds





Anatomical specificity increased with STIFT

Closely spaced seed points in neighbouring tracts result in well separated tracts using STIFT

- From the seedpoint within the OR³, the geniculostriate pathway was tracked, while seeding just outside the OR reconstructed the associative fibers of the IOFF¹ and ILF²; PICO tracked the IOFF and ILF to V1
- Different parts of the tapetum were tracked by PICO and STIFT; STIFT showed clearly separate fiber tracts for both seeds

Large veins and iron-rich subcortical structures can affect STIFT results negatively

At specific locations STIFT can be a valuable tool to increase specificity and accuracy of fiber tracking





Acknowledgements

VIP Brain Networks



Ministerie van Economische Zaken



Thanks to Dirk-Jan Kroon for the implementation of the edge-enhancing diffusion filter





Additional material





Methods: acquisition

DWI, GRE, and T1-weighted images in two healthy volunteers:

Table 1. Acquisition parameters

* subj1/subj2

	DWI	GRE	T1
<i>sequence</i>	SE-EPI	3D FLASH	MPRAGE
<i>field strength</i>	3T	7T	3T
<i>coil</i>	32-ch array	8-ch array	32-ch array
<i>TR/TE/TI</i>	8300/95/- ms	36/25/- ms	2300/3.03/1100 ms
<i>matrix size</i>	110 x 110	448 x 336	256 x 256
<i>FOV [mm]</i>	220 x 220	224 x 168	256 x 256
<i>slice thickness [mm]</i>	2.0	0.5	1.0
<i>Resolution [mm]</i>	2.0 x 2.0 x 2.0	0.5 x 0.5 x 0.5	1.0 x 1.0 x 1.0
<i>no of slices</i>	64	208/240*	192
<i>Bandwidth</i>		120 Hz/px	
<i>flip angle [°]</i>		15	
<i>diffusion volumes</i>	7 @ b = 0 s/mm ²	x	x
	61 @ b = 1000 s/mm ²	x	x
<i>acquisition time</i>	~10 min	~20 min	~10 min





Methods: preprocessing (1)

1. DWI artefact detection & realignment with PATCH¹
2. Brain extraction with FSL
3. Bias field correction with FSL
4. Coregistration
 - a) GRE → T1; with FSL using FAST-based weighting volumes
 - b) Mean b=0 DWI → T1; with constrained warping²
5. GRE image structure tensor with edge-enhancing diffusion³
6. Vessel enhancing diffusion with VED⁴
7. WM-GM Segmentation with FSL

¹ Zwiers, ISMRM 2009

² Visser, ISMRM 2010; Poster #3459

³ Kroon & Slump, IEEE-EMBS Benelux 2009

⁴ Koopmans, MRMP 2008

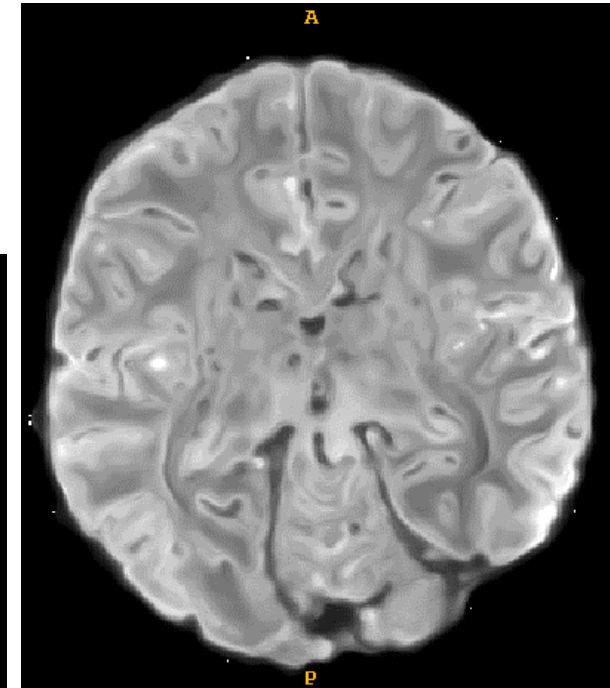
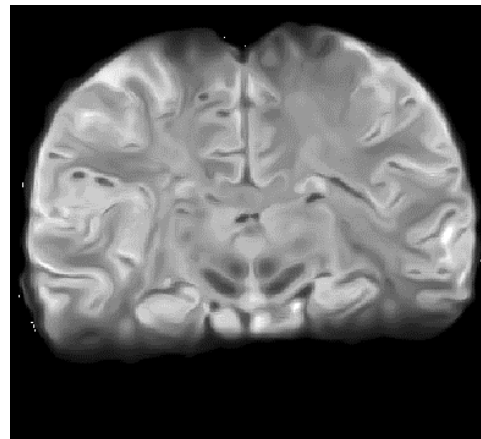
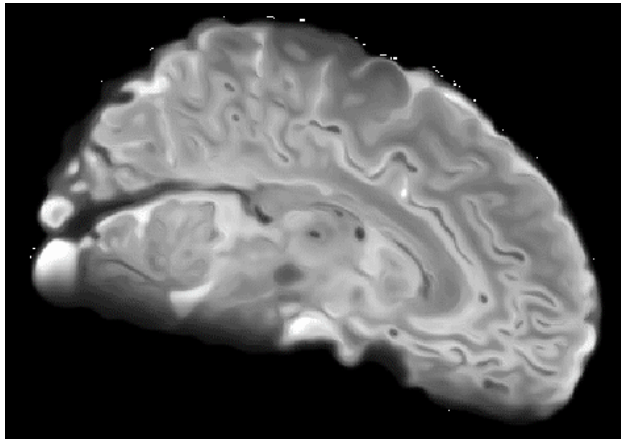




