

of Mental Health

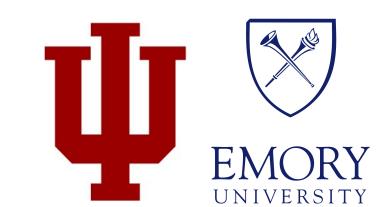
Highlight Results, Don't Hide Them: Improving **Reproducibility, with Applications to NARPS**





SUMA!

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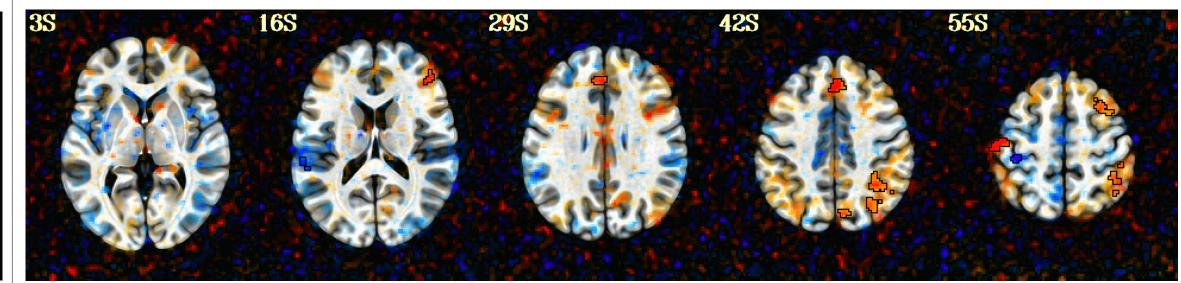
Introduction

• Standard thresholded activity maps amplify differences, distort interpretations of results and harm reproducibility evaluation.¹ •We show how the "highlighting" approach²

A) Standard "hiding": opaque threshold

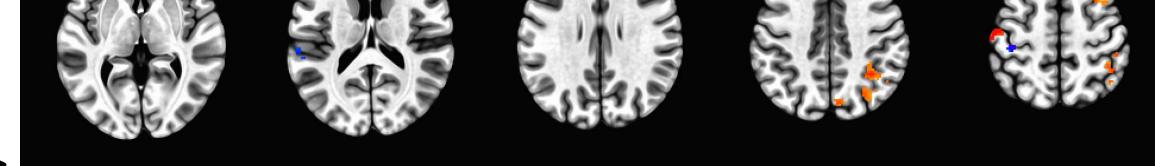


B) Improved "highlighting": translucent threshold



(focus on the strongest results, but also show the rest) greatly improves reporting for interpretation and meta analysis.³ •We present examples using the NARPS⁴

data, where >60 teams analyzed the same data with their own pipelines and compared results primarily using thresholded maps. • Highlighting demystifies the kind of variability seen among the NARPS teams' results: it is mainly a varied strength of agreement rather than disagreement.



- Hides information (good science presents evidence) • Wastes data (modeling occurred everywhere!) Mathematically promotes selection bias • *Neither* the brain *nor* BOLD are ON/OFF like this • Results are inherently sensitive to threshold parameter: tiny differences are magnified • Harms both understanding and reproducibility:
- similarities (or differences!) are hidden
- Presents more scientifically: show evidence instead of threshold-sensitive decisions (let reader see more) • Emphasize focal regions—but place in context • Improve quality control (QC): reduce lurking artifacts (even check outside brain!) and reveal poor modeling
- Reduce artificial and arbitrary threshold dependency
- •Allow for more meaningful comparisons across studies: better for evaluating reproducibility

Methods and Results

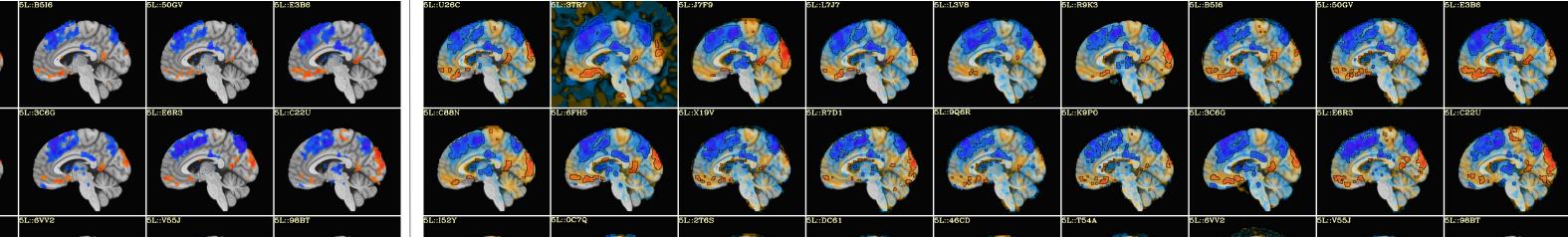
•We performed cross-study comparisons of the NARPS teams' results; outcomes were similar for each NARPS hypotheses.

• Compare seeing each NARPS teams' stats result for Hyp #1&3 in: hiding mode (A) vs in highlighting mode (B).

