## Localized Cluster Enhancement: TFCE revisited with valid error control

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### Overview

- 1. TFCE recap
- 2. What error rate (if any) does TFCE control?
- 3. Discuss why TFCE very strongly depends a threshold  $h_0$
- 4. Introduce a Localized TFCE: to provide precise inference and valid error control

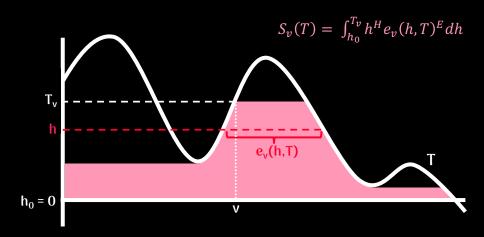
### TFCE recap

Suppose that we have a test-statistic  $T: \mathcal{V} \to \mathbb{R}$  at each voxel in the brain. The the TFCE transformation is defined as:

$$S_v(T) = \int_{h_0}^{T_v} h^H e_v(h, T)^E \mathrm{d}h.$$

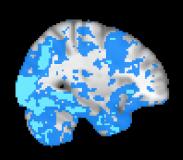
- Here h is the height and  $e_v(h,T)$  is the extent of the test statistic at the level h for the voxel v.
- In practice H = 2, E = 0 and  $h_0 = 0$  are the default parameters typically chosen.
- $S_v$  thresholded at a level based on permutation thresholding.
- Permuted TFCE test-statistics:  $S_{v,1}^*, \ldots, S_{v,P}^*$  are calculated.
- level  $t^*$  chosen based on the  $\alpha$  quantile of the permutation distribution of  $\max_{1 \leq p \leq P} S_{v,p}^*$ .

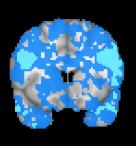
## Understanding the TFCE integral

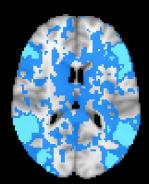


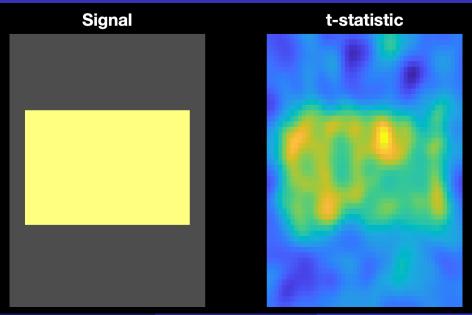
### TFCE illustration

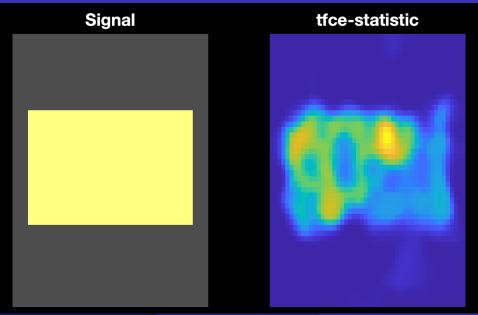
We apply TFCE to 20 subjects from the HCP to the primary social contrast.

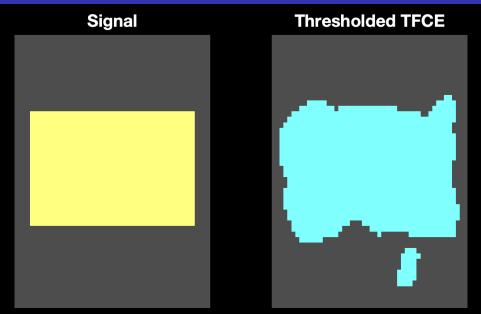


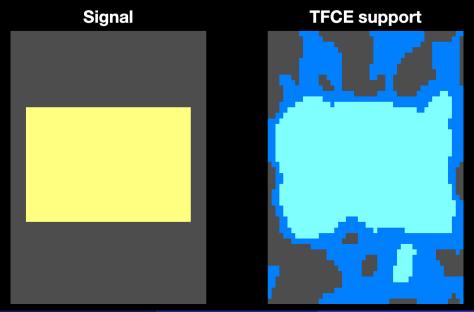


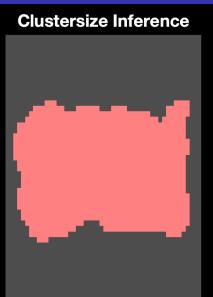




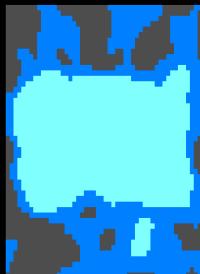




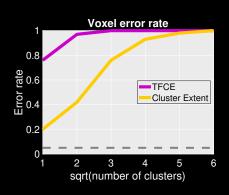


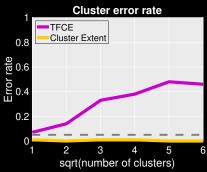


### Thresholded TFCE



## TFCE false positives





## TFCE increases the number of free paramters

One of the main motativations for TFCE was to reduce researcher degrees of freedom, to make it "threshold free". However the TFCE transformation is defined as:

$$S_v = \int_{h_0}^{T_v} h^H e_v(h)^E dh.$$
 (1)

- Unlike Clustersize inference TFCE thus actually has 3 free parameters.
- In particular  $h_0$  acts as a threshold.

