# Go Figure: Keep context in images to meaningfully interpret results

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## Introduction

**Question: What are "the results" of a study to share in figures?** Answer: It depends a lot, based on how you choose to threshold.

Both of the following images *clearly show the same suprathreshold clusters (p=0.001+cluster)*, but they have some important differences:

## An example of ambiguity from context loss

Standard all-or-nothing thresholding removes context around clusters. How does this lead to ambiguity and misinterpretation?

•When results are presented with standard, all-or-nothing thresholding, researchers and readers will interpret these results as "fully lateralized

### (old) All-or-nothing threshold



Hides results from 99% of brain

→ treats them like 0 activity

Removes context around clusters

→ creates ambiguity

Biases results

→ makes results unstable
 • Harms reproducibility
 → can't interpret or compare well
 The information loss here leads
 to misinterpretation and biases
 towards non-reproducibility.

## (new) Transparent threshold\*

response":



- Shows sub-threshold results
- $\rightarrow$  see gradation of effects/stats
- Keeps context around clusters
- $\rightarrow$  see networks, extent, etc.
- Appropriately stable results
- $\rightarrow$  less sensitive to arbitrariness
- Leads to informed comparisons



<u>Cluster interpretation</u>
 Lateralized response

• But that sparse result could have come from *any one of the following* results, which have very different biological interpretations:



Transparent thresholding shows the context that allows for a more accurate assessment (near left-right symmetry, *not* lateralization).
All-or-nothing over-reduces results → ambiguity and misinterpretation Transparency keeps context → clarity and richer understanding

Sidenote: Statisticians know this same lesson as Anscombe's Quartet (Anscombe, 1973).

These plots have the same summary stats (mean, stdev and correlation), but

→ more accurate reproducibility The evidence is presented more scientifically (but still digestibly) for more accurate evaluations.

\* Transparent thresholding displays the same above-threshold results as standard "all-or-nothing" thresholds, but it then also presents subthreshold results with increasing transparency. Here, we also put a boundary around the suprathreshold results, to further highlight them (see Allen et al., 2012).



Showing the plots provides useful info  $\rightarrow$  understand & evaluate the data.

# Transparent thresholding: Now available in a software package near you!



B) AFNI: macaque task FMRI (overlay: effect contrast, threshold: *t*-stat)



C) NiiVue: resting state FMRI (overlay: Pearson correlation, threshold: Pearson correlation)



#### E) bidspm: task FMRI (overlay: *t*-stat, threshold: *t*-stat)





F) Trends-Matlab & GIFT: task FMRI (overlay: effect contrast, threshold: *t*-stat)



G) BrainVoyager: task FMRI (overlay: *t*-stat, threshold: FDR *q*-values)



#### I) FSLeyes: task FMRI (overlay: Z-stat, threshold: Z-stat)



J) RMINC & MRIcrotome: rat structural MRI morphometry (overlay: *t*-stat, threshold: *t*-



K) RMINC: mouse structural MRI morphometry ROIs (overlay: *t*-stat, threshold: *t*-stat)



D) CIVET & minc-toolkit-v2: cortical thickness of structural MRI (overlay: t-stat, threshold: t-stat)





## Conclusions

Data visualization is an important analysis step. The method of thresholding is a key processing choice. Modern neuroimaging should use modern thresholding to present results, reducing bias, improving understanding and (of course!) improving reproducibility. For more discussion and examples (including with NARPS and the dead salmon!) please see here:

Go Figure: Transparency in neuroscience images preserves context and clarifies interpretation

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