

Validity of summary statistics-based mixed-effects group fMRI

Camille Maumet

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Validity of summary statistics-based mixed-effects group fMRI

Camille Maumet and Thomas Nichols

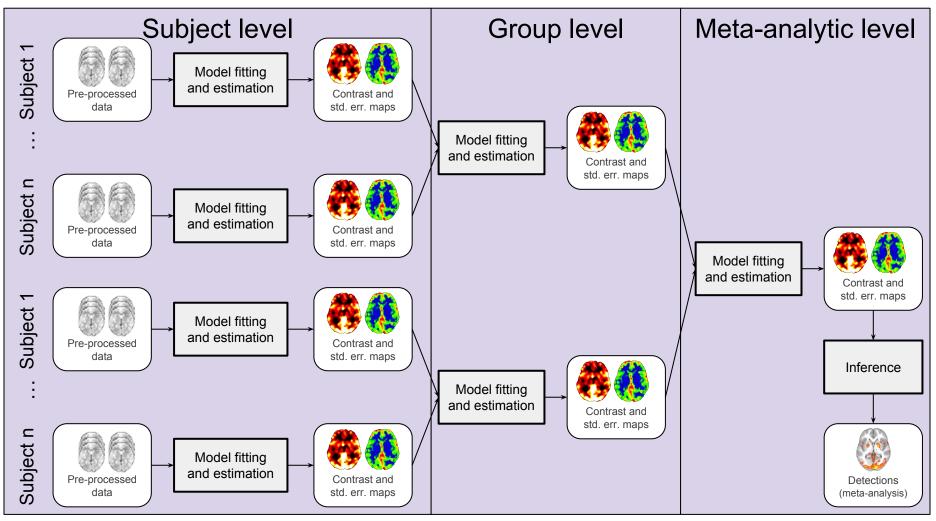
1. University of Rennes, Inria, CNRS, Inserm, IRISA, Rennes, France.

2. Oxford Big Data Institute, University of Oxford, Oxford, UK.



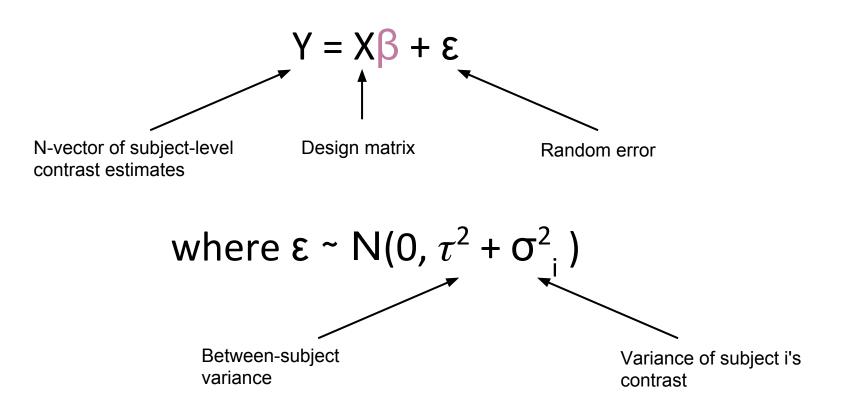


What is the link between meta-analysis and group fMRI?



Methods

GLM for group fMRI



Solving the GLM: OLS vs. FGLS

Y = Xβ + ε, where ε ~ N(0,
$$\tau^2$$
 + σ^2_i)

Assuming $\tau^2 + \sigma_i^2$ constant, **Ordinary Least Squares (OLS)** gives the following group statistic estimate:

$$T = \left(\sum_{i=1}^{N} \frac{Y_i}{\sqrt{N}}\right) \middle/ \widehat{\sigma_C^2}$$

where $\widehat{\sigma_C^2}$ is the usual one-sample variance. Under ${
m H_0}:~T\sim {\cal T}_{N-1}$

Solving the GLM: OLS vs. FGLS

Y = Xβ + ε, where ε ~ N(0,
$$\tau^2$$
 + σ^2_i)

