

Lesion segmentation and characterisation

Tuesday 30th January 2018 Carole SUDRE



White matter hyperintensity (WMH) segmentation

Technical impact Proposed solution Extension to longitudinal studies

Location characterisation

Available scale range and uncertainties Proposed solution Population level applications Patient level applications

WMH - Technical impact







- Errors in tissue volumes
- Errors in tissue borders
- Errors in parameters estimation







Automated WMH segmentation

UCL

Pathology

Clinical association



Technical impact





<u>à I</u> (

- Data modelled as weighted sum of Gaussian distributions
- Noise in magnitude image considered as Gaussian
- Each tissue modelled with Gaussian distribution

Data modelling





C. H. Sudre, M. J. Cardoso, and S. Ourselin, "Bilayered anatomically constrained split-and-merge expectation maximisation algorithm (BiASM) for brain segmentation," in SPIE Medical Imaging (S. Ourselin and M. A. Styner, eds.), vol. 9034, pp. 903411–903411–7, International Society for Optics and Photonics, International Society for Optics and Photonics, 2014.

C. H. Sudre, M. J. Cardoso, W. Bouvy, G. J. Biessels, J. Barnes, and S. Ourselin, "Bayesian Model Selection for Pathological Data," in MICCAI 2014 (P. G. Et al., ed.), LNCS 8673, pp. 323–330, Springer International, 2014.

C. Sudre, M. J. Cardoso, W. Bouvy, G. Biessels, J. Barnes, and S. Ourselin, "Bayesian model selection for pathological neuroimaging data applied to white matter lesion segmentation.," IEEE Transactions on Medical Imaging, vol. 34, pp. 2079–2102, apr 2015.

Post-processing

Model independent of application





WMH segmentation



≜UC

3 lesion-related components







Longitudinal extension



Images **Disease evolution in time** Intra-subject time Time point constraint consistency Average building Lesion segmentation Average model Average Constraint over individual time points Atlase Gaussian Mixture Model Model selection Association with





Longitudinal simulator

Longitudinal simulation





Progression patterns

	Linear_500	Linear_750	NonLinear_5	NonLinear_15
DSC	0.66	0.66	0.68	0.64
DSC	[0.27 0.76]	[0.34 0.76]	[0.39 0.76]	[0.41 0.74]
TDD	0.85	0.80	0.81	0.79
IFK	[0.68 0.89]	$[0.57\ 0.88]$	$[0.66\ 0.87]$	[0.67 0.85]
AD:4	1.92]	1.89	1.96	2.05
AvDist	[1.14 11.98]	[1.06 9.69]	[1.09 5.25]	[1.11 6.62]
OE/TatE	0.81	0.77	0.81	0.8
OE/ IOU	[0.47 0.89]	[0.53 0.88]	[0.67 0.90]	[0.57 0.88]
OEFP/FP	0.73	0.71	0.71	0.69
	[0.39 0.86]	[0.47 0.85]	[0.56 0.85]	[0.50 0.85]
OFEN/EN	0.96	0.93	0.96	0.94
OLIMPIN	[0.87 0.98]	$[0.78\ 0.97]$	[0.92 0.98]	$[0.90\ 0.97]$
FD/TotF	0.82	0.79	0.76	0.76
FI/IOUF	[0.60 0.88]	[0.59 0.86]	[0.59 0.85]	[0.59 0.82]

Acronyms expansion: DSC - Dice Similarity Coefficient; TPR - True Positive Rate; AvDist - Average Distance; FP/TotF - Proportion of false positives in the total of error; OE/TotF - Proportion of outline error in the total error; OEFP/FP - Proportion of false positive outline error in the false positives; OEFN/FN - Proportion of false negative outline error in the false negatives.

		Cross+	Long	Ref
	% change median	2.28	0.21	0.09
Flat_High	% change IQR	[-17.77 25.11]	[-8.19 9.29]	[-1.99 1.89]
	p-value	0.06	0.54	0.83
	Lin concordance	0.95	0.98	0.999
Flat_Low	% change median	6.53	1.53	0
	% change IQR	[-17.1 7.78]	[12.0 17.6]	[-1.96 1.46]
	p-value	0.0002	0.05	0.42
	Lin concordance	0.87	0.95	0.999

Application to ADNI data

APOE	Base	line	2 ye	sars	
33					Number
-4					% cha CI Overa Pairw

	АРОЕ		
	33	43	44
Number (A β)	164 (161)	108 (102)	24 (23)
% change	5.68	8.68	15.53
CI	[3.56 7.84]	[5.95 11.48]	[9.09 22.34]
Overall p		0.009	
Pairwise		33 vs 44 **	

UCL

Effect size

	33 vs 43	43 vs 44	33 vs 44
Cross+	0.05	0.21	0.28
Long	0.21	0.32	0.47

WMH Location characterisation



Continuity with ventricular lining

Periventricular?

Within absolute extent

Confluent lesions Concomitant atrophy Lobar separation

à l (ci

Continuity with absolute (mm) extent

Lobes and layers



Layers

Laplace equation between ventricular and cortical surface Normalised distance Independent of atrophy Discretised = 4 layers







Infographic representation

4 layers and 9 lobar regions



Sudre, C. H., Anson, B. G., Davagnanam, I., Schmitt, A., Mendelson, A. F., Prados, F., ... & Cardoso, M. J. (2017). Bullseye's representation of cerebral white matter hyperintensities. *Journal of Neuroradiology*.



≜UC

Lesion frequency: Proportion of a region occupied by WMH

Lesion distribution: Proportion of the WMH volume localised in a region

Population level - Risk factors in WMH **UCL**



Population level - Comparison



MS without OABS

MS with OABS



MS - Multiple sclerosis OABS - Overactive Bladder syndrome

Patient level - Diagnostic tool



CUSHING Disease



Blinded diagnosis

3hrs with images 5 min with Bullseyes

Conclusion - Ongoing work



WMH segmentation

Automated generic outlier data modelling Adaptable post-processing Longitudinal extension

Location characterisation

Systematic patient-specific coordinate frame Infographic tool summarising 3D distribution Application to patient or population level Other applications quality control, protocol comparison

And beyond...

Longitudinal location patterns Distinction between subtypes of lesion Beyond the brain onto spinal cord Analysis of lesion texture and shape...



Thank you for your attention !

Questions?

Frederik Barkhof **Josephine Barnes** M. Jorge Cardoso Nishi Chaturvedi Nick Fox Chris Frost **Beatriz Gomez Anson** Alun Hughes **Rolf Jäger** Marc Modat Sébastien Ourselin Ferran Prados Carrasco Lorna Smith Claudia Wheeler-Kingshott Xixi Yang



à ((

TRANSLATIONAL IMAGING GROUP