Methods: preprocessing (2)

Coregistration

- a) GRE \rightarrow T1; with FSL using FAST-based weighting volumes
- b) Mean $b=0$ DWI \rightarrow T1; with constrained warping in PE direction²



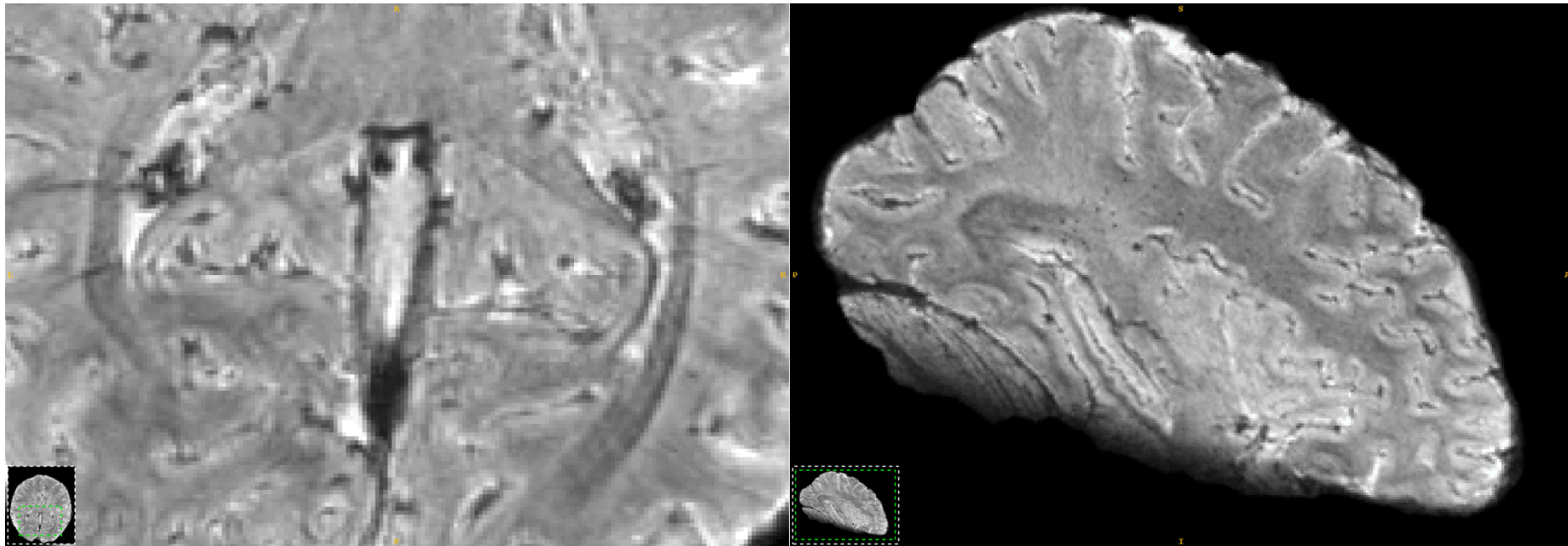
² Visser, ISMRM 2010; Poster #3459





Methods: preprocessing (3)

GRE image structure tensor with edge-enhancing diffusion³



³ Kroon & Slump, IEEE-EMBS Benelux 2009