Feasible Generalised Least Squares (FGLS) gives the following group statistic estimate:

$$T = \left(\sum \kappa_i Y_i\right) / \sqrt{\sum_{i=1}^N \kappa_i}$$
 where $\kappa_i = 1/(\hat{\tau}^2 + s_i^2)$

where s_i^2 is the study i's sampling variance. In large samples, under ${
m H_0}$: $T\sim {\cal T}_{N-1}$

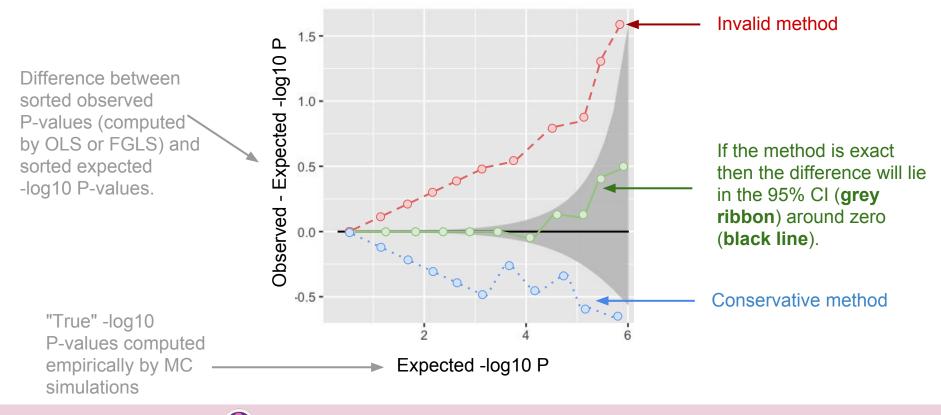
Solving the GLM: OLS vs. FGLS

	Statistic	Assumptions	Implementation
OLS	$T = \left(\sum_{i=1}^{N} \frac{Y_i}{\sqrt{N}}\right) \middle/ \widehat{\sigma_C^2}$	τ^2 + σ^2_{i} constant	SPM 'One-sample t-test' (default)
			'Mixed effects: Simple OLS'
			AFNII 3dttest++
FGLS	$T = \left(\sum \kappa_i Y_i\right) / \sqrt{\sum_{i=1}^N \kappa_i}$	Large samples	SPM 'Mixed-effects analysis'
			'Mixed effects: FLAME1' (default)
			AFNI 3dMEMA

Goal: assess the validity of OLS and FGLS under violation of their assumption

Method: Monte Carlo simulations

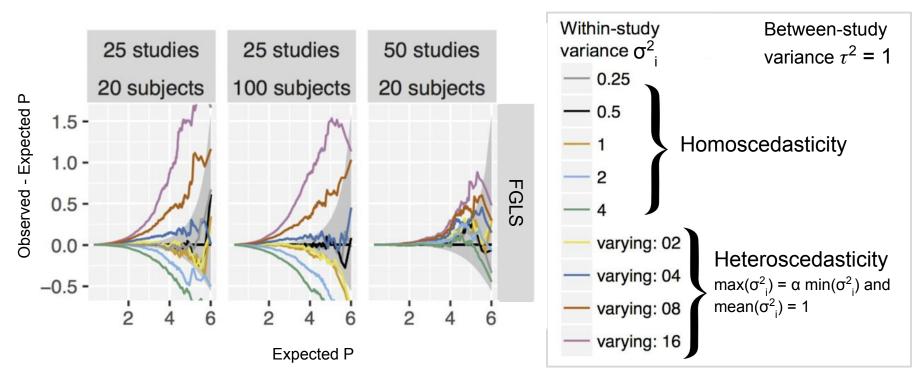
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And in meta-analyses?



Question: Does this issue also affects group fMRI studies?

