

# Cluster failure: Why fMRI inferences for spatial extent have inflated false positive rates

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Functional magnetic resonance imaging (fMRI) is 25 years old, yet surprisingly its most common statistical methods have not been validated using real data. Here, we used resting-state fMRI data from 396 healthy controls to conduct 2 million task group analyses. Using this null data with different experimental designs, we estimate the incidence of significant results while controlling for multiple testing. In theory, we should find 5% false positives (for a significance threshold of 5%) but instead we found that the most common software packages for fMRI analysis (SPM, FSL, AFNI) can result in false positive rates of up to 70% for cluster inference. These results question the validity of some 40,000 fMRI studies and may have a large impact on the interpretation of neuroimaging results.

## Methods

- Downloaded 396 resting state fMRI datasets from the 1000 functional connectomes project [1], Beijing 198 datasets, Cambridge 198 datasets
- Analyzed all 396 datasets in SPM, FSL and AFNI, scripts available at <https://github.com/wanderine/ParametricMultisubjectfMRI>
- Used 4 activity paradigms with this resting data, 2 block based (B1, B2) and 2 event related (E1, E2)
- Varied smoothing between 4 and 10 mm FWHM
- Ran 1,000 random group analyses (two-sample and one-sample t-tests) for each parameter combination
- Tested voxel inference as well as cluster inference, used 2 cluster defining thresholds ( $p = 0.01$  and  $p = 0.001$ )
- Tested SPM, FSL FLAME1, FSL OLS, AFNI 3dttest, AFNI 3dMEMA and a non-parametric permutation test
- Familywise error (FWE) rate estimated as number of significant results ( $p = 0.05$ , corrected) divided by number of group analyses (1,000)

## Results

Estimated FWE rates for cluster inference and two-sample t-tests, for the Beijing datasets, are given in Figure 1 (Cambridge results were similar). Corresponding results for voxel inference are given in Figure 2. Note how all permutation tests are within the 95% confidence interval; across all simulations only some one-sample t-tests generated invalid permutation tests, attributable to failures of the symmetry assumption.