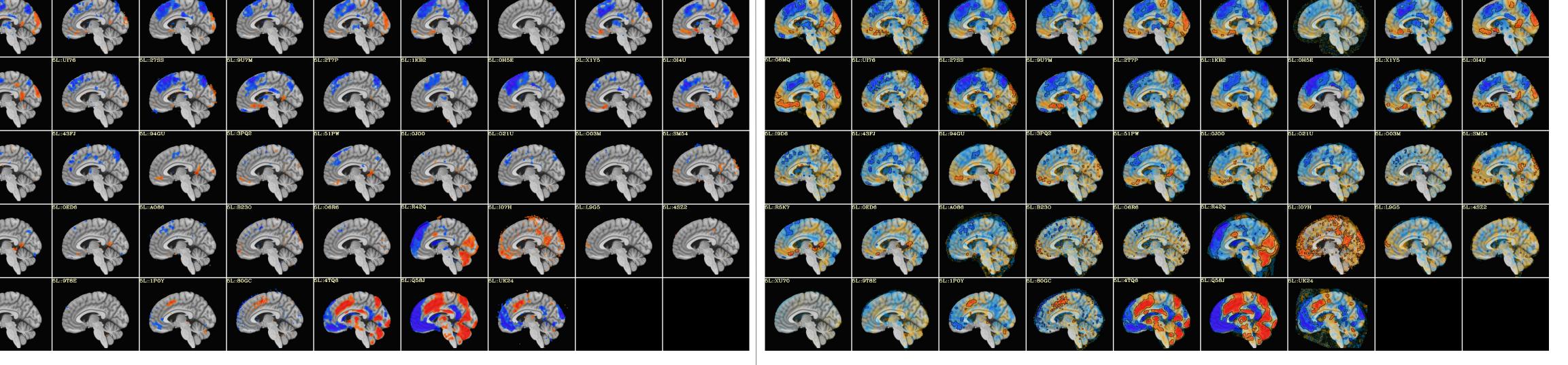
<u>NARPS teams' results comparison (Hyp #1 and 3, +gain, indif grp)</u>

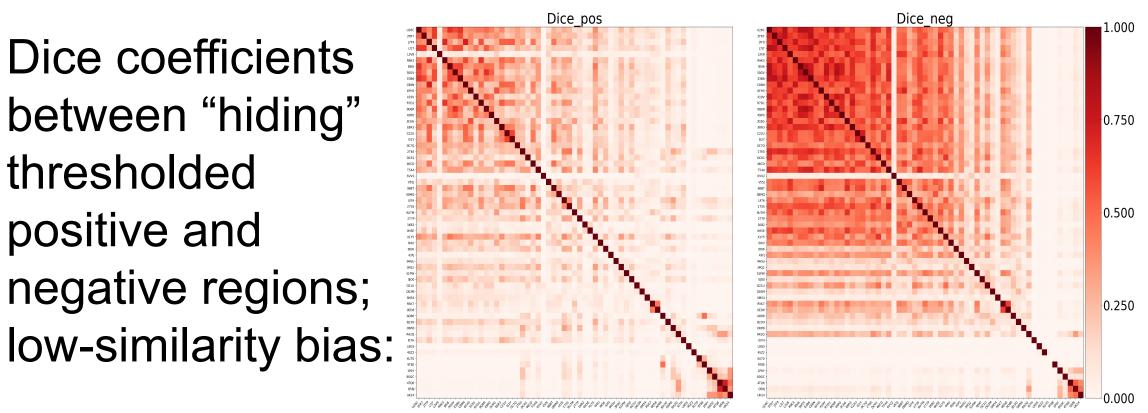
A) Standard "hiding": opaque threshold

B) Improved "highlighting": translucent threshold

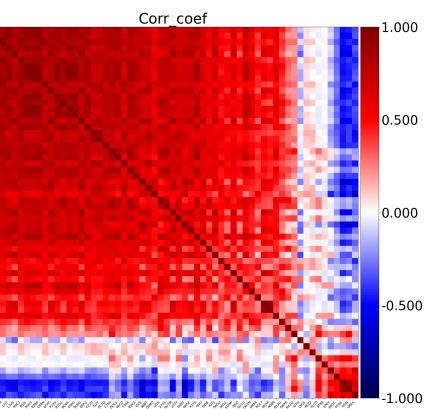


- "A" appears to show little agreement, which might be interpretable as wide variability or disagreement
- "B" shows more information and that teams' results actually have high agreement, simply with varied strength.
- This can be quantified with similarity matrices: degree of agreement in A is unsurprisingly lower, while B shows mostly excellent agreement, with a few results that are different (and a few with inverted signs).
- The results are analogous across all Hyps³: transparent thresholding shows that most <u>NARPS teams produce similar activation</u> patterns, just with varied strength.





Correlation coefficients between "highlighting" unthresholded statistical maps; most data with very high correlation, and more details about other cases:



Conclusions

While data sharing is useful, the thresholding recommendations made here are distinct and important on their own. Results should be presented informatively from the start, without the need for extra work, downloading, re-analyzing, etc. A study's interpretation and a reader's initial impression should be as accurate and complete as possible. This work has major implications for the neuroimaging community in presenting results and performing metaanalyses. Individual studies and meta-analyses should adopt a "highlighting" approach, in order to accurately assess results.

Tutorial Pages and Download Scripts

thresholded

positive and

 Download NARPS processing scripts, using AFNI⁵: https://github.com/afni/apaper highlight narps •AFNI command examples for scripting images: https://afni.nimh.nih.gov/pub/dist/doc/htmldoc/tutorials/auto_image/auto_%40chauffeur_afni.html • Transparent thresholding in AFNI GUI (AFNI Academy Bootcamp channel): https://www.youtube.com/watch?v=VT77zJ0zGnA&list=PL_CD549H9kgqwHr0EDtvAU8hylsOj30OK&index=3 • "Highlight" viewing in afni proc.py quality control (APQC) HTML: https://afni.nimh.nih.gov/pub/dist/doc/htmldoc/tutorials/apqc_html/apqc_ex1.html#qc-block-vstat

Acknowledgment & References

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