#### Localized Cluster Enhancement

Recall the TFCE statistic is:

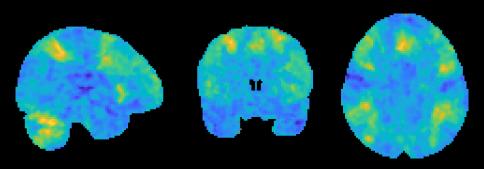
$$S_v(T) = \int_{h_0}^{T_v} h^H e_v(h, T)^E dh.$$

To localize this statistic, for a region  $R \subset \mathcal{V}$  let

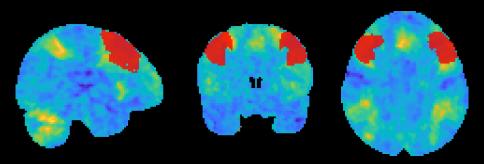
$$S_{v,R}(T) = S_v(T \times 1[R]) = \int_{h_0}^{T_v} h^H e_v(h, T \times 1[R])^E dh.$$

Let  $t^*$  be the  $(1 - \alpha)$  quantile of the permutation distribution of  $\max_{v \in \mathcal{V}} S_v(T)$ . Then we can say R contains at least one active voxel if  $S_{v,R}(T) > t^*$ .

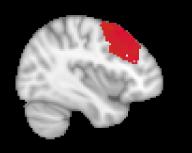
The t-statistic based on 20 subjects (for the HCP Gambling task)



Apply a mask of the Middle Frontal Gyrus



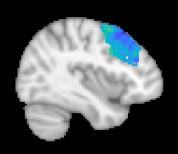
Apply a mask of the Middle Frontal Gyrus







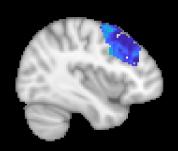
Apply a mask of the Middle Frontal Gyrus







#### Apply TFCE on that mask





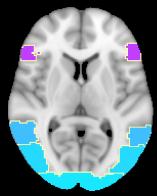


### Theory

Suppose that the data satisfies an exchangeability assumption. Then

**Theorem:**  $\mathbb{P}(S_{v,R}(T) < t^* \text{ for all inactive } R) \geq 1 - \alpha.$ 

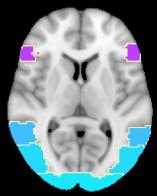
## Regional activation (HCP - Social)



(a) Localized TFCE,  $h_0 = 0$ 

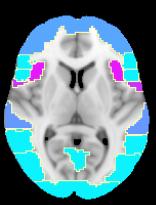
```
0.02
0.018
0.016
0.014
0.012
0.01
0.008
0.006
0.004
0.002
```

## Regional activation (HCP - Social)



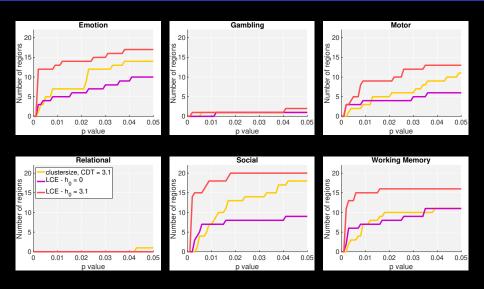
(a) Localized TFCE,  $h_0 = 0$ 





(b) Localized TFCE,  $h_0 = 3.1$ 

## Power comparison for a regional analysis



### Conclusions

- TFCE as classically used can have inflated voxel and clusterwise error rates.
- Localized Cluster Enhancement provably controls clusterwise and regional error rates and allows for increases in power and localization.
- TFCE is not threshold free as it (strongly) depends on a threshold  $h_0$ . The default choice of  $h_0 = 0$  means that TFCE typically can only make the weak global statement in practice.
- For localized cluster enchancement we recommend a threshold of  $h_0 = 3.1$  in line with the default for clustersize inference.

#### Thanks

- Slides for this talk are available on my website: sjdavenport.github.io/talks
- Code to implement LCE and a tutorial on TFCE are available in the StatBrainz MATLAB package available at: sjdavenport.github.io/software
- Further details available at my poster: 1871

### Global vs voxel vs cluster level inference

There are 3 types of inference statements typically used.

- 1. Voxel: Every highlighted voxel is active
- 2. Cluster: Every cluster contains at least one active voxel
- 3. Global: There is some voxel active somewhere in the brain.

### Classifying fMRI inference methods

There are 3 types of inference statements typically used.

1. Voxel: Every highlighted voxel is active

Voxelwise inference

2. Cluster: Every cluster contains at least one active voxel Clustersize inference

3. Global: There is some voxel active somewhere in the brain.

TFCE