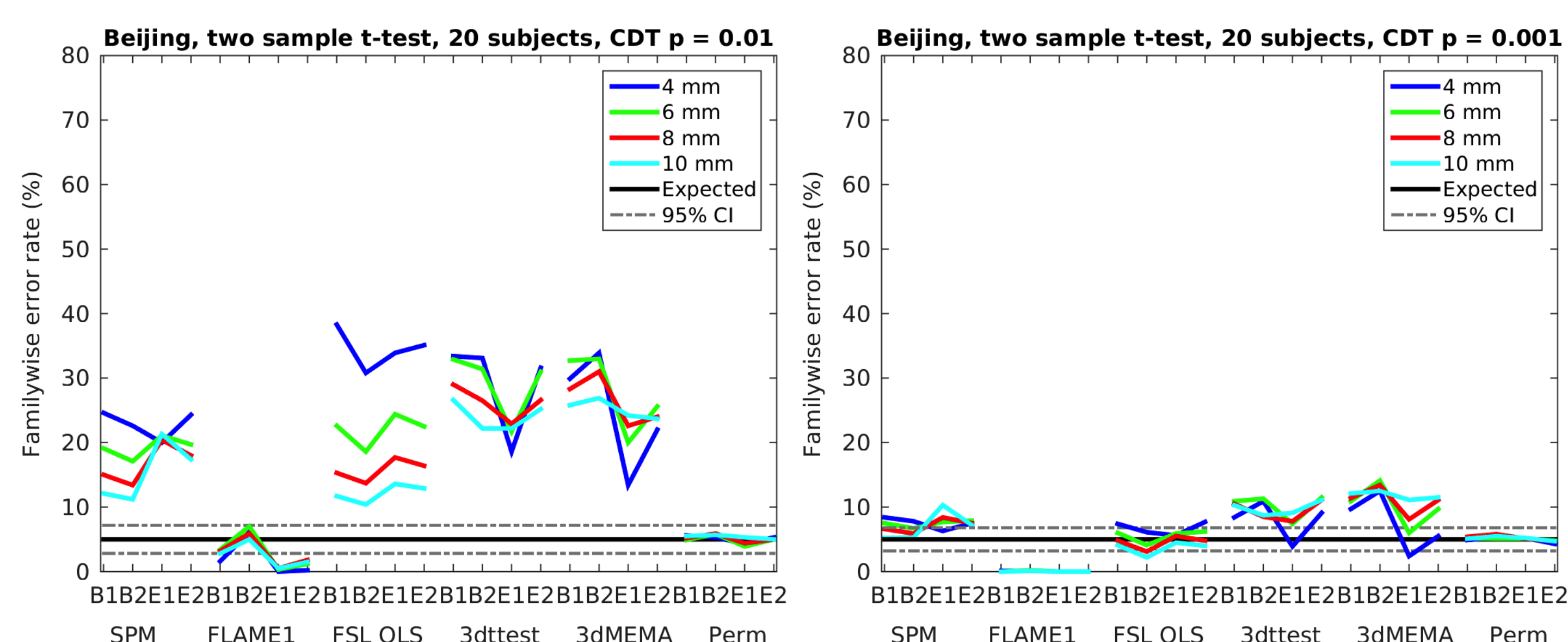


Figure 1

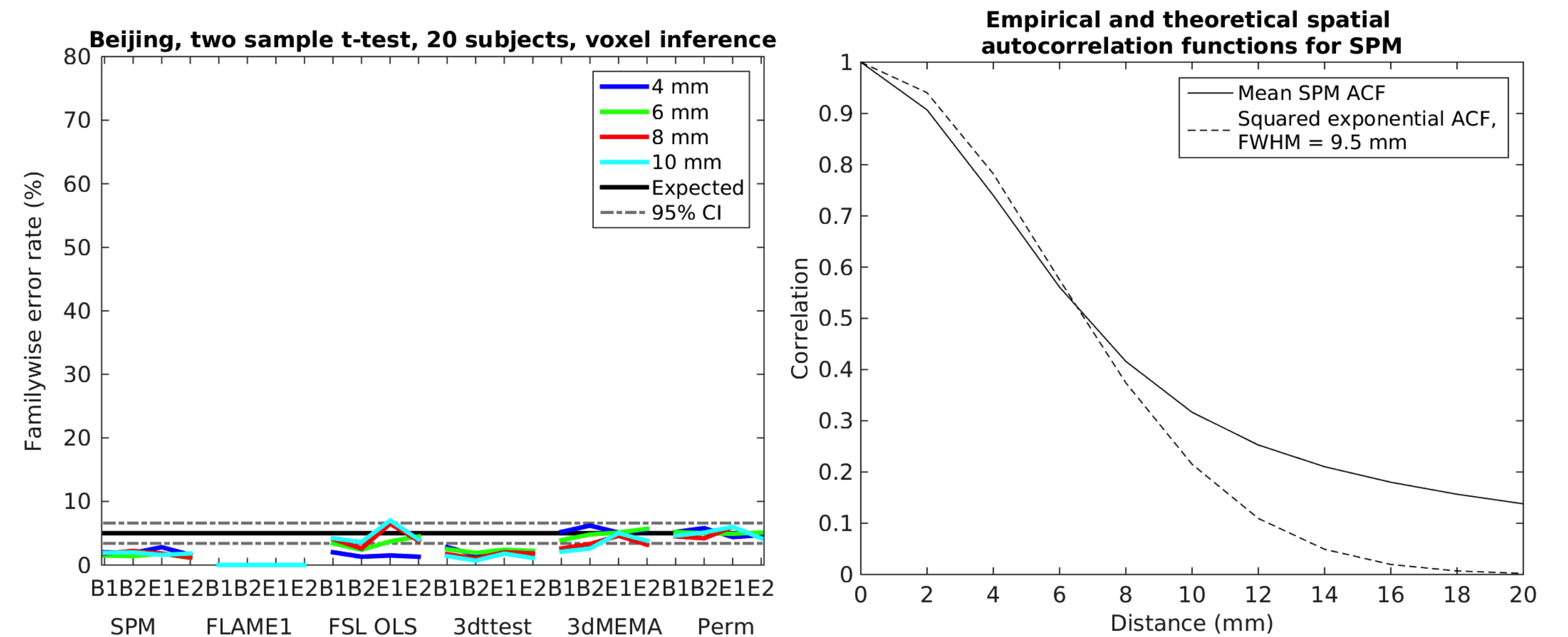


Figure 2

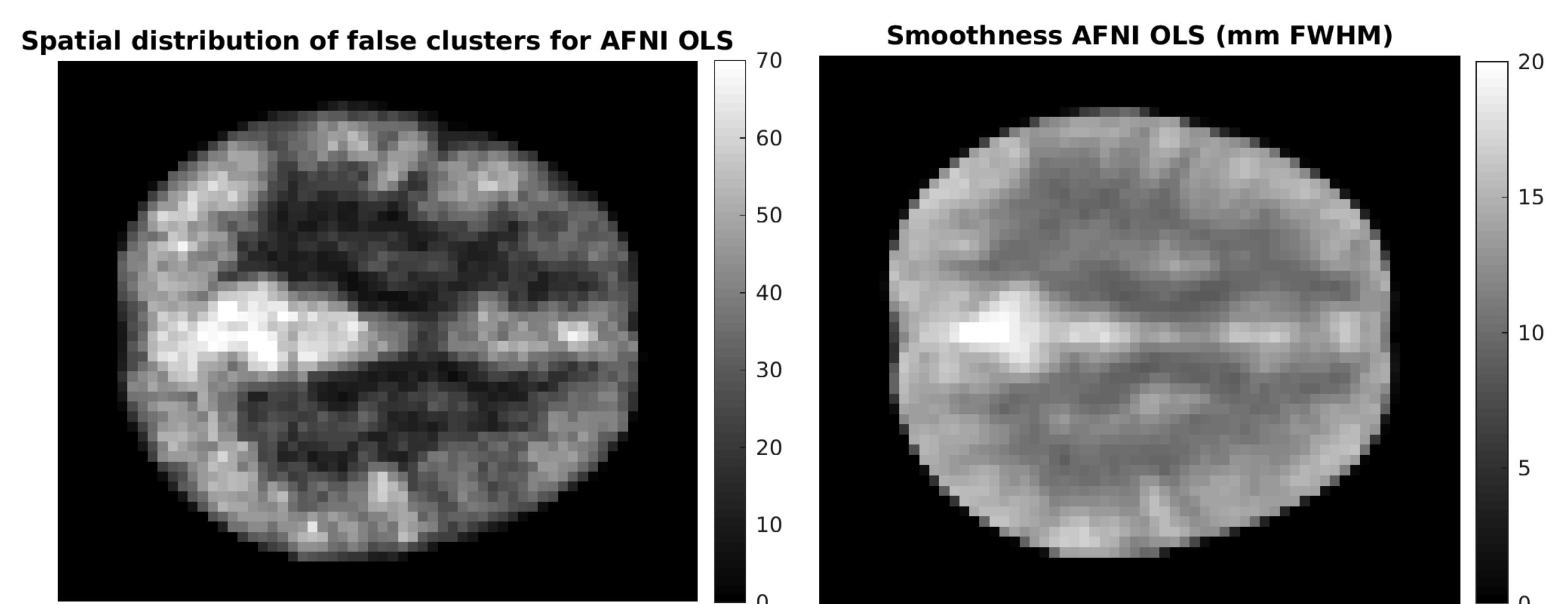


Figure 3

Figure 4 shows the ratio between non-parametric and parametric p-values for activity clusters obtained from four task datasets from OpenfMRI [2]. The non-parametric cluster p-value is always larger compared to the parametric p-value, especially for a cluster defining threshold of  $p = 0.01$ .