Simulations

Between- and within-subject variances in real fMRI data

21 studies investigating pain

Study	# subjects	# outliers	Mean τ^2	Mean σ^2_{in}	Max/Min σ² _{in} ratio	$\underset{\sigma^{2}_{out}}{\text{Mean}}$
01	25	2	354	100	4	282
02	25	1	680	100	5	317
03	20	2	800	100	5	281
04	20	3	886		9	347
05	9	0	1802		3	-
Summary		0-15%	18-170	100	2-9	164-300

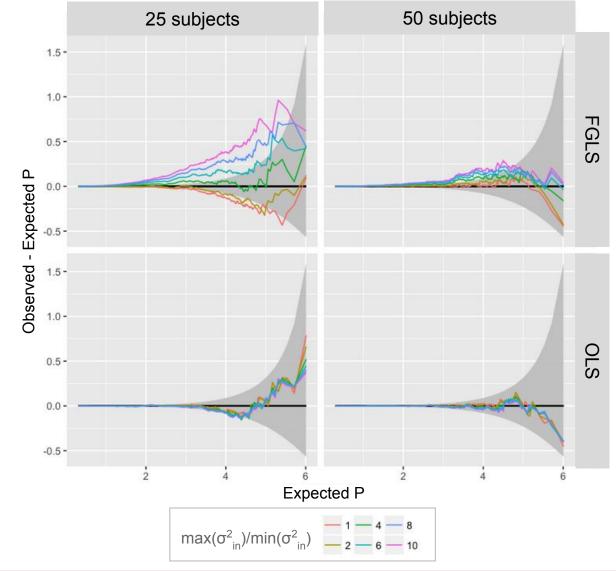
Simulation setup

- $N \in \{25, 50\}$ subjects
- $\tau^2 = 1$ and mean(σ^2_{in}) = 1
- Two settings for the within-subject variance:
 - $\max(\sigma_{in}^2) / \min(\sigma_{in}^2) \in \{1, 2, 4, 6, 8, 10\}$
 - 16% of outliers with $\sigma^2_{out} \in \{ 0.25, 0.5, 2, 4 \}$

10⁶ realisations

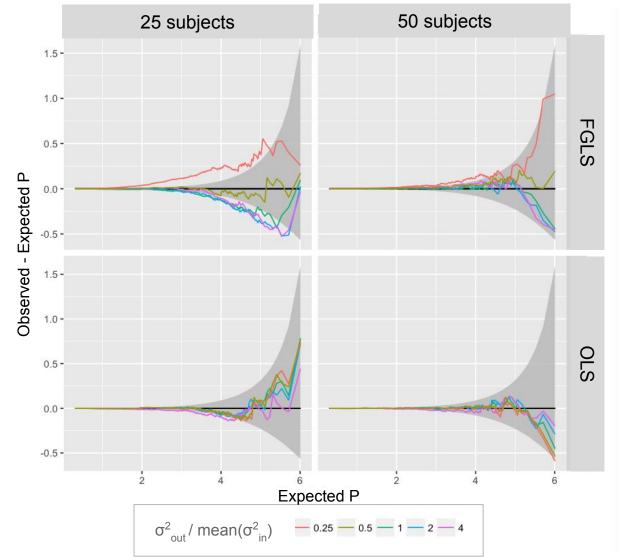
Results

Varying levels of heteroscedasticity



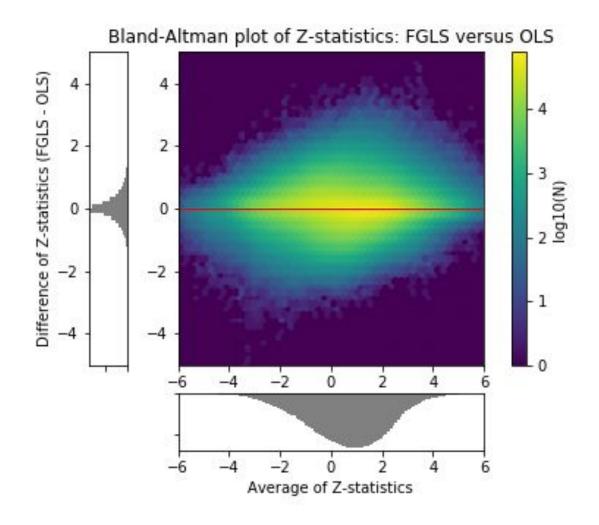
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In the presence of outliers



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Real data



Conclusions

FGLS can be invalid is small samples, especially in the presence of strong heteroscedasticity or 'low variance' outliers.

RFX remained valid in all studied settings.

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