

Evaluating Reconstruction Accuracy in Neural Decoding Models

Team: Motorized Marmots 🐹🐹

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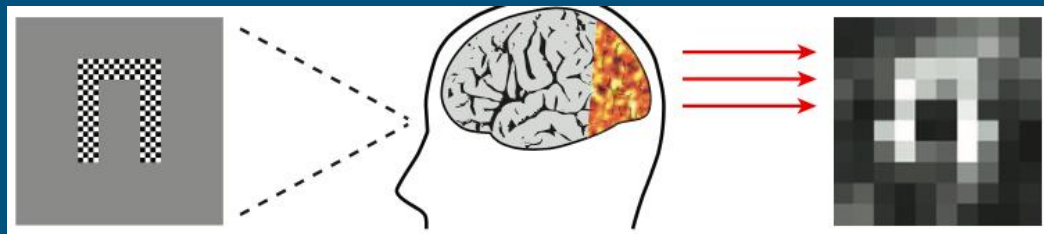
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Introduction

- Decoding models relate regional brain activity (e.g., voxel-wise BOLD response) to the stimulus:
- In human vision research, decoding models are evaluated for their ability to use fMRI responses to an image for predicting:
 - its stimulus class ('classification')
 - recognise it among distractors ('identification'),
 - or reconstruct the original image

- However....

The “one pixel attack”



From Kay, K., Gallant, J. Nat Neurosci (2009)

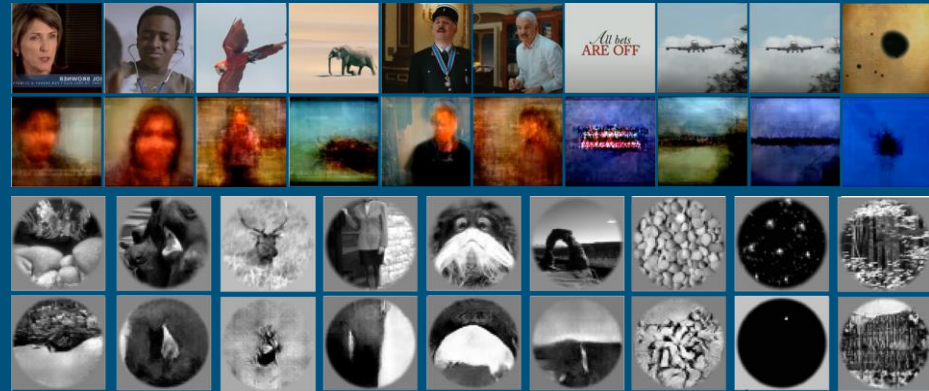
Introduction (contd.)

- There are multiple metrics for quantifying model performance on image reconstruction — leading to:
 - ad-hoc choices
 - less comparability among decoding models
- We propose that human judgement should serve as the golden standard; we, therefore, compared 2 metrics of reconstruction accuracy against 2 human judgement tasks

Study	Metrics Used
Bashivan et al., 2019	<ul style="list-style-type: none">- Distance in pixel-space- Euclidean distance
Miyawaki et al., 2008	<ul style="list-style-type: none">- Spatial correlation- Image identification
Naselaris et al., 2009	<ul style="list-style-type: none">- Spatial similarity (Brooks and Pappas, 2006)- Semantic similarity, at 4 different levels of specificity

Methods

- Human judgment:
 - Participants (N = 6, 3F, 3M; Age range: 23-27)
 - 2 Dimensions: accuracy; perceived similarity
- Conventional measures:
 - Metrics: RMSE (Root Mean Squared Error) and FSIM (Feature Similarity Index)
 - Implemented in Python using the image_similarity_metrics package (Müller et al., 2020)



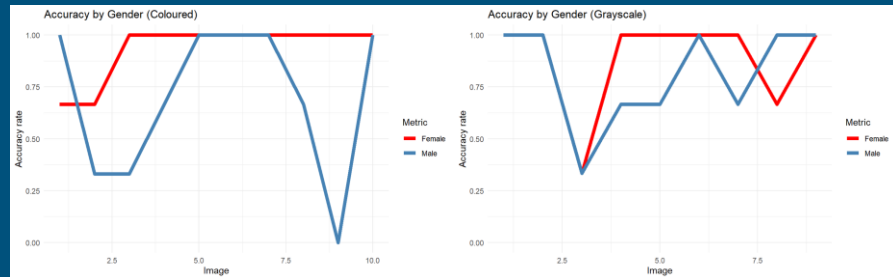
Materials:

- 10 colored pairs (Nishimoto et al., 2011) in 3 min;
- 9 grayscale pairs (K. Seeliger et al., 2018) in 2 min

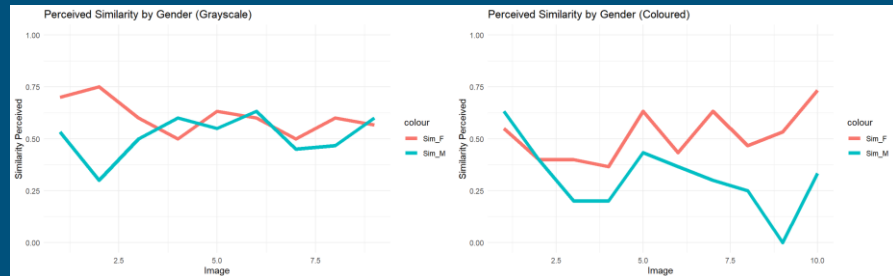
Results

Interesting results

- In general, females matched the reconstructed images with the original ones more accurately and more quickly
- Males perform better with grayscale images
- Mean similarity ratings were lower for males than females



Accuracy by Gender



Perceived Similarity by Gender
(correct trials only)

Discussion

- Our results suggest that human judgements may help improve comparability among decoding models
- Future empirical surveys should investigate:
 - Differences b/w category of objects (e.g. faces vs tools)
 - Graded judgements of similarity (e.g. shape vs taxonomic class)
 - The metrics for human judgment need further examination as well



	Matching Accuracy	Similarity Score
RMSE	$\rho = -0.12, p > 0.63$	$\rho = -0.44^*, p > 0.055$
FSIM	$\rho = -0.14, p > 0.55$	$\rho = -0.58^{**}, p < 0.01$

*Almost significant at $p < 0.05$

**Significant at $p < 0.01$