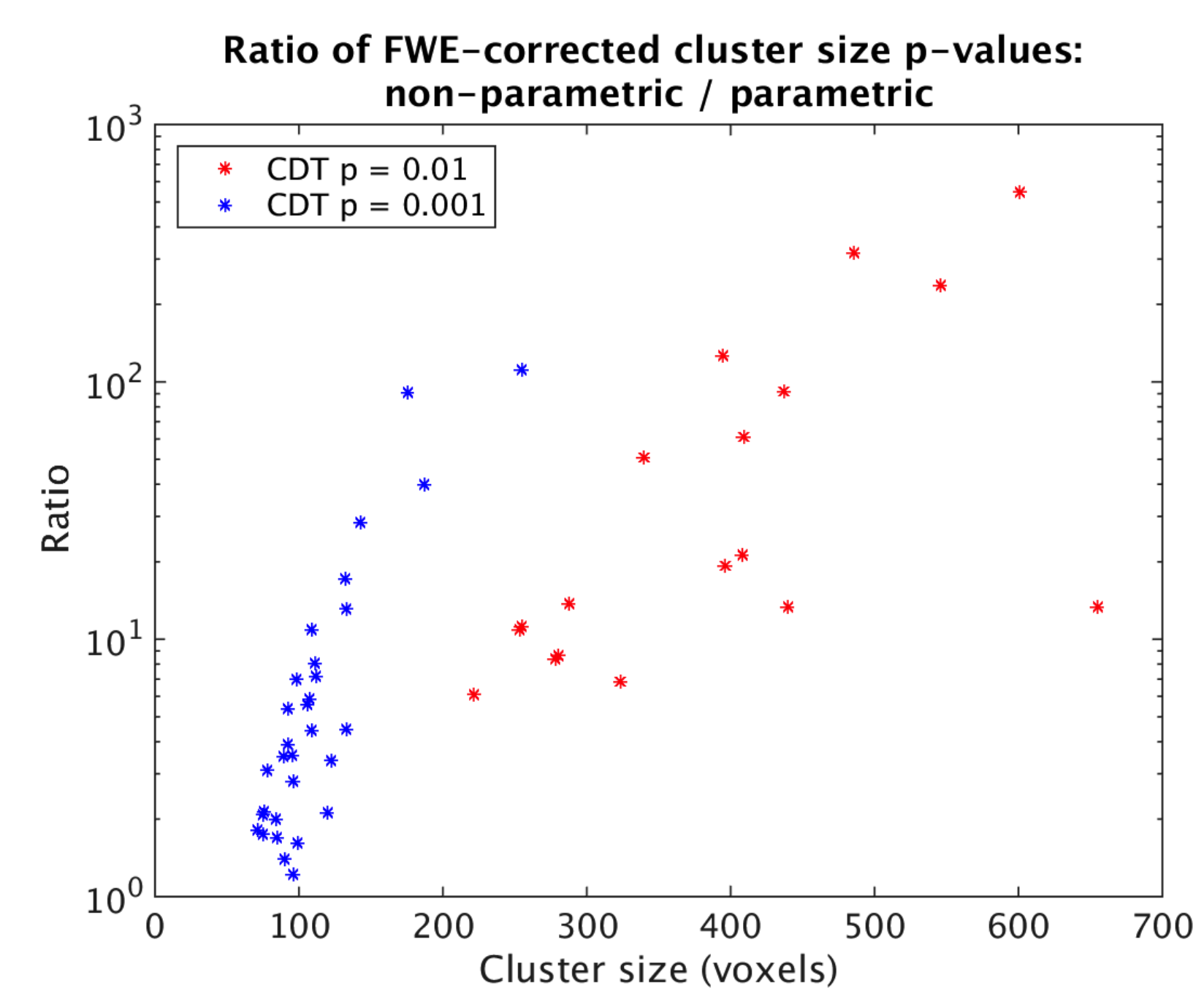


Figure 4

## Discussion

Using mass empirical analyses with task-free fMRI data we have found that the parametric statistical methods used for group fMRI analysis with the packages SPM, FSL and AFNI can produce FWE-corrected cluster p-values that are erroneous, being spuriously low and inflating statistical significance. This calls into question the validity of countless published fMRI studies based on parametric cluster-wise inference. See the full paper [3] for all the details.

## References

- [1] Biswal, B. et al., "Toward discovery science of human brain function," *PNAS*, vol. 107, pp. 4734–4739, 2010.
- [2] Poldrack, R. et al., "Toward open sharing of task-based fMRI data: the OpenfMRI project," *Frontiers in Neuroinformatics*, vol. 7, no. 12, 2013.
- [3] A. Eklund, T. Nichols, and H. Knutsson, "Cluster failure: why fMRI inferences for spatial extent have inflated false positive rates," *PNAS*, 2016.

The parametric statistical methods work well, if conservatively, for voxel inference, but not for cluster inference. The main problems for cluster inference are that the data have a non-Gaussian spatial autocorrelation function (Figure 2), and that the spatial smoothness is not constant in the brain (Figure 3).